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ABSTRACT

A method has been developed for studying the feasibility of reducing annoyance due to aircraft noise by modifying the flight trajectories over a community. Numerical optimization is used to compute the optimum flight paths, based upon a parametric form that implicitly includes some of the problem restrictions. The other constraints are formulated as penalties in the cost function. Various aircraft on multiple trajectories (landing and takeoff) can be considered. The modular design employed allows for the substitution of alternate models of the population distribution, aircraft noise, flight paths, and annoyance, or for the addition of other features (e.g., fuel consumption) in the cost function. A reduction in the required amount of searching over local minima has been achieved through use of the presence of statistical lateral dispersion in the flight paths.

Three annoyance models have been studied: Noise Impact Index (NII), Level-Weighted Population (LWP), and Highly Annoyed Population Number (HAPN). It is shown that NII is not suitable for comparing different annoyance reduction strategies in any one community. Use of either LWP or HAPN as the criterion indicates that significant reductions in community annoyance are possible through this approach.

The equivalent aircraft noise concept has been developed, by which the time needed for computation of the noise levels is reduced significantly when a large variety of aircraft is under consideration. This concept has also been shown to have applications in the problem of aircraft-trajectory assignments as it relates to noise reduction.

It is found that, generally, the population distribution does not have a dominant influence in determining the optimum set of flight paths when the mixture of aircraft on those paths changes; the optimum set may have to be recomputed for a change in the flight mix.

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LIST OF SYMBOLS

\underline{a}	Arbitrary vector
$A(\underline{x})$	Annoyance function
\underline{b}	Linear coefficient (n-vector) in quadratic cost function; arbitrary vector
c	Constant term in quadratic cost function
c_1, c_2	Aircraft noise level coefficients; constants in the aircraft dynamic constraints
$c_{1,eq}, c_{2,eq}$	Equivalent aircraft noise level coefficient.
c_3, c_4, c_5	Constants in the aircraft dynamic constraints
c_{xx}	Aircraft stability coefficients
dB	Decibel
D-F-P	Davidon-Fletcher-Powell (optimization algorithm)
\underline{d}_k	Search direction in the k-th iteration
$\hat{\underline{d}}_k$	Unit vector along \underline{d}_k
E	Energy of radiated sound; error in the approximate solution \underline{x}_k
E_0	Reference sound energy
EPNdB	Effective Perceived Noise decibel
EPNL	Effective Perceived Noise Level
$f(\underline{x})$	Cost function
$\tilde{f}(\underline{x})$	Cost function including penalties
$f_{F,TOF}$	Cost of fuel and time of flight
$\underline{F}(\underline{x})$	Hessian of the cost (n x n matrix)
g	Acceleration due to gravity
$\underline{g}(\underline{x})$	Gradient of the cost function (n-vector); m-vector of constraints

LIST OF SYMBOLS (Continued)

$\underline{h}(\underline{x})$	m-vector of constraints
HAPN	Highly Annoyed Population Number
I	Sound intensity
IAF	Initial Approach Fix
I_0	Sound intensity reference level
IPNL	Integrated Perceived Noise Level
k_0, k_1, k_2, k_3	Constants in the cubic fit line search
l	Length along a flight path
l_1, l_2	Lengths of adjacent trajectory segments
l_k	Length of a rectangular population grid cell
l_T	Total trajectory length
LWP	Level-Weighted Population
m	Mass of an aircraft; number of constraints
\underline{M}	Direction modifying matrix (n x n)
n	Number of unknown parameters; number of corner point coordinates; number of population grid cells
N	Number of noise level samples per day
$N_{a/c}$	Equivalent number of aircraft
N_f	Total number of flights per day
NII	Noise Impact Index
NII'	Modified Noise Impact Index
$p(L_{dn})$	Population distribution function
$P_i(\underline{x})$	Penalty function
$P(L_{dn})_i$	Population in grid cell i

LIST OF SYMBOLS (Continued)

PNL	Perceived Noise Level
P_{Total}	Total population included in annoyance calculation
q_k	n-vector used in modified D-F-P method
q_∞	Dynamic pressure
\underline{Q}	Quadratic weighting matrix (n x n)
r	Distance (range) from an aircraft to any point
R	Radius of curvature of a flight path's ground track
S	Wing reference area
\underline{S}	Approximation of the inverse Hessian, an (n x n) matrix
V_T	Magnitude of total aircraft velocity
\bar{V}	Magnitude of average total aircraft velocity
$W(L_{dn})$	Impact intensity weighting function
$w_{F,TOF}$	Fuel and time of flight weight
w_{HAP}	Weighting function used in HAPN
w_i	Day-night weighting factor
\underline{x}	n-vector of unknown parameters
\underline{x}^*	Optimum value of \underline{x}
x	Corner point, x-coordinate
x_f	Final trajectory point, x-coordinate
x_0	Initial trajectory point, x-coordinate
x_t	Tangent point on trajectory, x-coordinate
\hat{y}	Distance between trajectories along an arbitrary direction

LIST OF SYMBOLS (Continued)

y	Corner point, y-coordinate
y_f	Final trajectory point, y-coordinate
y_0	Initial trajectory point, y-coordinate
y_t	Tangent point on trajectory, y-coordinate
z	Corner point, z-coordinate
\bar{z}	Average altitude during a turn
z_f	Final trajectory point, z-coordinate
z_0	Initial trajectory point, z-coordinate
z_t	Tangent point on trajectory, z-coordinate
α_k	Search parameter in k-th iteration
α^*	Optimum value of α_k
γ	Flight path angle
γ_c	Climb angle
γ_d	Descent angle
γ_k	Scaling factor in modified D-F-P method
δ	Final opening angle
δa_{\max}	Maximum aileron deflection
δr_{\max}	Maximum rudder deflection
$\delta r_1, \delta r_2, \delta r_3$	Constants depending upon control surface deflection limits
θ	Angle of straight-line trajectory with respect to fictitious population distribution x-axis
θ_{disp}	Angle of dispersion with respect to mean flight path
μ	m-vector of penalty weights

LIST OF SYMBOLS (Continued)

σ_y	Standard deviation in the lateral dispersion of flight paths
σ_z	Standard deviation in the vertical dispersion of flight paths
ϕ_{\max}	Maximum roll angle
$\Phi(\underline{x})$	m-vector of penalty or barrier functions
ψ	Yaw angle
ψ_{\max}	Maximum allowed yaw rate
$\underline{\omega}$	m-vector of penalty or barrier weights

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CHAPTER I

INTRODUCTION

The question to be considered is this: Is it possible to achieve a substantial reduction in community annoyance due to aircraft noise through the use of trajectory modifications? While not a complete remedy for the problem of aircraft noise, such modifications will be shown to aid significantly in its solution. What is particularly appealing about this technique of noise control is that its cost is relatively low compared with other methods. This study has produced a system that may be used to analyze the noise effect of aircraft operations on a given population and to minimize that effect (the community annoyance) to the extent possible via modifications in the flight paths.

Historical Background

The problem of severe aircraft noise began in the 1950's with the advent of the commercial jet airliners. Complaints and litigation proceedings followed soon thereafter, with the result that numerous studies were begun. As the use of jet transportation became more extensive, so did the awareness of the effects of its noise upon the environment and people living near airport terminals. Today the aircraft noise issue is perceived by the airline industry as one of its most important problems.

Analyzing the noise effects of aircraft operations is not a wholly objective matter; the physical measurements of sound levels are only part of the analysis. Numerous physical measures of sound noise levels

have been defined, e.g. L_A , L_{dn} .¹ Some of these attempt to include the frequency response of the human ear, while others try to indicate the degree of perceived annoyance. Ref. [1] is a useful compendium of these measures and rating schemes.

Less easy to quantify, and open to many interpretations, are the subjective psychological and sociological effects of the noise, collectively referred to as "noise annoyance." Several attempts have been made to relate the physical measurement to the noise annoyance. Alexander [2] examined some of the relationship models used in studies of the aircraft noise problem. Tarnopolsky [3] has studied the mental health effects from extended exposure to aircraft noise. While Borsky [4] showed that a linear relationship exists between the percentage of people "highly annoyed" and the noise exposure level (L_{dn}) in the range 55 to 80 dB, Kryter [5, p.444] indicates that the relationship is nonlinear (power law) in terms of the maximum EPNL.² More recent studies by von Gierke [6] and Goldstein [7] have used a single quantity to represent the average degree of noise annoyance to a community. This measure, the Noise Impact Index (NII) has the properties of intensivity (includes the intensity of the annoyance) and extensivity (includes the

¹ Both of these are logarithmic measures of sound power intensity. L_A is the A-weighted sound level, with the weighting based upon the frequency response of the human ear. L_{dn} is the average day-night sound level, based upon the average power density (A-weighted) during a 24-hour period. Nighttime sound levels are weighted more heavily in the average. Chapter III elaborates on these measures.

² The Effective Perceived Noise Level (EPNL) is a measure of the sound power density averaged over a short period of time (8 seconds), with spectral weighting. See Appendix A.

extent of the annoyance, i.e. the numbers of people affected). The Noise Impact Index is defined as the weighted sum of all the people affected, divided by the total affected population. Schultz [8] has shown that the proper weighting to be used for the NII is a nonlinear function of the noise exposure level (in L_{dn}).

Methods for reducing the annoyance due to aircraft operations generally lie in one of three categories: 1) noise source modifications, 2) land use control, and 3) aircraft operational control. Noise source modification is more readily used in the design stage of future aircraft than it is in existing ones, where "retrofitting" the engine nacelles with sound absorption material is quite costly. For example, the estimated cost of retrofitting a Boeing 737 with sound absorption material is \$202,000, while the cost of replacing the engines' front fans with quieter ones is \$1,000,000 [9]. Land use control involves areas experiencing the most noise annoyance near an airport. Such land areas may be purchased by the airport, or perhaps zoned as industrial areas, thus reducing the numbers of people affected. This remedy is not suitable for metropolitan areas where land for housing is in short supply.

The third method for reducing noise annoyance, aircraft operational control, requires a method of predicting noise levels that will be experienced as a result of whatever operational changes are made. Considerable effort has been made in this regard. Padula and Liu [10] examined a mathematical model which treated an aircraft as a moving prolate spheroid that scattered acoustic waves from a trailing point source. Barkhana and Cook [11] presented a model for determining the region on the ground that would experience noise above 70 EPNdB.

Dunn and Peart [12] studied five basic aircraft types: turbojet, turbofan, turboprop, V/STOL, and the helicopter, developing models of both the noise sources and the noise levels contours. These have been incorporated into computer programs for noise level estimation by Crowley et al. [13]. The statistical lateral dispersion of flight paths about the nominally specified trajectories has been analyzed by Filotas [14], who finds that this has a significant effect on noise levels on the ground fairly near to the mean flight paths. The most sophisticated model devised for predicting sound levels is the FAA Integrated Noise Model. A variety of noise ratings may be estimated based upon actual atmospheric conditions and the mixture of aircraft landing and taking-off at a particular site.

The more common methods used to reduce aircraft noise effects through operational controls are the procedures of two-segment approach and thrust cut-back on takeoff. Zalovcik [15] empirically determined that a combination of the two -- a 6 degree slide slope (angle of the flight path with respect to the ground) followed by the standard 3 degree slope, and an associated thrust reduction -- could yield a decrease in sound pressure levels (SPL) of up to 14 dB. Glick et al. [16], simulating aircraft operations near the San Jose Municipal Airport, studied similar operational modifications, including alternate routes for approaches and landings; however, no systematic method for varying the flight paths was employed.

Jacob [17] presented an analytic solution to the problem of minimizing the average effective perceived noise level for landing aircraft confined to a vertical plane. Barkhana et al. [18] showed the

feasibility of computing trajectories in three dimensions, with noise impact as part of the performance measure being minimized. Their method, a combination of dynamic simulation and the use of steepest descents to determine the aircraft control inputs, required large amounts of computer time. A somewhat different method, parameter optimization, has been shown by Rader and Hull [19] to be relevant to the computation of optimum aircraft trajectories. They employed a gradient method to find the angle-of-attack history for supersonic aircraft that results in the minimum time-to-climb trajectory at full power. A fifth-order Tschebycheff polynomial series was used to approximate the angle-of-attack history, with the polynomial coefficients as the parameters to be determined by the optimization scheme.

Jacobson et al. [20] developed a system for determining optimum aircraft landing trajectories (via parametric optimization) with respect to the NII, for Boeing 707 and 727 aircraft. In the work reported here, the system has been expanded to include both takeoff and landing paths, and a variety of commercial aircraft.

Using this system, the current study has been directed to certain essential problems of noise impact reduction through the use of operational controls. First is the matter of which has the dominant influence in determining the noise impact on a community: the particular assignment of aircraft to the various trajectories, or the population distribution. This has an important bearing on the work of the air traffic controllers, who must make the flight path assignments; recomputing the optimum trajectories for a change in the aircraft mixture is, at present, a lengthy process.

Also examined has been the problem of aircraft takeoff. Because they are less severely constrained, trajectories used in taking off are more easily modified to reduce the noise impact than are landing trajectories. The endpoint of a takeoff path is not restricted to be a specific point in space, and the rate of climb is allowed to vary more than the rate of descent in the landing case.

Another problem inherent to trajectory modifications is that of restricting the fuel consumed and/or the time of flight. Although reduced noise effects are desired, it would not be acceptable to achieve them at great expense in terms of fuel or time. A study has been conducted on the effects of including the fuel and time of flight in the performance measure.

Previous work [21] relied on NII as the measure of noise annoyance to a community. While NII acts as a good indication of annoyance per person in comparing communities of different sizes, it does not reliably measure the change in annoyance for modified trajectories in any one community. This has prompted the examination of alternate measures and their effects on the optimum trajectories thus computed.

The parameter optimization approach works well for airports of modest size, where the number of trajectories and hence, the number of parameters needed to specify them, is small. For larger airports, the number of parameters required can result in memory requirements that exceed the capabilities of the computer being used. An alternative scheme has been devised for determining the solution in this situation. In addition, the concept of "equivalent aircraft type" has been developed, in which the noise characteristics of all the aircraft on a given

trajectory are lumped together into a single "equivalent aircraft." This transformation has yielded a considerable reduction in the time required to calculate the noise levels experienced by the populace, and as a byproduct, some insight into the sensitivity of these levels to changes in the numbers of aircraft on the flight paths.

Lastly, the penalty function method of including the various constraints in the problem formulation has been examined to determine the correct weighting for the penalties.

Chapter II is an outline of the problem solution method. Following it are detailed descriptions in Chapters III to VII of the models developed and employed in the optimization software. Chapter VIII contains the results, and Chapter IX, the conclusions and recommendations.

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CHAPTER II

THE APPROACH TO THE SOLUTION

To examine the feasibility of reducing noise impact on a community using operational changes in aircraft flight paths, several major components of the system must be considered. These include: specifying aircraft trajectories, incorporating passenger, pilot, aircraft, and regulatory restrictions, and searching for "best" paths that reduce the noise impact. In this chapter, each of these areas is addressed in order to show the rationale behind the approach developed for solving the problem.

Although specific mathematical models and functions have been employed in the development, it is not claimed that they are the sole choices. The modularity of the approach allows for perhaps more accurate replacements to be made in any of the areas subsequently discussed.

Any attempt to reduce noise in a community requires some quantitative evaluation (i.e. model) of the effects of the noise. From the model may be extracted a measure, a numerical function, that quantifies the noise effect.¹ Many such models (and measures) have been established, emphasizing different noise effects. For example, the Speech Interference Level (SIL) attempts to quantify the interference of noise with speech communication, while the Percent Hearing Loss (PHL)

¹ It is preferred that the measure be a scalar quantity, a single value, for purposes of ranking different solutions to a particular noise problem. A unique method of ranking different collections of values does not exist.

measure is intended to predict that physiological effect for a given range of average noise levels. For the model of overall annoyance of a community because of noise, there are two major considerations: the intensity of the sound levels present and the extent of those levels in terms of the numbers of people exposed to them. Among the measures that attempt to indicate both the intensity and extensity of the noise is the Noise Impact Index (NII), which has been the principal measure used in this study.

The model of annoyance upon which NII is based treats extensity and intensity in the following ways: For annoyance intensity, it is assumed the psychological response to noise depends upon the average noise levels experienced. Nighttime levels are weighted more heavily than are daytime levels. The intensity of annoyance is represented by a nonlinear weighting function of the average sound level. For annoyance extensity, the NII model attempts to indicate the annoyance of the entire community exposed, by including the number of people exposed at different levels. Intensity and extensity are combined in the NII by summing the number of people exposed to a given noise level, multiplied by the intensity weighting of that level. This sum is then normalized by the total population exposed to annoying levels, giving an indication of the average annoyance per person in the community.

Once the annoyance measure has been chosen, a practical way to evaluate it must be devised. In the use of NII, this entails the use of two mathematical models: a noise level model and a population distribution model. Under ideal conditions, it may be assumed with reasonable accuracy that the noise levels from aircraft far away (several hundred

meters) depend only upon the distance from the aircraft.² Since the details of that dependence (and the flight paths) are known for all the aircraft operating in the area, it is possible to compute the average noise level at any point on the ground. The spatial distribution of people around an airport may be approximated by dividing the region into a number of smaller areas (cells). United States Census Data can then be used to obtain the number of people in each cell. For given sets of trajectories and the associated aircraft on them, the average sound level in each cell can be calculated. It is then a straightforward matter to calculate the NII for those given conditions.

With a practical implementation of the annoyance measure established, it is then feasible to begin the search for trajectories that result in the minimum possible value of the measure. Traditionally, aircraft trajectory modification has been performed using the calculus of variations or dynamic programming to solve the equations of motion of the aircraft, subject to certain constraint and minimize some criterion referred to as the cost function. Typically the cost function has a physical significance, such as the total fuel consumed by the aircraft flying these trajectories. While this method of dynamic simulation works well for aircraft confined to a vertical plane (see for example Erzberger [22]), the complexity of the equations in three dimensions results in a large increase in the amount of time needed for computation.

² This also assumes given power and control settings on the aircraft.

An alternative technique has been applied here. Each trajectory is modelled as a specific functional in parametric form. The optimization process then consists of searching for the set of parameters that specifies the optimum set of trajectories, i.e., the set that results in the minimum value of the cost function. In this transformed version of the problem, the number of variables to be determined, and hence, the computation time, is smaller than in the original form, but the presence of various restrictions on the trajectories now creates some difficulty. Whereas these restrictions may be readily handled in the simulation technique and in fact are often automatically built into the problem formulation, they must be explicitly included when employing the parameter optimization scheme.

Two classes of optimization methods exist for solving this type of problem. The first, referred to as primal methods, searches for the set of parameters that will result in the optimum solution, but the search is conducted over only those sets of parameters that specify acceptable solutions, i.e., no restriction is violated. The second class, approximating methods, searches over any sets of parameters without restriction. Inclusion of the problem constraints is accomplished by adding terms (penalty or barrier functions) to the cost function. The penalty functions are such that the cost function increases when constraints are violated; if no violation occurs, then no penalties are added. The unrestricted optimization will then, ideally, avoid restricted sets of parameters in its search for the lowest value of the cost function. For reasons given in Chapter V, an approximating method was chosen for this work. The unconstrained optimization

method used was the one of Davidon, Fletcher, and Powell, modified by Luenberger [23] to improve its performance in a variety of problems.

As for the formulation of the cost function, clearly the minimization of annoyance is the true objective, given that the presence of the penalties is a necessity; however, other concerns may be included, weighted in proportion to their importance. For example, this work contains a study of the effects upon the solution if the total fuel consumed and time of flight are included.

A suitable parametric form for mathematically describing the trajectories is one that requires a small number of parameters, but that still has sufficient flexibility to satisfy all of the constraints. A good choice for such a trajectory model is a set of linear segments joined by smooth arcs. The types of flight paths that result from this model are similar to those currently flown by commercial airliners. This is desirable from the perspective of the pilots, who must not be burdened with overly complicated flight paths. A trajectory is completely specified once the coordinate of the intersections of its linear segments are determined. This is the result of the optimization process.

Included in the constraints are: the dynamic capabilities of the aircraft, considerations of passenger comfort, maintenance of safe separation distances between two trajectories, avoidance of frequent high noise level flights over any one area, and a limit to the number of maneuvers required of the pilots. If some constraints place different requirements upon some property of the trajectory set, then the most restrictive of the constraints prevails.

In the following chapters, the models and measures used in the solution method are treated in detail. Chapter III develops the annoyance model and considers alternative to NII. Chapter IV derives the trajectory model. The optimization method is described in Chapter V. Finally in this development, the cost function, including the formulation of the constraint penalties, is discussed in Chapter VI.

CHAPTER III

THE ANNOYANCE MODELS

As discussed in the previous chapter, a desirable community noise annoyance model includes both the intensity and extensity of the annoyance and has associated with it a scalar measure that quantifies the annoyance. Three measures are examined in the following sections: Noise Impact Index (NII), Level-Weighted Population (LWP), and Highly Annoyed Population Number (HAPN). First, however, models of aircraft noise level and population distribution, which are required by the annoyance models, are developed.

Aircraft Noise Model

For most types of aircraft, the sound level or intensity (power per unit area) in dB units may be expressed as a function of the distance r from the plane:

$$L = c_1 - c_2 \log_{10} r \quad (3-1)$$

where c_1 and c_2 are constants for a given type of aircraft operating with given power and control settings.

The human ear converts sound energy into nerve impulses with some filtering of certain frequency ranges. In order to account for this, a measure of sound level that includes an approximation of the ear's frequency response has been devised (see Ref. [1]). Known as the A-weighted level, L_A indicates the "instantaneously" perceived¹

¹ Actually, the time required to perceive a change in sound level is about 0.5 sec. See Kryter [5, p.3].

sound level. There are, however, other measures that incorporate the duration of the noise or the presence of high amplitude pure tones in predicting the judged sound level. Of particular interest here is a scale known as Effective Perceived Noise Level (EPNL), whose units are EPNdB. In developing the noise model for this work, data from the Federal Aviation Administration's Integrated Noise Model (see Ref. [24]) has been employed. These relate the noise level in EPNdB to the distance from a particular type of aircraft. Figure 3.1 shows an example of the data for a DC-8-30 jet approaching for landing on a 3° glide slope. The solid line is the least squares fit of the data. It has the relation

$$\text{EPNdB}_{\text{DC-8-30}} = 155.74 - 28.32 \cdot \log_{10} r. \quad (3-2)$$

Similar fits were obtained for other types of aircraft to be studied. The results are shown in Table 3.1.

All three of the annoyance indicators to be examined use L_A for calculating the annoyance intensity. The approximate conversion is

$$L_A \approx \text{EPNL} - 13\text{dB}. \quad (3-3)$$

This relation is derived in Appendix A, along with the procedure for calculation noise levels when multiple sources are present.

The Population Model

In order to assess the extensity of noise annoyance, a model of the population distribution is required. Some annoyance measures are cast in terms of population distribution as a function of noise level; however, for computational purposes, it is better to start with the spatial distribution of people. This is compatible with the noise model

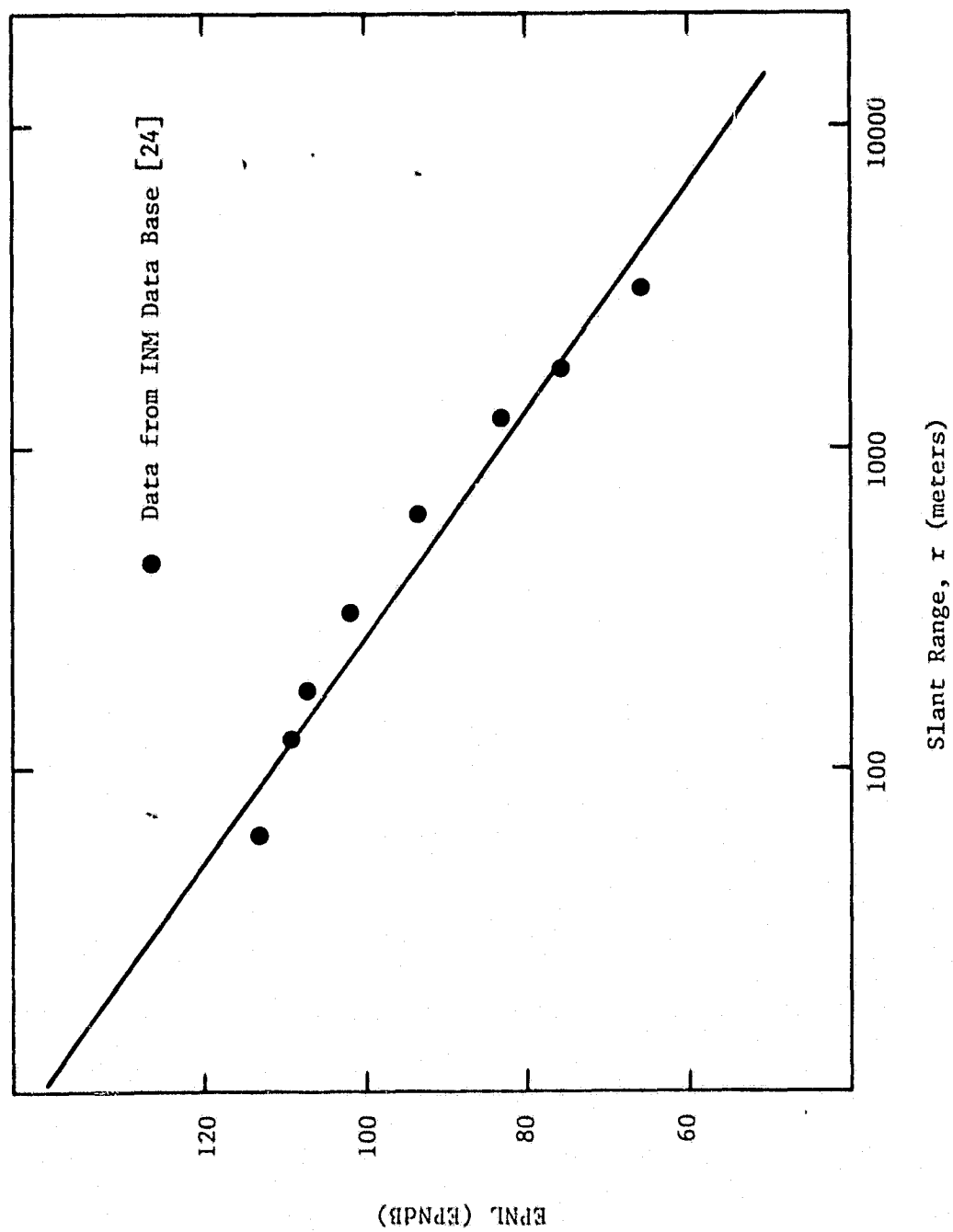


Figure 3.1 EPNL vs. Slant Range for DC-8-30, 3° Approach

Table 3.1

L_A vs. Slant Range for Some Commercial Aircraft[†]

$$L_A = c_1 - c_2 \cdot \log_{10} r, \quad r = \text{slant range (meters)}$$

3° Glide slope approaches

A/C Name	c_1	c_2
DC-8-30	155.74	28.32
DC-9 w/SAM	142.28	24.65
DC-10-10	146.22	28.80
707 w/SAM	135.11	23.60
720	140.88	21.95
727-200	139.47	22.66
727 w/SAM	124.28	18.74
737-100/200	154.74	29.52
737 w/SAM	147.31	26.96
747-200	139.89	24.78
L-1011	136.94	24.49
A-300	166.47	37.20
BAC-111	150.72	27.06
VC-10	142.65	22.95
CV-990	157.34	27.16
SST	136.22	15.96

[†]Data obtained from Ref. [24] and reduced by least square-error fits.

already developed (the two are appropriately combined later in this chapter) and has the advantage of being independently calculable (i.e. no noise calculations are required).

As described in Chapter II, the distribution is approximated by determining the number of people inside each cell of a grid that overlays a map of the community. The grid should contain a sufficient number of cells to insure good resolution of the distribution, but not so many cells that large numbers of calculations occur in the annoyance measure evaluation. A scheme that satisfies both of these conditions is one in which the size of the cells grows with distance from the airport. The small cells near the airport provide better resolution in a region where the noise levels are changing more rapidly over the ground.

As an example, Figure 3.2 shows the grid used for the Patrick Henry Airport at Hampton, Virginia. (Some of the grid lines in the center are not shown, so that the runways will be visible.) Here, the cells adjacent to the runways are 400 meters on a side. The length of a cell side grows both horizontally and vertically, according to the relation

$$l_{k+1} = 1.3 \cdot l_k \quad (3-4)$$

where l_k is the length of cell k in either direction. Covering a square region 32,000 meters on a side, the grid contains 576 cells, approximately 11 times fewer than if all the cells were 400 meters on a side.

Alternative grid schemes also may be used. Some examples are: concentric rings divided into sectors by radial lines (with the ring radii either constant, or growing with distance), and square cells of constant dimensions (if the number required is acceptable).

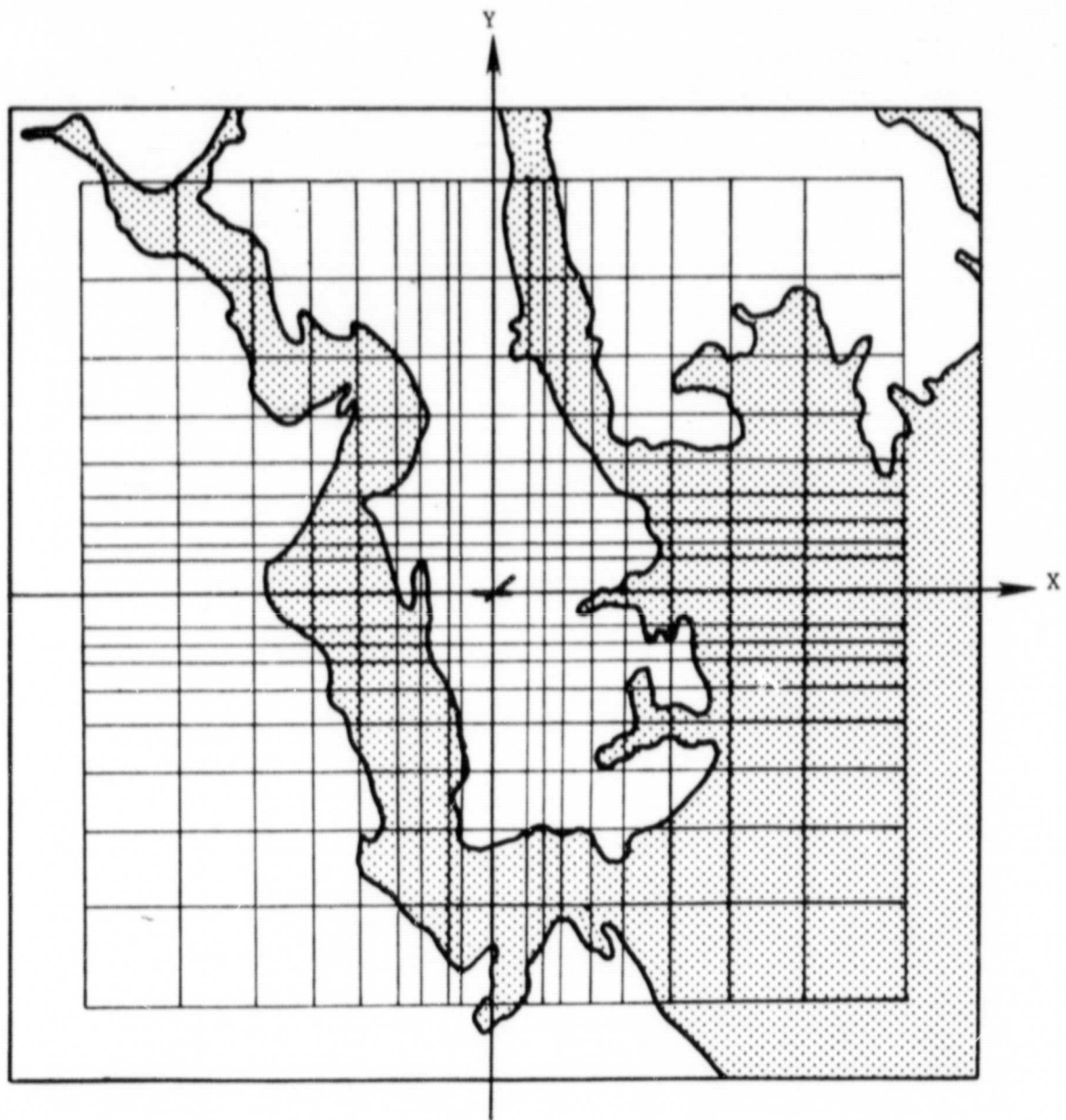


Figure 3.2 Population Grid at Patrick Henry Airport (Partial)

Once the grid scheme is chosen, the number of people in each grid cell may be determined using United States Census Data in a computer program called SITE II [25]. It requires as input a reference point in latitude and longitude, and coordinates with respect to the reference that specify the boundaries of each cell. The population in each cell is output, along with a variety of demographic information. Although only the population data have been used in this study, the other data could be useful in an annoyance measure that includes social and economic information.

In each grid cell, it is presumed that the population is concentrated at the geometric centroid of the cell. This is done to simplify the noise level calculations, i.e., all the people will receive the noise level present at the centroid. Chang [26, p.29] has shown that this is a good approximation in that the aircraft trajectories determined using this method are not significantly different from those calculated using more exact methods.

Level-Weighted Population (LWP)

This indicator of community annoyance, devised by the National Academy of Sciences Committee on Hearing, Bioacoustics and Biomechanics [27] assumes that the intensity of annoyance due to noise varies with the noise level. The relation between annoyance and noise level is characterized by the intensity weighting function $W(L_{dn})$, where L_{dn} is

the average day-night noise level. L_{dn} is defined as²

$$L_{dn} = 10 \times \log_{10} \frac{\sum_{t=1}^N w_t 10^{L_{A,t}/10}}{N} ; \quad (3-5)$$

$$w_t = \begin{cases} 1 & \text{for noise between 7 A.M. and 10 P.M.} \\ 10 & \text{for noise between 10 P.M. and 7 A.M.,} \end{cases}$$

N = Number of noise level samples taken in 24 hours,

$L_{A,t}$ = t -th sample of the noise level, A-weighted.

A plot of $W(L_{dn})$ is shown in Figure 3.3. The analytic expression for it is:

$$W(L_{dn}) = \frac{3.36 \times 10^{-6} \cdot 10^{0.103 \cdot L_{dn}}}{0.2 \cdot 10^{0.03 L_{dn}} + 1.43 \times 10^{-4} \cdot 10^{0.08 L_{dn}}} \quad (3-6)$$

This function is based upon a collection of social surveys of annoyance caused by various types of noise [8]. Although this synthesis represents possibly the best available data for annoyance prediction, its mathematical model has the property of being unbounded. This is acceptable, though, if there is a bound to the largest value of L_{dn} likely to be encountered.

Level-Weighted Population is defined by

$$LWP = \int W(L_{dn}) p(L_{dn}) d(L_{dn}) \quad (3-7)$$

where $p(L_{dn})$ is the number of people receiving noise between L_{dn} and $L_{dn} + d(L_{dn})$. The integration is performed over the range of L_{dn} considered to be annoying. By convention, the lower limit is 55 dB;

² L_{dn} relies upon the concept of energy addition, which is developed in Appendix A.

SOUND LEVEL WEIGHTING FUNCTION
FOR OVERALL IMPACT ANALYSIS

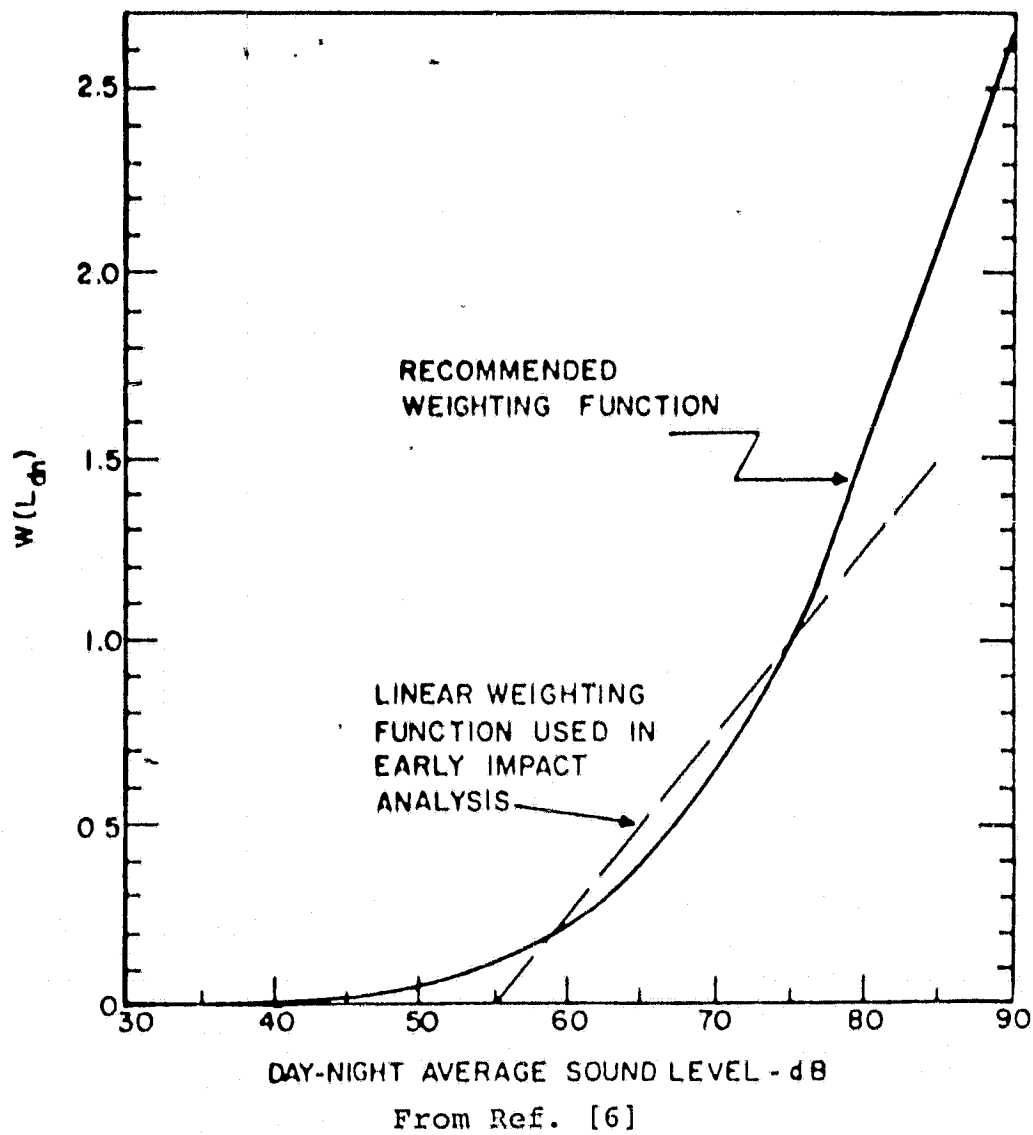


Figure 3.3 Impact Intensity Weighting Function

the upper limit is the highest level present. Evaluation of this integral is facilitated by approximating it as a discrete sum

$$LWP \simeq \sum_{i=1}^n W(L_{dn})_i P(L_{dn})_i \quad (3-8)$$

where i is the index of a population group, $P(L_{dn})_i$ is the number of people in group i , L_{dn} is the noise level they receive and $W(L_{dn})_i$ is the intensity weighting for that L_{dn} .

Noise Impact Index (NII)

This measure, also developed by the National Academy of Sciences, is defined as

$$NII = \frac{LWP}{\int P(L_{dn}) d(L_{dn})} \quad (3-9)$$

where the symbols and limits of integration are the same as in Equation (3-7). Again, a discrete sum is usually used to approximate the integral:

$$\int P(L_{dn}) d(L_{dn}) \simeq \sum P(L_{dn})_i$$

= total population exposed to annoying noise levels.

The purpose of the denominator is to normalize LWP, making NII an indicator of the average annoyance per person in the community. Such an indicator is quite useful when comparing noise problems in communities with widely different populations: (LWP is not a good choice for use in such comparisons because it tends to grow with population size.) There is, however, a serious problem with NII regarding its use in comparing different strategies for noise reduction in any one community. As an example of the problem, consider a simple population distribution of just two clusters of people, as shown in Figure 3.4. The people

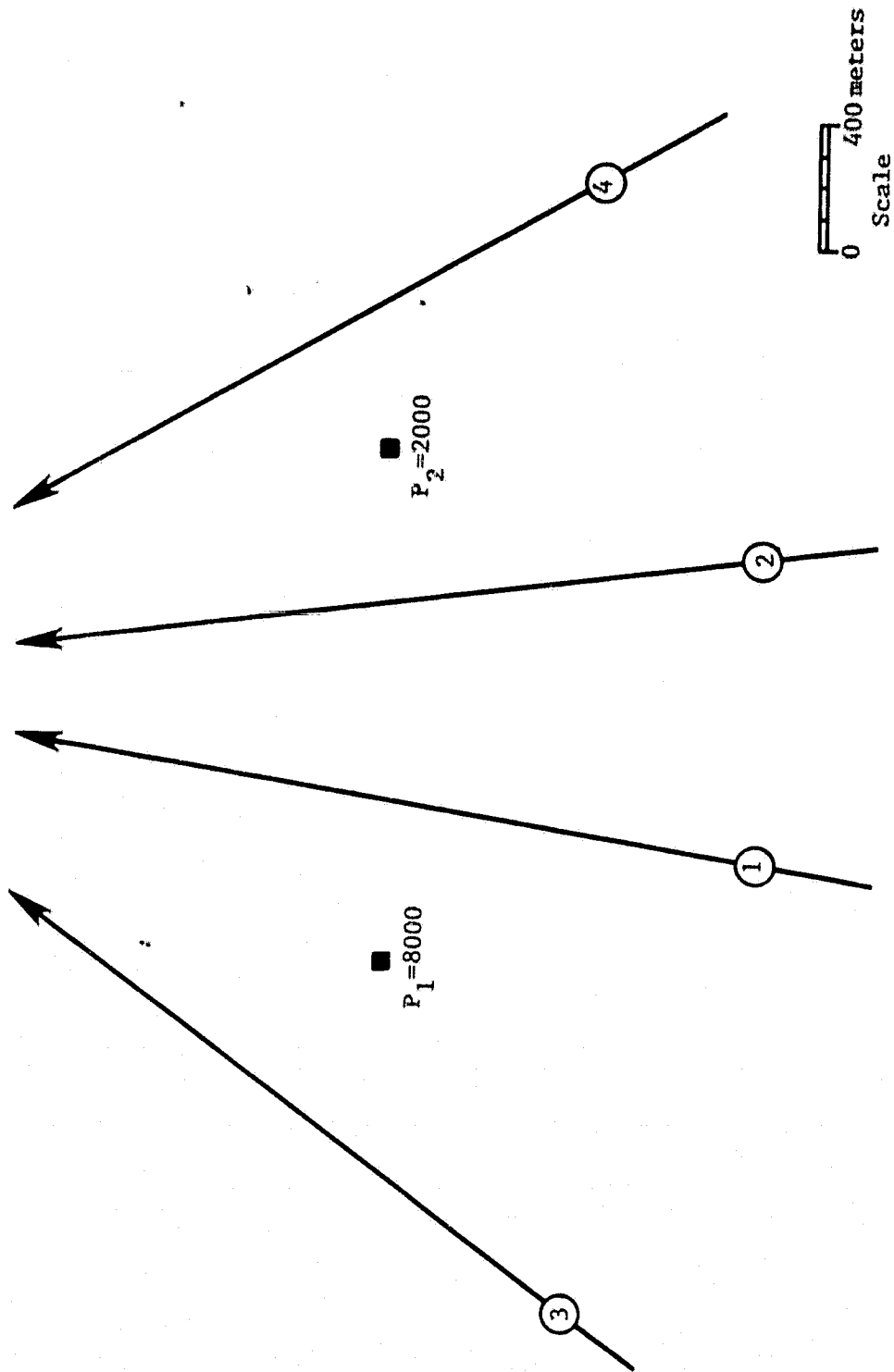


Figure 3.4 A Simple Population Distribution and the Problem With NII

receive noise from aircraft flying on one of the tracks (ground projections) shown. Table 3.2 gives the results for each of the tracks.

Tracks 2, 3, and 4 all have lower LWP than Track 1, but the NII values rank differently. Track 2 has the lowest NII value and Track 4, with the lowest LWP value, has the highest NII. The problem occurs because both LWP and the denominator in Equation (3-9) both change, but by different factors. Clearly, the use of Track 4, where 2,000 people receive an L_{dn} of 67dB is superior (in terms of total community annoyance) to the use of Track 3, where 8,000 people receive an L_{dn} of 64dB. The LWP measure indicates this intuitive choice, while the NII shows the opposite. In Chapter VIII, it will be demonstrated that a similar situation occurs for a real population distribution.

Highly Annoyed Population Number (HAPN)

Perhaps a more tangible indicator of community annoyance would be desirable. An example of such an indicator is HAPN, which attempts to measure the total number of people who are highly annoyed.

The model of annoyance intensity used by HAP differs from that of LWP and NII in two ways: the sound level scale used is maximum dB(A), and the intensity weighting function is:³

$$w_{HAP} = \begin{cases} 6.5885 \times 10^{-12} \cdot (L_{Amax})^{5.957603} \\ 6.5885 \times 10^{-12} \cdot L_{Amax} \geq 75.3 \text{ dB(A)} \end{cases} \quad (3-10)$$

³ This weighting was obtained by a least square-error fit of data in Kryter [5, p.444].

Table 3.2
Comparison of NII vs. LWP (Simple
Population Distribution)

Track	1	2	3	4
	L_{dn}	L_{dn}	L_{dn}	L_{dn}
$P_1 = 8000$	64	56	64	49.5
$P_2 = 2000$	56	64	49	.67
LWP	3244	1874	2960	1005
NII	0.3244	0.1874	0.3701	0.5026
P_{TOT}^{\dagger}	10,000	10,000	8,000	2,000

† Total population receiving annoying levels (>55 dB)

where

$$L_{Amax} = \begin{cases} \text{maximum A-level received, 7 A.M. to 10 P.M.} \\ \text{maximum A-level received} + 10\text{dB, 10 P.M. to 7 A.M.} \end{cases}$$

W_{HAP} is interpreted as the fraction of people receiving the noise who will be highly annoyed.

The definition is

$$HAPN = \frac{\sum_{i=1}^n \{W_{HAP}(L_{A,max,day})_i \cdot P_i + W_{HAP}(L_{A,max,night})_i \cdot P_i\}}{P_{Total}} \quad (3-11)$$

where the day-night notation is as indicated in Equation (3-5), P_i is the i -th population group, n is the number of groups, and P_{Total} is the total population being considered. Note that daytime and nighttime levels are treated separately, so that it is possible for some population groups to be included twice. Dividing by the total population does not present the same problem here as does the denominator of Equation (3-5), since P_{Total} is a constant for a given community.

CHAPTER IV

THE TRAJECTORY MODEL

Having established in Chapter II that each trajectory should be determined by specifying the unknowns in a general parametric expression, the next matter is to choose that expression, using a form that is compatible with the restrictions involved in the problem. In developing a suitable expression, or model, it is helpful first to examine the sorts of flight paths that are currently used in the region near an airport terminal (within 30,000 meters, where the noise levels are significant). Next, all of the restrictions that pertain directly to the trajectories must be inspected and cast in mathematical terms. Finally, a choice of the parametric form of the flight paths may be made, and the restrictions restated in terms specific to that form.

Typical Flight Paths

A. Landing

In general, an airport will have several runways and be receiving flights from a number of directions. For reasons of safety and organization (from the aspect of air traffic controllers), the number of paths available into the runways is limited by requiring that each aircraft first fly to a reporting point (also referred to as an arrival fix or an entry point). From there it proceeds to its assigned runway, with the requirement that as it passes over a point called the inner marker, its heading must be toward the runway and remain so until touchdown. The direction of all traffic, taking off and landing, on a

given runway is fixed at any given time, determined chiefly by the wind velocity in the area. Figure 4.1 is a reproduction of an approach "plate" used by pilots with instrument landing systems (ILS) in the aircraft. The projection of the trajectory onto the ground is referred to as the "ground track." This particular plate, for Patrick Henry International Airport at Hampton, Virginia, indicates two initial approach fixes (IAF's), Swing and Franklin, for use with the major runway. The entry points and inner markers conveniently serve as endpoints¹ for the landing trajectories that are to be computed.

B. Takeoff

Takeoff paths are considerably less restricted than are landing paths. An aircraft taking off may execute a turn before crossing over the inner marker. It may then proceed along a path that is not required to pass through a reporting point (or "exit fix"). Figure 4.2 shows a typical takeoff path (ground track) for Lambert-St. Louis International Airport.

Except for their different endpoint conditions, landing and takeoff ground tracks are seen to be rather similar: each consists of a combination of straight segments and smooth turns. As for the vertical projection of each type of path (called the "profile"), landing aircraft usually do not descend with a glide slope of more than three degrees, while those taking off may climb at angles as great as fifteen degrees.

¹ A fixed entry point is not required for aircraft with more sophisticated navigation equipment, such as VOR/DME (VHF Omni-Range/Distance Measuring Equipment). The flight path model should therefore allow for moveable entry points.

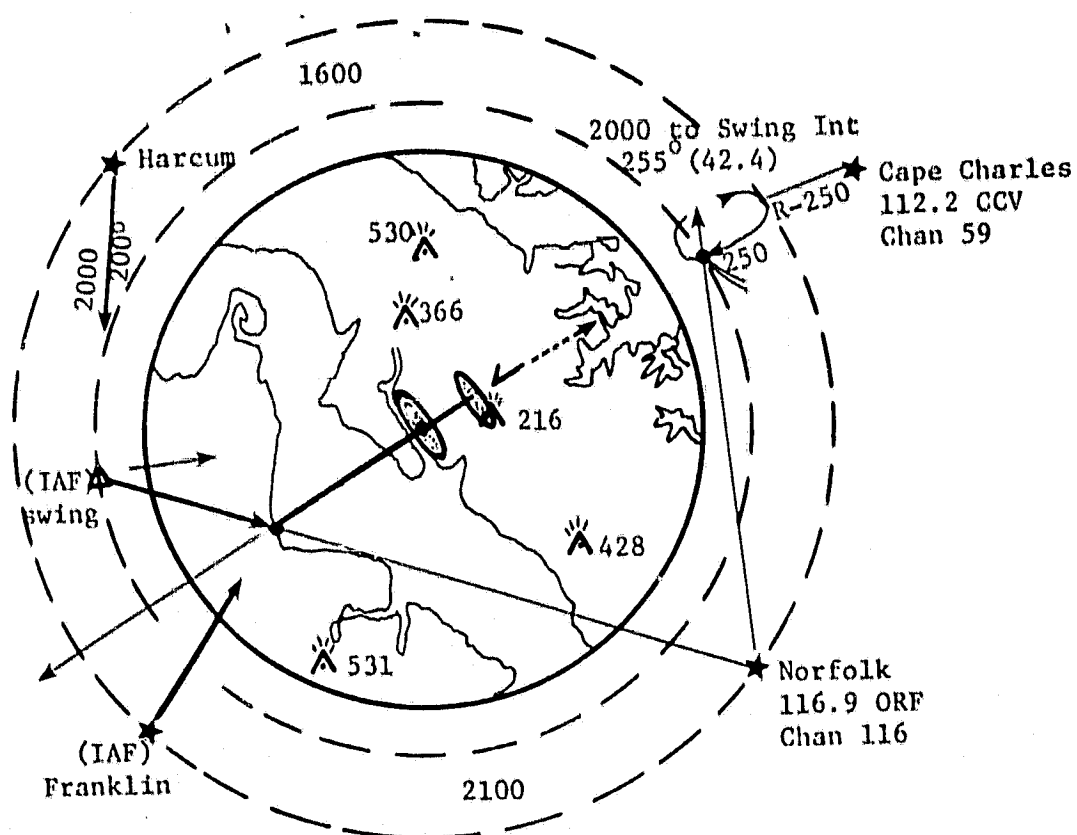


Figure 4.1 ILS Landing Approach at Patrick Henry Airport

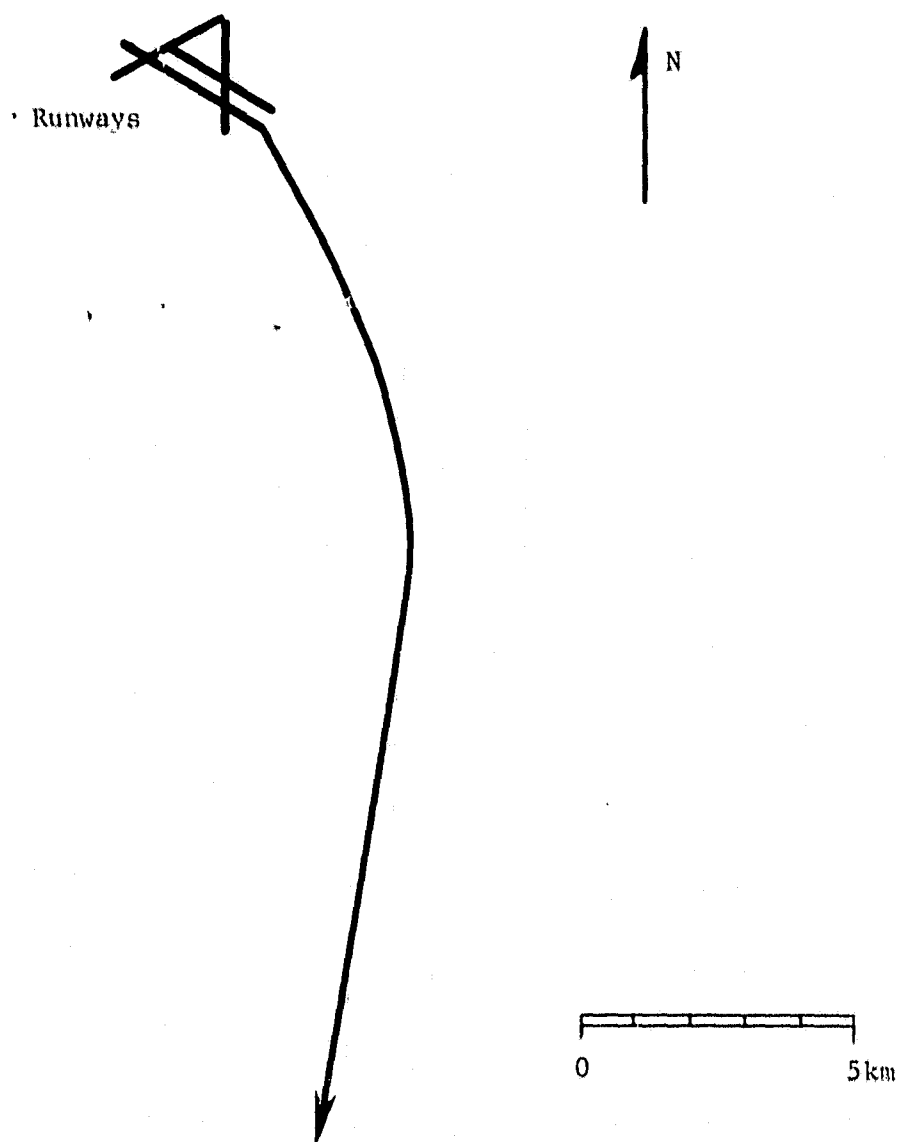


Figure 4.2 Typical Takeoff Trajectory (Ground Track)

Restrictions

The restrictions that pertain directly to the feasibility of the trajectories can be grouped into four categories: aircraft dynamics, passenger comfort, statutory regulations, and piloting considerations. A fifth constraint, concerning the maximum allowable number of high-level noise events for a given population grid cell, has been included in this work.

A. Aircraft Dynamic Constraints

The aircraft dynamic constraints limit a trajectory to one which given types of aircraft are capable of flying. The restrictions may be stated in terms of fixed properties of the aircraft and maximum allowable control surface deflections. In the event of several restrictions on any one quantity, the most severe will prevail. The following expressions, developed from linear equations of aircraft motion [28], are divided into lateral (parallel to the locally flat ground) and longitudinal (in a vertical plane) constraints.

1. Lateral Constraint

$$\dot{\psi}_{\max} \leq (c_1 + c_2 c_3) \min(\delta r_1, \delta r_2, \delta r_3) \quad (4-1)$$

where

$$\delta r_1 \leq \frac{\phi_{\max}}{c_4 + c_2 c_5}$$

$$\delta r_2 \leq \delta r_{\max}$$

$$\delta r_3 \leq \frac{\delta a_{\max}}{c_2}$$

$$c_1 = - \frac{c_{1\delta r} c_{n\beta} - c_{n\delta r} c_{1\beta}}{c_{1\beta} \bar{c}_{nr} - c_{n\beta} \bar{c}_{1r}}$$

$$c_2 = - \frac{C_{n\delta r} C_{l_r} + C_{l\delta r} C_{n_r}}{C_{n\delta a} C_{l_r} + C_{l\delta a} C_{n_r}}$$

$$c_3 = - \frac{C_{l\delta a} C_{n_\beta} - C_{n\delta a} C_{l_\beta}}{C_{l_\beta} \bar{C}_{n_r} - C_{n_\beta} \bar{C}_{l_r}}$$

$$c_4 = \{C_{y\delta r} (\bar{C}_{l_r} C_{n_\beta} - C_{l_\beta} \bar{C}_{n_r}) + C_{l\delta r} [C_{y_p} \bar{C}_{n_r} + C_{n_\beta} (\bar{m} - \bar{C}_{y_r})] \\ + C_{n\delta r} [C_{l_\beta} (\bar{m} - \bar{C}_{y_r}) + C_{y_\beta} \bar{C}_{l_r}]\} / \frac{mg}{q^\infty S} (C_{l_\beta} \bar{C}_{n_r} - C_{n_\beta} \bar{C}_{l_r})$$

$$c_5 = \{C_{y\delta a} (\bar{C}_{l_r} C_{n_\beta} - C_{l_\beta} \bar{C}_{n_r}) + C_{l\delta a} [C_{y_\beta} \bar{C}_{n_r} + C_{n_\beta} (\bar{m} - \bar{C}_{y_r})] \\ + C_{n\delta a} [C_{l_\beta} (\bar{m} - \bar{C}_{y_r}) + C_{y_\beta} \bar{C}_{l_r}]\} / \frac{mg}{q^\infty S} (C_{l_\beta} \bar{C}_{n_r} - C_{n_\beta} \bar{C}_{l_r})$$

with

$\dot{\psi}_{\max}$ = maximum yaw rate

ϕ_{\max} = maximum roll angle

δr_{\max} = maximum rudder deflection

δa_{\max} = maximum aileron deflection

m = mass of aircraft

g = acceleration due to gravity

S = wing reference area

q^∞ = dynamic pressure

C 's = aircraft stability coefficients²

\bar{C} 's = normalized stability coefficients²

$\bar{m} = \frac{mV_T}{q^\infty S}$, V_T = total velocity

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² For an explanation of these stability coefficients, see Ref. [29].

The right-hand side of Eqn. (4-1) will be a constant for a given type of aircraft, and the left side may be written in terms of the function that specifies the trajectory. In Figure 4.3a, the ground track is described in Cartesian coordinates as $y = f(x)$. With respect to the arbitrary x -axis, the slope of the flight path is dy/dx , with corresponding angle

$$\psi = \tan^{-1} dy/dx \quad (4-2)$$

however the quantity of interest is the rate at which that angle is changing (also the yaw rate of the aircraft):

$$\dot{\psi} = \frac{d}{dt} \tan^{-1} dy/dx = \frac{\partial}{\partial x} (\tan^{-1} \frac{dy}{dx}) \frac{dx}{dt} \quad (4-3)$$

But dx/dt is the x -component of the aircraft's velocity (with respect to the ground-fixed coordinate system):

$$V_x = \frac{dx}{dt} = \bar{V} \cos \psi = \bar{V} \left(\frac{1}{[1 + (dy/dx)^2]^{1/2}} \right), \quad (4-4)$$

where \bar{V} is the average, total velocity, so that

$$\dot{\psi} = \frac{(d^2y/dx^2)\bar{V}}{[1 + (dy/dx)^2]^{3/2}} = \frac{\bar{V}}{R(x)} \quad (4-5)$$

where $R(x)$ is the radius of curvature. The constraint (4-1) may now be written

$$\frac{[1 + (dy/dx)^2]^{3/2}}{d^2y/dx^2} = R(x) \geq \frac{\bar{V}}{(c_1 + c_2 c_3) \min(\delta r_1, \delta r_2, \delta r_3)} \quad (4-6)$$

2. Longitudinal Constraint

$$\gamma_{d_{\max}} \leq \gamma \leq \gamma_{c_{\max}} \quad (4-7)$$

where γ is the angle of the flight path with respect to the horizontal, and $\gamma_{d_{\max}}$ and $\gamma_{c_{\max}}$ are the maximum allowable angles of descent and

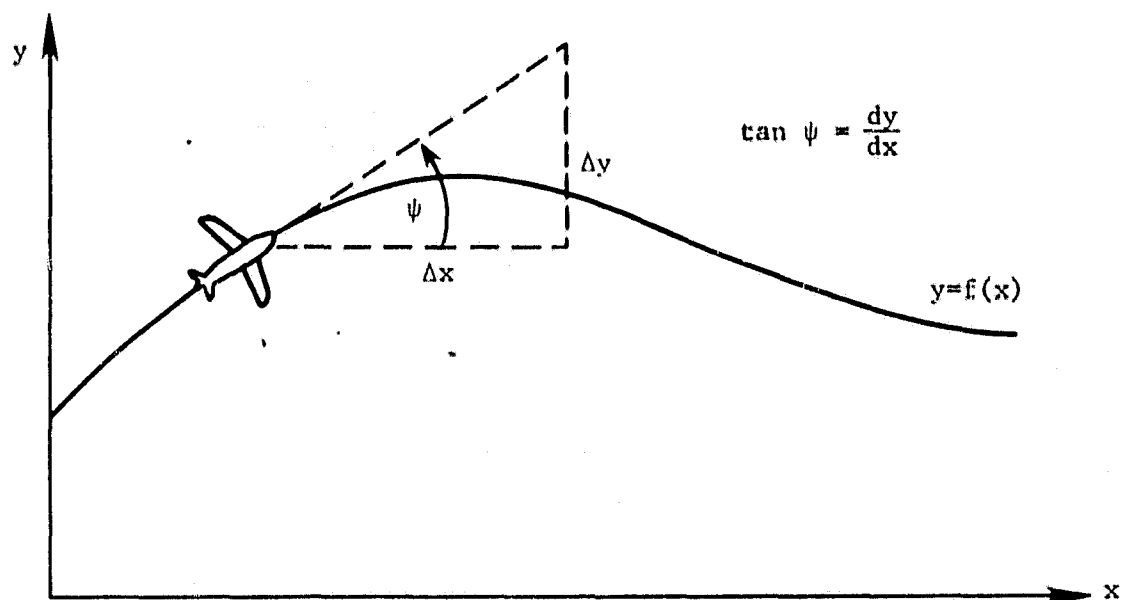


Figure 4.3a Horizontal Flight Path Angle

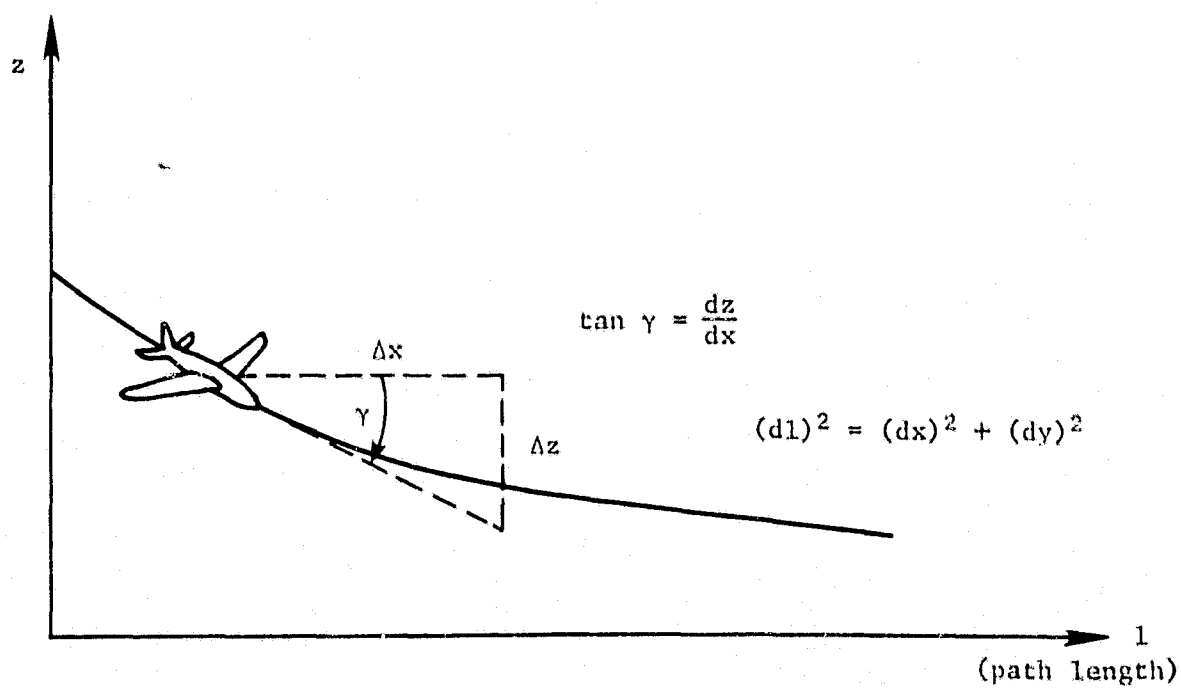


Figure 4.3b Vertical Flight Path Angle

climb, respectively. Referring to Figure (4-3b), it is seen that, in cartesian coordinates

$$\tan \gamma = \frac{dz}{dl} \quad (4-8)$$

so that Eqn. (4-7) may be rewritten

$$\tan \gamma_{d_{\max}} \leq \frac{dz}{dl} \leq \tan \gamma_{c_{\max}} \quad (4-9)$$

B. Passenger Ride Quality Constraints

The passenger comfort, or ride quality, constraints limit the operation of an aircraft to those maneuvers found statistically to be acceptable to the passengers. An expression of these limitations (see Ref. [30]) for an aircraft executing a coordinated turn³ with climb or descent is

$$0.0665 |\phi_{\max}| + 0.07 |p_{\max}| + 0.098 |\theta_{\max}| - 0.118 |\delta_{\max}| \leq C, \quad (4-10)$$

$$C = \begin{cases} 0.76 & \text{for 90\% passenger satisfaction} \\ 1.76 & \text{for 80\% passenger satisfaction,} \end{cases}$$

where ϕ_{\max} = maximum roll angle (degrees)
 p_{\max} = maximum roll rate (degree/sec)
 θ_{\max} = maximum pitch angle (degrees)
 γ_{\max} = maximum flight path angle (degrees).

For a coordinated turn and considering only the roll angle part of Eqn. (4-10), the constraint may be written

$$\frac{\bar{V}^2}{R} \leq g \frac{\sin \phi}{\cos \phi} = C_2 g \quad (4-11)$$

³ A coordinated turn is one in which no sideslip occurs, i.e. the component of gravitational force along a wing provides the centripetal force necessary to maintain the particular turning radius being used.

or, using derivatives of the trajectory function:

$$\frac{[1 + (dy/dx)^2]^{3/2}}{d^2y/dx^2} \geq \frac{\bar{V}^2}{C_2 g} \quad (4-12)$$

where $C_2 = \begin{cases} 0.19 & \text{for 90\% passenger satisfaction} \\ 0.50 & \text{for 80\% passenger satisfaction} \end{cases}$

\bar{V} = average total velocity

g = acceleration due to gravity

ϕ = roll angle.

C. Regulatory Constraints

Regulations affecting allowable flight paths may pertain either to safety, governmental security, or noise annoyance. Safety restrictions involve single trajectories (e.g. avoidance of low altitude obstacles) and multiple paths (e.g. maintenance of safe separating distances); military installations often prohibit general and commercial aviation flights from passing too closely for security reasons; and there may be certain areas (e.g. schools and hospitals) which are particularly sensitive to noise from nearby aircraft.

Two safety restrictions that apply to every airport have been included in this study. They are:

1. A safe distance should be maintained between any two trajectories as far as is possible. Because several trajectories may share a single runway, the region in which this restriction applies often excludes an area within several miles of the airport. The minimum allowed separating distance used in this work was 800 meters.

2. No landing aircraft should be required to execute a turn of more than 45 degrees, as it crosses the inner marker, in order to come into alignment with the runway. This is referred to here as the final heading requirement.

Restrictions that are related to special areas to be avoided, for security or annoyance reasons, are particular to the given airport community being considered. Although not included in this study, these constraints could be formulated and contained in the problem as described in Chapter VI.

D. Pilot Operating Constraints

It is possible that the optimum trajectories computed would require more maneuvering of an aircraft than a pilot can or is willing to perform, especially in the near terminal area, where his attention is required on a number of matters. The constraints then, are to limit the number and complexity of the turning maneuvers on the flight paths.

E. Threshold Noise Constraint

A difficulty with all three of the annoyance measures discussed in Chapter III (and other ratings) is that an "improvement," as determined by these measures, can be made by exchanging a large number of people exposed to medium noise levels for a small group exposed to extremely high levels. To avoid this possibility, a constraint is imposed that prohibits any area⁴ (grid cell) from receiving

⁴ For some cells near an airport, the high levels may be unavoidable due to runway locations. These cells must be excluded from the restriction.

maximum noise levels ($L_{A,max}$) during a flyover, at or above a "threshold" level, more than N_{max} times per day. The threshold level and value of N_{max} chosen should be as small as possible, without overly restricting the possible trajectories.

Flight Path Model

In selecting an appropriate parametric form for specifying a trajectory, it is necessary to weigh the benefits of a large number of parameters --more degrees of freedom and hence, more flexibility -- against the attendant increase in computation needed to determine the values of those parameters. Chang [26], for example, reported that using only five terms in a Fourier series representation of aircraft landing trajectories (the unknown parameters were the Fourier coefficients) led to overly complicated flight paths, while increasing the number of terms would have resulted in unacceptable computational requirements without necessarily simplifying the paths.

The model chosen for a flight path in this work avoids that problem by directly relating the maximum number of turning maneuvers to the number of parameters to be determined. Each trajectory is represented by a chain of linear segments joined by helical arcs (circular in the ground plane, linear in the profile). Figures 4.4a and 4.4b show an example of a path with three segments. The two intersections of the linear segments in Figure 4.4a are called "corner points;" the cartesian coordinate of these points, along with those of the other trajectories, make up the set of parameters to be determined. In the ground plane, a circular arc with minimum radius (determined by the dynamic and

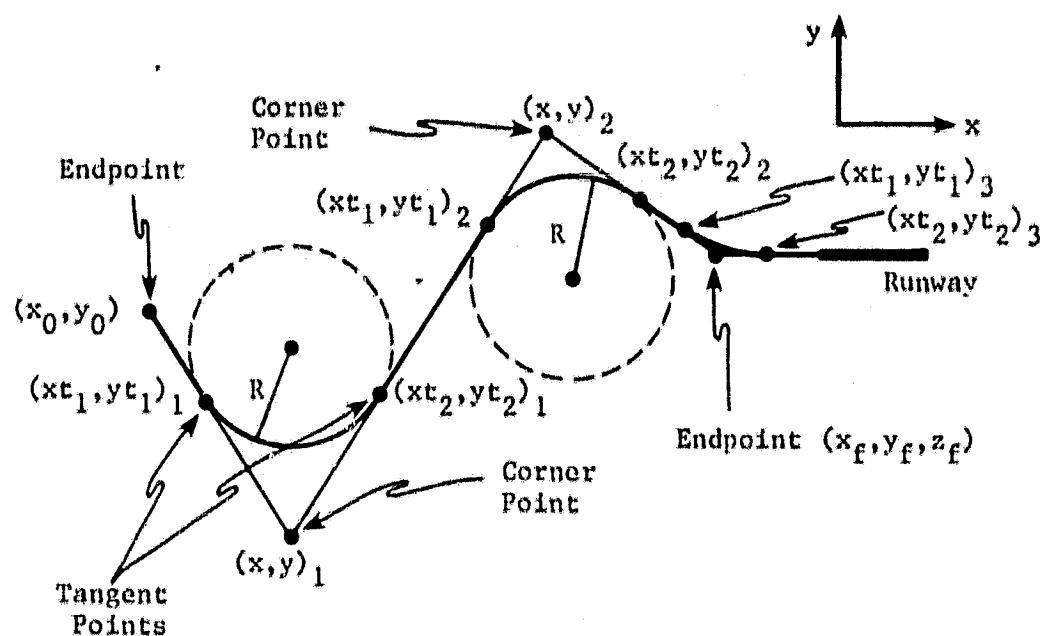


Figure 4.4a Ground Track, Linear Segment Representation

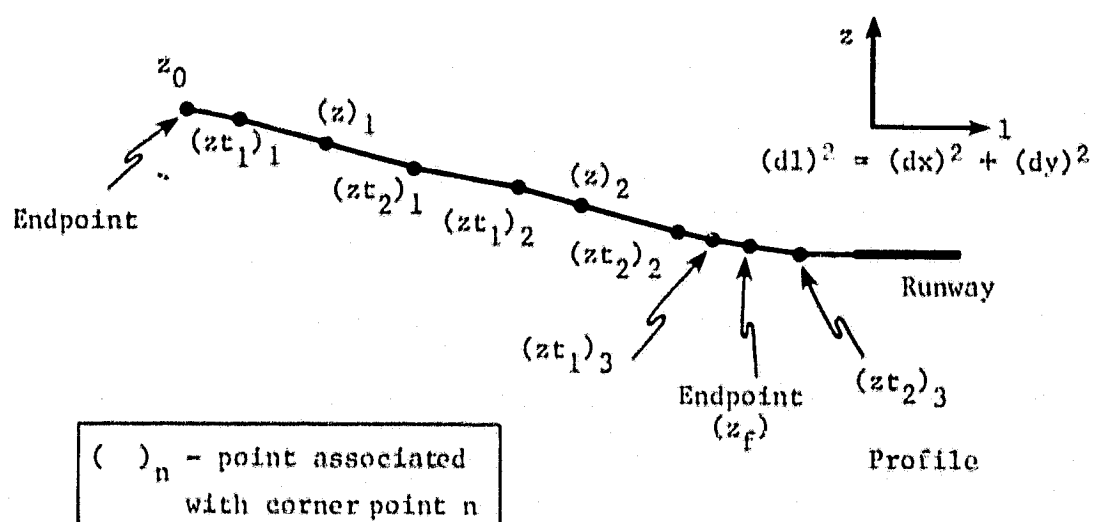


Figure 4.4b Profile, Linear Segment Representation

passenger comfort constraints) is inscribed in the angle between any two linear segments. The arc, tangent to both segments, describes the ground track of the path flown during a turn at that corner. Between tangent points, as shown in Figure 4.4b, the profile of the path is approximated by a linear segment. The resulting turn, in three dimensions, is a helical arc. Since each corner point (requiring three coordinates) may have at most one turn, the complexity of the flight path is limited in a natural way by the allowed flexibility, i.e. the number of corner points used.

Also treated in an implicit way are the endpoint restrictions. Since only the corner points' coordinates are variables, the endpoints of a trajectory may be placed in the appropriate (constant) positions without need of an explicit restriction to hold them constant. For example, in Figure 4.4a, a landing path begins at the entry point (arrival fix) and ends at the inner marker, after which the aircraft flies straight to the runway. For takeoffs (and landings not required to report to the entry point), the endpoint away from the runway may be made variable as shown in Figure 4.5. Here, a pseudo-segment has been added so that the true endpoint becomes a variable corner point. The pseudo-segment must be long enough to prevent a violation of the dynamic and passenger comfort constraints (subsequently discussed), and no noise calculations should involve this fictitious segment.

Explicit statements of the remaining constraints are required for this particular trajectory model. The lateral dynamic restriction, Eqn. (4-6) and passenger comfort constraint Eqn. (4-12) are both inequalities involving the radius of curvature of the ground track; they may be combined to yield

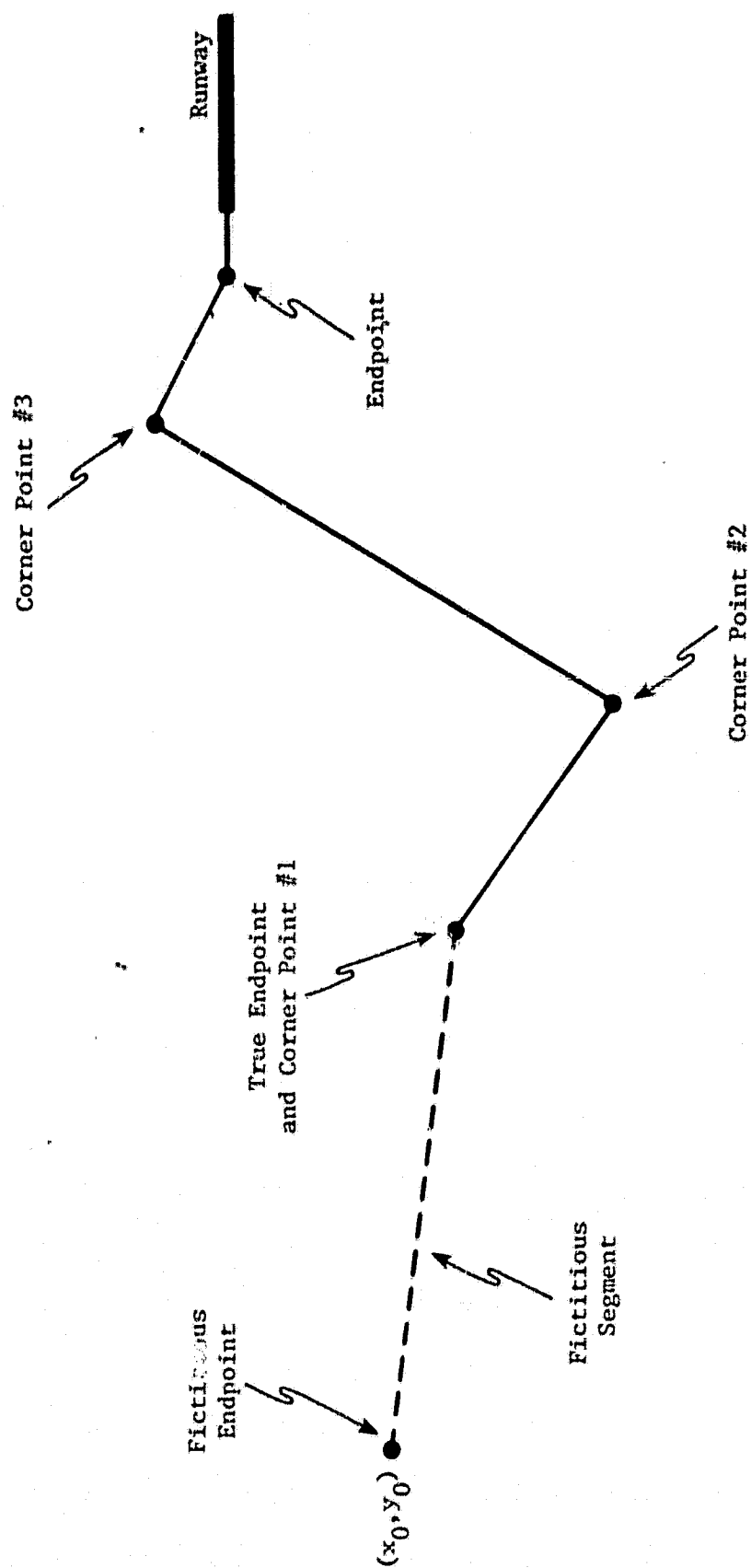


Figure 4.5 Variable Endpoint

$$R \geq \max \left\{ \frac{\bar{V}^2}{C_2 g}, \frac{\bar{V}}{(C_1 + C_2 C_3) \min(\delta x_1, \delta x_2, \delta x_3)} \right\}, \quad (4-13)$$

where R is the radius of curvature of any point on the ground track. For the linear segments of a path, $R \rightarrow \infty$; for the turns at the corner point, R is just the radius of the circular arc used to describe a turn in the ground track. Generally, the passenger comfort term (first on the right in Eqn. (4-13)) is the larger, so that a constant value of R (R_{\min}) that satisfies Eqn. (4-13) may be employed for all turns once the maximum value of \bar{V} is known for a trajectory.

Incorporation of the longitudinal dynamic constraint is equally straightforward. Eqn. (4-9) involves only the vertical slope (the slope of the profile with respect to the ground plane), a quantity that is readily calculated once the coordinates of the tangent points are known (see Figures 4.4a,b). Appendix C gives a derivation of the formulae used to obtain these coordinates.

Because of the form of the flight path model used here, some additional restrictions on the feasibility of any trajectory must be included. As shown in Figure 4.6, it is possible to specify corner points that result in nonsensical trajectories for a given value of R_{\min} . Figure 4.7 indicates the pertinent quantities to use in the restrictions. The first restriction prevents a path from becoming disconnected between any two corner points by requiring that

$$\left| \frac{R_{\min}}{\tan \frac{\alpha}{2}} \right| + \left| \frac{R_{\min}}{\tan \frac{\beta}{2}} \right| \leq l_2. \quad (4-14)$$

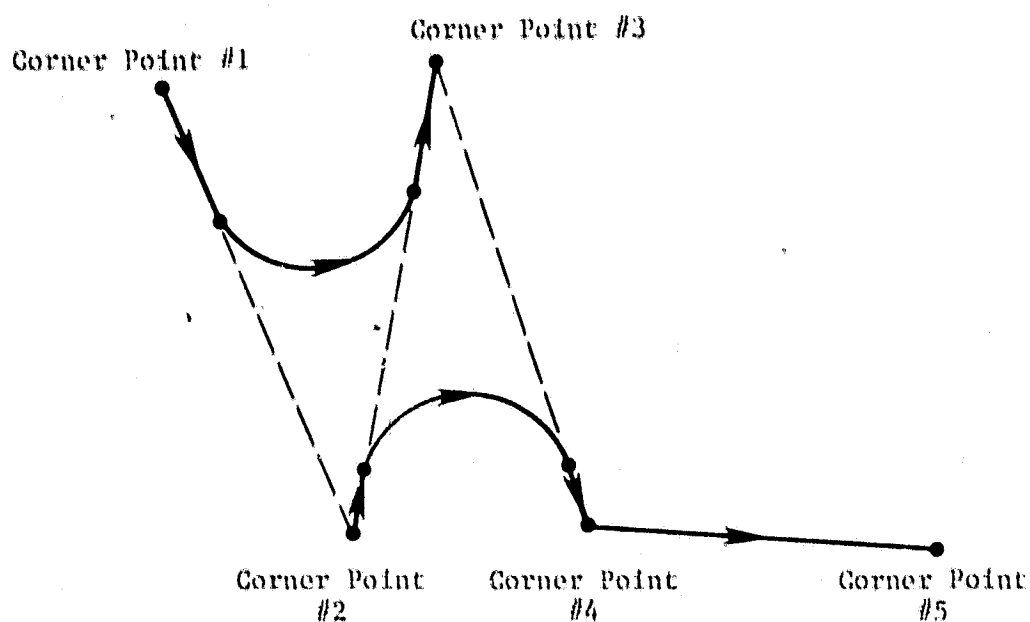


Figure 4.6 Corner Points that Specify an Impossible Trajectory

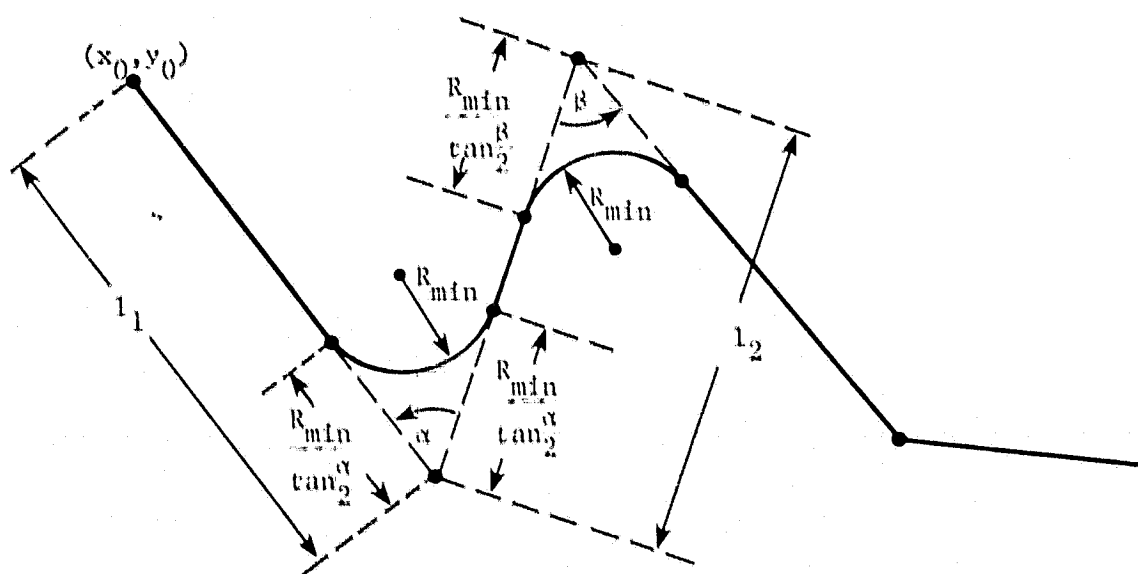


Figure 4.7 Relations Needed to Specify Feasible Flight Paths

Second, for the outermost segment in the chain, an explicit requirement that the circular arc have a tangent point on the segment⁵ must be stated, i.e.,

$$\frac{R_{\min}}{\tan \frac{\alpha}{2}} \leq l_1. \quad (4-15)$$

Maintaining a safe distance between any two trajectories may be accomplished by sampling the distance between their ground tracks at a number of locations, as shown in Figure 4.8a. This test makes two assumptions: that the profiles of the two paths are sufficiently similar to allow these sampled distances in the ground plane to be good approximations of the actual distances, and that enough samples can be taken to obtain an accurate representation of spacing between the two paths. The first assumption is not an object of concern because it leads, at most, to an underestimation of the spacing. As for the number of samples needed, it has been found in this work that as few as ten measurements will suffice; however, as shown in Figure 4.8b, the successive values \hat{y}_i must be tested not only for minimum magnitude, but also for a change in sign, indicating a crossing of the paths. The two constraints are then:

$$\hat{y}_i \geq d_{\min}, \quad i = 1, s \quad (4-16)$$

$$\hat{y}_i \hat{y}_{i+1} > 0, \quad i = 1, s-1 \quad (4-17)$$

where s is the number of samples. As stated previously, when two or more flight paths share the same runway, or use runways whose ends

⁵The assumption of tangency is necessary for determining the expression for the arc, as developed in Appendix C. On the other segments, Eqn. (4-14) or (4-18), (4-19) assure this condition.

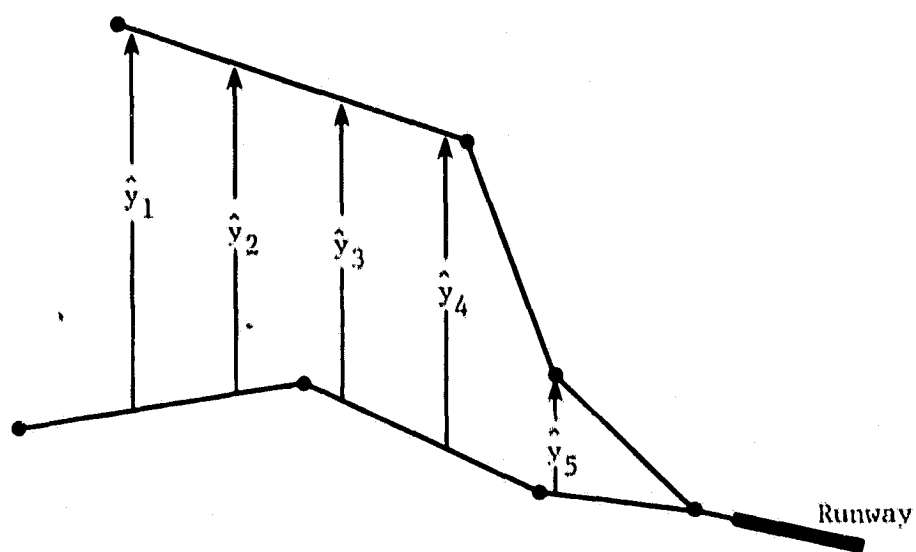


Figure 4.8a Sampling the Distances Between Two Paths

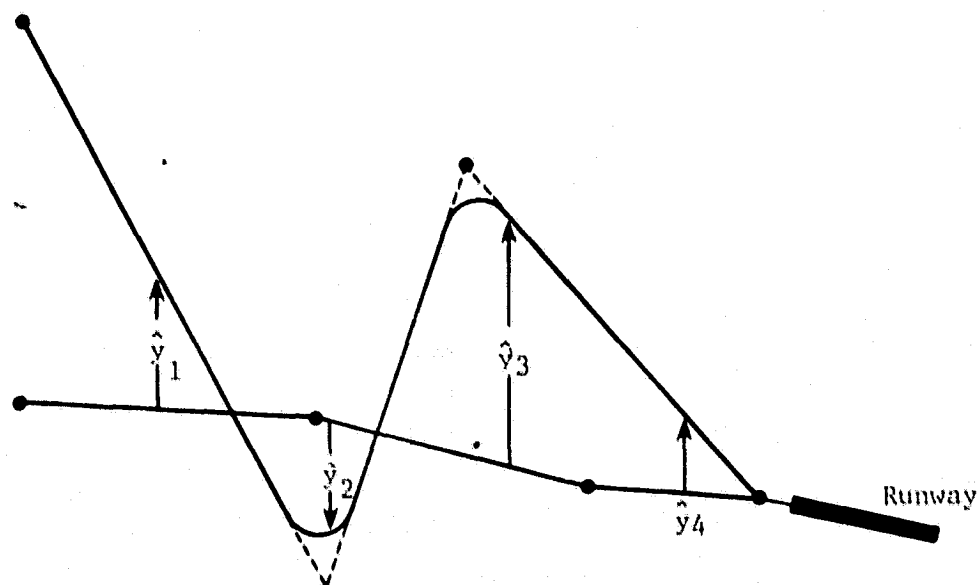


Figure 4.8b Two Crossing Trajectories with $y_i \geq d_{\min}$

are close together, the separation constraint, Eqn. (4-16), should not be applied near these runways where the close spacing between trajectories is unavoidable. Further, at larger airports with sophisticated navigational and air traffic control systems, both the separation and crossover restrictions, Eqns. (4-16) and (4-17) may not be required.

The final heading constraint is modelled by requiring that the angle δ , and length l_1 , shown in Figure 4.9, satisfy the following:

$$\delta \geq 135^\circ, \quad (4-18)$$

$$l_1 \leq \frac{1}{2}l. \quad (4-19)$$

Together, these guarantee that a landing aircraft will have a path aligned with the runway before reaching it, and that any final turn will be no greater than 45° .

Finally, the threshold noise constraint (which is independent of the trajectory model) requires an analytic form. In each grid cell to which the constraint applies, the maximum A-level ($L_{A,max}$) experienced during each aircraft flyover may not exceed a threshold level ($L_{A,Thresh}$) more than N_{max} times per day. This may be expressed as

$$\sum_{i=N^*}^{N_f} \max[0, L_{A,max_i} - L_{A,Thresh}]_{j \in J} \leq 0 \quad (4-20)$$

where $N^* =$ index of the first flyover after N_{max} flights with $L_{A,max_i} \geq L_{A,Thresh}$

$N_f =$ number of flights during a 24 hour period,

$j =$ index for the grid cells to be included in this constraint, i.e., a set denoted J .

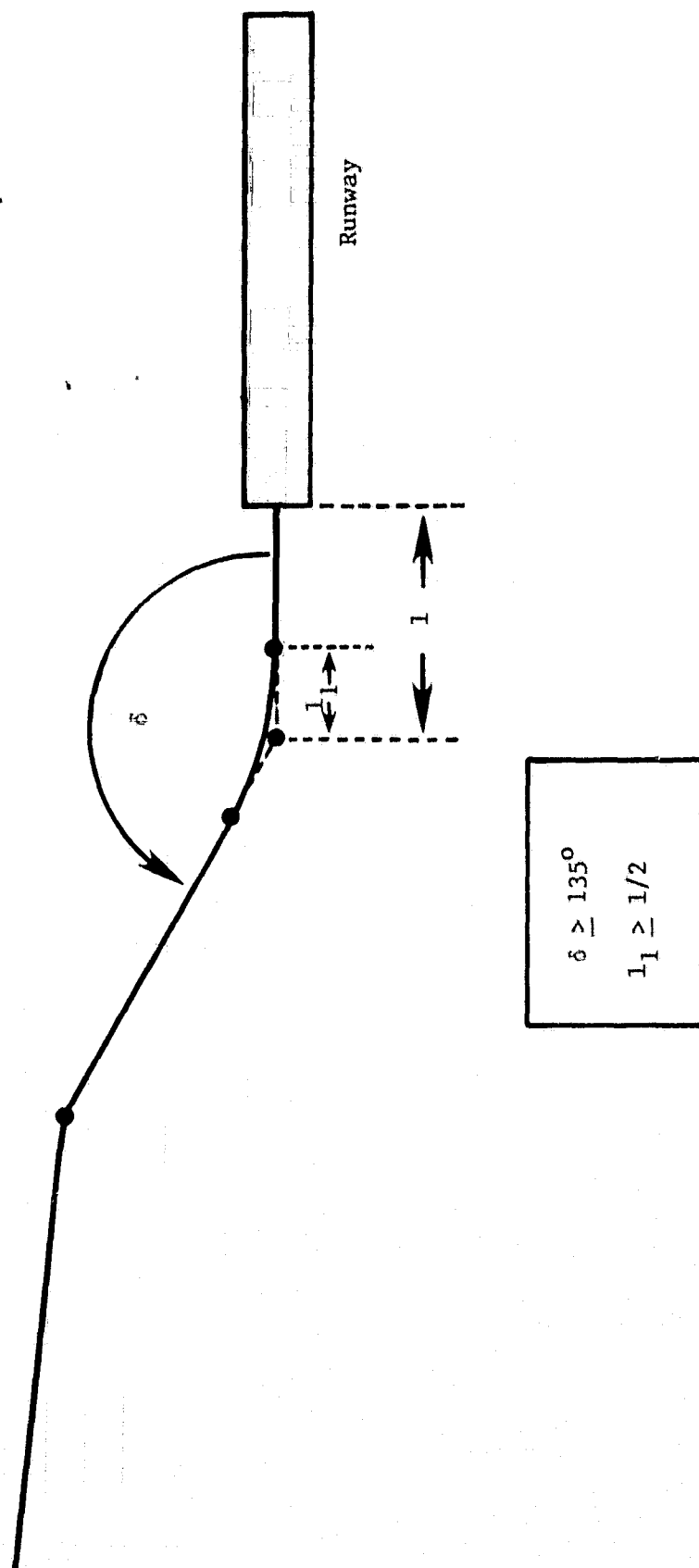


Figure 4.9 The Final Heading Constraints

CHAPTER V

THE OPTIMIZATION SCHEME

In Chapter II, the notion of representing each aircraft trajectory in parametric form was introduced, the idea being to reduce the amount of computation required for determining the optimum set of flight paths. This chapter develops the numerical algorithm that will be used to determine the optimum parameter values. In fact, several algorithms are described in the development, giving a good representation of available methods that are compatible with the solution approach adopted in this work.

All of the optimization methods to be discussed require a criterion, referred to here as the cost function, by which to judge the improvement being made in a system. The cost function must depend, perhaps implicitly, only upon the parameters needed to describe the system (in this case, those that specify the trajectories). As formulated in Chapter VI, the cost will be a function (explicitly) of the noise annoyance, the solution restrictions (constraints), and if desired, other attributes of the system. All of these components, however, depend upon the trajectories, which in turn, are functions of the parameters.

Most optimization algorithms are iterative descent methods. They approximate the solution to a problem usually in several steps, with each successive step making use of previous approximations. This search progresses in such a way that the value of the cost function at each iteration is lower the previous value; the search is then said to be

descending.¹ All of the optimization methods suitable for this work involve derivatives of the cost function with respect to the unknown parameters. Because the cost function includes the annoyance model, which lacks a tractable analytic expression, derivatives of the cost are best determined numerically. This strongly influences the choice of which class of optimization methods to employ.

Of the two suitable for use in this work, primal and approximating methods, the former relies more upon extremely accurate derivative calculations. These methods search over only those parameter combinations not prohibited by the restrictions. Although primal algorithms offer the advantage of "suboptimal" solutions at each iteration² (reduced cost and all constraints satisfied), the necessity of large numbers of calculations to compute the derivatives accurately outbalances the attractive features.

The approximating methods begin with a problem in the form

$$\begin{aligned} &\text{minimize } f(\underline{x}) \\ &\quad \underline{x} \in E^n \\ &\text{subject to: } \underline{h}(\underline{x}) \leq \underline{c} \end{aligned} \tag{5-1}$$

where \underline{x} is an n -vector whose components are the parameters to be determined $f(\underline{x})$ is the cost function, $\underline{h}(\underline{x})$ is an m -vector-valued function that includes all the constraints, \underline{c} is a constant m -vector, and E^n

¹ It is always possible to cast an optimization problem in a form that seeks a minimum value of the cost.

² If for some reason the iterative process must stop before reaching the optimal solution, these sub-optimal solutions offer at least an improvement over the initial condition.

is Euclidean n -space. With the appropriate transformation, the problem becomes unconstrained:³

$$\begin{aligned} &\text{minimize } \bar{f}(\underline{x}) = f(\underline{x}) + \underline{w}^T \underline{\phi}(\underline{x}) \\ &\underline{x} \in E^n \end{aligned} \quad (5-2)$$

where \underline{w} is a m -vector of constants, called the weighting or controlling parameter, and $\underline{\phi}(\underline{x})$ is an m -vector-valued function that depends upon the problem constraints.

Two forms are popular for use in constructing $\underline{\phi}(\underline{x})$: barriers and penalties. Barrier functions are not useful in the work for two reasons. First is their requirement that the initial point in the search be a feasible point (satisfy all the constraints). Determining a feasible starting point can be quite difficult for cases with many trajectories having several segments. More important though, is a fundamental deficiency of barrier methods. Theoretically, the term $\underline{w}^T \underline{\phi}(\underline{x})$ would create a "barrier" in the n -dimensional surface defined by the modified cost function in Eqn. (5-2). Presumably, the solution point \underline{x}_k (obtained in the k -th iteration) would not move near the barrier, incurring a high cost, in the next iteration. This, however, is not the case if large changes in \underline{x} are allowed; it is possible for the solution to "jump" the barrier, giving an \underline{x}_{k+1} that has a lower cost, but that is infeasible. Without explicitly checking the solution at each iteration, this condition would go "undetected" by the optimization algorithm.

³ Formally, the problem is constrained by the requirement $\underline{x} \in E^n$, but computationally, the solution is unrestricted in real n -space.

Fortunately, the penalty methods are somewhat more robust in this respect. Although they also permit infeasible solutions to pass for feasible ones, the condition is detectable and automatically corrected by the standard procedure for implementing the penalty methods.

A popular form for the penalty function is (see Ref. [23])

$$P_i(\underline{x}) = (\max[0, (h_i(\underline{x}) - c_i)])^2, \quad (5-3)$$

with the weighting parameters chosen as

$$\underline{w}^T = [\mu_1, \mu_2, \dots, \mu_m], \mu_i \geq 0. \quad (5-4)$$

The problem is then expressed as (c.f. Eqn. (4-2)):

$$\underset{\underline{x} \in E^n}{\text{minimize}} \quad f(\underline{x}) + \sum_{i=1}^m \mu_i P_i(\underline{x}). \quad (5-5)$$

To remove totally the approximation in the constraints, introduced by the penalties, the problem must be solved a number of times, with increasing values of the μ_i . A sequence $\{\underline{\mu}^j\}$, with $\mu_i^{j+1} \geq \mu_i^j$, is used to determine a sequence of solutions $\{\underline{x}^j\}$. While some of the \underline{x}^j may violate constraints, it may be shown (see Luenberger [23, pp. 278-280]) that any limit point⁴ of the sequence $\{\underline{x}^j\}$ associated with the sequence $\{\underline{\mu}^j\}$ is a solution to the original problem, Eqn. (5-1).

As a practical matter, though, it is necessary only to use a value of $\underline{\mu}$ that results (at worst) in acceptably small violations of the constraints. Some experimentation with the particular problem being solved may be required to determine the components of $\underline{\mu}$ that will give acceptable results. Too large values tend to slow the convergence to the

⁴ If there exists a subset K of the positive integers such that the subsequence $\{\underline{x}^k\}_{k \in K}$ converges to \underline{x} , then \underline{x} is a limit point of $\{\underline{x}^k\}$.

solution, while small values may result in constraint violations that are excessive (by whatever approximation standards apply).

With the constrained problem formulated as an unconstrained one, a suitable optimization method must be selected. The following sections develop several traditional algorithms, leading to the technique chosen for this study.

As with most nonlinear problems, the task of finding the minimum of a nonlinear function $f(\underline{x})$ lacks a general, closed-form solution. Iterative technique, however, can be used to make successive approximations of the solution, until the calculated value is judged to be reasonably close to the actual one.

Once a starting point is chosen, new points are generated by a recursion of the form

$$\underline{x}_{k+1} = \underline{x}_k + \alpha_k \underline{d}_k \quad (5-6)$$

where \underline{d}_k is the "direction" vector and α_k is a scalar that determines how far to move along the direction \underline{d}_k in order to reach \underline{x}_{k+1} . The values of α_k and \underline{d}_k vary in the methods subsequently discussed.

Steepest Descent

Perhaps the most "short-sighted" of its type, this method often gives satisfactory results, but more important, it serves as a foundation for the more sophisticated algorithms.

Given a function $f(\underline{x})$ and a point \underline{x}_0 , the task is to find a good direction (in n -space) in which to search for a new point \underline{x}_1 , such that $f(\underline{x}_1) \leq f(\underline{x}_0)$. The first approach is conveniently provided by a property of the gradient operator

$$\nabla = \frac{\partial}{\partial x_i} [\hat{a}_i]^T \quad (\text{a row vector}), \quad (5-7)$$

namely that $\underline{g}(\underline{x}) = \nabla^T f(\underline{x})$ is a vector that points in the direction in which $f(\underline{x})$ is increasing most rapidly. Since the problem is to find the minimum value of $f(\underline{x})$, the direction in which to search is along $-\underline{g}(\underline{x})$. This will result in the most rapid decrease of $f(\underline{x})$ for that iteration. (Later methods sacrifice this immediate improvement for other advantages.)

As stated earlier, it is not practical to obtain an analytic expression for $\underline{g}(\underline{x})$ in the problem under consideration. An approximate numerical form

$$[\underline{g}(\underline{x})]_i \simeq \left[\frac{f(x_1, x_2, \dots, x_i + \Delta x, \dots, x_n) - f(\underline{x})}{\Delta x} \right]_i, \quad (5-8)$$

$i = 1, n$

has been employed. In the limit $\Delta x \rightarrow 0$, Eqn. (5-8) yields the exact gradient; small values of Δx should therefore be used to obtain accurate approximations.

The method of steepest descent is stated as the following algorithm:

Steepest Descent Algorithm

Step 0: Choose an initial point \underline{x}_0 , set $k=0$.

Step 1: Compute $\underline{d}_k = -\underline{g}(\underline{x}_k)$

Step 2: Minimize (with respect to $\alpha_k > 0$)

$$f(\underline{x}_k + \alpha_k \underline{d}_k)$$

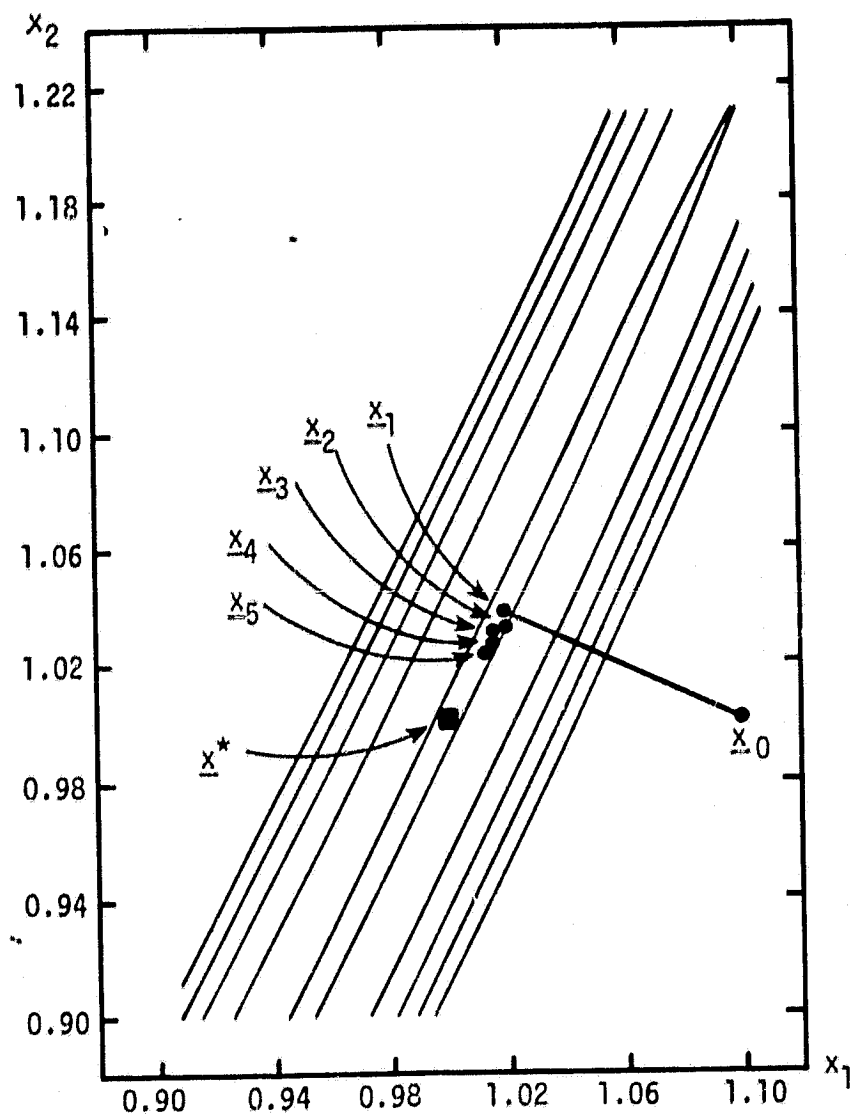


Figure 5.1 Steepest Descent Example

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Iterative values of \underline{x}_k up to $k=4$ are shown as circles, connected by line segments to indicate the progression. This example demonstrates a major problem of steepest descent, namely, its slow convergence properties. Its initial step to \underline{x}_1 is directly "downhill," causing a large decrease in the value of $f(\underline{x})$; however, once in the shallow region, it oscillates ineffectually, making slow progress. The curvature along the valley prevents $-\underline{g}(\underline{x})$, the search direction, from pointing directly to the solution. A more rigorous explanation of the convergence properties of this algorithm may be found in Luenberger [23, pp. 148-155].

Newton's Method

A more sophisticated approach to the general problem of minimizing a function f is to approximate that function with a quadratic form in the region of the solution:

$$f(\underline{x}) \simeq c - \underline{b}^T \underline{x} + \frac{1}{2} \underline{x}^T \underline{Q} \underline{x}. \quad (5-10)$$

This approximate function has a minimum at

$$\underline{x}^* = \underline{Q}^{-1} \underline{b} \quad (5-11)$$

provided that \underline{Q} is a positive definite symmetric matrix.⁵ Taylor's Theorem⁶ may be used to expand $f(\underline{x})$ about \underline{x}_k , keeping only up to quadratic terms:

$$f(\underline{x}) \simeq f(\underline{x}_k) + \underline{\nabla} f(\underline{x}_k)(\underline{x} - \underline{x}_k) + \frac{1}{2}(\underline{x} - \underline{x}_k)^T \underline{E}(\underline{x}_k)(\underline{x} - \underline{x}_k), \quad (5-12)$$

where $\underline{E}(\underline{x}_k)$ the Hessian of f , is the $n \times n$ matrix of second-partial derivatives of $f(\underline{x})$, evaluated at \underline{x}_k ,

⁵ In R^n , a matrix \underline{Q} , is positive definite if $\underline{x}^T \underline{Q} \underline{x} > 0$ for all $\underline{x} \neq \underline{0}$.

⁶ See Pierre [32].

$$\underline{F}(\underline{x}) = \left[\frac{\partial^2 f(\underline{x})}{\partial x_i \partial x_j} \right] \quad (5-13)$$

Comparing terms of (5-12) with (5-11) gives a minimum⁷ at

$$(\underline{x} - \underline{x}_k) = -[\underline{F}(\underline{x}_k)]^{-1} \underline{g}(\underline{x}_k)^T, \quad (5-14)$$

from which follows the recursion

$$\underline{x}_{k+1} = \underline{x}_k - [\underline{F}(\underline{x}_k)]^{-1} \underline{g}(\underline{x}_k)^T. \quad (5-15)$$

The algorithm for Newton's method is then:

Newton's Method

Step 0: Choose an initial point \underline{x}_0 , set $k = 0$.

Step 1: Compute $\underline{g}(\underline{x}_k)$, $[\underline{F}(\underline{x}_k)]^{-1}$.

Step 2: Set $\underline{x}_{k+1} = \underline{x}_k - [\underline{F}(\underline{x}_k)]^{-1} \underline{g}(\underline{x}_k)$.

Set $k=k+1$

Go to Step 1.

As before, this procedure continues until a stopping criterion is met.

No line search is required, but computing, inverting, and storing the inverse Hessian requires more computing resource than does the steepest descent algorithm. Newton's method does, however, converge rapidly when \underline{x}_k is near the solution \underline{x}^* . This depends somewhat upon how nearly quadratic $f(\underline{x})$ is: a scaling factor α_k , and a line search may be included for cases where the quadratic approximation is not sufficient. The recursion then becomes

$$\underline{x}_{k+1} = \underline{x}_k - \alpha_k [\underline{F}(\underline{x}_k)]^{-1} \underline{g}(\underline{x}_k), \quad (5-16)$$

⁷ The requirement that \underline{Q} (and \underline{F}) be positive definite is based upon the necessary and sufficient conditions for \underline{x}^* to be a minimum point. (See Luenberger [23, pp. 110-114]).

and a line search to minimize $f(\underline{x}_k + \alpha_k [F(\underline{x}_k)]^{-1} \underline{g}(\underline{x}_k))$ must be added to Step 2 of the algorithm.

Newton's Method: Example 2

Figure 5.2 shows the results of applying Newton's method to the same problem as in Example-1. The pure form of the method, without the search parameter α_k , was used with excellent results. The seemingly odd behavior exhibited in iteration 1 demonstrates how this method, in sacrificing the initial rapid progress of steepest descent, converges swiftly on the solution after only five iterations.

Conjugate Gradient Methods

Of the descent techniques examined thus far, one converges quite slowly when near the solution, and the other, quite rapidly, but at the expense of large increases in computation. A good compromise is provided by the conjugate gradient methods. These are developed qualitatively here; theoretical treatments may be found in Ref. [23] and [32].

Conjugate gradient techniques are a subset of the more general conjugate direction method. As the name implies, gradient information is used in computing the search directions; it is the special conjugacy of these directions that distinguishes the methods. The quadratic problem considered previously serves to illustrate the notions involved.

A quadratic function

$$f(\underline{x}) = \frac{1}{2} \underline{x}^T \underline{Q} \underline{x} - \underline{b}^T \underline{x} \quad (5-17)$$

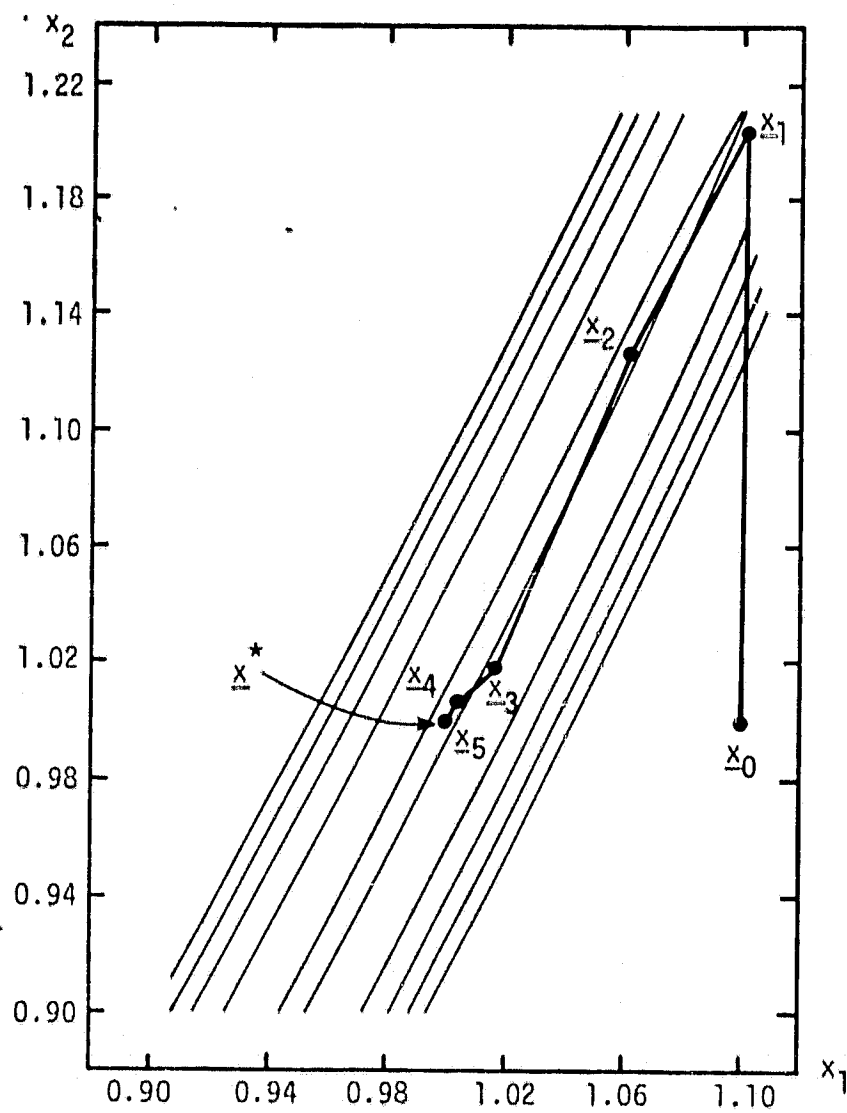


Figure 5.2 Newton's Method Example

has a minimum at

$$\underline{x}^* = \underline{Q}^{-1}\underline{b} \quad (5-18)$$

if \underline{Q} is positive definite. This solution may be expanded, using any linearly independent set of vectors $\{\underline{d}_i\}_{i=0}^{n-1}$, as

$$\underline{x}^* = \sum_{i=0}^{n-1} \alpha_i \underline{d}_i. \quad (5-19)$$

If the \underline{d}_i are \underline{Q} -orthogonal (or \underline{Q} -conjugate), i.e. $\underline{d}_j^T \underline{Q} \underline{d}_i = 0$ for $i \neq j$, then the following results⁸:

$$\underline{Q}^{-1}\underline{b} = \sum_{i=0}^{n-1} \alpha_i \underline{d}_i, \quad (5-20)$$

$$\underline{Q} \underline{Q}^{-1} \underline{b} = \sum_{i=0}^{n-1} \alpha_i \underline{Q} \underline{d}_i, \quad (5-21)$$

$$\underline{d}_j^T \underline{Q} \underline{Q}^{-1} \underline{b} = \sum_{i=0}^{n-1} \alpha_i \underline{d}_j^T \underline{Q} \underline{d}_i, \quad (5-22)$$

$$\underline{d}_j^T \underline{b} = \alpha_j \underline{d}_j^T \underline{Q} \underline{d}_j, \quad (5-23)$$

$$\alpha_j = \frac{\underline{d}_j^T \underline{b}}{\underline{d}_j^T \underline{Q} \underline{d}_j}. \quad (5-24)$$

The solution, in terms of the conjugate directions, is then

$$\underline{x}^* = \sum_{i=0}^{n-1} \frac{\underline{d}_i^T \underline{b}}{\underline{d}_i^T \underline{Q} \underline{d}_i} \underline{d}_i. \quad (5-25)$$

⁸ Proof that the \underline{Q} -conjugate set $\{\underline{d}_i\}$ are linearly independent:

Assume that the \underline{d}_i are not linearly independent. Then there exist

$\alpha_i \neq 0$ such that $\sum_{i=0}^{n-1} \alpha_i \underline{d}_i = \underline{0}$. Multiplying by $\underline{d}_j^T \underline{Q}$ yields (for each i)

$\alpha_i \underline{d}_i^T \underline{Q} \underline{d}_i = 0$. Since \underline{Q} is positive definite, $\underline{d}_i^T \underline{Q} \underline{d}_i > 0$ and $\alpha_i \equiv 0$,

contradicting the assumption.

All that remains is to specify the \underline{d}_i in terms of \underline{Q} and \underline{b} , the known quantities in the problem.

What the conjugate gradient algorithm does is to compute the \underline{d}_i iteratively, using linear combinations of the gradients in all previous iterations.

Conjugate Gradient Algorithm

Step 0: Set $k = 0$, $\underline{x}_k = \underline{x}_0$, $\underline{d}_0 = -\underline{g}_0 = \underline{b} - \underline{Q}\underline{x}_0$.

$$\text{Step 1: Compute } \alpha_k = \frac{-\underline{g}_k^T \underline{d}_k}{\underline{d}_k^T \underline{Q} \underline{d}_k} \quad (5-26)$$

$$\underline{x}_{k+1} = \underline{x}_k + \alpha_k \underline{d}_k \quad (5-27)$$

$$\text{Step 2: } \beta_k = \frac{\underline{g}_{k+1}^T \underline{Q} \underline{d}_k}{\underline{d}_k^T \underline{Q} \underline{d}_k} \quad (5-28)$$

$$\underline{d}_{k+1} = -\underline{g}_{k+1} + \beta_k \underline{d}_k \quad (5-29)$$

$$(\underline{g}_k = \underline{Q}\underline{x}_k - \underline{b}) \quad (5-30)$$

Step 3: Set $k = k+1$ and go to Step 1.

These iterative calculations are exact for a purely quadratic problem and the solution, Eqn. (5-25) is obtained in n steps.

In general, the function being minimized is not quadratic. Near a minimum point, the Taylor expansion is

$$f(\underline{x}) = f(\underline{x}^*) + \underline{\nabla} f(\underline{x}^*)(\underline{x} - \underline{x}^*) + \frac{1}{2}(\underline{x} - \underline{x}^*)^T \underline{F}(\underline{x}^*)(\underline{x} - \underline{x}^*) + O(\|\underline{x} - \underline{x}^*\|^3). \quad (5-31)$$

A necessary condition for \underline{x}^* to be an unconstrained minimum point is that $\underline{\nabla} f(\underline{x}^*) \equiv \underline{0}$. The fourth term becomes negligible as $\underline{x} \rightarrow \underline{x}^*$, leaving only the quadratic term in \underline{x} . This leads to the following extension of the conjugate gradient algorithm in which \underline{Q} is replaced by \underline{F} .

General Conjugate Gradient Algorithm

$$\text{Step 0: Set } k = 0, \underline{d}_0 = -\underline{g}_0 = \nabla^T f(\underline{x}_0) \quad (5-31)$$

$$\text{Step 1: Compute } \alpha_k = \frac{-\underline{g}_k^T \underline{d}_k}{\underline{d}_k^T \underline{F}(\underline{x}_k) \underline{d}_k} \quad (5-32)$$

$$\underline{x}_{k+1} = \underline{x}_k + \alpha_k \underline{d}_k \quad (5-33)$$

$$\text{Step 2: Compute } \underline{g}_{k+1} = \nabla^T f(\underline{x}_{k+1}) \quad (5-34)$$

If $k \neq mn-1$ (m an integer),

$$\text{compute } \beta_k = \frac{\underline{g}_{k+1}^T \underline{F}(\underline{x}_k) \underline{d}_k}{\underline{d}_k^T \underline{F}(\underline{x}_k) \underline{d}_k} \quad (5-35)$$

$$\text{set } \underline{d}_{k+1} = -\underline{g}_{k+1} + \beta_k \underline{d}_k \quad (5-36)$$

$$k = k+1$$

Else,

$$\text{Set } \underline{d}_{k+1} = -\underline{g}_{k+1} \quad (5-37)$$

$$k = k+1$$

Go to Step 1.

In Step 2, the "else" branch is called a spacer (or restart) step. It is inserted to aid in the convergence to the solution by reinitiating the conjugate direction calculations every n steps.

A disadvantage of the above algorithm is its requirement of calculating $\underline{F}(\underline{x})$. Even if storage of the $n \times n$ matrix is not an obstacle, numerical accuracy (along with that of $\underline{g}(\underline{x})$) is still a concern. Errors may grow during the iterative construction of the \underline{d}_k to the point that even the spacer step cannot "recover" the search and no progress is made.

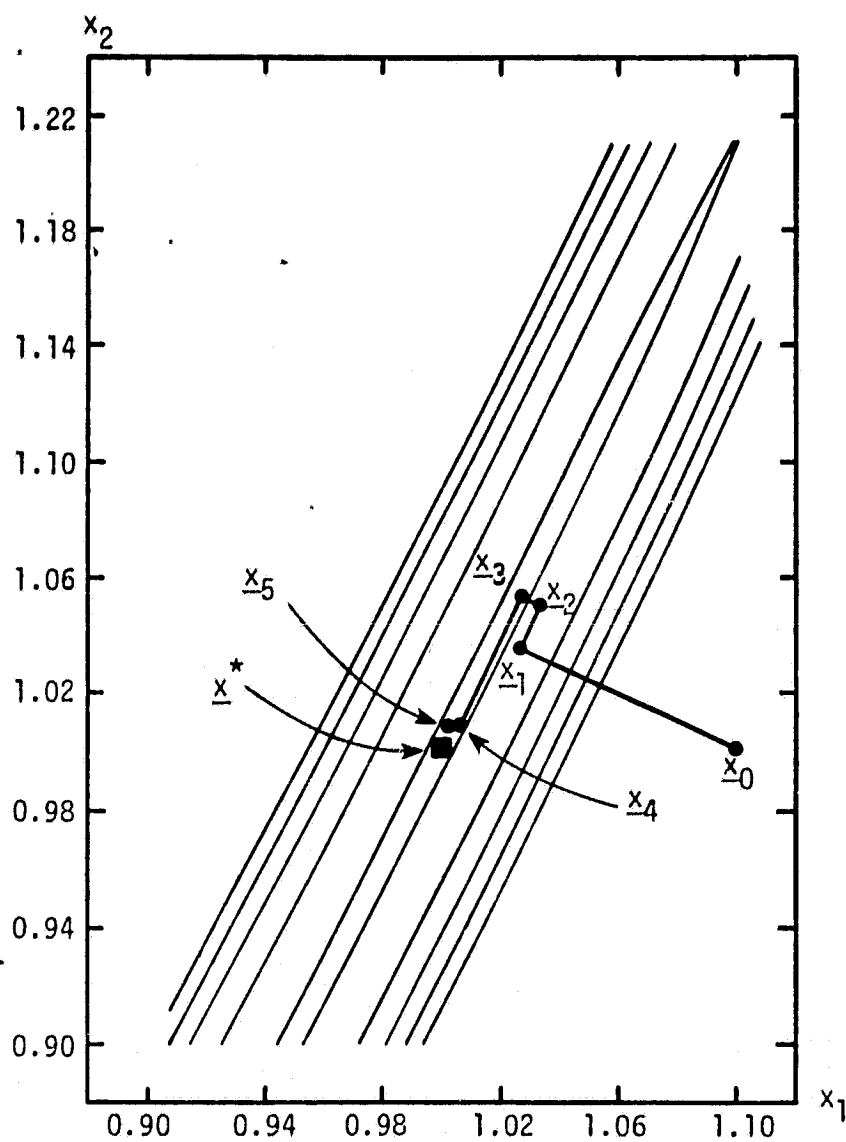


Figure 5.3 Conjugate Gradients Example

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$$\text{Step 2: Set } \underline{d}_k = -\underline{S}_k \underline{g}_k. \quad (5-40)$$

$$\text{Step 3: Minimize } f(\underline{x}_k + \alpha_k \underline{d}_k) \text{ with respect to } \alpha \geq 0 \quad (5-41)$$

$$\text{Set } \underline{x}_{k+1} = \underline{x}_k + \alpha_k \underline{d}_k \quad (5-42)$$

$$\underline{p}_k = \alpha_k \underline{d}_k \quad (5-43)$$

$$\underline{q}_k = \underline{g}_{k+1} - \underline{g}_k \quad (5-44)$$

(Note: α_k must be selected so that $\underline{p}_k^T \underline{q}_k > 0$)

Step 4: If $k \neq mn$ (m an integer), set

$$\underline{S}_{k+1} = \left[\underline{S}_k - \frac{\underline{S}_k \underline{q}_k \underline{q}_k^T \underline{S}_k}{\underline{q}_k^T \underline{S}_k \underline{q}_k} \right] \frac{\underline{p}_k^T \underline{q}_k}{\underline{q}_k^T \underline{S}_k \underline{q}_k} + \frac{\underline{p}_k \underline{p}_k^T}{\underline{p}_k^T \underline{q}_k} \quad (5-45)$$

Set $k = k+1$.

Go to Step 2.

Else,

Go to Step 1.

A. Requirement for step-wise descent

The above algorithm, along with steepest descent and Newton's method, belongs to a class of methods that use the recursion

$$\underline{x}_{k+1} = \underline{x}_k - \alpha_k \underline{M}_k \underline{g}_k \quad (5-46)$$

to compute the minimum of $f(\underline{x})$, where $\alpha \geq 0$, and \underline{M}_k is an $n \times n$ matrix that modifies the direction $-\underline{g}_k$. For steepest descent, $\underline{M}_k \bar{\nabla} \underline{I}$, for Newton's Method, $\underline{M}_k = [\underline{F}(\underline{x}_k)]^{-1}$, and for this method, $\underline{M}_k = \underline{S}_k \simeq [\underline{F}(\underline{x}_k)]^{-1}$ (a point which remains to be proved). A desirable property of such iterative methods is that each step produces a lower value of $f(\underline{x})$. The requirement on \underline{M}_k that insures this is now derived.

Using Taylor's theorem to expand $f(\underline{x})_{\underline{x}=\underline{x}_{k+1}}$ about \underline{x}_k gives:

$$f(\underline{x}_{k+1}) = f(\underline{x}_k) + \underline{\nabla} f(\underline{x}_k)(\underline{x}_{k+1} - \underline{x}_k) + O(\|\underline{x}_{k+1} - \underline{x}_k\|^2), \quad (5-47)$$

into which Eqn. (5-46) may be inserted to yield:

$$f(\underline{x}_{k+1}) = f(\underline{x}_k) - \alpha_k \underline{g}_k^T \underline{M}_k \underline{g}_k + O(1\alpha^2). \quad (5-48)$$

For small α , the trailing term becomes negligible compared to the second term on the right. If $f(\underline{x}_{k+1})$ is to be less than $f(\underline{x}_k)$ for small $\alpha_k > 0$, then

$$-\underline{g}_k^T \underline{M}_k \underline{g}_k > 0. \quad (5-49)$$

This can be insured for all $\underline{g}_k \neq \underline{0}$ by requiring the \underline{M}_k be positive definite. (Clearly, for steepest descent this is the case, but for Newton's method, $\underline{M}_k = [\underline{F}(\underline{x}_k)]^{-1}$ is not guaranteed to be positive definite except near a minimum.)

B. Positive Definiteness of \underline{S}_k

Given the algorithm for generating the \underline{S}_k , it will be shown that the positive definiteness of \underline{S}_{k+1} follows from that of \underline{S}_k . For $\underline{x} \in E^n$,

$$\underline{x}^T \underline{S}_{k+1} \underline{x} = \gamma_k \underline{x}^T \underline{S}_k \underline{x} - \gamma_k \frac{(\underline{x}^T \underline{S}_k \underline{q}_k)^2}{\underline{q}_k^T \underline{S}_k \underline{q}_k} + \frac{(\underline{x}^T \underline{p}_k)^2}{\underline{p}_k^T \underline{q}_k} \quad (5-50)$$

where

$$\gamma_k = \frac{\underline{p}_k^T \underline{q}_k}{\underline{q}_k^T \underline{S}_k \underline{q}_k}.$$

Substituting $\underline{a} = \underline{S}_k^{-1/2} \underline{p}_k$, $\underline{b} = \underline{S}_k^{-1/2} \underline{q}_k$ gives

$$\underline{x}^T \underline{S}_{k+1} \underline{x} = \gamma_k \left[\frac{(\underline{a}^T \underline{a})(\underline{b}^T \underline{b}) - (\underline{a}^T \underline{b})^2}{\underline{b}^T \underline{b}} \right] + \frac{(\underline{x}^T \underline{p}_k)^2}{\underline{p}_k^T \underline{q}_k}. \quad (5-51)$$

By definition, $\underline{q}_k = \underline{q}_{k+1} - \underline{q}_k$, so that

$$\underline{p}_k^T \underline{q}_k = \underline{p}_k^T \underline{q}_{k+1} - \underline{p}_k^T \underline{g}_k. \quad (5-52)$$

Since \underline{x}_{k+1} is the minimum of $f(\underline{x})$ along $\underline{p}_k = \alpha_k \underline{d}_k$, then

$$\frac{\partial f}{\partial \alpha}(\underline{x}_k + \alpha \underline{d}_k) = \frac{\nabla f(\underline{x}_{k+1}) \underline{d}_k}{\underline{d}_k^T \underline{d}_k} = \frac{\alpha_k \underline{g}_{k+1}^T \underline{p}_k}{\underline{d}_k^T \underline{d}_k} = 0 \quad (5-53)$$

Therefore Eqn. (5-52) yields

$$\underline{p}_k^T \underline{g}_k = -\underline{p}_k^T \underline{g}_k \quad (5-54)$$

By definition, $\underline{p}_k = \alpha_k \underline{d}_k = -\alpha_k \underline{S}_k \underline{g}_k$, so that

$$\underline{p}_k^T \underline{g}_k = \alpha_k (\underline{S}_k \underline{g}_k)^T \underline{g}_k = \alpha_k \underline{g}_k^T \underline{S}_k \underline{g}_k. \quad (5-55)$$

Eqn. (5-51) may now be written

$$\underline{x}^T \underline{S}_{k+1} \underline{x} = \gamma_k \left[\frac{(\underline{a}^T \underline{a})(\underline{b}^T \underline{b}) - (\underline{a}^T \underline{b})^2}{\underline{b}^T \underline{b}} \right] + \frac{(\underline{x}^T \underline{p}_k)^2}{\alpha_k \underline{g}_k^T \underline{S}_k \underline{g}_k} \quad (5-56)$$

By the Cauchy-Schwarz inequality,⁹ the first term on the right of Eqn. (5-56) is positive semi-definite, as is the second term, by virtue of the definition of inner product. To demonstrate that both terms cannot vanish simultaneously, assume that $\underline{a} = \xi \underline{b}$, for which case the first term vanishes. Then $\underline{x} = \xi \underline{g}_k$ and

$$\underline{p}_k^T \underline{x} = \xi \underline{p}_k^T \underline{g}_k = \xi \alpha_k \underline{g}_k^T \underline{S}_k \underline{g}_k \neq 0 \quad (5-57)$$

since \underline{S}_k is positive definite. Therefore, $\underline{x}^T \underline{S}_{k+1} \underline{x} > 0$ for all $\underline{x} \neq 0$. Since the algorithm starts with $\underline{S}_0 = \underline{I}$, all of the \underline{S}_k will be positive definite. Q.E.D.

An extremely useful piece of information in this proof regards the determination of α_k . The magnitude of $\underline{p}_k^T \underline{g}_k$ in Eqn. (5-55) is not paramount; only its sign affects the definiteness of \underline{S}_{k+1} . For this

⁹ Cauchy-Schwarz Inequality.

$$(\underline{a}^T \underline{a})(\underline{b}^T \underline{b}) \geq (\underline{a}^T \underline{b})^2$$

reason, the algorithm requires that the line search in Step 3 be accurate enough only to give $\underline{p}_k^T \underline{q}_k > 0$. (When applied to nonquadratic problems, the requirement that α_k minimizes $f(\underline{x})$ along $\underline{x}_k + \alpha_k \underline{d}_k$ becomes more stringent).

C. Convergence of \underline{S}_k to \underline{F}^{-1}

For the modified Davidon-Fletcher-Powell method, it may be shown that if $f(\underline{x})$ is quadratic (\underline{F} constant and positive definite), then

$$\underline{p}_i^T \underline{F} \underline{p}_j = 0, \quad 0 \leq i < j \leq k \quad (5-58)$$

and

$$\underline{S}_{k+1} \underline{F} \underline{p}_i = \underline{p}_i, \quad 0 \leq i \leq k, \quad (5-59)$$

from which the convergence of the algorithm in a finite number of steps will be proved, along with the fact that $\underline{S}_n = \underline{F}^{-1}$.

A Taylor expansion of $g(\underline{x}) = \underline{x}^T \underline{F} \underline{x}$ gives

$$g(\underline{x}) \simeq g(\underline{x}_k) + \underline{F}(\underline{x} - \underline{x}_k). \quad (5-60)$$

Using Eqn. (5-42) - (5-44) in the above yields

$$\underline{q}_k = \underline{g}_{k+1} - \underline{g}_k = \underline{F} \underline{x}_{k+1} - \underline{F} \underline{x}_k = \underline{F} \underline{p}_k. \quad (5-61)$$

From (5-45)

$$\underline{S}_{k+1} \underline{F} \underline{p}_k = \underline{S}_{k+1} \underline{q}_k = \underline{p}_k \quad (5-62)$$

The proof of (5-58) and (5-59) is by induction: They are true for $k = 0$ by virtue of (5-61) and (5-62). Assuming that they are true for iteration $k-1$, it will be shown that they are true for iteration k .

Using (5-61), \underline{q}_k may be constructed from previous iterations:

$$\underline{q}_k = \underline{g}_{i+1} + \underline{F}(\underline{p}_{i+1} + \dots + \underline{p}_{k-1}) \quad (5-63)$$

From (5-53) and (5-58)

$$p_i^T g_k = p_i^T g_{i+1} = 0, \quad 0 \leq i < k. \quad (5-64)$$

Transposing (5-62) and substituting into (5-64) yields

$$p_i^T F S_k g_k = 0, \quad i < k. \quad (5-65)$$

From the definition of p_k , the above may be rewritten

$$p_i^T F p_k = 0, \quad i < k, \quad (5-66)$$

proving (5-58) for k .

Still assuming that (5-61) is true for $k-1$,

$$S_k F p_i = p_i, \quad 0 \leq i < k, \quad (5-67)$$

which combined with (5-62) gives

$$S_{k+1} F p_i = p_i, \quad 0 \leq i \leq k. \quad \text{Q.E.D.}$$

Since the p_k are F -orthogonal, they are linearly independent (see footnote 8). If the matrix P whose columns are the p_k is constructed

$$P = [p_0 | p_1 | \dots | p_{n-1}], \quad (5-68)$$

then (5-59) yields

$$S_n F P = P. \quad (5-69)$$

The linear independence of the p_k insures that P^{-1} exists, so that

$$S_n F = P P^{-1} = I \quad (5-70)$$

and

$$S_n = F^{-1}. \quad (5-71)$$

Finally, this method is seen to be a conjugate direction method, by virtue of the F -conjugacy of the p_k . For the quadratic case used in the preceding analyses, it will converge in n steps or less. The presence of the scaling factor γ_k alters the eigenvalues in $S_k \approx F^{-1}$ at each iteration such that they span a small range. This leads to a

better rate of convergence than was generally obtained in the original method. For a development of this feature, see Ref. [23]. A pure steepest descent step every n steps insures that the conjugate direction process will be initiated in the vicinity of the minimum, where its convergence is fastest.

Modified D-F-P Method: Example 4

Returning to the problem (5-9) used in the previous examples, the last algorithm is employed with excellent results. Figure 5.4 shows that after five iterations, the approximate solution is comparable to that of the conjugate gradient method in Example 3.

The last algorithm was chosen for use in this study for several reasons. First, it has convergence properties superior to those of steepest descent. Second, it requires only first order derivative information, which may be generated practically. Finally, the search directions employed are guaranteed to decrease the cost function at each iteration, an assurance that the conjugate direction methods lack for nonquadratic problems.

Until now, the problem of nonconvex (or multimodal) cost functions has not been considered. The preceding search methods all refer to finding "a minimum" of the cost, but in nonconvex cases, there is more than one minimum. The lowest valued of these so-called relative or local minima is referred to as the absolute or global minimum. There is no way to guarantee that any of the standard iterative techniques will converge to the global minimum of a nonconvex function.

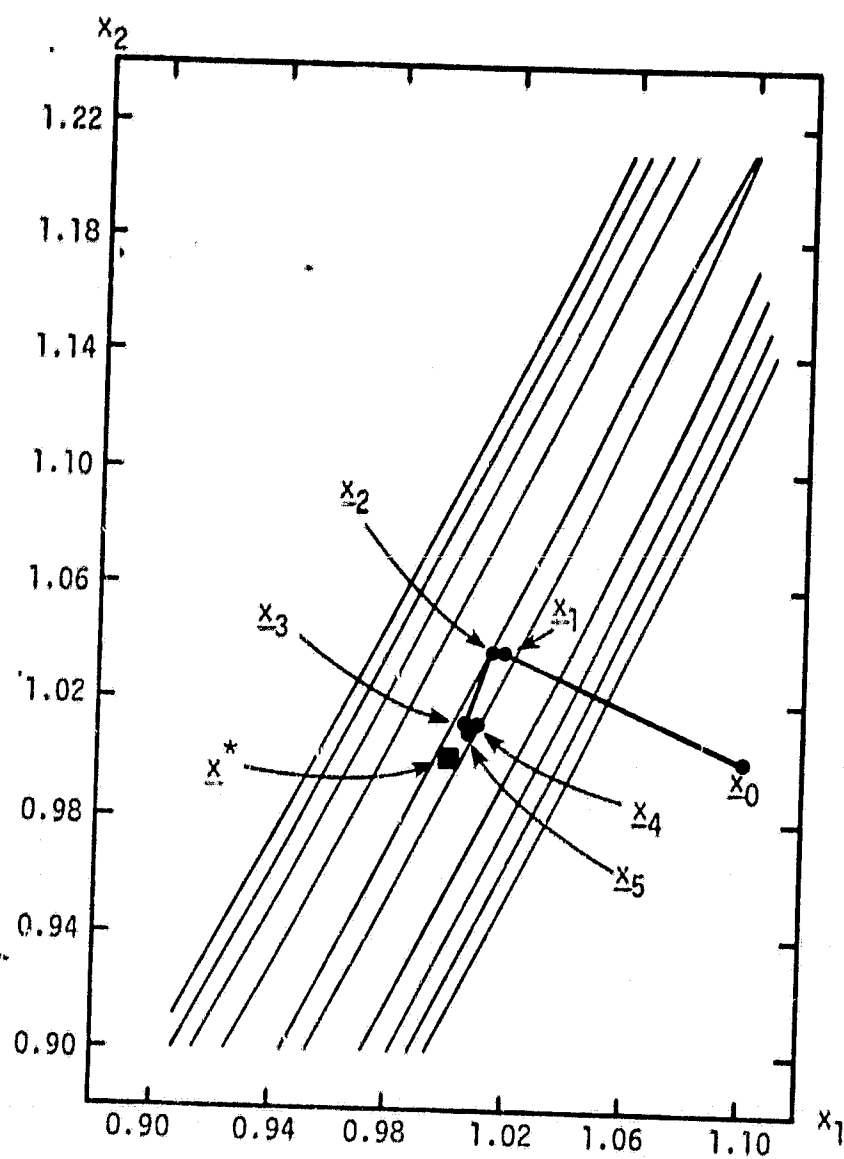


Figure 5.4 Modified D-F-P Example

For simple functions, different starting points may be used to find all of the local minima and hence, the global one. In Chapter VII, it will be seen that the cost function employed in this work has far too many local minima for this technique to be practical. A scheme for examining ranges of the variables will be shown to be applicable.

CHAPTER VI

THE COST FUNCTION

With analytic expressions for the noise annoyance (Chapter III) and the system constraints (Chapter IV), it is possible to formulate a mathematical criterion, the cost function, by which the amount of any change may be judged. As explained in the preceeding chapter, including the constraints as penalties in the cost function allows the use of standard, unconstrained optimization techniques, which are more practical for this work than are the so-called primal methods. While the annoyance is the primary quantity to be minimized (with the penalties as a necessary adjunct of the cost), it may be desired to include other properties of the system in the cost, such as the fuel and time of flight required for each trajectory.

The Penalty Functions

All of the constraints developed in Chapter IV, Eqns. (4-9) and (4-13) through (4-20) are in the form

$$h_i(\underline{x}) \leq c_i \quad (6-1)$$

as required by Eqn. (5-1), even though the unknown \underline{x} does not appear explicitly in any of them. Penalties in the form of Eqn. (5-3) may then be written for each constraint as follows:

[Dynamic and Passenger Comfort Constraint, Eqns. (4-9), (4-13)]

$$P_1 = \sum_{\text{traj.}} \sum_{\text{corner points}} (\max[0, \max\{\frac{\bar{v}^2}{c_2 g}, \frac{\bar{v}}{(c_1 + c_2 c_3) \min(\delta r_1 \delta r_2 \delta r_3)}\} - R])^2 \quad (6-2)$$

$$P_2 = \sum_{\text{traj. segments}} (\max[0, \tan \gamma_{\text{dmax}} - \frac{dz}{dl}, \frac{dz}{dl} - \tan \gamma_{\text{cmax}}])^2 \quad (6-3)$$

[Geometric Feasibility Constraints, Eqns. (4-14), (4-15)]

$$P_3 = \sum_{\text{traj. corner points}} (\max[0, \left[\frac{R_{\min}}{\tan \frac{\alpha}{2}} \right] + \left[\frac{R_{\min}}{\tan \frac{\beta}{2}} \right] - l_2])^2 \quad (6-4)$$

$$P_4 = \sum_{\text{traj.}} (\max[0, \frac{R_{\min}}{\tan \frac{\alpha}{2}} - l_1])^2 \quad (6-5)$$

[Separation and Crossover Constraints, Eqn. (4-16), (4-17)]

$$P_5 = \sum_{\substack{\text{all traj.} \\ \text{pairs}}} \sum_{i=1}^s (\max[0, d_{\min} - \hat{y}_i])^2 \quad (6-6)$$

$$P_6 = \sum_{\substack{\text{all traj.} \\ \text{pairs}}} \sum_{i=1}^{s-1} (\max[0, -\hat{y}_i \hat{y}_{i+1}])^2 \quad (6-7)$$

[Final Heading Constraint, Eqns. (4-18), (4-19)]

$$P_7 = \sum_{\text{traj.}} (\max[0, 135^\circ - \delta])^2 \quad (6-8)$$

$$P_8 = \sum_{\text{traj.}} (\max[0, l_1 - \frac{1}{2}l])^2 \quad (6-9)$$

[Threshold Constraint, Eqn. (4-20)]

$$P_9 = \sum_{j \in J} (\max[0, \{ \sum_{i=N^*}^{N_f} \max(0, L_{A, \max, i} - L_{A, \text{Thresh}}) \}])^2 \quad (6-10)$$

The Total Cost

Adjoining the penalties to the desired annoyance measures, say NII, gives the total cost function:

$$f(\underline{x}) = w_A A(\underline{x}) + \underline{\mu}^T \underline{P}(\underline{x}) \quad (6-11)$$

where $A(\underline{x})$ is the annoyance (NII), w_A is the annoyance weight, $\underline{\mu}$ is the m-vector of penalty weights, and $\underline{P}(\underline{x})$ is the m-vector of penalties (there are m constraints). As before, the variable \underline{x} is present only in a formal sense; it does not appear explicitly in these terms in this work.

The annoyance weight w_A governs the relative importance of the annoyance, keeping it from being dominated by the penalty terms. This is most important in cases where the optimal solution is on a constraint boundary. In such a region, $\underline{\mu}^T \underline{P}(\underline{x})$ will in general be nonzero, but as discussed in Chapter V, small values are acceptable. The purpose of w_A then is to insure that whatever reduction can be made in the total cost corresponds to a decrease in the annoyance, rather than a decrease in penalties that represent tolerable constraint violations. When the constraints are violated excessively, the attendant large penalties naturally should comprise the dominant portion of the cost.

Experimentally, the following values of w_A and the μ_i have been found to give satisfactory results when NII is used for the annoyance measure:

$\mu_1 = 10^4 - 10^6$	(lateral dynamic, passenger comfort)
$\mu_2 = 10^2 - 10^6$	(longitudinal dynamic)
$\mu_3 = 10^4 - 10^6$	(geometric feasibility)
$\mu_4 = 10^4 - 10^6$	(geometric feasibility, first segment)

$\mu_5 = 10^3$	(separation)
$\mu_6 = 10^3$	(crossover)
$\mu_7 = 10^4 - 10^6$	(final heading)
$\mu_8 = 10^4 - 10^6$	(final heading)
$\mu_9 = 10^3$	(threshold noise).

The rather wide range in μ_2 merits some explanation. It has been often found¹ in this work that the optimum trajectories will have the maximum vertical slopes allowed by the longitudinal dynamic constraint, Eqn. (4-9). This is intuitively expected, since the maximum slopes give the highest average altitude, the largest average distances from all the populace, and hence, the lowest average noise levels. Because of this tendency in the solutions, the associated penalty (P_2) generally is non-zero. Starting with a small value of μ_2 allows the optimization to work on the ground tracks without hinderance from penalties on the profiles (vertical components). After several iterations, μ_2 may be increased if excessive longitudinal constraint violations are occurring.

Modifying the Cost Function: Inclusion of Fuel and Time of Flight

An example of altering the cost function to include other attributes is a study conducted to examine the relationship between the noise

¹ In cases where a trajectory segment passes over a region of low population density, increasing its slope does not significantly lower the noise levels and hence, the total community annoyance. The optimization stops when the change in cost between successive iterations is less than some criterion. Therefore such a segment may not be exactly optimum, but practically it is so. Fuel consumption rates that vary with climb and descent rates are not included in this analysis.

annoyance, and the fuel and time of flight needed. It is assumed that the latter two vary linearly with the flight path length only. That is, constant power settings and noise propagation characteristics are used. The resulting term to be added to the cost is

$$f_{F,TOF}(\underline{x}) = \sum_{\text{traj.}} W_{F,TOF} l_T \quad (6-12)$$

where $W_{F,TOF}$ is the fuel and time of flight weight, and l_T is the trajectory length. Eqn. (6-11) becomes

$$f_1(\underline{x}) = w_A A(\underline{x}) + f_{F,TOF}(\underline{x}) + \mu^T P(\underline{x}). \quad (6-13)$$

Using $f_1(\underline{x})$ as the cost function, a relation between the annoyance and $f_{F,TOF}$ was obtained by varying $w_{F,TOF}$. This investigation is reported in Chapter VIII.

CHAPTER VII

INTEGRATION OF THE MODELS

The annoyance, flight path, and cost function models are combined with the chosen optimization method to create a system for analyzing the aircraft noise annoyance problem and the feasibility of remedying it through trajectory modifications. While the components of the system are essentially modular (and hence, capable of being altered individually without replacing the entire system), some linkage among them is required. Further, some special considerations must be given to larger airports and to some population distributions.

The Flyover/Noise Simulation: Small Airports

At any airport, there are, in general, a number of flight paths (both for takeoffs and landings) in use on a given day, with different types and numbers of aircraft using these trajectories. In this work, it is assumed that for the period of time under consideration, the type of each path (takeoff or landing) and the runway to which it is assigned do not change. In addition, the noise level data used is that for landing aircraft with constant power and control settings.¹

Calculation of the annoyance measure requires information about the distribution of the sound levels (L_{dn}) on the ground; for the current population distribution model, only the sound levels at the cell

¹ These assumptions are not fundamental to the operation of the system. A more sophisticated noise model, for instance, that employs thrust dependent noise data could be used instead of the present one.

centroids are needed. First, the position of each type of aircraft, on each trajectory, is sampled at equal time intervals (10 sec.). Then Eqn. (3-1); along with data from Table 3.1, is used to compute the sound level (L_A) at each centroid due to each type of aircraft, from every sample position and trajectory. This computational approach is practical so long as the number of flight paths and aircraft types is small. Referring to Appendix A, these levels for all of the aircraft are added on an energy basis,² and the average day-night level (L_{dn}) at each centroid may then be calculated from Eqn. (3-5).

Knowledge of the L_{dn} distribution, along with the population distribution, allows the calculation of either LWP (c.f. Eqn. (3-8)) or NII (c.f. Eqn. (3-9)); however, evaluation of HAPN is not necessarily as straightforward. This measure requires that the sound level (L_A) be calculated at each centroid, at each sampling instant, and that the maximum values (day and night) be retained. At smaller airports, where the flight operations are dispersed and non-overlapping in time, the L_A values depend upon only one aircraft at any sampling instant. For a larger number of operations where the overlap (in time) is unavoidable, the sequence of flight operations must be explicitly included in the L_A calculations, since at every sampling instant the types of

² It is not necessary to compute L_A (at each centroid) for every sampling instant, since these levels will themselves be averaged (using energy addition) to get L_{dn} . Only the energy term, Eqn. (A-4), must be computed and accumulated (correctly weighted, c.f. Eqn. (3-5)) into the total centroid energy for every sampling instant. If n aircraft of the same type fly on a give trajectory, then their total energy contribution at any centroid is n times the energy added by one aircraft.

aircraft flying will determine the value of L_A at any centroid. In this study, only a situation of the former type has been considered for the purpose of comparing HAPN with LWP and NII.

Large Airports and the Equivalent Aircraft Concept

Summing (on an energy basis) over all trajectories, sample points, and aircraft types may become impractical as the number of any of these grows, depending upon the computing resources available. At large airports, such as the Lambert-St. Louis facility in St. Louis, Missouri, the number of trajectories and aircraft (fifteen paths and five hundred operations per day) precludes the straightforward approach of the last section. In such a case, the summation over aircraft types may be removed by replacing the mixture of planes on each trajectory with a single equivalent aircraft, whose sound level distribution is the same as if all the aircraft assigned to that path were flying together on it. For this treatment to be valid, it is necessary that either NII or LWP be used as the annoyance measure. The reason is that both of them are functions of the L_{dn} distribution, which depends only on the total sound energy distribution, integrated over time (one 24-hour period) and not on the time history of the energy contributions (except for the distinction between day and night occurrences).

The exact expression for the total sound level (using any power intensity related scale) contributed from n sources is (see Appendix A)

$$L_{TOT} = 10 \cdot \log_{10} \sum_{i=1}^n 10^{L_i/10} \quad (A-6)$$

with, from Eqn. (3-1)

$$L_i = c_{1,i} - c_{2,i} \cdot \log_{10} r, \quad (7-1)$$

where the subscript i is the aircraft index and n is the total number of aircraft. A simple, approximate expression for Eqn. (A-6) may be obtained in the following way. For all the aircraft on a given trajectory, the appropriate L_i 's are added (on an energy basis, and using the L_{dn} convention which weights nighttime energy terms with a factor of ten), i.e. an equivalent level L^* may be written

$$L_k^*(r) = 10 \cdot \log_{10} \sum_{i=1}^n w_i 10^{L_i(r)/10} \quad (7-2)$$

where the subscript k denotes that this is the L^* contribution from the k -th trajectory, and w_i is the same time-of-day factor as in Eqn. (3-5). If $L_k^*(r)$ vs. r is plotted for a variety of values of r , an excellent approximation of the resulting data is the expression

$$L_k^* \simeq c_{1,k,eq} - c_{2,k,eq} \cdot \log_{10}(r), \quad (7-3)$$

where $c_{1,k,eq}$ and $c_{2,k,eq}$ are the equivalent aircraft noise coefficients for the k -th trajectory. An example of this is shown in Figure 7.1. The largest value of r was chosen to be 60,000 meters, since that is the largest value of the range that would be encountered in the near zone area (32,000 meter radius) of an airport. In this example, the estimated correlation coefficient squared in the least square-error fit (dotted line) was $r^2 = 0.9995$. (Spacing the abscissa values such that $r_{j+1} = ar_j$, where a is a constant, results in their being equally spaced on a logarithmic scale, and gives a better curve fit.)

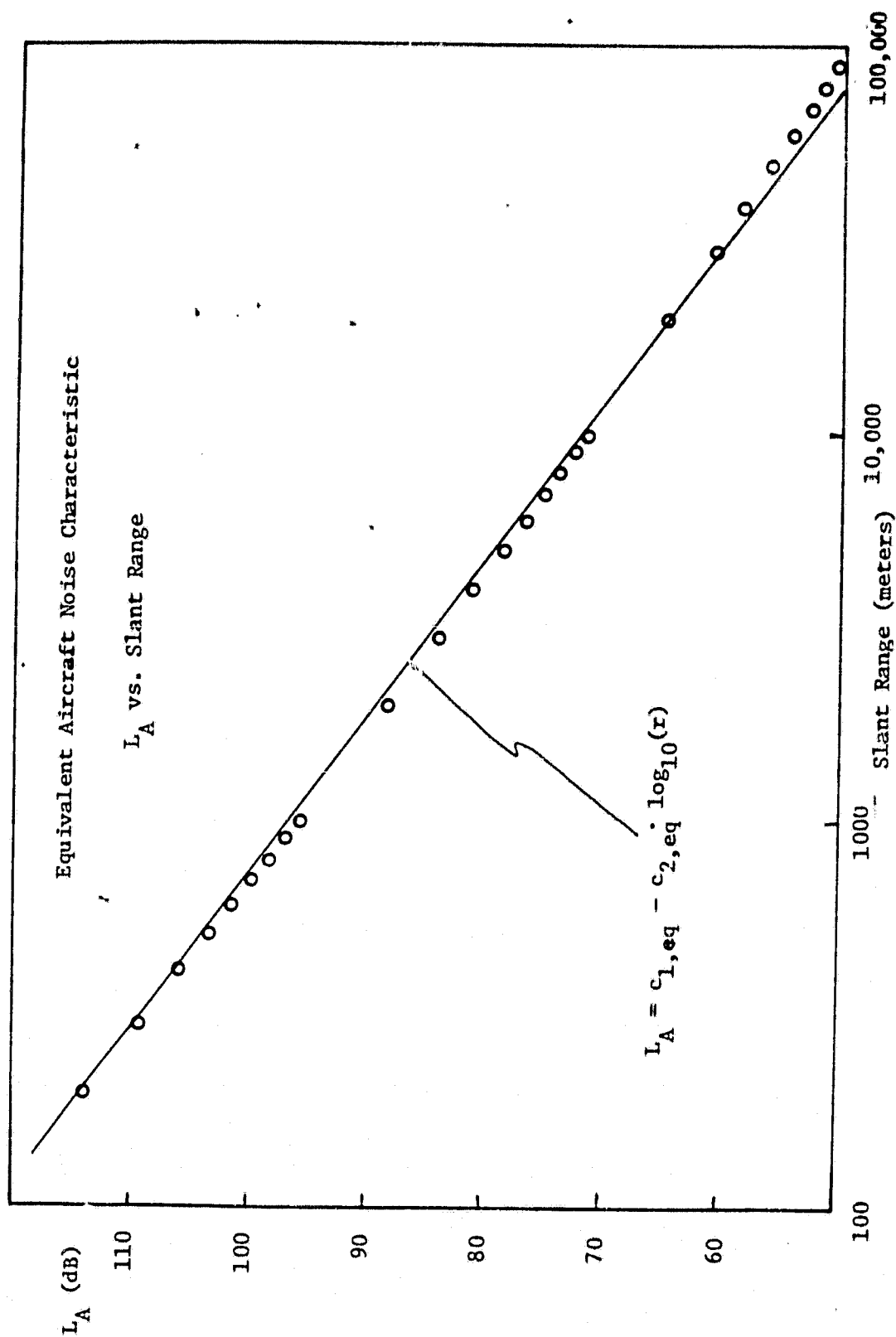


Figure 7.1 Equivalent Aircraft Noise Data Fit

With the coefficients $c_{1,k,eq}$ and $c_{2,k,eq}$ for all the flight paths, the total L^* at any point on the ground, at each sampling instant is just the sum (on an energy basis):

$$L_t^* = 10 \cdot \log_{10} \sum_{k=1}^{N_T} 10^{L_{k,t}^*/10}, \quad (7-4)$$

where N_T is the number of trajectories, and t is the sampling index. Since the time-of-day weighting has already been included in L_t^* , the expression for L_{dn} becomes (c.f. En. (3-5))

$$L_{dn} = 10 \cdot \log_{10} \frac{\sum_{t=1}^N 10^{L_t^*/10}}{N}, \quad (7-5)$$

where N is the number of noise level samples.

This method of using equivalent aircraft coefficients reduces the number of calculations needed to determine L_{dn} at each grid cell by a factor approximately equal to the number of aircraft types that are operating in an area. Since at many airports this number is frequently on the order of ten, a considerable savings in computation time may be achieved. (Minimal computation time is required to determine the equivalent aircraft noise coefficients.)

As it is the method relies upon the assumption that $c_{1,i}$ and $c_{2,i}$ are constant for each aircraft type. If varying thrust levels and control settings are to be considered, it might be possible to make these coefficients functions of time, and for a given airport, predict their values (and hence, those of the equivalent aircraft coefficients) by estimating the time histories of the thrust and controls. A second disadvantage of the method is that L_A values are not computed at each

sampling instant. Thus, the threshold noise constraint either must be omitted or altered so that the limitation is in terms of L_{dn} in each grid cell, rather than the maximum A-level experienced. Finally, the errors introduced in the L_{dn} values by the equivalent aircraft method are small (≤ 1 dB); however, the nonlinear intensity weighting function, $W(L_{dn})$, may amplify these errors, giving NII and LWP values that differ by several percent from those calculated exactly. The amount of error in NII and LWP changes with the sets of trajectories, but the locations of the local minima do not differ significantly in comparisons of the exact and approximate calculations.¹

Equivalent Aircraft Coefficients and the Assignment Problem

An interesting feature of the equivalent aircraft method is that the noise level coefficient $c_{1,k,eq}$ (k is the trajectory index) has a strong dependence only on the total number of aircraft on a particular flight path, as long as there is a uniform mixture of aircraft types present. A plot of $c_{1,k,eq}$ vs. the L_{dn} -equivalent number² of aircraft per trajectory (for the Lambert/St. Louis International Airport) is shown in Figure 7.2. The dotted line is the least square-error fit and has the form

¹ For example, using the Phoenix Sky Harbor Airport case, the error in NII for the nominal (currently specified) trajectories is 2.1% and for the optimum set, 0.2%. The corner pts. (equivalent aircraft method) differ from the optimum ones (exact calculations) by no more than 100 meters.

² The L_{dn} convention of weighting a nighttime sound level with a factor ten is equivalent to adding the sound levels, on an energy basis, of ten identical sound level occurrences. Hence, one nighttime flight is equivalent (in L_{dn} calculations) to ten daytime flights.

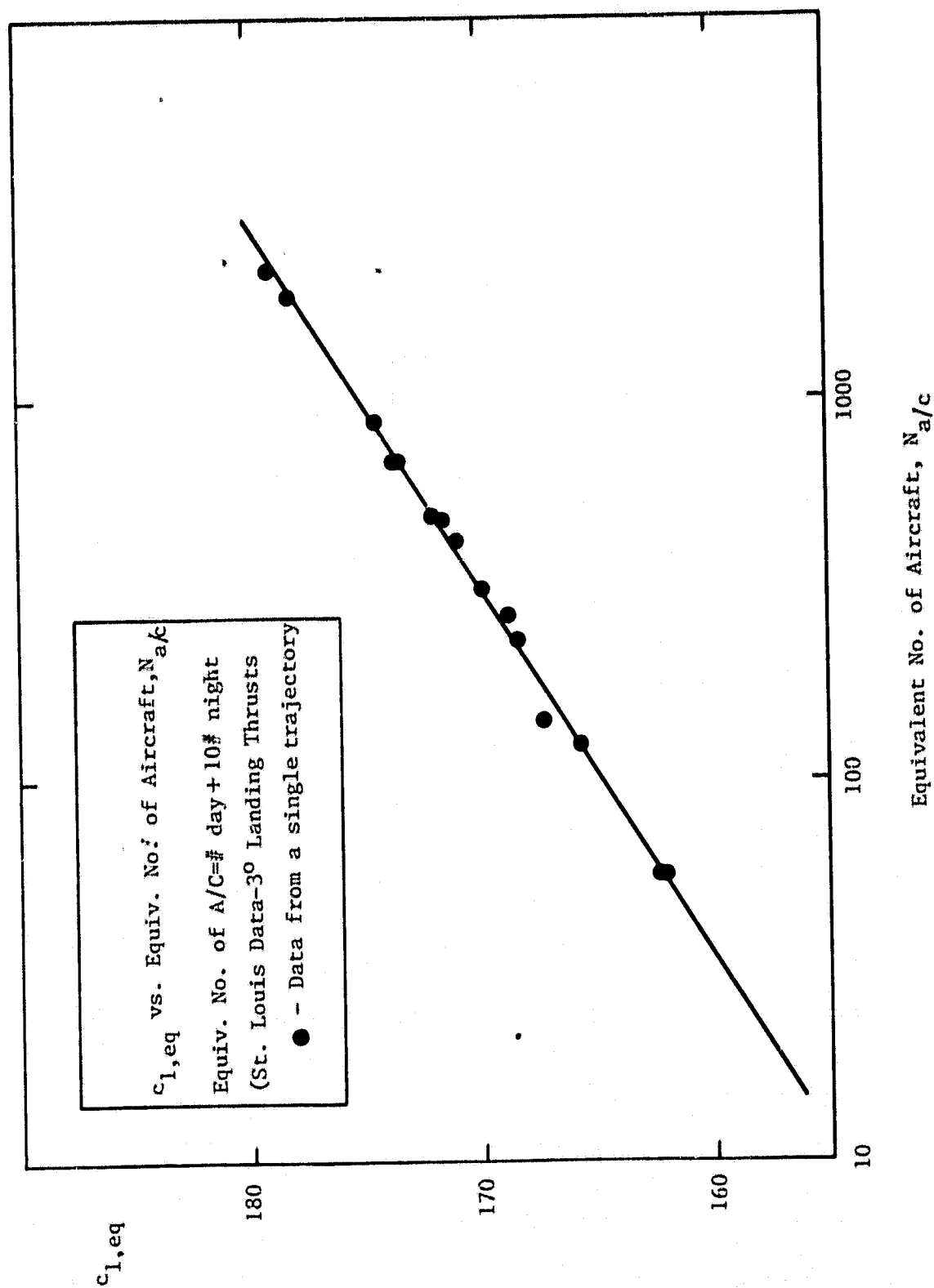


Figure 7.2 Plot of $c_{l,eq}$ vs. Equivalent Number of Aircraft

$$c_{1,k,eq} = a + b \cdot \log_{10} N_{a/c} , \quad (7-6)$$

where $N_{a/c}$ is the L_{dn} -equivalent number of aircraft per trajectory and in this case, $a = 144.65$ and $b = 10.16$. This fit has an estimated correlation coefficient squared, $r^2 = 0.993$. As for the coefficient $c_{2,k,eq}$, the data (also for the Lambert-St. Louis case) are not as strongly correlated with $N_{a/c}$. Figure 7.3 shows a log-log plot of $c_{2,k,eq}$ vs. $N_{a/c}$, with the least square-error fit (dotted line) for $N_{a/c} \geq 250$. The functional form is

$$c_{2,k,eq} = a N_{a/c}^b , \quad (7-7)$$

where $a = 23.44$ and $b = 7.89 \times 10^{-3}$; the estimated correlation coefficient squared is $r^2 = 0.894$. Restricting $N_{a/c}$ to values greater than 250 allows for a good fit of the data with a simple functional form. For $N_{a/c} < 250$, Eqn. (7-7) will not accurately predict $c_{2,k,eq}$, and an individual fit of the data, as in Figure 7.1, will be necessary.

These relations between the coefficients and $N_{a/c}$, Eqns. (7-6) and (7-7), have a bearing on the problem of how to assign aircraft to the different trajectories, with the community annoyance from noise as the criterion.

In this work, the assignment problem is not considered (the assignments are assumed to be known and remain constant for a given airport); however, it will be a necessary addition in subsequent refinements of this study. The aforementioned relations may be useful in formulating a performance criterion to be used in the initial assignment of aircraft, or in altering an existing plan.

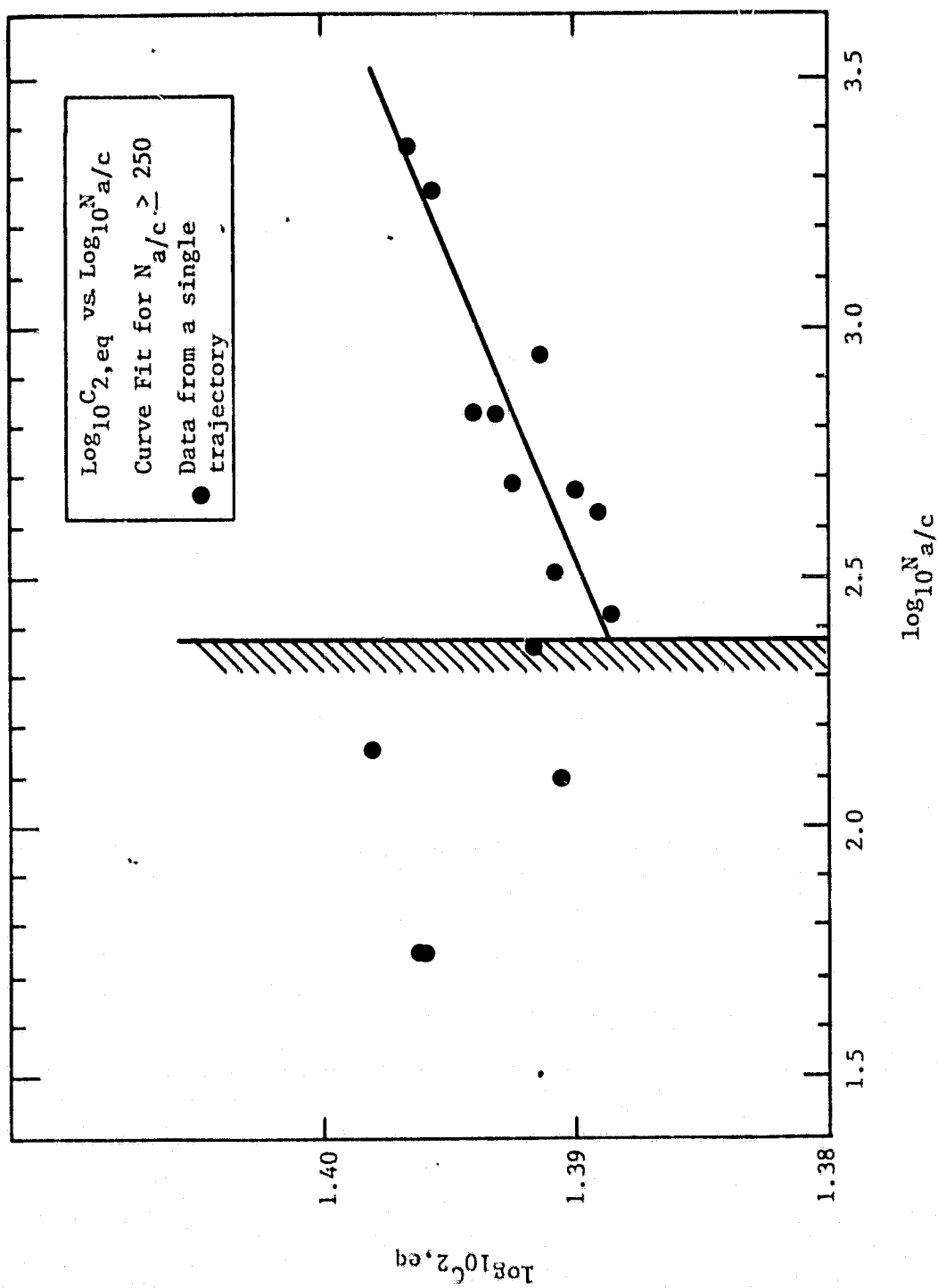


Figure 7.3 Plot of $\log_{10} C_{2,eq}$ vs. $\log_{10} N a/c$

The Integrated System

A general summary of the systems operation is provided by the flowchart in Figure 7.4.

The input required is: 1) the population distribution data, 2) the number of each aircraft type and their time-of-day designations (day or night), 3) the noise level characteristics (either by individual aircraft type or equivalent aircraft noise coefficients), 4) the constraint data (minimum turning radius, threshold noise level, etc., and penalty weights), 5) an initial set of flight paths (specified by their end points and corner points), and 6) the stopping criteria (the maximum number of iterations, and the lower limit on relative changes in the cost function).

For the initial trajectories, the total cost must be evaluated, before starting the iterative search. (If the system is being used only for evaluation of the trajectories, then no search is performed, and the process stops after this step.) First, the annoyance measure (e.g., NII, LWP, HAPN) is evaluated. Then the constraints are tested: a violated constraint results in a positive penalty; otherwise, the penalty is zero. Finally, the total cost is computed by adding the penalties to the annoyance. As shown in Chapter VI, other cost terms may be added to the total; the results of one such modification appear in Chapter VIII.

Now the optimization loop begins. After each search step, a test is made to determine if any of the stopping criterion is satisfied. If not, the iteration continues; if so, the process stops and the "minimum" annoyance trajectories are determined. In fact, the set determined may

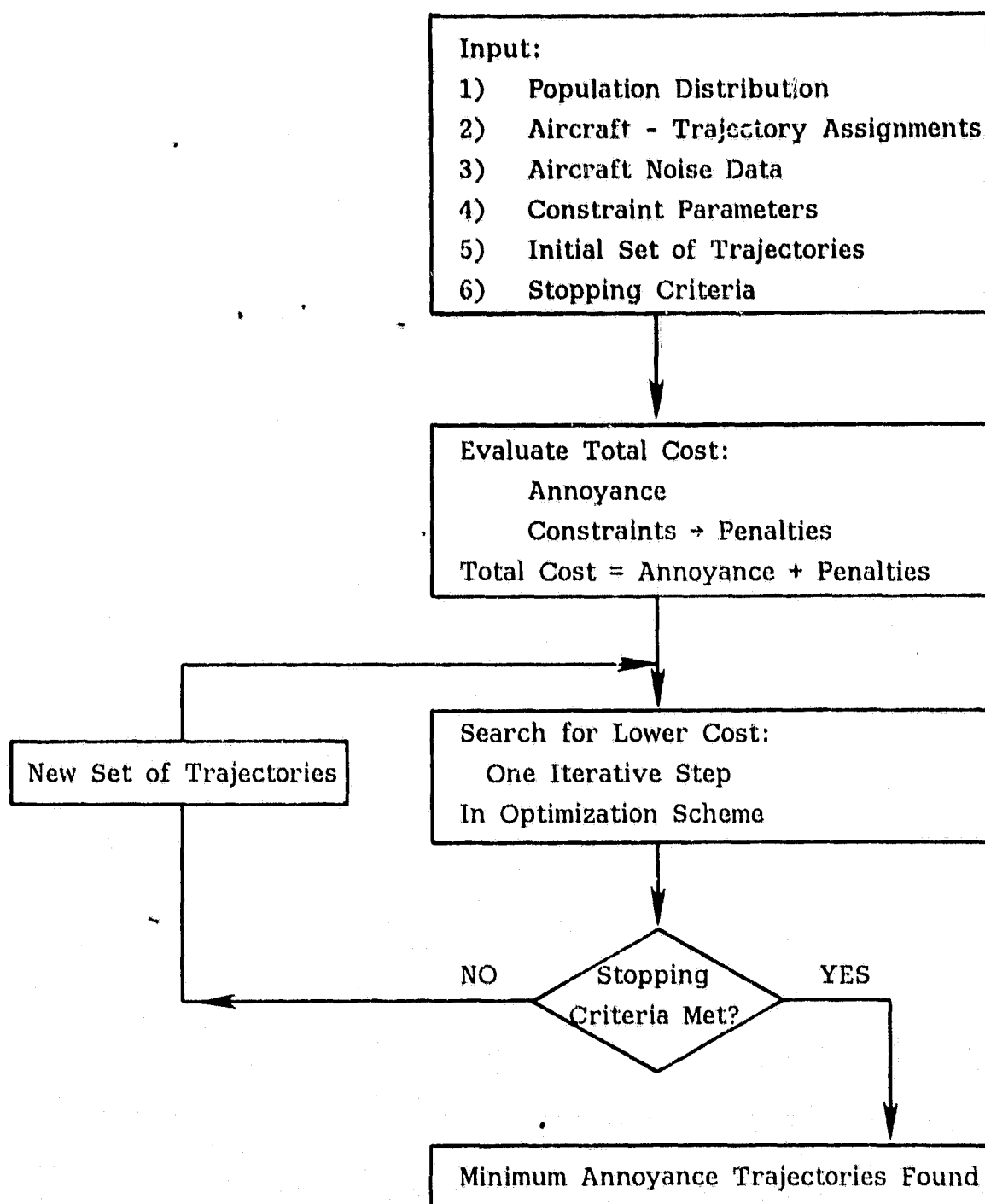


Figure 7.4 System Flowchart

be only a locally minimum set (a problem discussed in the next section), or perhaps not even that (the maximum limit on the number iterations allowed may have been reached); however, because of the type of optimization method used (Chapter V), each successive set of paths will result in a lower total cost (and a lower annoyance, if the penalties are zero).

Besides a limit on the number of iterations, the stopping criteria include the tests: 1) Is the gradient of the cost function zero? (i.e., the search has found a local minimum), and 2) Is the relative difference between successive values of the cost negligible? (i.e., the search has come close enough³ to a local minimum to warrant no further searching).

Local Minima and the Population Distribution

A problem with the optimization, already expressed in Chapter V, is that of local minima in the cost function. As an example of the problem, consider a cost function of two variables, for which the contours are plotted in Figure 7.5. Points A, B, and C are the local minima; point B corresponds to the lowest value of the cost, and hence, is the global minimum. The difficulty associated with this situation is that of finding point B by use of any of the optimization methods discussed in Chapter V. Given an initial point from which to search, each

³ The error in the solution, $E = \underline{x}^k - \underline{x}^*$, where \underline{x}^k is the k-th iterative solution and \underline{x}^* is the exact solution, may in fact be large. If g_k is small, though, and the relative difference between successive costs is small, then further searching may not lead to a significant reduction in the cost. Lacking other information, it may be assumed that, practically, a local minimum has been located.

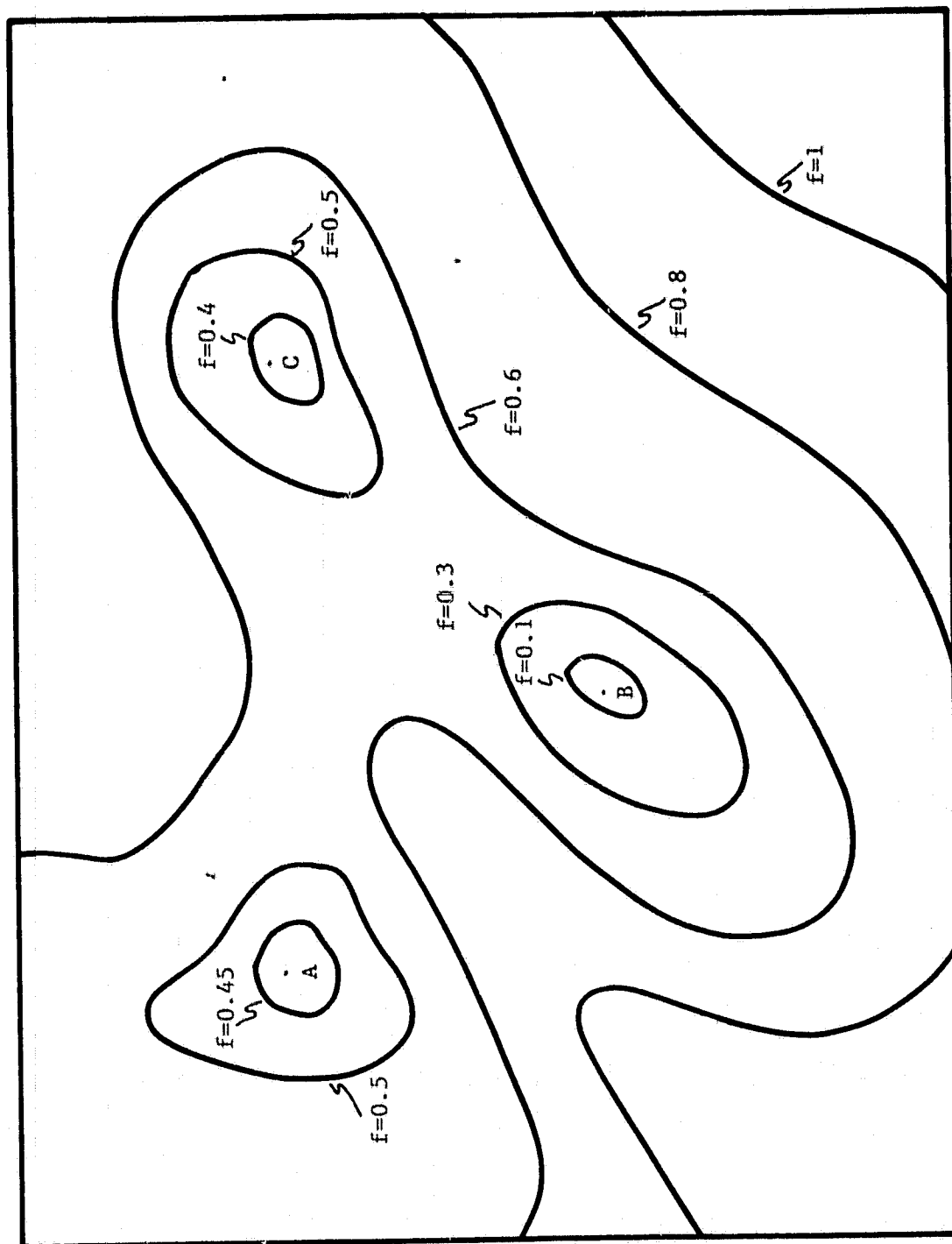


Figure 7.5 Several Local Minima

of those methods will converge to one of the local minima, but will not automatically find the global minimum. Which local minimum is found depends upon the starting point in the search. The conventional technique used to find the global minimum is to perform the optimization a number of times, from different initial points. In cases where the cost function has only a few minima, this approach may indeed allow all of them to be found. If the total number of local minima is unknown, but judged to be small, it is often possible to estimate the probability of overlooking a number of minima. That probability may then be made acceptably small by repeating the optimization with a large enough number of starting points. Ref. [35] and [36] are good examples of this approach.

The cost function used in this work has an extremely large number of local minima. This is demonstrated by considering the population distribution shown in Figure 7.6a along with the simple straight-path trajectory. This path is at a constant altitude (300 m.). There are 1000 people at each dot on the map. In Figure 7.6b, the resulting NII values versus θ are plotted (a typical mixture of commercial aircraft is assumed for generating the noise levels).

Even though NII is computed from noise levels experienced by the entire community, the local minima exhibited in Figure 7.6b demonstrate the influence of high levels that occur locally within the community. For instance, at $\theta_1 = 20^\circ$, the trajectory passes close to one population center and almost directly over another; the resulting noise level at these points, and hence, their contributions to the NII will be high: a peak occurs in Figure 7.6b at $\theta_1 = 20^\circ$. Conversely, at $\theta_2 = 26^\circ$, the

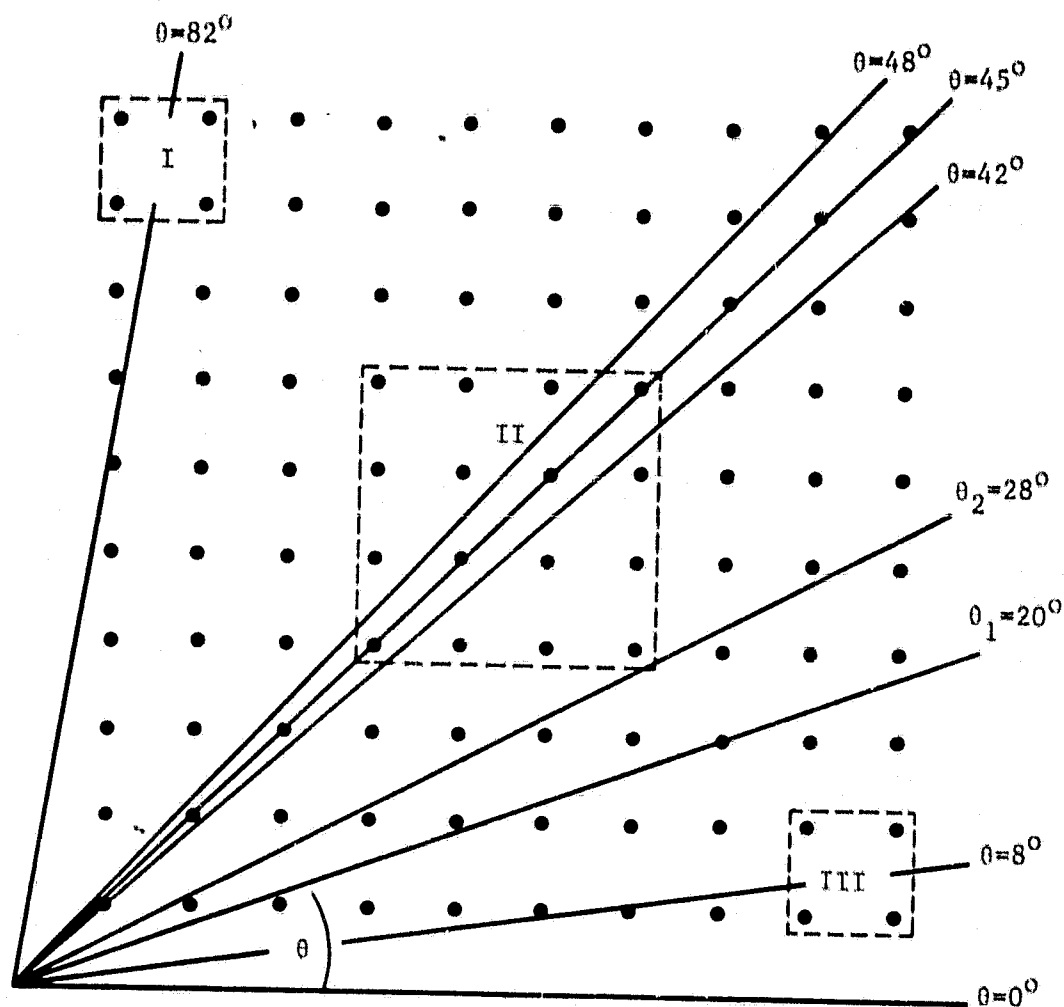


Figure 7.6a Uniform Discrete Population Distribution and Straight Flight Paths

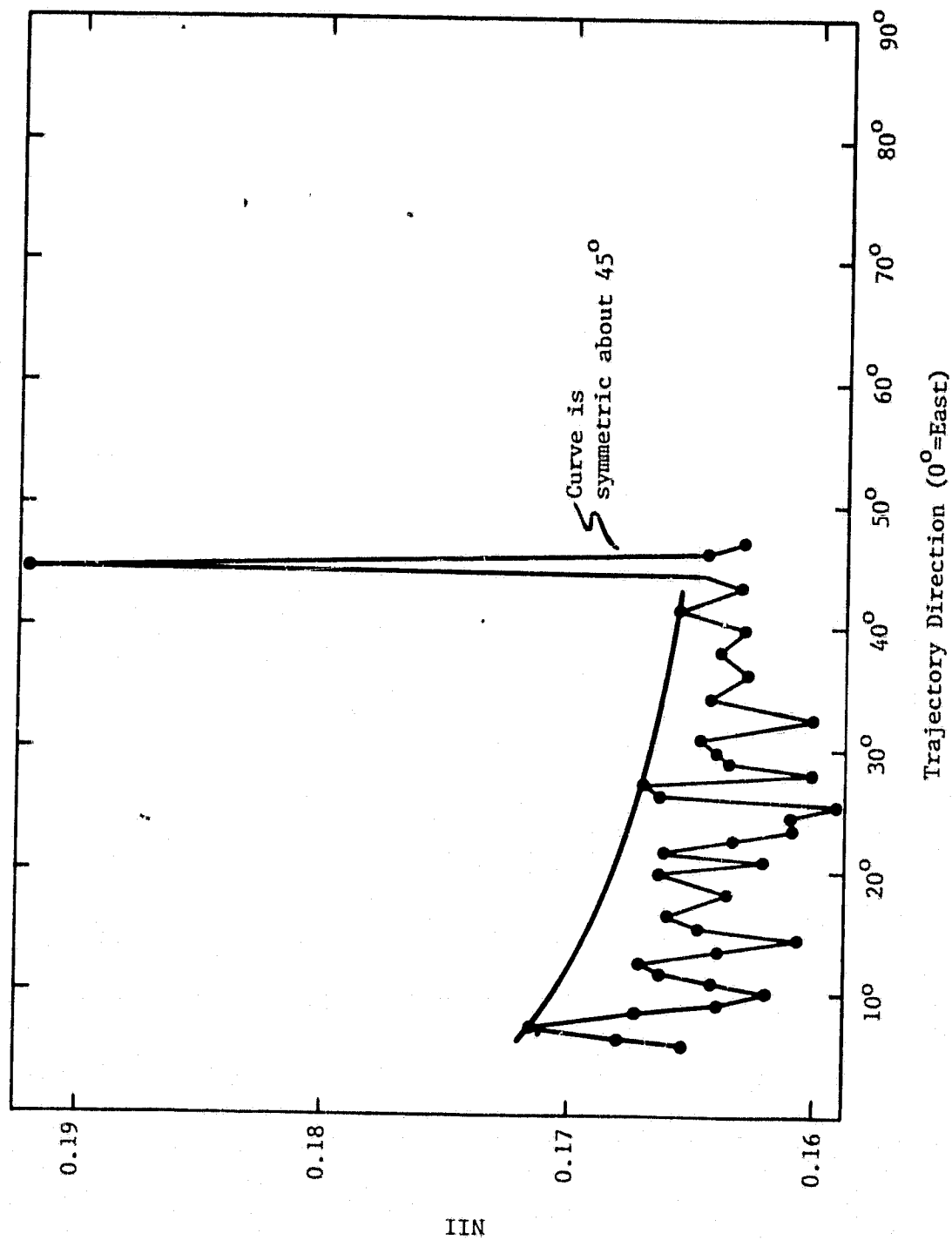


Figure 7.6b NII vs. Trajectory Direction

population points nearest to the trajectory are farther away from it than those in the first case. The noise levels they experience are lower, and a minimum in the NII occurs at $\theta_2 = 26^\circ$. (At $\theta = 45^\circ$, the sharp peak is a local effect caused by the large number of population centers directly beneath the path.)

This example demonstrates two important problems. First, there are many local minima in the cost function, caused mostly by the use of a particular representation of the population distribution.⁴ In a realistic case, there are more population centers and trajectories, which lead to an even larger number of local minima. Second, the arbitrary lower limit on L_{dn} values to consider (55 dB) for use in the NII has altered the results that would be expected intuitively. The envelope of NII peaks in Figure 7.6b is seen to decrease as θ increases, i.e., as more of the population receives higher noise intensities, the community annoyance decreases. This occurs because the denominator in Eqn. (3-9) increases, in this case, faster than LWP. Once again, the inappropriateness of NII for use in comparing the annoyance from different trajectories in a given community is in evidence.

A dramatic change in the number of local minima in this example occurs if NII is calculated without a lower limit on the L_{dn} interval (c.f. Eqn. (3-9)) but with the total population fixed. This modified version of NII will be denoted NII' . Figure 7.6c shows the results

⁴ In each population grid cell, the people are assumed to be concentrated at the geometric centroid of the cell. The overall result is a discontinuous (discrete) density distribution.

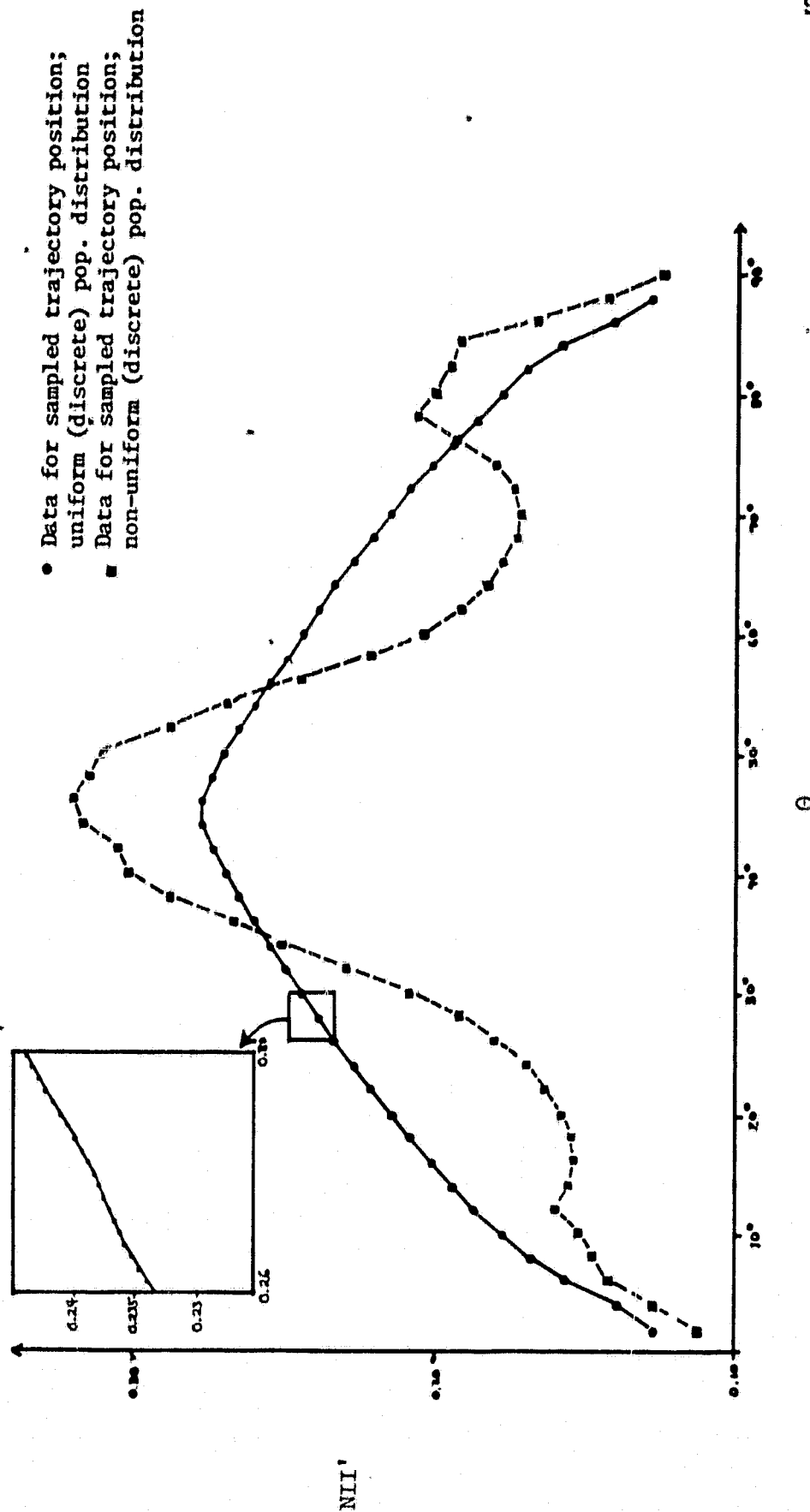


Figure 7.6c NII' vs. Trajectory Direction

(solid line). The complete disappearance of all the local minima is fortuitous in this case (the influence of local effects still persists, though, as seen in the inset); it occurs as a consequence of the uniform (discrete) distribution of people and also because of the particular degree of freedom given to trajectory variations here.⁵

There remains the dilemma of local minima, caused by macroscopic variations in the population density. Again, referring to Figure 7.6a, assume that in Region I, there are 20,000 people at each population center; in Region II, 25,000 people at each center; in Region III, 10,000 at each center; all other centers have 1,000 people. The resulting values of NII' vs. θ appear in Figure 7.6c (dotted line). At $\theta = 16^\circ$ and $\theta = 70^\circ$, there are local minima, caused by the lack of large numbers of people close to these paths. Because the truly large clusters of people are in Regions I, II and III, it is reasonable to conclude that these minima are real effects of the population distribution, and not artifices of its approximate representation. Note that at $\theta = 8^\circ, 82^\circ, 42^\circ, 48^\circ$, there are slight dips in NII' , caused by the flight path's avoidance of all the nearest population centers.

The possibility of artificial minima remains, though. Note that at $\theta = 8^\circ, 42^\circ, 48^\circ$ and 82° , there are inflections in NII' . Although not minimum points, there correspond closely to minima in Figure 7.6b. Further, no such behavior is seen in Figure 7.6c (solid line) for the

⁵ Only θ may vary, which limits the possible paths. In the next section, it will be seen that trajectories with several segments still have local minimum problems with this type of population distribution model.

uniform distribution. It may be suspected then that artificially produced minima may occur even with the use of NII', depending upon the actual distribution of people and the use of this population distribution model.

Global Searching

Regardless of the cause of the local minima, some procedure for determining the optimum solution must be developed. First, it must be recognized that finding the exact optimal set of flight paths is not as important as determining a set (if one exists) that produces a significant reduction in the annoyance measure, compared with the set presently used (nominal set). A given set will be considered optimum if there is not another set, significantly different, with a lower cost function value.

The definition of "significantly different" is developed in terms of the statistical lateral dispersion of individual flights from the specified trajectories. Galloway [37] states that the lateral dispersion of landing aircraft is approximately 750 m. at a distance of 16,000 m. from the airport. This corresponds to an angular dispersion of approximately 3°. For takeoffs, where the paths are less tightly controlled, the lateral dispersion is given in terms of the standard deviation (σ) of the side-line distance (y) from the specified path. For straight-out takeoffs, the standard deviation grows linearly with the distance l and is given by

$$\sigma_y = 0.08 l . \quad (7-8)$$

After a turn, the dispersion is expressed as

$$\sigma_y = 3.5 \times 10^3 \sigma_z / \bar{z} , \quad (7-9)$$

where σ_z/\bar{z} (the standard deviation in altitude dispersion divided by the average altitude during the turn) is generally a constant; the value $\sigma_z/\bar{z} \simeq 0.2$, for variable weight transports, will be used here.

Eqn. (7-8) may be used to estimate the angular dispersion on straight segments of a takeoff by assuming that all of the flights will lie within $2\sigma_y$ of the specified paths (a Gaussian distribution would contain 95% of the paths within $2\sigma_y$ of the mean path). The outermost path, to either side of a straight segment, will then form an angle, θ_{disp} , with, the segment, given by

$$\theta_{\text{disp}} = \tan^{-1} \frac{0.16}{1} = 9^\circ. \quad (7-10)$$

A set of trajectories will be considered "significantly different" from another set only if any path lies outside the dispersion limits of the corresponding path in the other set. As an example of how this may be used to restrict the amount of searching required in the optimization process, the flight paths at the Lambert-St. Louis International Airport will be considered. Figure 7.7a shows the twelve nominal takeoff paths and three nominal landing paths. The shaded regions represent the predicted dispersion about each path. Note that the trajectories with turns possess more average dispersion, c.f. Eqn. (7-9); this is due to the extra drifting during a turn.

Searching for the optimum trajectory set within the dispersion limits of this set is unprofitable since such an optimum would not by definition be significantly different from the nominal set. Of course, it may happen that the nominal set is the optimum, but regardless, any searching must occur "outside" the dispersion boundaries. Different

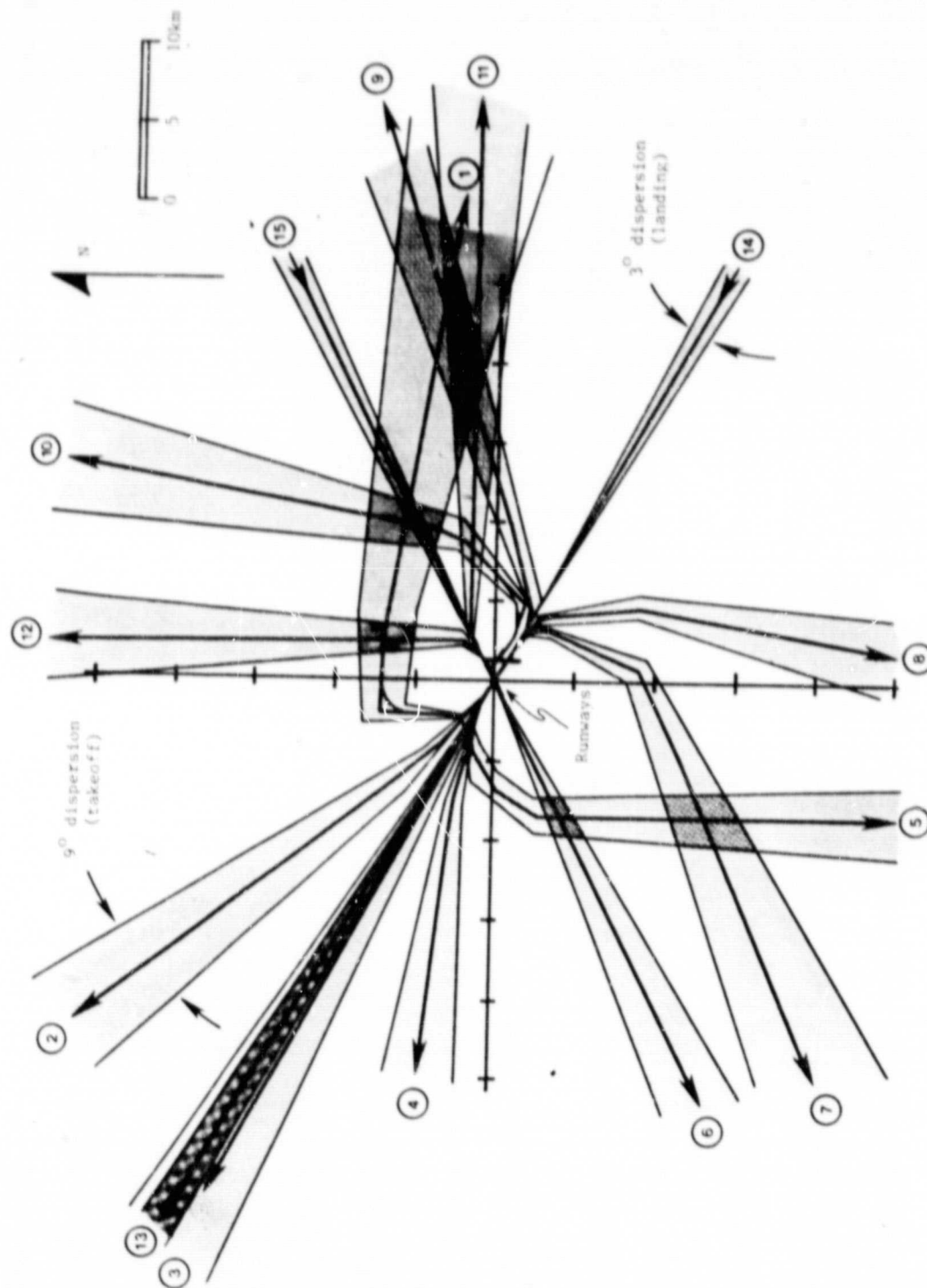


Figure 7.7a Lateral Dispersion in the Nominal Flight Paths at Lambert-St. Louis International Airport

starting points (representing significantly different trajectory sets) are used to perform the optimization a number of times. The lowest valued local minimum is then taken to be the global minimum.

Choices for the starting point are limited by two factors: the lateral dispersion on these paths and the constraints (particularly the aircraft dynamic and passenger comfort restrictions). For example as shown in Figure 7.7b, there are only five choices for track #7 starting points. Moving track #7 farther to the east would cause it to overlap track #8, in which case the two paths would be replaced by one. Consideration of cases farther to the north involves the same result with track #3. Thus, the presence of many trajectories limits the required amount of global searching, an important saving since the amount of computation needed for the optimization (local) increases with the number of flight paths. At smaller airports, where the number of trajectories is small, more starting points will be allowed, but the local optimization will require less computation.

In situations where the search has converged to a local minimum not significantly different from the starting point (but the existence of one is suspected), restarting the optimization after perturbing the corner points by some amount (~1,000 m.) may free the search from the region of the local minimum. Figure 7.8 shows an example of such a situation using a fictitious population distribution. Here, the short-range noise effects, as before, create a local minimum for the nominal landing path directly over the two lower population clusters. Repeated perturbation of the y-coordinates of the corner points eventually allows the optimization to find the global minimum. It is well outside of the

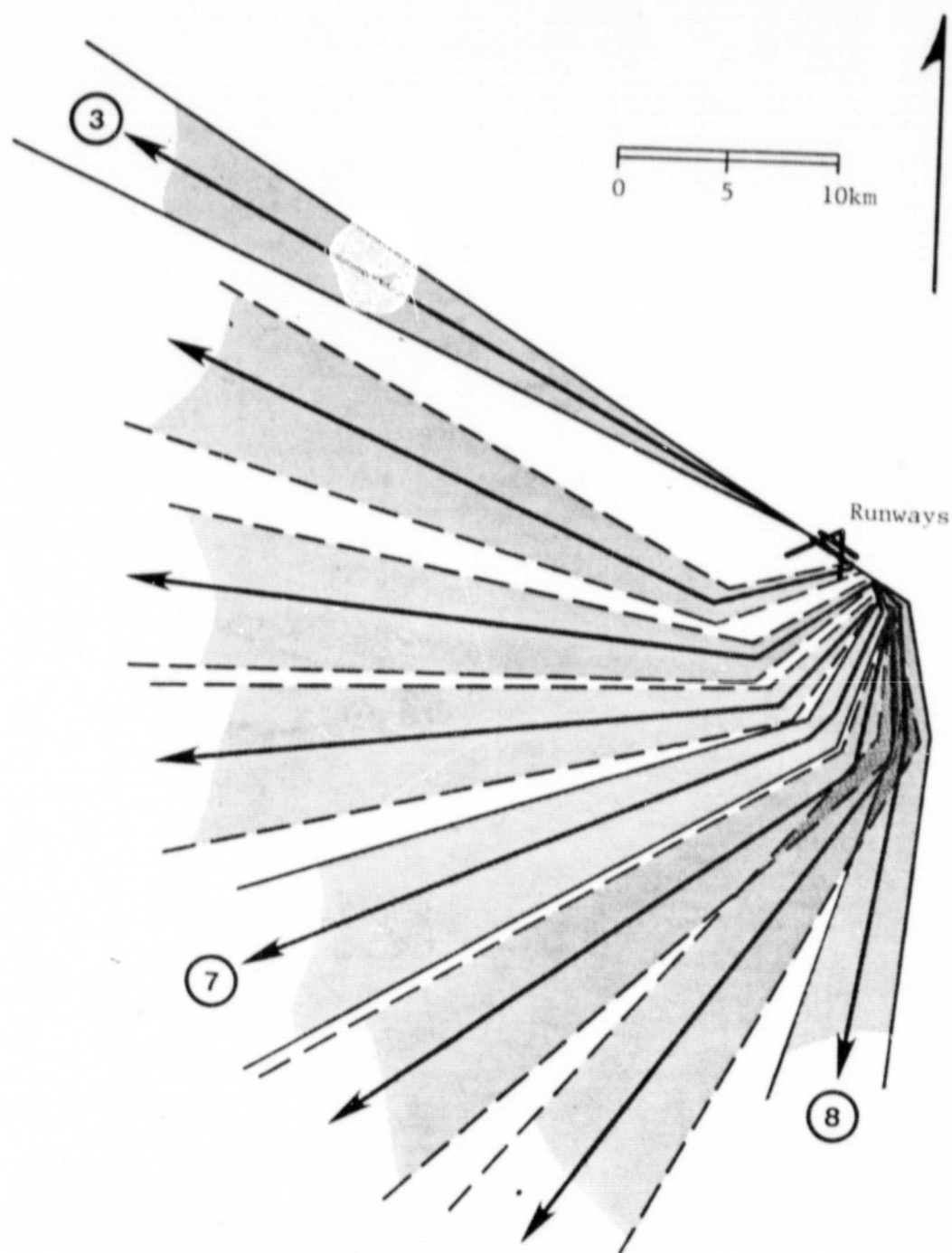


Figure 7.7b Starting Points for Track #7

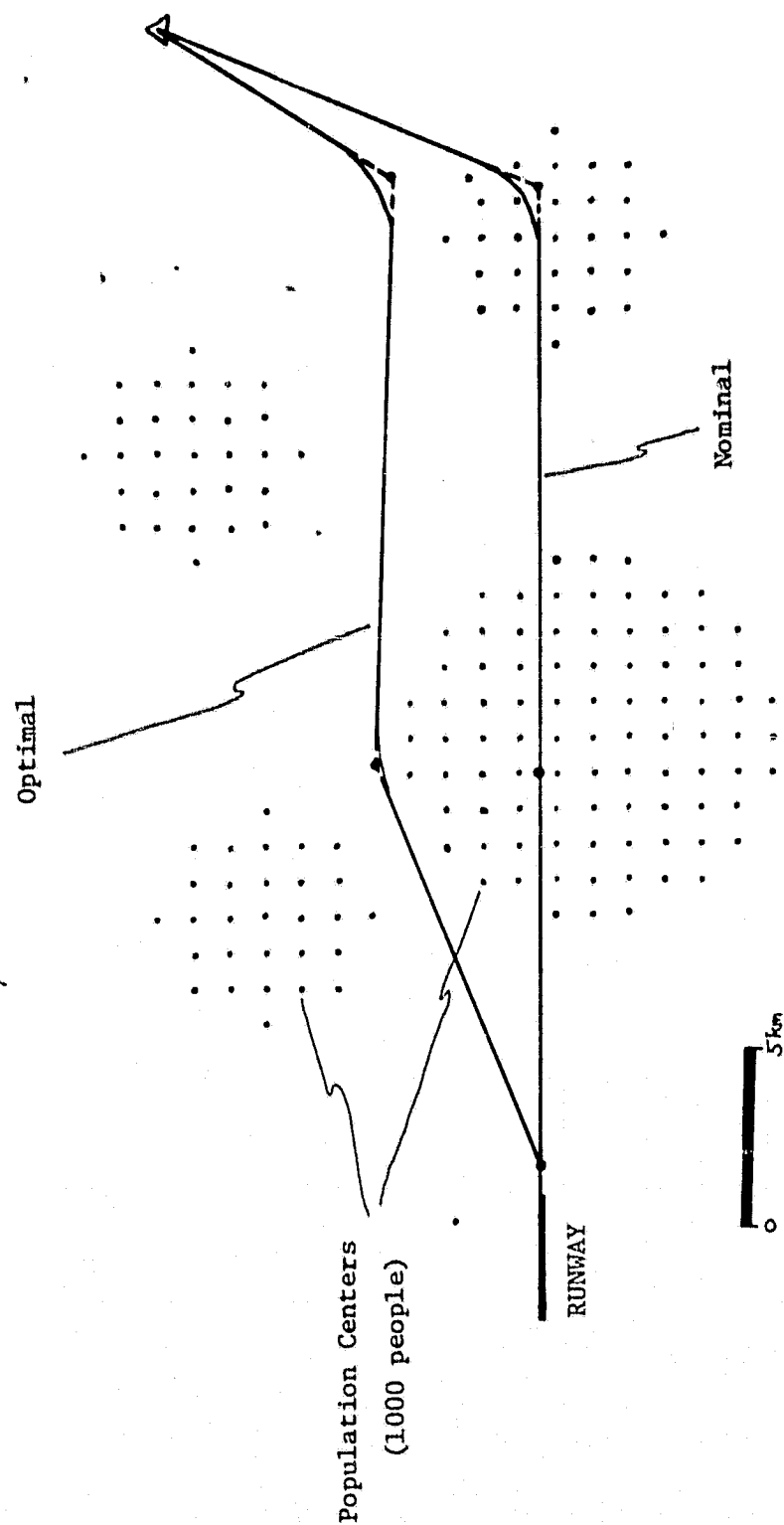


Figure 7.8 Local Minima in a Fictitious Population Distribution

dispersion limits of the nominal path and the reduction in NII' is 39%. The simplicity of the contrived problem provides for an easy decision about which way to perturb the corner points. In a real case, the direction may not be as obvious.

CHAPTER VIII

RESULTS

Two computer programs for implementing the system described in Chapter VII have been developed. The first, MANIP2, is a revision of MANIP (see Ref. [26]), with the following refinements: 1) optimization of a flight path's vertical profile is included, 2) takeoff and landing trajectories may be optimized, simultaneously, 3) a larger variety of commercial aircraft noise data is included, and 4) fuel and time of flight have been included in the cost function (as described in Chapter VI). The second program, CYCLIC, has the same features as MANIP2, except that 1) it employs the equivalent aircraft concept for computing noise levels, 2) the optimization scheme is modified to reduce storage requirements for large airports (i.e., many trajectories), and 3) fuel and time of flight are not included in the cost. A more complete description of CYCLIC is given in the Cyclic Optimization section of this chapter. Both programs were written in FORTRAN IV for use on CDC Cyber 170 series machines. Appendix D contains the source code listings.

The results of several investigations appear in the following sections. Two airports were used in these studies: Phoenix Sky Harbor International and Lambert-St. Louis International. Their respective population distributions are depicted in Figures 8.1a and 8.1b. In each grid section in Figure 8.1a (Phoenix) the number shown is the population count for that sector; smaller cell sizes in the St. Louis data necessitate (for legibility) the use of population density shading in Figure 8.1b. The body of results is divided as follows: 1) takeoff

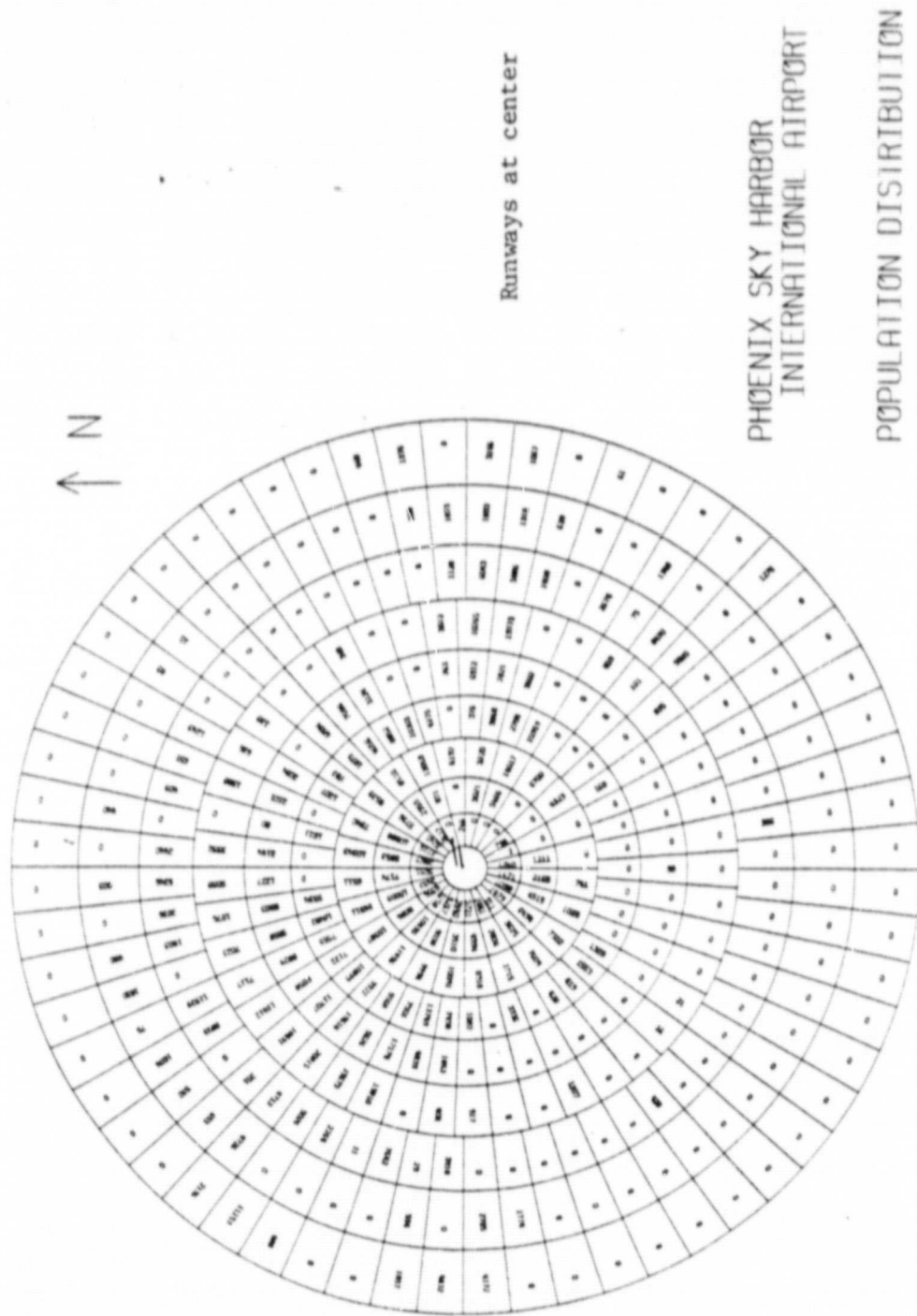


Figure 8.1a Phoenix (Sky Harbor International Airport) Population Distribution

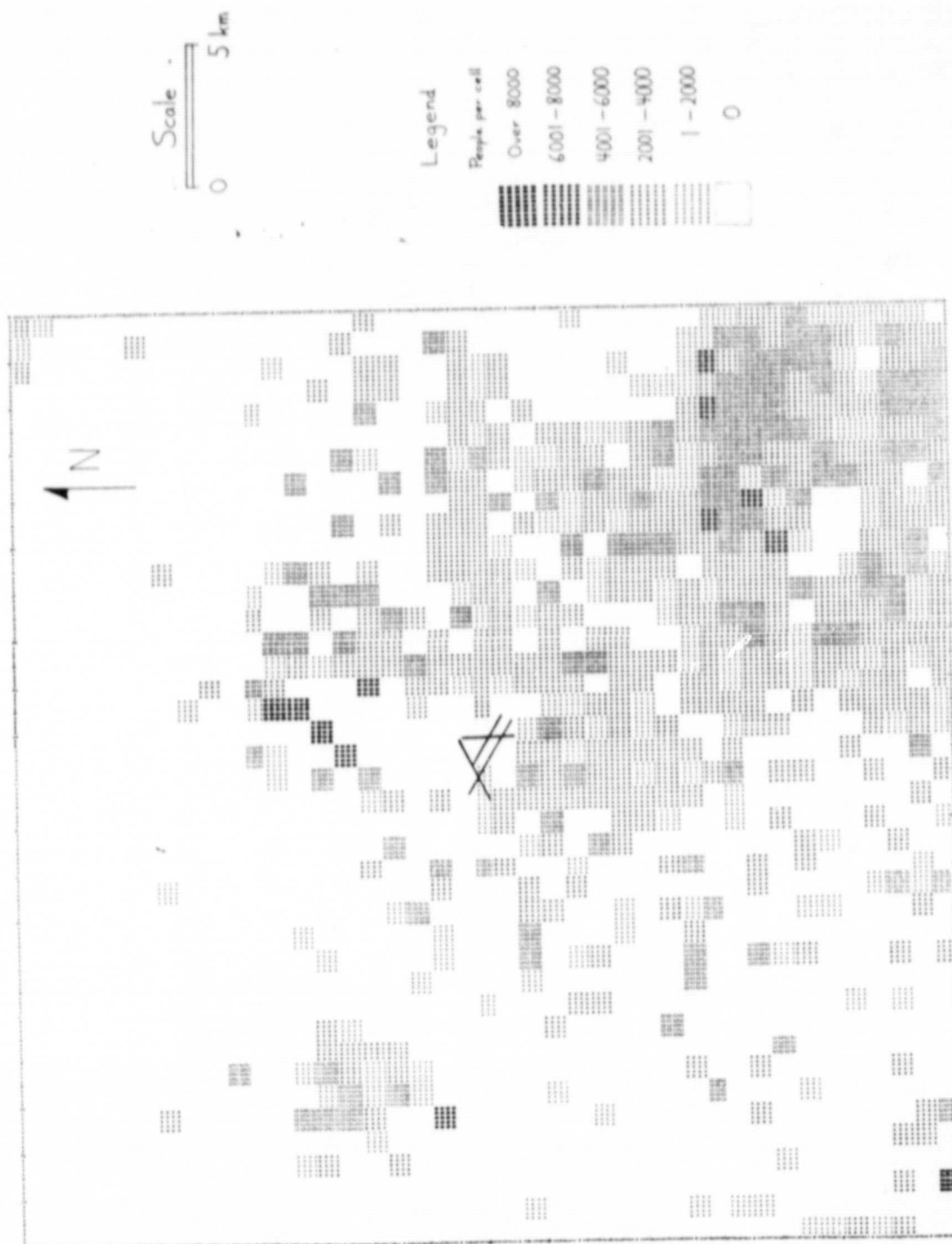


Figure 8.1b St. Louis (Lambert-St. Louis International Airport) Population Distribution

trajectories, 2) a comparison of optimum landing paths using NII, LWP, and HAPN as the annoyance criterion, 3) many takeoff and landing paths -- cyclic optimization, 4) a study of the effects of including the fuel and time of flight costs in the total cost function, and 5) the role of the population distribution in determining the optimum paths.

In all of the following sections, the flight paths are represented by their ground tracks, with the flight path angles, γ_c -climb, and γ_d -descent, indicated for each segment.

Takeoff Trajectory

Using the Lambert-St. Louis Airport community as a test case, a single trajectory was optimized¹, with LWP as the annoyance criterion. Both the nominal and optimum paths as shown in Figure 8.2. The nominal path was approximated by a five segment trajectory, with a 5° flight path angle (γ); the optimum path was restricted to three segments and $\gamma_{\max} = 6^\circ$. Nearly all of the variation between the two paths is in the ground tracks. This phenomenon, seen throughout the results that follow, occurs because 1) variations in the population density distribution occur only in the ground plane, and 2) the average distances from the flight paths to the population centers change more rapidly due to horizontal variations in the flight paths than due to vertical variations (owing to the small angles between the ground and the trajectories). The exception to this is, of course, when a segment

¹ Aircraft-trajectory assignment data for St. Louis were provided by the NASA-Langley Research Center [38].

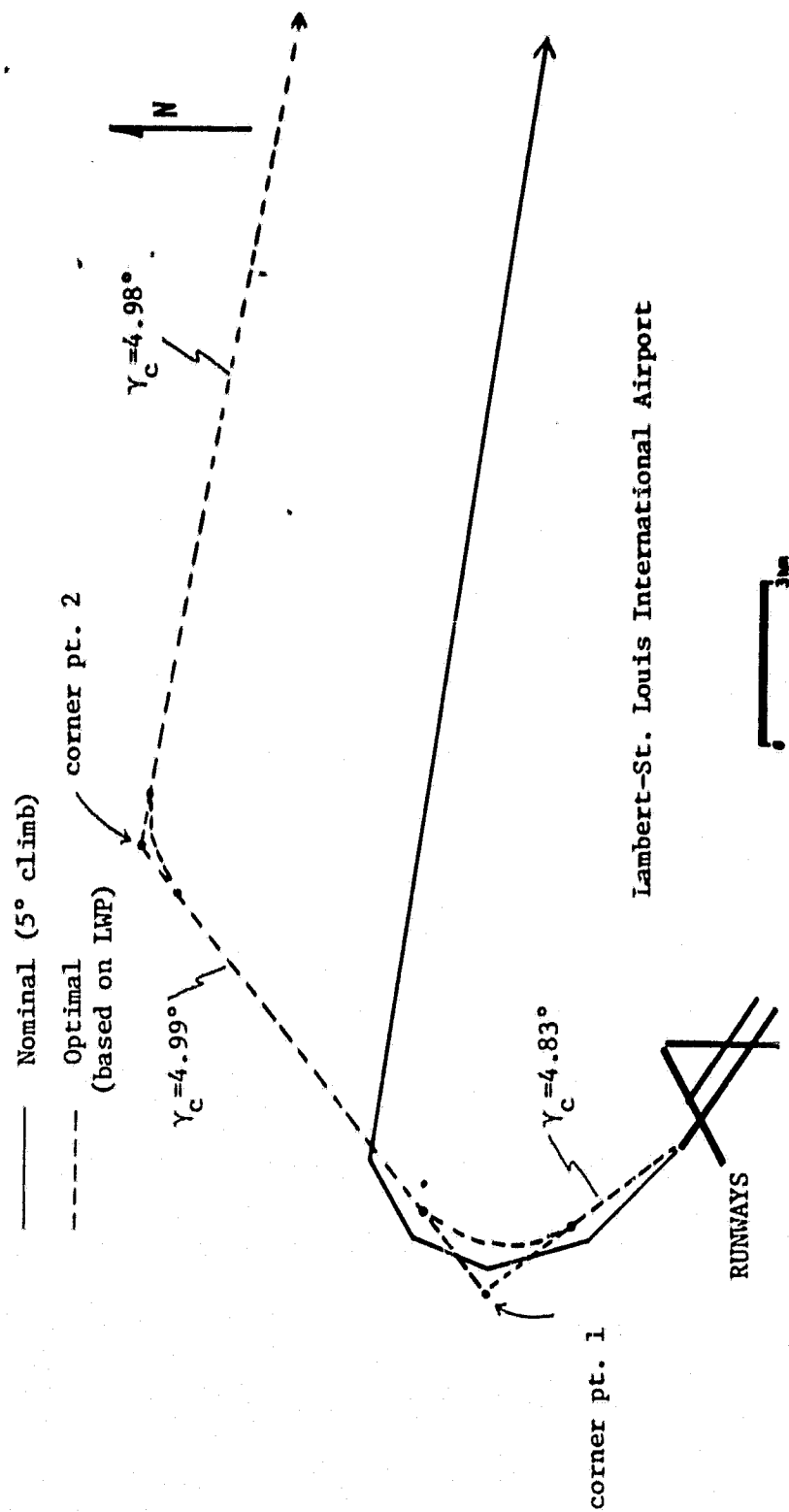


Figure 8.2 Single Takeoff Trajectory - Nominal and Optimum

near the runway (and closer to the ground) passes close to a population center. Then an increase in the flight path angle γ will produce a rapid decrease in the noise level at that center.

A point approximately 90 km. east of the airport was chosen as the outermost endpoint. This allowed the position of corner point #2 to vary without being restricted by the geometric constraints. The comparison of annoyance measures is as follows: nominal-LWP = 2571.1, NII = 0.161; optimum-LWP = 1267.9 (50.6% lower than nominal), NII = 0.228 (42% higher than nominal). This increase in NII, while LWP is decreasing, is another recurring phenomenon. It is caused by the change in total population used to normalize NII (i.e. people receiving $L_{dn} \geq 55\text{dB}$).

Comparison of Annoyance Models

In Chapter III, it was shown that for a simple population distribution (2 concentrated clusters of people), and a straight-line flight path, the measures NII and LWP indicated significantly different optimum positions for the path. Here, NII, LWP, and HAPN will be compared for general trajectories and a real population distribution (Phoenix Sky Harbor Airport community). This section also serves to illustrate the calculation of optimum landing paths.

For landings on the runway indicated in Figure 8.3a there are two entry points (initial approach fixes). This figure shows the nominal ground tracks; the glide slopes are 3° . An estimate of the mixture of aircraft types operating at this airport was obtained from data in the Official Airline Guide [39]. Because the nominal paths are joined

$\gamma_d = 3.0^\circ$ (all segments)

3 segments/path

RUNWAYS

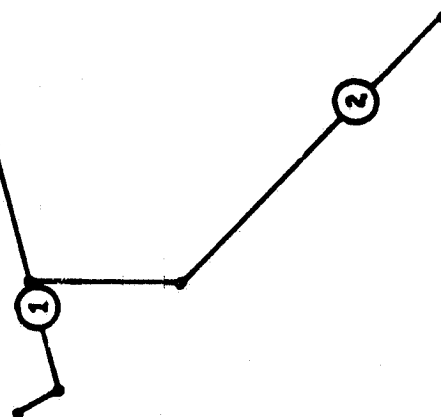


Figure 8.3a Nominal Trajectories at Phoenix Sky Harbor Airport

together over a large portion of their lengths, it is assumed that separation for safety purposes is not a requirement for traffic control here; therefore the separation constraint is bypassed. Note that there are three segments on each path, with some overlapping.

With NII as the annoyance measure, the optimum pair of trajectories obtained is that shown in Figure 8.3b. Values of NII, LWP and HAPN for this pair are given in Table 8.1, along with comparisons with the corresponding nominal values.

With either LWP or HAPN as the annoyance measure, the optimum pair is radically different, as seen in Figure 8.3c. The annoyance measures associated with these flight paths are given in Table 8.1. It is seen that the improvement in the annoyance based upon LWP or HAPN is greater than that of NII. Further, NII increases when LWP and HAPN decrease, and vice versa, another indication of the normalization problem in NII.

Fuel and Time of Flight Study

As an example of how the system may be used to study associated problem, the cost of fuel and time of flight ($c_{F,TOF}$) has been incorporated in the total cost function. In this investigation, the relation between $c_{F,TOF}$ and NII is examined.

Using the modified cost function, Eqn. (6-13), the optimization was repeated for increasing values of $w_{F,TOF}$ (the fuel and time of flight weighting factor); the airport used was Phoenix Sky Harbor. Figure 8.4a summarizes the results, showing the relation between the NII, and the fuel and time of flight cost ($c_{F,TOF}$). Accompanying it are Figures

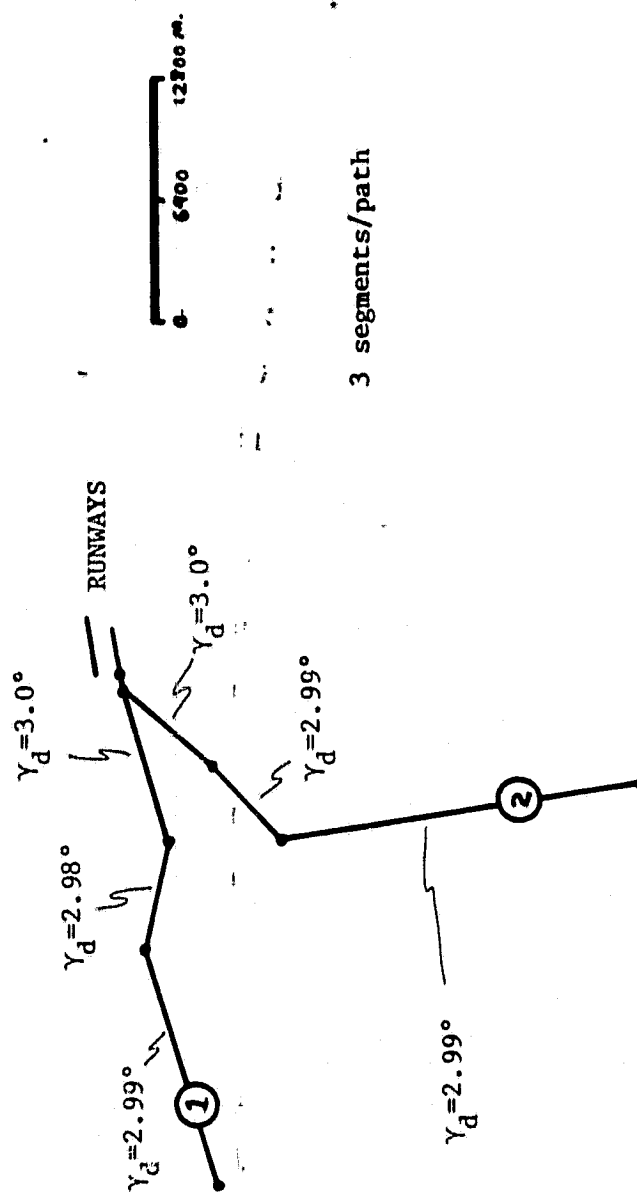


Figure 8.3b Optimum (NII) Trajectories at Phoenix Sky Harbor Airport

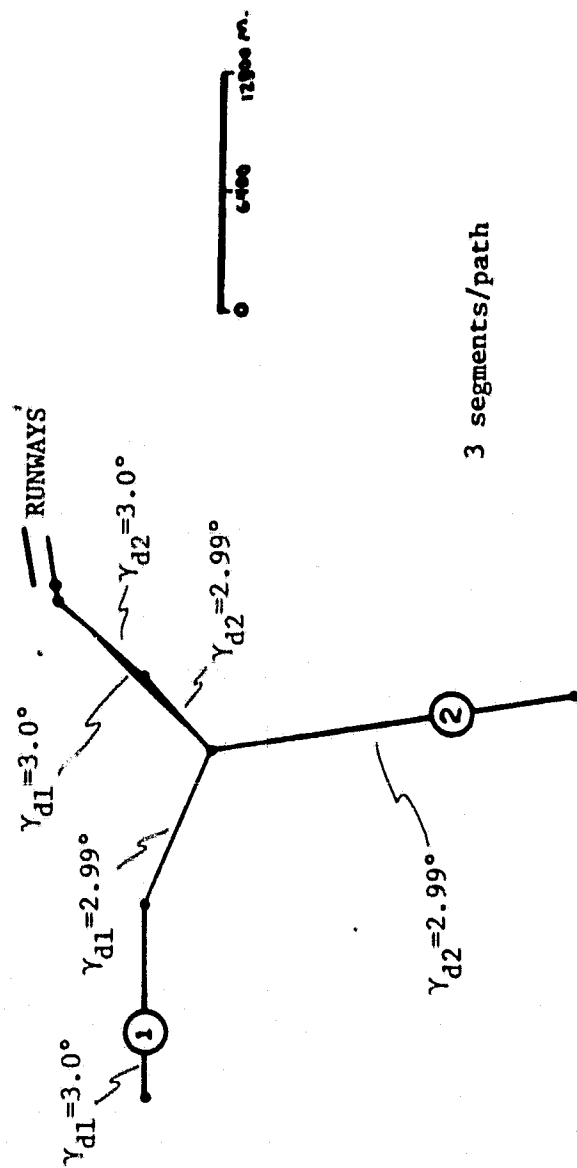
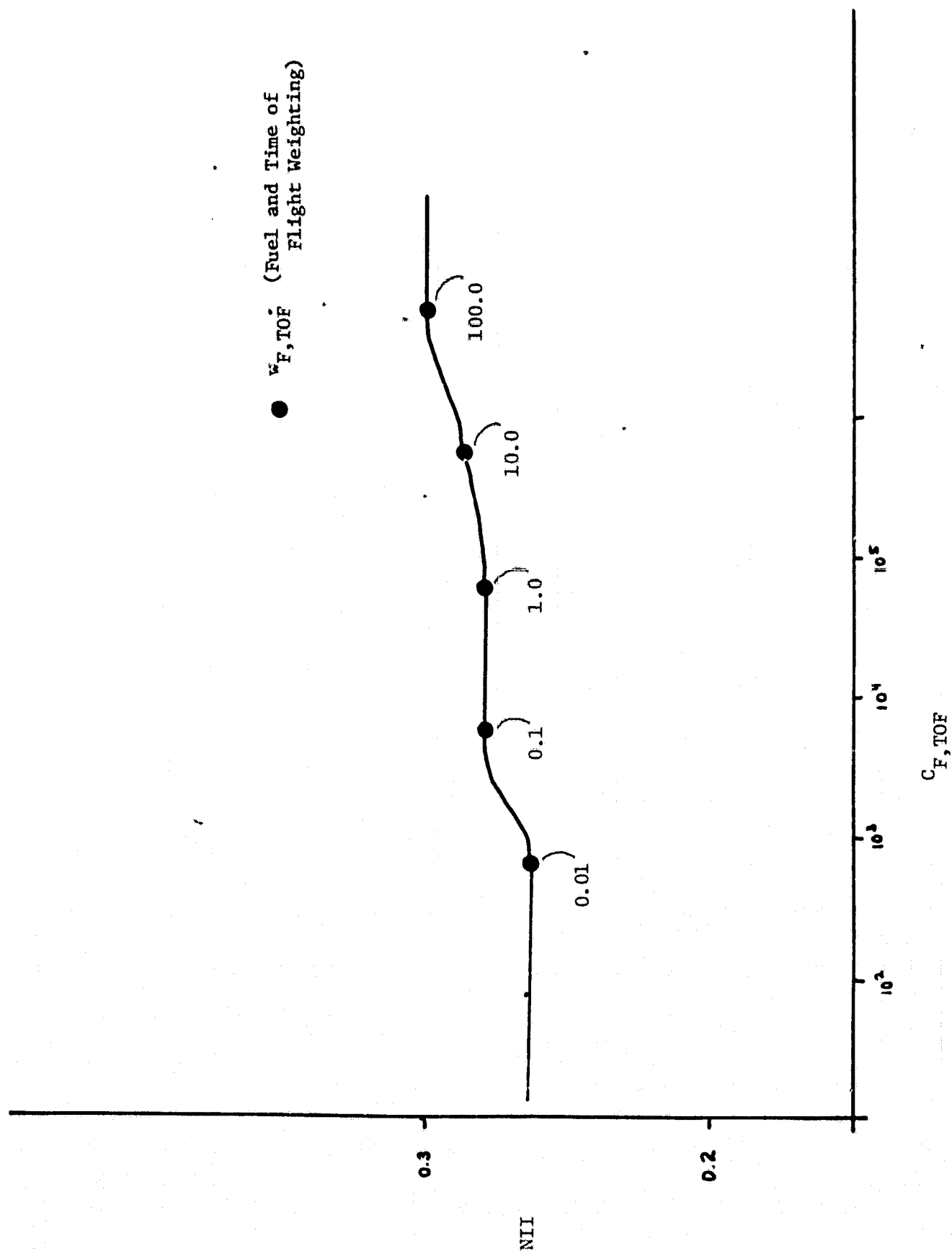


Figure 8.3c Optimum (LWP or HAPN) Trajectories at Phoenix Sky Harbor Airport

Table 8.1
Comparison of Annoyance Measures
(For Phoenix Sky Harbor)

Trajectory Set	NII	$\dagger \Delta\%$	LWP	$\dagger \Delta\%$	HAPN	$\dagger \Delta\%$	Figure Showing Optimum
Nominal	0.281		30566		0.491		8-3a
NII optimum	0.264	-6	31993	4.7	0.524	6.7	8-3b
LWP optimum	0.290	3	25192	-17	0.445	-9.3	8-3c
HAPN optimum	0.290	3	25192	-17	0.445	-9.3	8-3c

\dagger Percentage difference from nominal value.

Figure 8.4a NII vs. Fuel and Time of Flight Cost ($C_{F,TOF}$)

8.4b through 8.4f, which show the optimum landing trajectories for each value of $w_{F,TOF}$ used.

As expected, the flight paths become shorter (straighter) as $c_{F,TOF}$ increases. Trajectory #1 shortens rapidly ($w_{F,TOF} = 0.1$) to a straight path, but trajectory #2 is hindered by two factors: a large cluster of people southeast of the runways (see Figure 8.1a), and the final angle of approach restriction. For $w_{F,TOF} \leq 3$ shortening this path will result in NII increases that are greater than any decrease in the fuel and time of flight cost; hence, there is a flat region in Figure 8.4a for $0.1 \leq w_{F,TOF} \leq 3$. Larger values of $w_{F,TOF}$ increase the relative importance of $c_{F,TOF}$, and the total cost is reduced at the expense of NII. It is impossible, though, for trajectory #2 to become completely linear because of the final angle constraint. The effect of $c_{F,TOF}$ is then saturated at $w_{F,TOF} = 100$. The extrapolation for very small $w_{F,TOF}$ is based upon the value obtained for NII in the Comparison of Annoyance Measures section (where $w_{F,TOF} = 0$).

Cyclic Optimization

At large airports such as Lambert-St. Louis International, the large variety of aircraft that must be considered necessitates the use of the equivalent aircraft concept in order to keep the computation time manageable. Further, the number of trajectories being optimized (15 at St. Louis) will require a considerable increase in storage requirements, compared with those needed for Phoenix. If the storage required is in excess of the allocated resources, the dilemma may be remedied by optimizing the flight paths one at a time, cyclically. The power density

Figure 8.4b Optimum Trajectories,
 $w_{F, TOF} = 0.01$

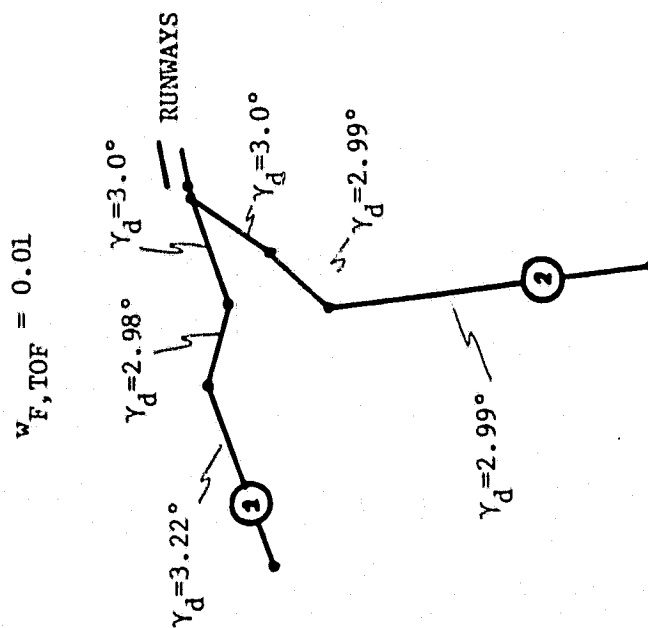


Figure 8.4c Optimum Trajectories,
 $w_{F, TOF} = 0.1$

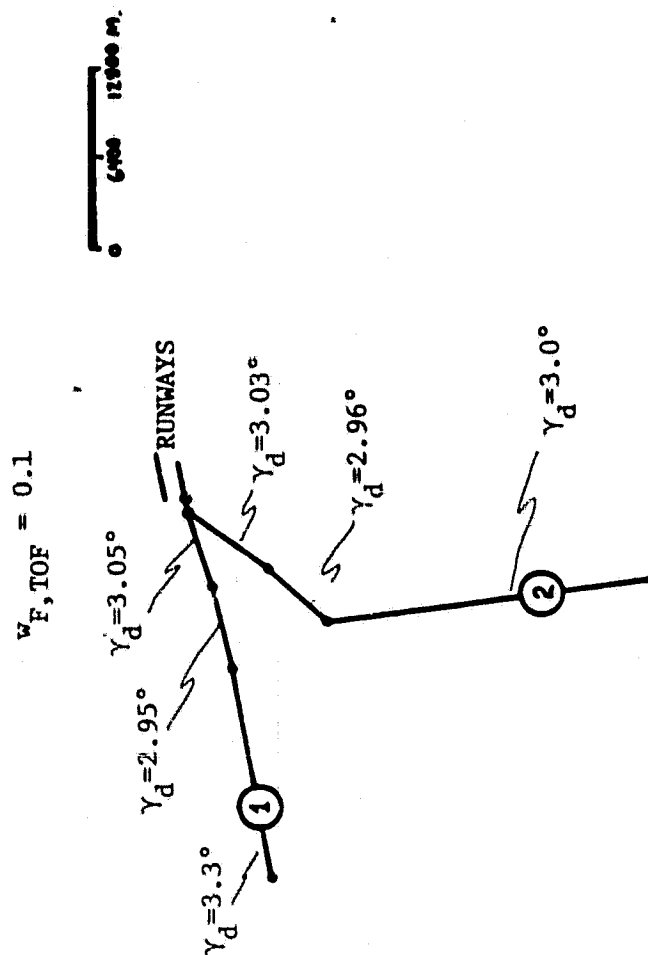


Figure 8.4d Optimum Trajectories,

$$w_{F, TOF} = 1.0$$

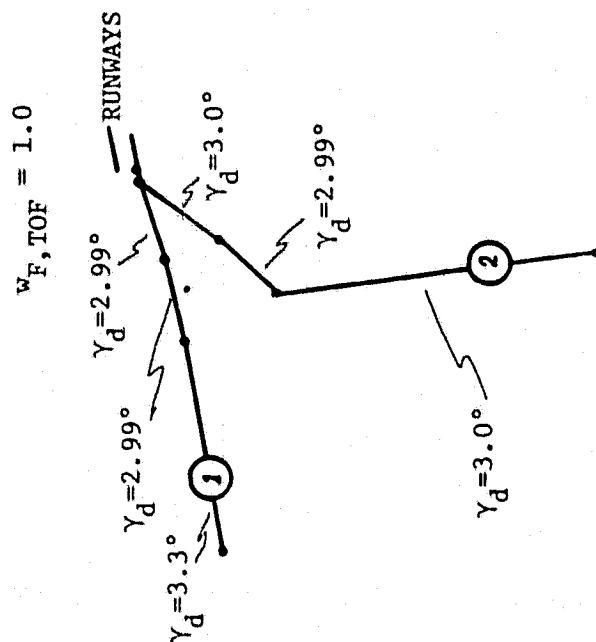
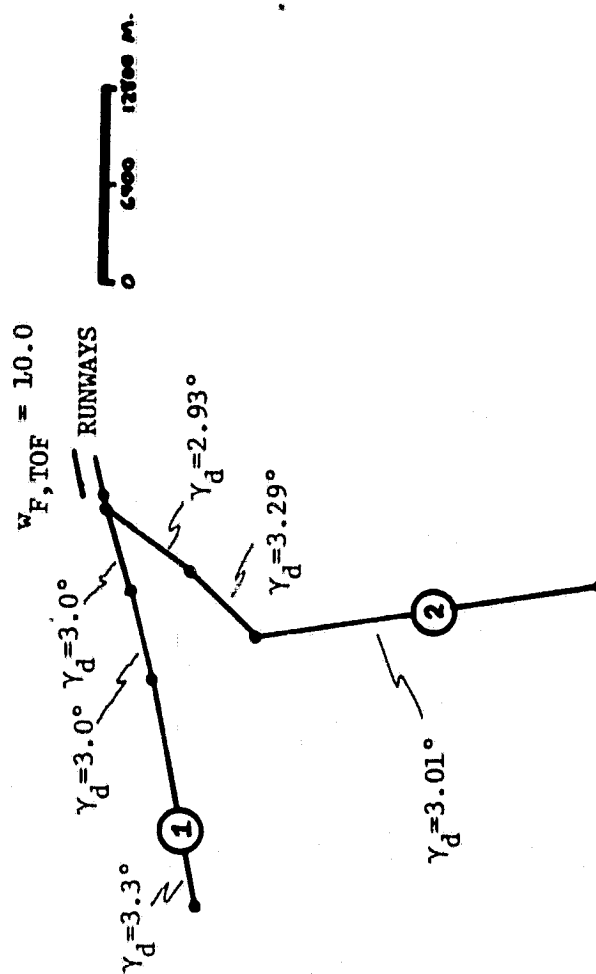


Figure 8.4e Optimum Trajectories,

$$w_{F, TOF} = 10.0$$



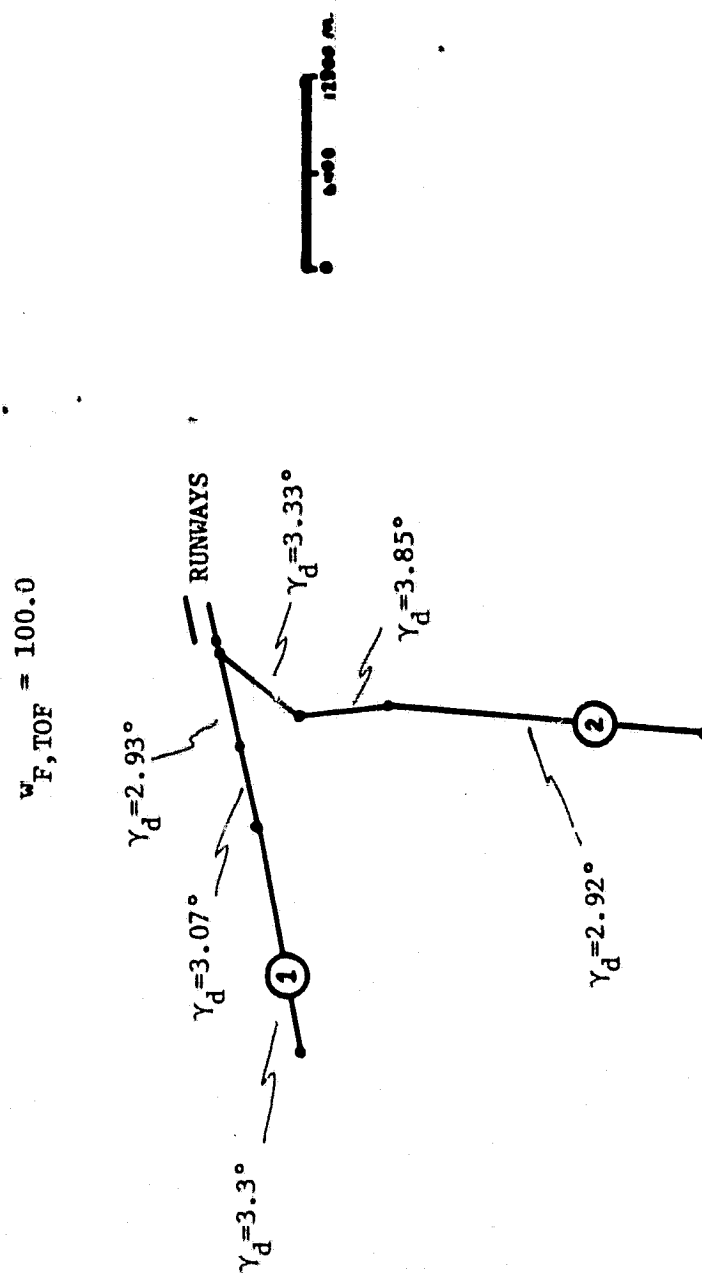


Figure 8.4f Optimum Trajectories, $w_{F, TOF} = 100.0$

St. Louis Nominal Trajectory Set

5 segments/trajectory

NII = 0.2076

LWP = 105230

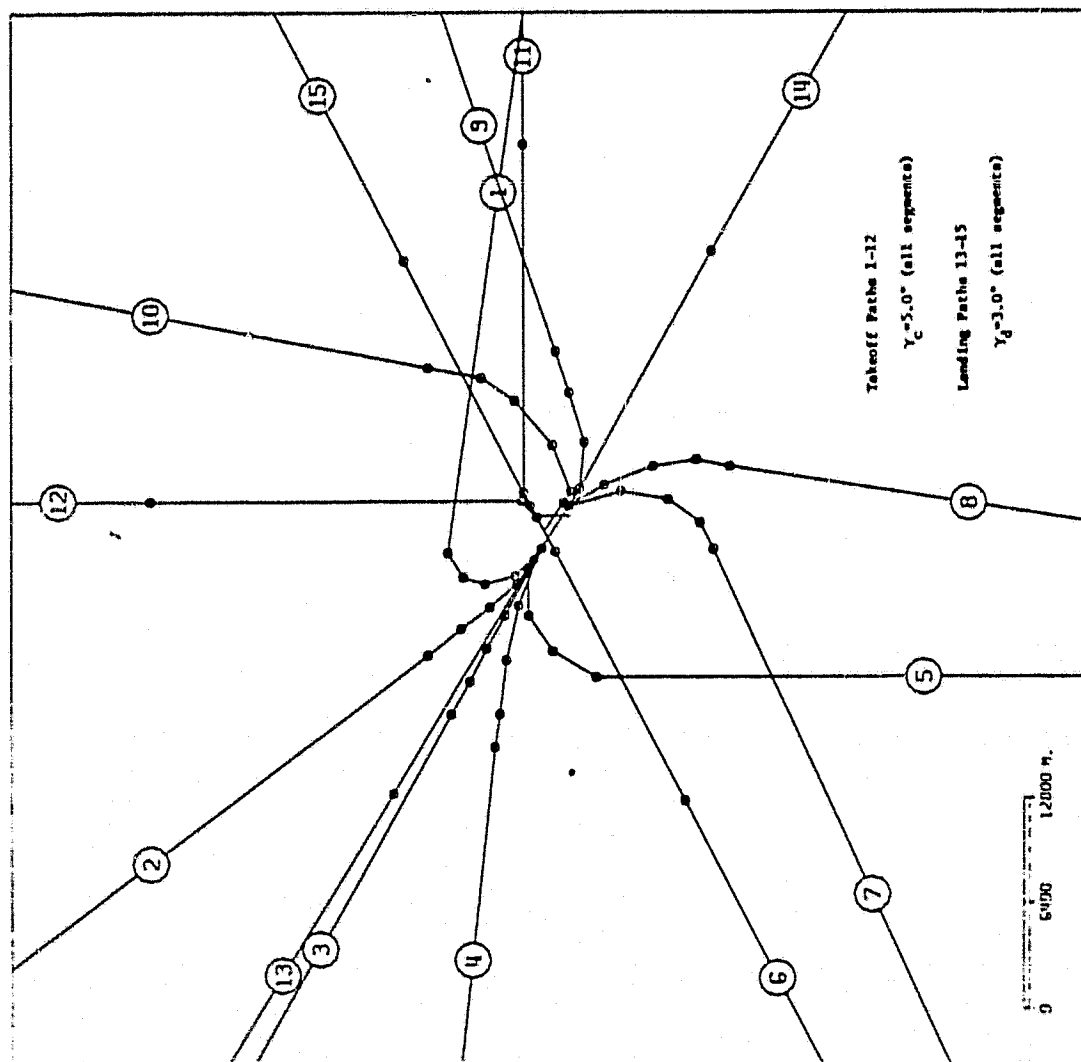
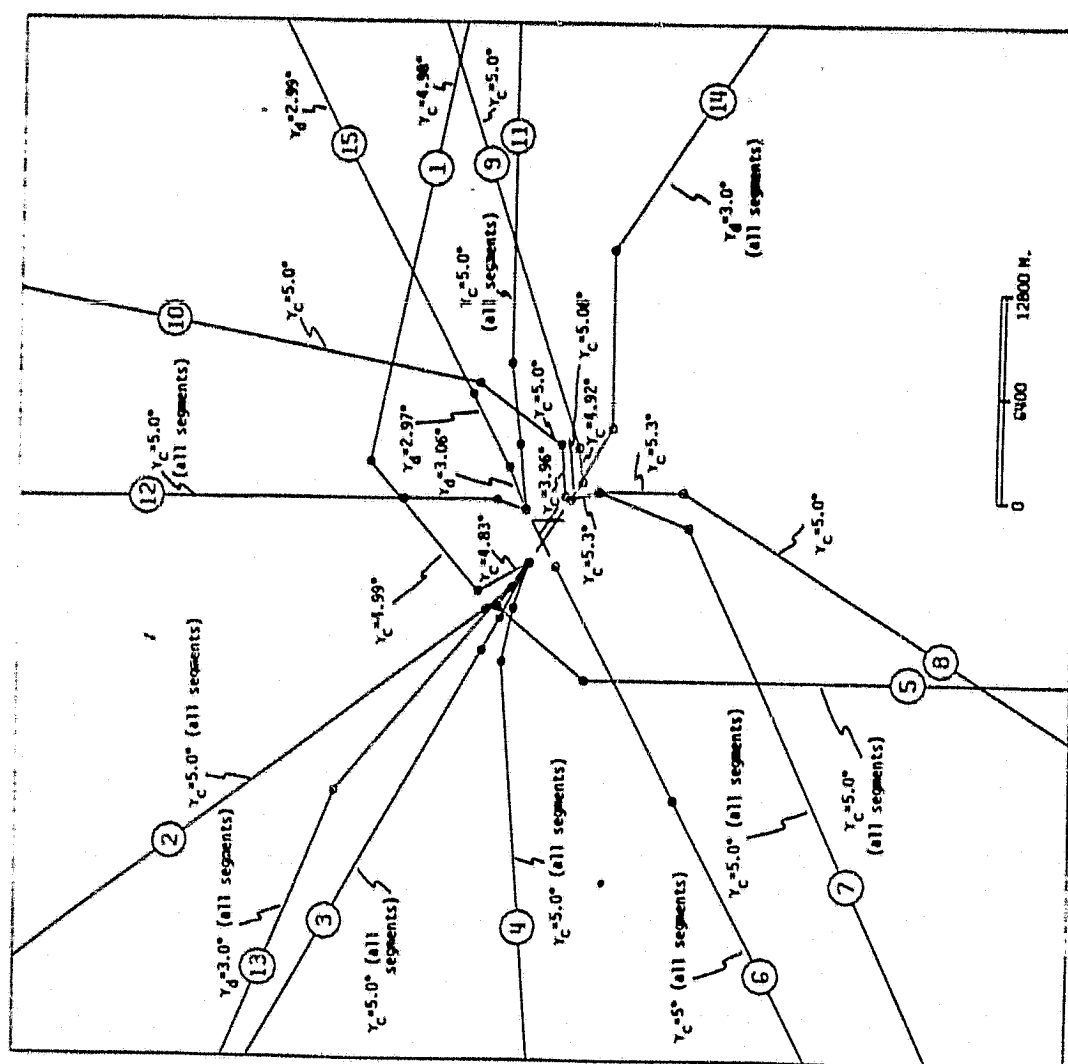


Figure 8.5a St. Louis Nominal Trajectory Set



St. Louis Optimum Trajectory Set
3 segments/trajectory

NII = 0.2108

LMP = 77849

Figure 8.5b St. Louis Optimum Trajectory Set

contributions from the (temporarily) constant trajectories are stored and then summed with those of the path being optimized in order to determine the L_{dn} value in each grid cell. As a result, the optimization storage requirements are for only one trajectory, but the annoyance measure is for the entire set of paths.

It should be noted that this approach does not guarantee convergence to a minimum; however, the results for the St. Louis case are excellent. Figure 8.5a shows the nominal paths at Lambert-St. Louis International (tracks 1-12: takeoff, 5° ; track 13-15: landing, 3°). Each trajectory is approximated with 5 segments. Judging by the population distribution (Figure 8.1b), it was decided that paths 1,2,7, 8,9,13,14, and 15 could give the most improvement in the annoyance (LWP). Using the program CYCLIC to cyclically optimize these paths (see Appendix D), and allowing a maximum of three segments per trajectory, the results obtained were: LWP decreased 26%, NII increased 1.5%, from the nominal values. Figure 8.5b shows the improved trajectories.

Population Distribution and Aircraft Mix

In order to investigate the relative importance of the population distribution vs. the aircraft mix (i.e., mixture of different aircraft types) on the optimum flight paths, a study was conducted using two different population distributions (one fictitious, one real), and two different aircraft mixes.

Using the fictitious population distribution shown in Figure 8.6, a single optimum landing trajectory (3°) was computed, using the CYCLIC

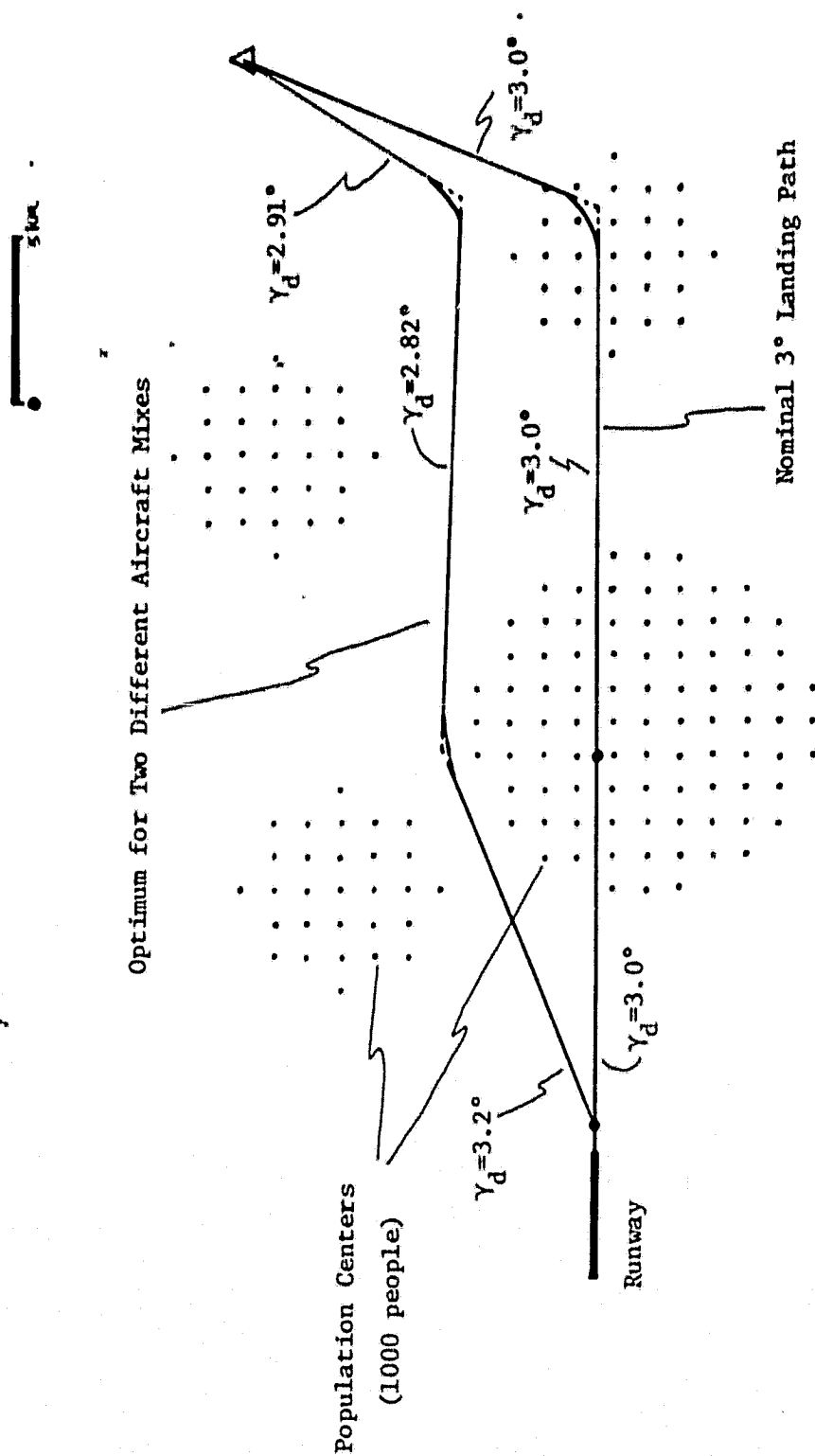


Figure 8.6 Fictitious Population Distribution - The Effect of Different Aircraft Mixes

program.² Two different mixtures of aircraft types were investigated; the degree of difference is evinced by the difference in their equivalent aircraft noise coefficients:

$$\text{Mix \#1: } c_{1,eq} = 170.70 \quad c_{2,eq} = 24.23$$

$$\text{Mix \#2: } c_{1,eq} = 166.00 \quad c_{2,eq} = 28.60.$$

The optimum paths for the two mixes were nearly identical; no two corresponding corner points differ by more than 36 m.

With a real population distribution, though, a strong dependence upon the aircraft mix is evident. For the Phoenix airport, two mixes were examined:

$$\text{Mix \#1: } c_{1,eq} = 169.47 \quad c_{2,eq} = 24.37$$

$$\text{Mix \#2: } c_{1,eq} = 166.00 \quad c_{2,eq} = 28.60,$$

Mix #1 corresponds to the true mix at Phoenix, which was used in the Comparison of Annoyance Models section and for which the optimum results appear in Figure 8.3b (NII) and 8.3c (LWP). Mix #2 is all DC-10 aircraft, with the equivalent amount of passenger seating of Mix #1. In the case of Mix #2, the optimum paths are significantly different, with either LWP or NII as the annoyance criterion. Figure 8.7a and 8.7b depict the results.

The consequence of this example is that the relative importance of the population distribution vs. the aircraft mix in determining the optimum trajectories cannot be stated in general. For a simple distribution of people, corridors for optimum trajectories that are

² Although only one trajectory, with three segments, was being optimized, CYCLIC was used so that the equivalent aircraft concept could be employed.

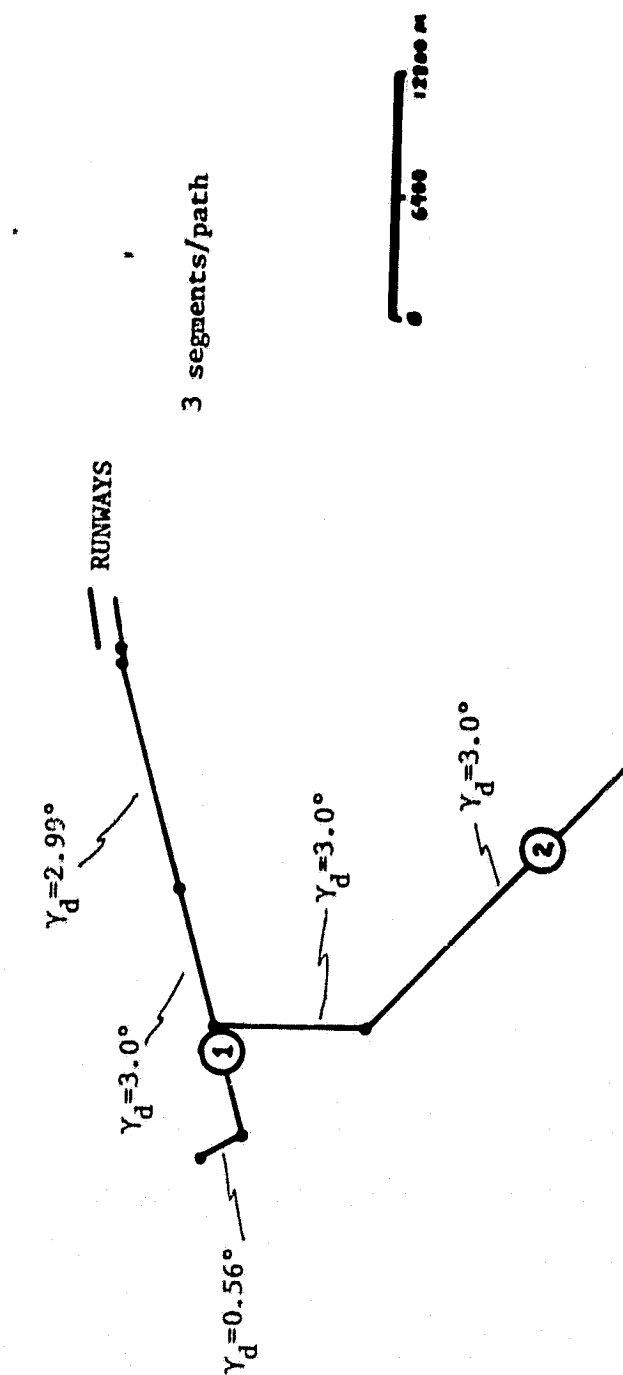


Figure 8.7a Optimum (NII) Landing Trajectories at Phoenix Sky Harbor Airport (DC-10 mix)

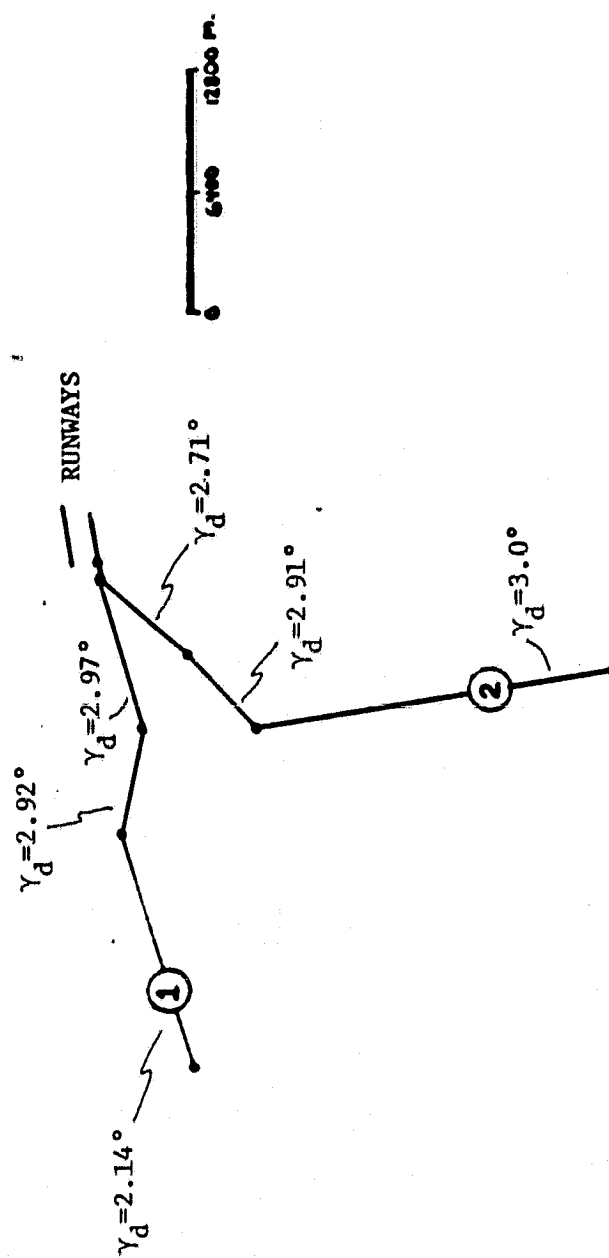


Figure 8.7b Optimum (LWP) Landing Trajectories at Phoenix Sky Harbor Airport (DC-10 mix)

independent of aircraft mix may exist (such appears to be case in the fictitious distribution in this example). A real population distribution, though, is usually more dispersed, which results in more people (on the average) being closer to the flight paths. Because of the nonlinearities in the noise and annoyance models, changes in the aircraft mix result in a change in the annoyance of unpredictable magnitude. Therefore, new optimum flight paths must be computed when the mix changes.³

³ For a particular situation, however, experience may indicate that small changes in the mix do not necessitate recomputation of the trajectories.

CHAPTER IX

CONCLUSIONS AND RECOMMENDATIONS FOR FURTHER STUDIES

This study has developed a method to assess the ability to reduce, through operational controls, the annoyance caused by aircraft noise in a community. Test cases, using a computer implementation of the method and actual airport situations, have shown the feasibility of using trajectory modifications in the solution to the problem of aircraft noise. The amount of reduction that can be gained depends upon the airport, aircraft mix, constraints, and model of annoyance being used.

A comparison study of the annoyance measures NII, LWP, and HAPN has shown that, in addition to resulting in different percentage improvements, the optimum trajectories that result from using each of these measures can be significantly different. Because of its mathematical definition, NII has been shown to be an inferior indicator of the amount of change in annoyance for a given community.

The method has a modular structure which allows for straightforward modification of any of its components, e.g., the flight path model, the aircraft noise model, the optimization algorithm, etc. In addition, the method may be expanded to include related problems in the optimization. A particular example of this was demonstrated by including the "costs" of fuel and time of flight in the cost function. The modified system was then used to examine the influence of these factors (along with the annoyance) on the resulting optimum flight paths.

Of particular importance is the finding that the population distribution in a given community does not necessarily determine a unique set of optimum trajectories; the mixture of aircraft strongly influences the solution. This is unfortunate from a computational aspect, since the mixtures on the flight paths generally will vary over time, requiring that new optimum sets be calculated.

Two concepts that may have applications elsewhere were developed in this work. First, the complexity of computing the flight paths was reduced by assuming a parametric solution and explicitly including restrictions based upon analyses of the linear dynamic equations of motion of the aircraft. This technique may be useful in situations where the computing resources available prohibit the customary variational approach to flight path optimization. Second, the equivalent aircraft concept not only reduces the time required for calculation of noise levels due to multiple sources, but also provides a means for predicting noise levels as a result of changes in the number of aircraft on any trajectory.

The difficulty associated with searching over many local minima was overcome by making use of the statistical lateral dispersions in the trajectories. These dispersions limit the size of the alternative regions (in the horizontal plane) in which to search for significantly different optimum paths.

Finally, this approach to the problem of aircraft noise has shown that annoyance can be reduced significantly via operational control. The method provides a tool for those who must decide what action to take in remedying the problem. It can also be used to optimally site an airport in a community to minimize future noise problems.

Recommendations

It is felt that further work should employ a population distribution model that does not artificially create optimum pathways. An analytic representation of the distribution (e.g. spline functions, or low order polynomials) would likely aid in the operation of the method, without necessarily increasing computational requirements.

Even with such a refinement, the difficulty of local minima in the cost function, caused by real, macroscopic variations in the population density will persist. Use of the statistical lateral dispersion in flight paths has been instrumental in restricting the global search and should be automated.

The assumption of constant runway designations, i.e., takeoff or landing, which was made in this study, is not generally true. Future work should include the possibility of having to compute an alternate set of flight paths, perhaps because of a change in the wind direction during the day. The cumulative power density distribution from the previous optimum configuration would have to be retained and combined with that of the new set in order to include the total effect over the period of time in question. It is envisioned that air traffic controllers could then use this method to compute alternate optimum flight paths in real time.

Finally, community land use planning could employ this system by restricting the maximum noise levels in proposed sensitive areas (schools, hospitals, etc.). The cost function would be adjoined by penalties for violations of these restrictions. Industrial areas where high noise levels from aircraft would have little effect, would be

weighted less heavily in the annoyance calculation. As the final step, the coordinates of the airport centroid would be included as unknown parameters to be determined (with some restrictions) in the optimization procedure. The result would be not only the optimum set of trajectories, but also an optimally positioned airport.

APPENDIX A

SPECIAL CONSIDERATIONS FOR THE NOISE LEVEL CALCULATIONS

Energy Addition

Sound intensity (power/area) is usually expressed in the form

$$L = 10 \cdot \log_{10} \frac{I}{I_0} \quad (A-1)$$

where I is the intensity being characterized and I_0 is a reference intensity (usually taken as the lowest audible intensity at 1,000 Hz, approximately 10^{-16} watt/cm²). The units are called decibels (dB) and provide a convenient scale for the extremely wide range of intensities that may occur.

For a given volume of space (in which the intensity of sound is constant), the acoustic energy received is proportional to the intensity and the time of exposure. More precisely,

$$E = aI\Delta t \quad (A-2)$$

where Δt is an increment of time during which I does not change appreciably, E is the increment of energy received, and a is a constant that depends upon the volume of space being considered. The sound level may then be expressed as

$$L = 10 \cdot \log_{10} \frac{E}{E_0} \quad (A-3)$$

and

$$E = E_0 \cdot 10^{L/10} \quad (A-4)$$

where E_0 is the energy received from the reference sound.

This provides a powerful tool for "adding" sound from different sources. Suppose that n different sources produce sound with levels L_i at a particular location. The total acoustic energy per unit volume of space is

$$E_{TOT} = \sum_{i=1}^n E_i = E_0 \sum_{i=1}^n 10^{L_i/10} \quad (A-5)$$

Eqn. (A-3) is used to calculate the intensity level in dB:

$$L_{TOT} = 10 \cdot \log_{10} \sum_{i=1}^n 10^{L_i/10} \quad (A-6)$$

In Chapter III, the L_{dn} measure uses the energy addition method to obtain a weighted average of noise levels occurring over a period of time. The use of energy addition is often referred to as adding on an "energy basis," although it is actually power densities that are being added.

Conversion of EPNL to L_A

The measure Effective Perceived Noise Level (EPNL) is based upon Perceived Noise Level (PNL) in the following way (see Ref. [5], pp. 287-300.):

$$EPNL = IPNL - 12 + (\text{onset corrections}) + (\text{impulse corrections}) \quad (A-7)$$

where IPNL is the Integrated Perceived Noise Level

$$IPNL = 10 + \log_{10} \sum_{i=1}^{16} 10^{PNL_i/10} \quad (A-8)$$

The values of PNL are taken every 0.5 sec. for 8 sec., hence the upper limit of 16 in the summation.

Neglecting the onset and impulse corrections (they are not significant for aircraft noise) gives

$$EPNL \approx 10 + \log_{10} \left(\sum_{i=1}^{16} 10^{PNL_i/10} \right) - 12 \quad (A-9)$$

$$\approx 10 \cdot \log_{10} \left(\frac{\sum_{i=1}^{16} 10^{PNL_i/10}}{16} \right) \quad (A-10)$$

The last expression is just the average PNL (on an every basis).

So for a constant PNL,

$$EPNL \approx PNL \quad (A-11)$$

Finally, for nonimpulsive sound

$$PNL \approx L_A + 13 \text{ dB}, \quad (A-12)$$

giving

$$L_A \approx EPNL - 13 \text{ dB}. \quad (A-13)$$

APPENDIX B

THE LINE SEARCH

In Chapter V, several of the optimization methods require a one-dimensional, or line, search in each iteration. The task is to find the value of α that minimizes the cost function f , i.e.

$$\underset{\alpha}{\text{minimize}} \ f(\underline{x}_k + \alpha \hat{\underline{d}}_k)$$

where \underline{x}_k is the current position (at the k -th iteration) in E^n and $\hat{\underline{d}}_k$ is a unit vector that defines the search direction, relative to \underline{x}_k .

A successful method for finding α , which has been employed in this work, is that of assuming a polynomial form in α for $f(\underline{x}_k + \alpha \hat{\underline{d}}_k)$, the exact minimum of which is determined by the calculus of one variable. The form used here is cubic, i.e.,

$$f(\underline{x}_k + \alpha \hat{\underline{d}}_k) = k_0 + k_1\alpha + k_2\alpha^2 + k_3\alpha^3, \quad (\text{B-1})$$

for which the minimum occurs at α^* , such that

$$\frac{\partial f}{\partial \alpha} \bigg|_{\alpha=\alpha^*} = k_1 + 2k_2\alpha^* + 3k_3\alpha^{*2} = 0, \quad (\text{B-2})$$

and

$$\frac{\partial^2 f}{\partial \alpha^2} \bigg|_{\alpha=\alpha^*} = 2k_2 + 6k_3\alpha^* > 0. \quad (\text{B-3})$$

The last equation eliminates one of the two roots of Eqn. (B-2), so that a unique minimum is given once the k_i are specified. This is accomplished by calculating f and $\frac{\partial f}{\partial \alpha}$ at two values of α . Denoting $f(\alpha_0)$, $f(\alpha_1)$, $\frac{\partial f}{\partial \alpha}(\alpha_0)$, and $\frac{\partial f}{\partial \alpha}(\alpha_1)$ by f_0 , f_1 , f'_0 , and f'_1 , respectively,

the minimum, α_2 , of the cubic function is given by the following:¹

$$u_1 = f'_0 + f'_1 - 3 \left[\frac{f_0 - f_1}{\alpha_0 - \alpha_1} \right] \quad (B-4)$$

$$u_2 = [u_1^2 - f'_0 f'_1]^{\frac{1}{2}} \quad (B-5)$$

$$\alpha_2 = \alpha_1 - (\alpha_1 - \alpha_0) \left[\frac{f'_1 + u_2 - u_1}{f'_1 - f'_0 + 2u_2} \right] \quad (B-6)$$

Calculation of the directional derivative $\frac{\partial f}{\partial \alpha}$ is normally performed by projecting the gradient $\underline{g}(\underline{x}_k) = \nabla^T f(\underline{x})$ onto the direction \underline{d}_k , i.e.

$$\frac{\partial f}{\partial \alpha} = \nabla^T f(\underline{x}) \cdot \hat{\underline{d}}_k, \quad (B-7)$$

where (\cdot) denotes the inner product in E^n . Since derivatives of f must be approximated numerically in this work, though, it is simpler to use the definition of derivative to obtain

$$\frac{\partial f}{\partial \alpha} \approx \frac{f(\underline{x}_k + \Delta \alpha \hat{\underline{d}}_k) - f(\underline{x}_k)}{\Delta \alpha} \quad (B-8)$$

where $\hat{\underline{d}}_k$ is a unit vector along \underline{d}_k and $\Delta \alpha$ is a small increment.

Choosing values of α_0 and α_1 for use in the cubic fit presents no problems if f is cubic: the fit will be exact regardless of what values of α are used. With a non-cubic f (as in this work), the value of α_2 given by Eqns. (B-4) - (B-6), may not be the true minimum along $\hat{\underline{d}}_k$; however, as stated in Chapter V for the modified D-F-P method, α_2 need only be accurate enough to insure that $\underline{p}^T \underline{q} > 0$. A logical choice for α_0 is zero, since then $f_0 = f(\underline{x}_k)$, a value known from the previous

¹ From Luenberger [23 p. 142].

iteration. If α_1 can be chosen such that it and α_0 "bracket" α^* , the true minimum, is greatly increased. A clever way of achieving this² (if f is not varying rapidly) is to make α_1 depend upon f_0 and f'_0 in the following manner:

$$\alpha_1 = \min [-2f_0/f'_0, \alpha_{\max}], \quad (\text{B-9})$$

where α_{\max} is "a constant," limiting value. The expression $-2f_0/f'_0$ is seen in Figure B.1 to represent the distance from $\alpha_0 = 0$ to the point of intersection of the tangent to f at α_0 and the α axis. Taking twice that distance for α_1 gives a point on the other side of the true minimum α^* . The resulting cubic fit is shown as the dotted curve, with α_2 as its minimum. In cases where α_2 is not sufficiently close to α^* to give $\underline{p}^T \underline{q} > 0$, the cubic fit is repeated, with $\alpha_0 = \alpha_2$, and α_1 computed as before. The convergence is quite rapid, generally requiring no more than two fits.

Finally, it should be noted that the use of $\hat{\underline{d}}_k$ rather than \underline{d}_k in Eqn. (B-8) means that $\Delta\alpha$ is the magnitude of the increment ($\Delta\alpha\hat{\underline{d}}_k$), which should be small to obtain accurate derivatives. Use of \underline{d}_k , whose norm varies between iterations, would give unpredictable results.

² Fletcher and Powell [34] first proposed this scheme.

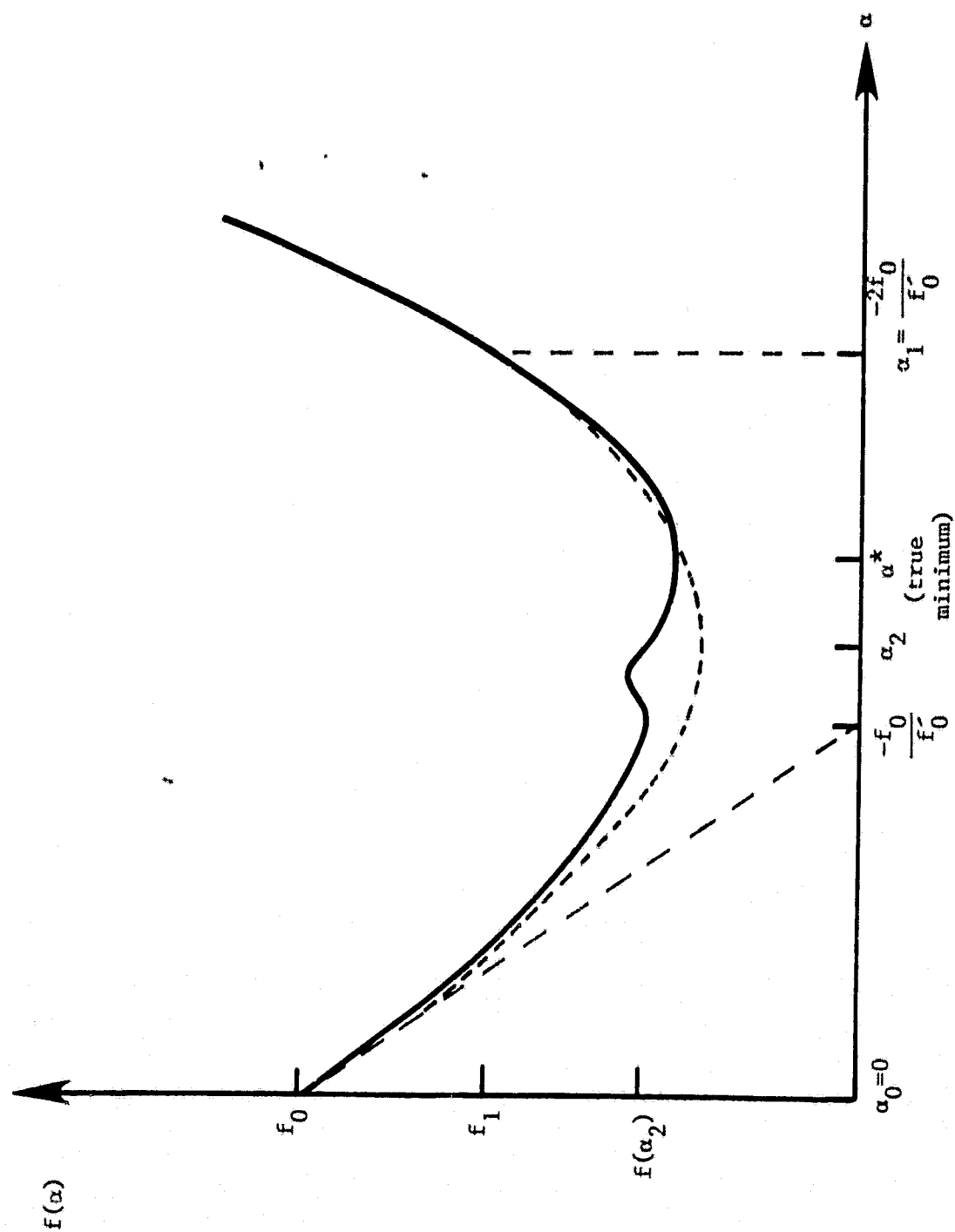


Figure B.1 Line Search by Cubic Fit

APPENDIX C

DETERMINATION OF TANGENT POINTS AND CENTER OF CIRCLE AT A CORNER POINT

In order to evaluate some of the constraint equations in Chapter IV, and for purposes of sampling an aircraft's position as described in Chapter VII, it is necessary to know the cartesian coordinates of the tangent points on a trajectory (see Figure C.1). The value of R_{\min} is determined by the dynamic and passenger comfort constraints (Eqn. 4-13), and the corner point coordinates are the unknowns that the optimization scheme determines.

Ground Plane

First, the value of α is given (by the Law of Cosines) as:

$$\alpha = \cos^{-1} \left[\frac{a^2 + b^2 - c^2}{2ab} \right] \quad (C-1)$$

where

$$a = \overline{P_1 P_2}, \quad b = \overline{P_2 P_3}, \quad \text{and} \quad c = \overline{P_1 P_3}.$$

The distance l from the corner point of interest (x_2, y_2) is

$$l = \frac{R_{\min}}{\tan \frac{\alpha}{2}}, \quad (C-2)$$

so that the tangent points may then be expressed as

$$x_{t1} = \frac{l}{\overline{P_1 P_2}} (x_1 - x_2) + x_2 \quad (C-3a)$$

$$y_{t1} = \frac{1}{P_1 P_2} (y_1 - y_2) + y_2 \quad (C-3b)$$

$$x_{t2} = \frac{1}{P_2 P_3} (x_3 - x_2) + x_2 \quad (C-3c)$$

$$y_{t2} = \frac{1}{P_2 P_3} (y_3 - y_2) + y_2, \quad (C-3d)$$

where x_i, y_i are the x and y coordinates of the i -th corner point.

The linearity of the segments requires that

$$\frac{z_{t1} - z_2}{z_1 - z_2} = \frac{1}{P_1 P_2} \quad (C-4a)$$

$$\frac{z_{t2} - z_2}{z_3 - z_2} = \frac{1}{P_2 P_3}, \quad (C-4b)$$

from which the values of z_{t1} and z_{t2} may be obtained (z_i is the z coordinate of the i -th corner point).

Although the coordinates of the tangent points in the ground plane and R_{\min} specify the circular arc, it is more convenient to express it in terms of R_{\min} and the coordinates of the circle's center, (x_c, y_c) . This point is just the place of intersection of the two radii shown in Figure C.1. The coordinates are given by

$$x_c = \frac{b_2 - b_1}{m_1 - m_2} \quad (C-5a)$$

$$y_c = \frac{m_1 b_2 - m_2 b_1}{m_1 - m_2} \quad (C-5b)$$

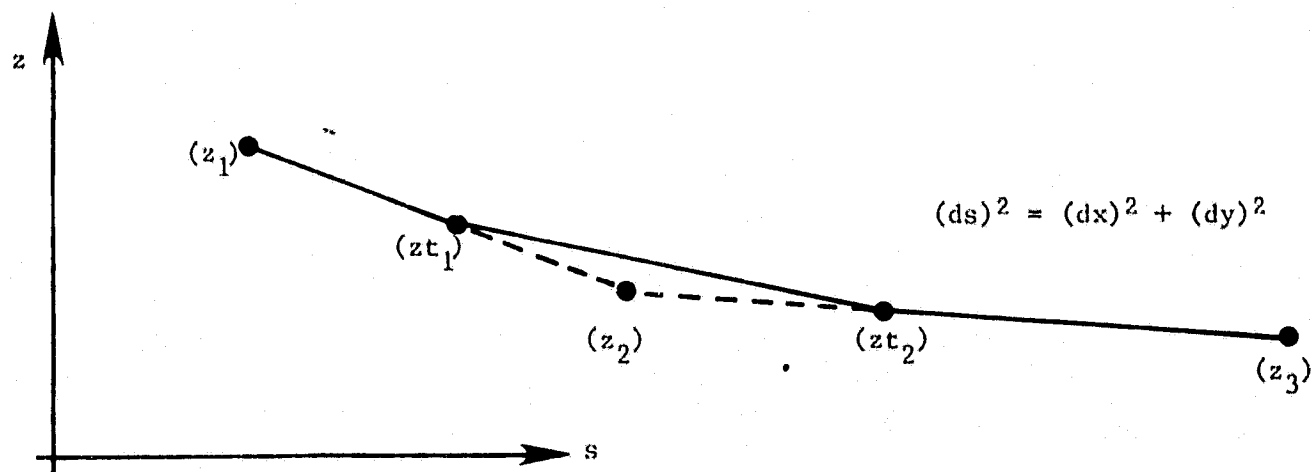
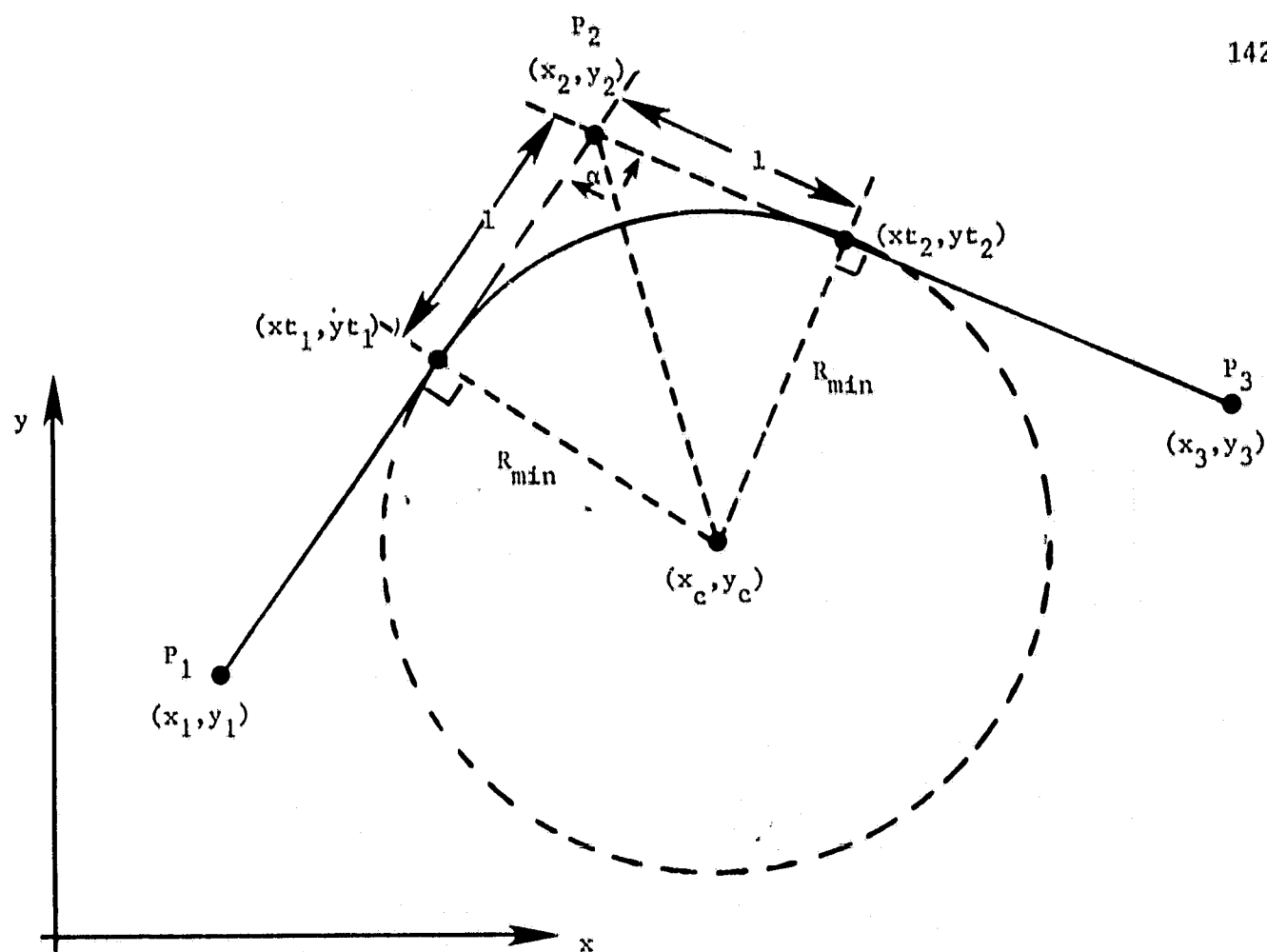


Figure C.1 The Tangent Points

where

$$M_1 = \frac{x_1 - x_{t1}}{y_{t1} - y_1} \quad (C-5c)$$

$$m_2 = \frac{x_3 - x_{t2}}{y_{t2} - y_3} \quad (C-5d)$$

$$b_1 = y_{t1} - m_1 x_{t1} \quad (C-5e)$$

$$b_2 = y_{t2} - m_2 x_{t2} \quad (C-5f)$$

APPENDIX D
MANIP2 AND CYCLIC PROGRAM LISTINGS

C			MAT	650
C		MAT	660
C			MAT	670
C	INPUT PARAMETERS:		MAT	680
C	IENTRY = "FIXED" FOR FIXED ENTRY PTS.		MAT	690
C	= "VARIAN" FOR VARIABLE ENTRY PTS.		MAT	700
C	ICNTINU = 0, INITIAL TRAJ'S ARE STRAIGHT LINES		MAT	710
C	= 1, INITIAL TRAJ'S ARE USER INPUT		MAT	720
C	SAMPT = TIME IN SECONDS BETWEEN NOISE LEVEL SAMPLES		MAT	730
C	PERIOD = TIME IN HRS. OVER WHICH THE SIMULATION OCCURS		MAT	740
C	ACSPD = A/C SPEED IN METERS/SEC.		MAT	750
C	NTRAJ = NUMBER OF FLIGHT TRAJECTORIES		MAT	760
C	NMAP = NUMBER OF POPULATION POINTS ON MAP		MAT	770
C	NSEG = NUMBER OF LINE SEGMENTS ON EACH TRAJECTORY		MAT	780
C	MAXIT = MAXIMUM NO. OF ITERATIONS ALLOWED		MAT	790
C	IF ONLY RESULT OF INITIAL CONDITION IS NEEDED, SET MAXIT = 0		MAT	800
C	DELH = PERTURBATION (METERS) IN X,Y DIRECTIONS AT CORNER		MAT	810
C	POINTS FOR GRADIENT CALCULATION		MAT	820
C	STOPCHG = STOP CRITERION FOR SUCCESSIVE COST CHANGE		MAT	830
C	XC,YC,ZC,XF,YF,ZF = STARTING AND FINAL POINTS OF TRAJECTORIES		MAT	840
C	XFCRT,YFCRT = AIRPORT LOCATION (METERS)		MAT	850
C	XP,YP,ZP (I,J) = COORD'S OF J-TH CORNER PT. ON I-TH TRAJ.		MAT	860
C	PWEIT1 = PENALTY WEIGHT ON DYN. CONSTR. ON TRAJ'S		MAT	870
C	PWEIT2 = PENALTY WEIGHT ON CONSTR. ON FINAL ANGLE		MAT	880
C	RADIUS = MIN. ALLOWED TURNING RADIUS (METERS)		MAT	890
C	FATCFW = FUEL AND TIME OF FLIGHT WEIGHT		MAT	900
C	XMIN,XMAX,YMIN,YMAX = BOUNDARIES OF AREA EXCLUDED		MAT	910
C	FROM MAX. NOISE (THRESHOLD) CONSTR.		MAT	920
C	ALMAX = MAX. ALLOWED A-LEVEL NOISE		MAT	930
C	RATIO = FRACTION OF FLIGHTS ALLOWED TO VIOLATE MAX. A-LEVEL		MAT	940
C	TWEIT1 = PENALTY WEIGHT ON MAX. NOISE AND NUMBER CONSTR.		MAT	950
C	TWEIT2 = NOT USED NOW, BUT SOME VALUE MUST BE INPUT		MAT	960
C	XBFGIN,XFINAL = BOUNDARIES ON X-AXIS WHERE SEPARATING		MAT	970
C	CONSTR. IS IMPOSED		MAT	980
C	YDIS = MIN. SEPARATING DISTANCE (Y-AXIS) BETWEEN TRAJ'S.		MAT	990
C	CWEIT1,CWEIT2 = PENALTY WEIGHTS ON SEP. AND CROSCOVER		MAT	1000
C	CONSTR'S. RESPECTIVELY		MAT	1010
C	IPROF = 0, NO PROFILE OPTIMIZATION		MAT	1020
C	= 1, PROFILE OPTIMIZATION OCCURS		MAT	1030
C	NOTE: IPROF IS ALWAYS SET = 1 IN THIS PROGRAM		MAT	1040
C	ZWEIT1 = PENALTY WEIGHT ON VERTICAL SLOPE CONSTR.		MAT	1050
C	GMAPAX = MAX. ALLOWED VERTICAL ANGLE ON EACH SEG.		MAT	1060
C	GMAPIN = MIN. ALLOWED VERTICAL ANGLE ON EACH SEG.		MAT	1070
C	PENCRIT = MIN. SIGNIFICANT DYN. PENALTY LEVEL		MAT	1080
C	AWEIT = SCALING ON NTI		MAT	1090
C			MAT	1100
C	ACHIX = NAMELIST: ACTYPE (I,1) = NO. OF ACTYPE ON TRAJ. I		MAT	1110
C	IN DAYTIME		MAT	1120
C	ACTYPE (I,2) = NO. OF ACTYPE ON TRAJ. I		MAT	1130
C	AT NIGHT		MAT	1140
C		MAT	1150
C			MAT	1160
C	CALL DATE (DAT)		MAT	1170
C	ENCODE (10,9010,DAT) DAT		MAT	1180
C	CALL REMARK (ICAT)		MAT	1190
C	CALL JPARAMS (JN)		MAT	1200
C	ENCODE (7,9020,JOBNAME) JN(1)		MAT	1210
C	CALL REMARK (JOBNAME)		MAT	1220
C	WRITE (6,9110)		MAT	1230
C	WRITE (6,9030)		MAT	1240
C	WRITE (6,9040) JOBNAME,DAT		MAT	1250
C	NPLANE = 0		MAT	1260
C			MAT	1270
C		MAT	1280

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C . INPUT ALL THE INFORMATION . MAT 1290
C . . MAT 1300
C . . MAT 1310
C ..... MAT 1320
C ..... MAT 1330
C READ (7,9120) (LABEL(I),I=1,8) . MAT 1340
C READ (7,9050) IENTRY . MAT 1350
C IF (IENTRY.NE."VARIAN".AND.IENTRY.NE."FIXED") CALL ERROR (4,"ILLEGAL . MAT 1360
C $L ENTRY POINT TYPE, CHECK DATA") . MAT 1370
C ..... MAT 1380
C ..... MAT 1390
C ..... MAT 1400
C ISEG1 = 1 : FIXED ENTRY PTS., 1ST SEG. INCLUDED IN ALL CALCS . MAT 1410
C ISEG1 = 2 : VARIABLE ENTRY PTS., 1ST SEG. EXCLUDED . MAT 1420
C ..... MAT 1430
C ..... MAT 1440
C ..... MAT 1450
C ISEG1 = 1 . MAT 1460
C IF (IENTRY.EQ."VARIAN") ISEG1 = 2 . MAT 1470
C READ (7,*) ICNTINL,SAMPT,PERICD,ACSPD . MAT 1480
C READ (7,*) NTRAJ,NMAP,NSEG,MAXIT . MAT 1490
C READ (7,*) DELT,STOPCHG . MAT 1500
C SAMLTH = SAMPT*ACSPC . MAT 1510
C DC 30 I = 1,NTRAJ . MAT 1520
C READ (7,*) XC(I),YO(I),ZO(I) . MAT 1530
C READ (7,*) XF(I),YF(I),ZF(I) . MAT 1540
C READ (7,*) XPORT(I),YPORT(I) . MAT 1550
C ..... MAT 1560
C ..... MAT 1570
C ..... MAT 1580
C CALCULATE INITIAL CORNER POINTS . MAT 1590
C ..... MAT 1600
C ..... MAT 1610
C ..... MAT 1620
C XM(I,1) = XC(I) . MAT 1630
C YM(I,1) = YO(I) . MAT 1640
C ZP(I,1) = ZO(I) . MAT 1650
C XM(I,NSEG+1) = XF(I) . MAT 1660
C YM(I,NSEG+1) = YF(I) . MAT 1670
C ZP(I,NSEG+1) = ZF(I) . MAT 1680
C XM(I,NSEG+2) = (XPORT(I)+XF(I))/2. . MAT 1690
C YM(I,NSEG+2) = (YPORT(I)+YF(I))/2. . MAT 1700
C ZM(I,NSEG+2) = ZF(I)/2. . MAT 1710
C IF (ICNTINL.EQ.1) GO TO 20 . MAT 1720
C DC 10 J = 2,NSEG . MAT 1730
C XM(I,J) = XC(I)+(FLOAT(J-1)/FLOAT(NSEG))*(XF(I)-XC(I)) . MAT 1740
C YM(I,J) = YO(I)+(FLOAT(J-1)/FLOAT(NSEG))*(YF(I)-YO(I)) . MAT 1750
C ZM(I,J) = ZO(I)+(FLOAT(J-1)/FLOAT(NSEG))*(ZF(I)-ZO(I)) . MAT 1760
10 CCNTINUE . MAT 1770
C GO TO 30 . MAT 1780
20 READ (7,*) (XM(I,J),YM(I,J),ZP(I,J),J=2,NSEG) . MAT 1790
30 CCNTINUE . MAT 1800
C DC 40 I = 1,NMAP . MAT 1810
C DC 40 J = 4,6 . MAT 1820
C ARRAY(I,J) = 0. . MAT 1830
40 CCNTINUE . MAT 1840
C READ (7,*) PWEIT1,PWEIT2,RADIUS,FATCFW . MAT 1850
C READ (7,*) XMIN,XMAX,YMIN,YMAX . MAT 1860
C READ (7,*) ALMAX,RATIO,TWEIT1,TWEIT2 . MAT 1870
C READ (7,*) XBEGIN,XFINAL,YOIS,CWEIT1,CWEIT2 . MAT 1880
C READ (7,*) IPRCF,ZWEIT1,CHAMIN,CHAMAX,PENCRIT,AWEIT . MAT 1890
C READ (10,*) ((ARRAY(I,J),J=1,3), (ARRAY(I,J),J=6,9),RMO(I),I=1,NMAP) . MAT 1900
C READ (7,ACMIX) . MAT 1910
C ..... MAT 1920

```

```

      IPRCF = 1
      DO 50 I = 1,NTPAJ
        DO 50 L = 1,32
          NPLANE = NPLANE+NAC(I,L)
50    CONTINUE
      DO 60 I = 1,NMAP
        ARRAY(I,10) = 1.
        IF (XMIN.LE.ARRAY(I,1).AND.ARRAY(I,1).LE.XMAX.AND.YMIN.LE.ARRAY(I,1)
          & I,2).AND.ARRAY(I,2).LE.YMAX.AND.ARRAY(I,3).NE.0.) ARRAY(I,10) =
          & 0.
60    CONTINUE
      MAXFLIT = IFIX(NPLANE*RATIO*0.5)

C .....
C .....
C ..... PRINT INFORMATION INPUT .....
C .....
C .....
C .....
      WRITE (6,9130) (LABEL(I),I=1,8)
      WRITE (6,9140) MAXIT,NTRAJ,NSEG,NMAP,DELH,STPCMG
      WRITE (6,9060) SAMPT,PERIGD,ACSPD
      IF (IENTRY.EQ."FIXED") WRITE (6,9070)
      IF (IENTRY.EQ."VARIABLE") WRITE (6,9080)
      IF (IPRCF.EQ.1) WRITE (6,9090)
      IF (IPRCF.EQ.0) WRITE (6,9100)
      WRITE (6,9150) ALMAX,PAXFLIT,XMIN,XMAX,YMIN,YMAX
      WRITE (6,9160) XBEGIN,XFINAL,YDIS
      WRITE (6,9170) RADILS,PWEIT1,PWEIT2,ZWEIT1,TWEIT1,TWEIT2,CHEIT1,CHEIT2,GMAMIN,GMAMAX,PENCRIT,AWEIT,FATOFW
      SEIT2,GMAMIN,GMAMAX,PENCRIT,AWEIT,FATOFW
      DO 90 I = 1,NTRAJ
        WRITE (6,9180) I,XO(I),YO(I),ZO(I),XF(I),YF(I),ZF(I)
        NSEG1 = NSEG+1
        DO 70 J = 1,NSEG1
          WRITE (6,9200) J,XM(I,J),YM(I,J),ZM(I,J)
70    CONTINUE
        WRITE (6,9210)
        DO 80 L = 1,31,2
          IF (NAC(I,L).NE.0.OR.NAC(I,L+1).NE.0) WRITE (6,9220) NAMAC(L),
          & NAMAC(L+1),NAC(I,L),NAC(I,L+1)
80    CONTINUE
        WRITE (6,9190) XPCPT(I),YPORT(I)
90    CONTINUE
        N = 2*(NSEG-1)*NTRAJ
        NVAR = 2
        IF (IPRCF.EQ.1) N = 3*(NSEG-1)*NTPAJ
        IF (IPRCF.EQ.1) NVAR = 3
        DO 100 I = 1,NTRAJ
          NSEG1 = NSEG-1
          DO 100 J = 1,NSEG1
            DO 100 K = 1,NVAR
              L = (I-1)*NVAR+NSEG1+(J-1)*NVAR+K
              IF (K.EQ.1) XNCW(L,1) = XM(I,J+1)
              IF (K.EQ.2) XNCW(L,1) = YM(I,J+1)
              IF (K.EQ.3) XNCW(L,1) = ZM(I,J+1)
100    CONTINUE
          DO 110 I = 1,N
            DELTAX(I) = DELH
110    CONTINUE

C .....
C .....
C ..... START OPTIMIZATION .....
C .....

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ORIGINAL PAGE IS
OF POOR QUALITY

```

C ..... MAI 257C
C CALL ONEWTON (MAXIT,STCPCG,N,XNOW,DELTAX,30,PENCRIT) MAI 258C
C STOP MAI 259C
C MAI 260C
C MAI 261C
901C FCRMAT (A10) MAI 262C
902C FCRMAT (A7) MAI 263C
903C FCRMAT (//," NO AMBIENT NOISE INCLUDED",//) MAI 264C
904C FCRMAT (1H1,9X,".....",//,10X,".",11X,".",//,10X,".",2X,A7,2MAI 265C
9X,".",//,10Y,".",11Y,".",//,10X,".",A10,1X,".",//,10X,".",11X,".",//,1MAI 266C
50X,".....",//) MAI 267C
905C FCRMAT (A5) MAI 268C
906C FCRMAT (5X,"FOR LDN CALCULATION, SAMPLE TIME = ",F4,C," SEC.",//,5XMAI 269C
5,"A/C LANDING PATTERN REPEATS EVERY ",F5.1," MPS.",//,5X,"A/C SPEEDMAI 270C
5 = ",F4.2," M/M/S",//) MAI 271C
907C FCRMAT (//,4X,"FIXED ENTRY POINTS",//) MAI 272C
908C FCRMAT (//,4X,"VARIABLE ENTRY POINTS",//) MAI 273C
909C FCRMAT (//,4X,".. PFCFILES OPTIMIZED ..",//) MAI 274C
910C FCRMAT (//,4X,".. ONLY CROUND TRACKS ARE OPTIMIZED ..",//) MAI 275C
911C FCRMAT (1H1,10H*****",//) MAI 276C
912C FCRMAT (EAL0) MAI 277C
913C FCRMAT (1X,EAL0) MAI 278C
914C FCRMAT (///,5X,"MAXIMUM ITERATION SETI ",I2,//,5X,"NUMBER OF TRAJECTORIES: ",I2,//,5X,"NUMBER OF SEGMENTS ON EACH TRAJECTORY: ",I1,MAI 279C
1X,"NUMBER OF POPULATION POINTS ON THE MAP: ",I4,//,5X,"PERTURB MAI 280C
5X,"TRAJECTORIES IN X AND Y DIRECTIONS ",F10.5," METERS FOR CALCULATINGMAI 281C
5 GRADIENTS",//,5X,"EXIT CRITERION FOR CNEWTON: ",1PE16.9,//) MAI 282C
915C FCRMAT (5X,"NO BLOCK MAY RECEIVE ",F5.1," DB (OR HIGHER) A-LEVEL",MAI 283C
5X,"NOISE MORE THAN ",I4," TIMES A DAY",//,3X,"..THIS CONSTRAINT ",MAI 284C
5EXCLUDES BLOCKS WITH CENTROIDS INSIDE THE RANGES: ",F8.1," TO ",F8MAI 285C
5.1," (X) AND ",F8.1," TO ",F8.1," (Y)",//) MAI 286C
916C FCRMAT (5X,40HWITHIN THE AREA OF X-COORDINATES BETWEEN,F8.1,5X ANCHMAI 287C
5 ,F8.1,41H, THE SEPARATING DISTANCE BETWEEN TRAJECT,5HOPIES,//,5X,1MAI 288C
5PSSHOLD BE AT LEAST,F8.1,7H METERS,//) MAI 289C
917C FCRMAT (//,5X,"OTHER CONSTRAINT PARAMETERS",//,4X,"MINIMUM RADIUS",MAI 290C
5 CF CURVATURE (IN X-Y PLANE PROJECTION) = ",1PE9.2,//,4X,"DYNAMIC MAI 291C
5PENALTY WEIGHTS: PWEIT1 = ",E9.2," PWEIT2 = ",E9.2," ZWEIT1 = ",MAI 292C
5E9.2,//,4X,"MAXIMUM NOISE LEVEL (THRESHOLD) ",MAI 293C
5EIT1 = ",E9.2,"TWEIT2 = ",E9.2,//,4X,"SEPARATION PENALTY WEIGHTS: MAI 294C
5CWEIT1 = ",E9.2," CWEIT2 = ",E9.2,//,4X,"CHAMIN = ",OPF5.1," GPAMAPMAI 295C
5X = ",F5.1,"PENCRIT = ",1PE9.2," AWEIT = ",E9.2,5X,"FATOFW = ",E9.2MAI 296C
52,///,5X,"INFORMATION OF EACH INITIAL TRAJECTORY:",//) MAI 297C
918C FCRMAT (10X,15HFLIGHTPATH NC: ,I1,//,12X,13HINITIAL X,Y,Z,1X,23HCOCPMAI 298C
5ORDINATES IN METERS: ,3(F8.1,3X),//,12X,11HFINAL X,Y,Z,1X,23HCOCPDINMAI 299C
5ATES IN METERS: ,3(F8.1,3X),//,12X,14HINITIAL CORNER,12H POINT POSIMA I 300C
5TIONS: ,//,14X,10HCCORNER NC: ,6X,1HX,9X,1HY,9X,1HZ,//) MAI 301C
919C FCRMAT (10X,"RUNWAY LOCATION: ",F8.1,10X,F8.1,//) MAI 302C
920C FCRMAT (18X,I1,6X,F8.1,3X,F8.1,3X,F8.1) MAI 303C
921C FCRMAT (15X,"A/C MIX",T45,"NO. OF A/C",//,T40,"DAYTIME",3X,"NIGHTTMAI 304C
5IME",//) MAI 305C
922C FCRMAT (T20,2A10,T42,I3,T52,I3/) MAI 306C
END MAI 307C
MAI 308C

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SUBROUTINE COST (IGRAD,TCTAL,ANTI,PNALTY,CLOSE,THRESH) CCS 10
COMMON NTRAJ,NMAP,NSEG,XM(3,6),YM(3,6),ZM(3,6),ARRAY(576,9),SPOSITCCS 20
5(3,200,3),XO(3),XF(3),YO(3),YF(3),ZC(3),ZF(3),NPOSTS(3),PHC(576),CCS 30
5ISEG1,IPOCF,MAXIT,PENCRIT,AWEIT,TVICLA,ALWP,FATQFC,FATQFW CCS 40
COMMON /CCORNER/ ANGLE(3,3),POSIT(3,200,3),XCENPT(3,3),YCENPT(3,3),CCS 50
5ZCENTR(3,3),XT1(3,5),YT1(3,5),ZT1(3,5),XT2(3,5),YT2(3,5),ZT2(3,5),CCS 60
5NPOST(3),DIS(3) CCS 70
COMMON /PRINT/ SYCENTP(3,3),SYCENTR(3,3),SXT1(3,5),SYT1(3,5),SZT1(CCS 80

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13,5),SXT2(3,5),SYT2(3,5),SZT2(3,5),SANGLE(3,5)
EATOF=0.
TCTAL=FATOF
ANII=TOTAL
CLOSE=ANII
THRESH=CLOSE
PNALTY=THRESH
DO 10 I = 1,NMAP
  DO 10 J = 4,5
    ARRAY(I,J) = 0.
10 CONTINUE
CALL OWNARID (PNALTY)
IF (PNALTY.LT.PENCRIT) GO TO 20
TCTAL = PNALTY
IF (MAXIT.NE.0) PETLPH
20 CALL NOISEL (THRESH)
CALL NII (ANII)
IF (NTRAJ.EQ.1) GO TO 30
CALL CRCSCVR (NPCSIT,POSIT,CLCSE)
30 TCTAL = ANII+THRESH+CLOSE+PNALTY+FATOF
IF (IGRAD.EQ.1) RETURN
DO 60 I = 1,NTRAJ
  SXT1(I,1) = XT1(I,1)
  SXT2(I,1) = XT2(I,1)
  SYT1(I,1) = YT1(I,1)
  SYT2(I,1) = YT2(I,1)
  SZT1(I,1) = ZT1(I,1)
  SZT2(I,1) = ZT2(I,1)
  DO 40 J = 2,NSEG
    SXCENTR(I,J-1) = XCENR(I,J-1)
    SYCENTR(I,J-1) = YCENTR(I,J-1)
    SXT1(I,J) = XT1(I,J)
    SYT1(I,J) = YT1(I,J)
    SXT2(I,J) = XT2(I,J)
    SYT2(I,J) = YT2(I,J)
    SZT1(I,J) = ZT1(I,J)
    SZT2(I,J) = ZT2(I,J)
    SANGLE(I,J-1) = ANGLE(I,J-1)
  40 CONTINUE
  NPOSITS(I) = NPCSIT(I)
  IPCST = NPOSITS(I)
  DO 50 J = 1,IPCST
    SPOSIT(I,J,1) = PCSIT(I,J,1)
    SPOSIT(I,J,2) = PCSIT(I,J,2)
    SPOSIT(I,J,3) = PCSIT(I,J,3)
  50 CONTINUE
60 CONTINUE
RETURN
END

```

```

CCS  40
CCS  100
CCS  110
CCS  120
CCS  130
CCS  140
CCS  150
CCS  160
CCS  170
CCS  180
CCS  190
CCS  200
CCS  210
CCS  220
CCS  230
CCS  240
CCS  250
CCS  260
CCS  270
CCS  280
CCS  290
CCS  300
CCS  310
CCS  320
CCS  330
CCS  340
CCS  350
CCS  360
CCS  370
CCS  380
CCS  390
CCS  400
CCS  410
CCS  420
CCS  430
CCS  440
CCS  450
CCS  460
CCS  470
CCS  480
CCS  490
CCS  500
CCS  510
CCS  520
CCS  530
CCS  540
CCS  550
CCS  560
CCS  570

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SUBROUTINE CCST1 (IGRAD,N,F,X,ANII,PNALTY,CLCSE,THRESH,NDEC)
COMMON NTRAJ,NMAP,NSEG,XH(3,6),YH(3,6),ZH(3,6),ARRAY(576,9),SPCSITCC1
(3,200,3),XC(3),XF(3),YC(3),YF(3),ZC(3),ZF(3),NPCSITS(3),PHQ(576),CC1
$1SEG1,IPRCF,MAXIT,PENCRIT,ALBIT,TVICLA,ALWP,FATOF,CLCSE,FATOFW
DIMENSION X(NDEC,1)
NVAR = 2
IF (IPRCF.EQ.1) NVAR = 3
DO 10 I = 1,NTRAJ
  NSEG1 = NSEG-1
  DO 10 J = 1,NSEG1
    DO 10 K = 1,NVAR

```

```

CC1  10
CC1  20
CC1  30
CC1  40
CC1  50
CC1  60
CC1  70
CC1  80
CC1  90
CC1  100
CC1  110

```

```

      L = (I-1)*NVAR+NSEG1+(J-1)*NVAR+K
      IF (K.EQ.1) YP(I,J+1) = Y(L,1)
      IF (K.EQ.2) YP(I,J+1) = X(L,1)
      IF (K.EQ.3) ZP(I,J+1) = X(L,1)
10  CONTINUE
      CALL COST (IGRAD,F,ANII,PNALTY,CLOSE,THRESH)
      PETURN
      END

```

CC1	120
CC1	130
CC1	140
CC1	150
CC1	160
CC1	170
CC1	180
CC1	190

```

      FUNCTION AACOS (A,B,C)
      X = (A**2+B**2-C**2)/(2.*A*B)
      IF (ABS(X).LT.1.1) GO TO 10
      WRITE (6,9010) X,A,B,C
      STOP
10  IF (X.GT.1.) X = 1.
      IF (X.LT.-1.) X = -1.
      AACOS = ACOS(X)
      RETURN
C
9010 FCFMAT (29H TROUBLE IN AACOS, X,A,B,C = ,4(IPE16.9,3X))
      END

```

AAC	10
AAC	20
AAC	30
AAC	40
AAC	50
AAC	60
AAC	70
AAC	80
AAC	90
AAC	100
AAC	110
AAC	120

```

      SUBROUTINE MCNIT (IT,Y,N,F,ANII,PNALTY,CLOSE,THRESH,NDEC)
      COMMON NTRAJ,NMAP,NSEG,XM(3,6),YM(3,6),ZM(3,6),APPAY(576,9),SPCSITS(576),
      & (3,200,3),XO(3),YF(3),YC(3),YF(3),ZO(3),ZF(3),NPCSITS(3),RMC(576),MCN
      & NSEG1,IPOFF,MAXIT,PENCPIT,AWEIT,TVICLA,ALWP,FATGFC,FATOFW
      COMMON /PRINT/ SXCENTR(3,3),SYCENTR(3,3),SXT1(3,5),SYT1(3,5),SZT1(
      & 3,5),SXT2(3,5),SYT2(3,5),SZT2(3,5),SANGLE(3,3)
      DIMENSION X(NDEC,1)
      PI = ATAN(1.)*4.
      NVAR = 2
      IF (IPPCF.EQ.1) NVAR = 3
      DO 10 I = 1,NTRAJ
      NSEG1 = NSEG-1
      DO 10 J = 1,NSEG1
      DO 10 K = 1,NVAR
      L = (I-1)*NVAR+NSEG1+(J-1)*NVAR+K
      IF (K.EQ.1) XM(I,J+1) = X(L,1)
      IF (K.EQ.2) YP(I,J+1) = X(L,1)
      IF (K.EQ.3) ZP(I,J+1) = X(L,1)
10  CONTINUE
      WRITE (6,80) IT,F,ANII,PNALTY,CLOSE,THRESH,TVICLA,ALWP,FATGFC
      DO 30 I = 1,NTRAJ
      WRITE (6,40) I,XP(I,1),YM(I,1),ZM(I,1)
      DO 20 J = 2,NSEG
      WRITE (6,50) XM(I,J),YM(I,J),ZM(I,J),SXCENTR(I,J-1),SYCENTR(I,
      & J-1),SXT1(I,J),SYT1(I,J),SZT1(I,J),SXT2(I,J),SYT2(I,J),SZT2(I,
      & J),SANGLE(I,J-1)*180./PI
20  CONTINUE
      WRITE (6,60) XM(I,NSEG+1),YM(I,NSEG+1),ZM(I,NSEG+1)
      NPCSITI = NPCSITS(I)
      WRITE (8,70) (ISPCSI(I,J,K),K=1,3),J=1,NPCSITI)
30  CONTINUE
      RETURN
C
C
40 FCFMAT (12Y,"FLIGHT PATH NO: ",I1,/,T20,"CORNER PTS.",T40,"CENTER MON
      & OF CIRCLE",T74,"TANGENTIAL PTS.",T110,"ANGLE (DEGREE)",/,14X,(" ",FMN

```

MCN	10
MCN	20
MCN	30
MCN	40
MCN	50
MCN	60
MCN	70
MCN	80
MCN	90
MCN	100
MCN	110
MCN	120
MCN	130
MCN	140
MCN	150
MCN	160
MCN	170
MCN	180
MCN	190
MCN	200
MCN	210
MCN	220
MCN	230
MCN	240
MCN	250
MCN	260
MCN	270
MCN	280
MCN	290
MCN	300
MCN	310
MCN	320
MCN	330
MCN	340
MCN	350
MCN	360

	87.0,M,F7.0,M,F7.0,M)	PEN	420
9C	FOPMAT (14X,"M","F7.C","M","F7.O","M","F7.O","M"),M("M","F7.C","M","F7.O","M"),ZPCN	360	
	S("M","F7.C","M","F7.C","M","F7.O","M"),3X,F6.1))	PEN	390
6C	FOPMAT (14X,"M","F7.C","M","F7.C","M","F7.O","M")	PEN	400
7O	FOPMAT (3(FB.1,ZX))	PEN	410
	END	PEN	420
	SUBRCUTINE RESLTY (IT,X,N,F,ANII,PNALTY,CLOSE,THRESH,NDEC) PFS	10	
	COMMON NTRAJ,NMAP,NSEG,XM(3,6),YM(3,6),ZM(3,6),ARRAY(576,9),SPCSITRES	20	
	(3,200,3),YO(3),XF(3),YC(3),YF(3),ZC(3),ZF(3),NPOSITS(3),PHQ(576),PFS	30	
	\$ISEG1,IPOOF,MAXIT,PENCRT,AWEIT,TVICLA,ALWP,FATCFE,FATOW	PES	40
	COMMON /PRINT/SYCENIP(3,3),SYCENTR(3,3),SXTI(3,5),SYTI(3,5),SZTI(PES	50	
	3,5),SXT2(3,5),SYT2(3,5),SZT2(3,5),SANGLE(3,3)	PES	60
	DIMENSION X(NDEC,1)	PES	70
	PI = ATAN(1.)*.4.	PES	80
	NVAR = 2	PES	90
	IF (IPFCF.EC.1) NVAR = 3	PES	100
	DC 10 I = 1,NTRAJ	PES	110
	NSEG1 = NSEG-1	PES	120
	DO 10 J = 1,NSEG1	PES	130
	DO 10 K = 1,NVAR	PES	140
	L = (I-1)*NVAR+NSEG1+(J-1)*NVAR+K	PES	150
	IF (K.EQ.1) XM(I,J+1) = Y(L,1)	PES	160
	IF (K.EQ.2) YM(I,J+1) = X(L,1)	PES	170
	IF (K.EQ.3) ZM(I,J+1) = X(L,1)	PES	180
10	CONTINUE	PES	190
	WRITE (6,9010) IT,F,ANII,PNALTY,CLOSE,THRESH,TVICLA,ALWP,FATOCF	PES	200
	DO 30 I = 1,NTRAJ	PES	210
	WRITE (6,9020) I,XM(I,1),YM(I,1),ZM(I,1)	PES	220
	DO 20 J = 2,NSEG	PES	230
	WRITE (6,9030) XM(I,J),YM(I,J),ZM(I,J),SXCENTR(I,J-1),SYCENTR(PES	240	
	I,J-1),SXTI(I,J),SYTI(I,J),SZTI(I,J),SXT2(I,J),SYT2(I,J),SZT2(PES	250	
	I,J),SANGLE(I,J-1)*180./PI	PES	260
20	CONTINUE	PES	270
	WRITE (6,9040) XM(I,NSEG+1),YM(I,NSEG+1),ZM(I,NSEG+1)	PES	280
	WRITE (6,9050)	PES	290
	NPOSITI = NPCSITI(I)	PES	300
	WRITE (6,9060) ((SPOSIT(I,J,K),K=1,3),J=1,NPOSITI)	PES	310
	WRITE (8,9070) ((SPOSIT(I,J,K),K=1,3),J=1,NPOSITI)	PES	320
30	CONTINUE	PES	330
	WRITE (9,9080) ((ARRAY(I,J),J=4,5),I=1,NMAP)	PES	340
	RETURN	PES	350
C		PES	360
9010	FOPMAT (1H1,9X,11ITERATION:,12/,12X,12HTOTAL CCST:,1PE16.9,,1PES	370	
	12X,17HANNOYANCE (NI)),1PE16.9,,12X,31MPENALTY ON DYNAMIC CNSTRPS	PES	380
	SAINTI,1PE16.9,,12X,34MPENALTY ON SEPARATING CONSTRAINTI,1PE16.GPF	PES	390
	1/,12X,2HPENALTY ON THRESHOLD NOISE:,1PE16.9,,12X,THRESH TIVCLAPF	PES	400
	TION = ",E16.9,,12X,LWP = ",E16.9,,12X,MFATCFE = ",E16.9,/)	PES	410
9020	FOPMAT (12X,"FLIGHT PATH NO:",11,,T20,"CORNER PTS.",T40,"CENTER PES	420	
	"OF CIRCLE",T74,"TANGENTIAL PTS.",T110,"ANGLE (DEGREE)",/,14X,"(",PES	430	
	87.0,"M","F7.C","M","F7.C","M")	PES	440
9030	FOPMAT (14X,"(",F7.C,"M","F7.O","M","F7.O","M"),("M","F7.C","M","F7.O","M"),2RES	450	
	S("M","F7.O","M","F7.C","M","F7.O","M"),3X,F6.1)	PES	460
9040	FOPMAT (14X,"(",F7.C,"M","F7.O","M","F7.O","M")	PES	470
9050	FOPMAT (/,,12X,36HFLIGHT TRAJECTORY,X,Y,Z, IN METEPS,/,19X,1PX,1PES	480	
	14X,1HY,14X,1HZ,/))	PES	490
9060	FOPMAT (14X,1PE10.3,5X,E10.3,5X,E10.3,5X)	PES	500
9070	FOPMAT (3(FB.1,ZX))	PES	510
9080	FOPMAT (2(1PE10.3,1X))	PES	520
	END	PES	530


```

9C   IF (KK.EQ.1) CPOSS(1+I,J) = YINTP(CPOSS(1,J),POSIT(1,1,1),PCSCPS  620
S   IT(1,1,2),PCSIT(1,2,1),POSIT(1,2,2),POSIT(1,3,1),POSIT(1,3,2),CPS  630
S   POSIT(1,4,1),PCSIT(1,4,2)) CPS  640
S   IF (KK.NE.1,AND,KK.NE.NP1) CROSS(1+I,J) = YINTP(CROSS(1,J),PCCPS  650
S   SIT(1,KK-1,1),POSIT(1,KK-1,2),POSIT(1,KK,1),POSIT(1,KK,2),PCSICPS  660
S   T(1,KK+1,1),POSIT(1,KK+1,2),POSIT(1,KK+2,1),POSIT(1,KK+2,2)) CPS  670
S   IF (KK.EQ.NP1) CPOSS(1+I,J) = YINTP(CROSS(1,J),POSIT(1,KK-2,1),CPS  680
S   ),POSIT(1,KK-2,2),PCSIT(1,KK-1,1),PCSIT(1,KK-1,2),POSIT(1,KK,1),CPS  690
S   ),POSIT(1,KK,2),POSIT(1,KK+1,1),PCSIT(1,KK+1,2)) CPS  700
100  CONTINUE CPS  710
110  CONTINUE CPS  720
C ..... CPS  730
C ..... CPS  740
C ..... CPS  750
C ..... NCV TEST THE NEARNESS OF OR THE CROSSOVER BETWEEN TRAJECTORIES CPS  760
C ..... CPS  770
C ..... CPS  780
C ..... CPS  790
C ..... CPS  800
C ..... CPS  810
C ..... DO 170 I = 2,NTRAJ CPS  820
C ..... DO 160 J = 3,NACD1 CPS  830
C ..... - IF (I.GE.J) GO TO 160 CPS  840
C ..... CO 140 K = 1,1C CPS  850
C ..... DIS1 = CROSS(I,K)-CROSS(J,K) CPS  860
C ..... IF (ABS(DIS1).GE.YDIS) GO TO 120 CPS  870
C ..... FX = FX+CWEIT1*(YDIS-DIS1)**2 CPS  880
120  IF (K.EQ.10) GO TO 140 CPS  890
C ..... DIS2 = CROSS(I,K+1)-CROSS(J,K+1) CPS  900
C ..... IF ((DIS1*DIS2).GE.0.) GO TO 140 CPS  910
C ..... I1 = 1 CPS  920
130  IF (K+1+I1.GT.10) GO TO 150 CPS  930
C ..... DIS3 = CROSS(I,K+1+I1)-CROSS(J,K+1+I1) CPS  940
C ..... IF (DIS2*DIS3.LE.0.) GO TO 150 CPS  950
C ..... IF (ABS(DIS2).LT.ABS(DIS3)) DIS2 = DIS3 CPS  960
C ..... I1 = I1+1 CPS  970
C ..... GO TO 130 CPS  980
140  CONTINUE CPS  990
C ..... GO TO 160 CPS  1000
150  FX = FX+CWEIT2*DIS3**2 CPS  1010
160  CONTINUE CPS  1020
170  CONTINUE CPS  1030
C ..... RETLPM CPS  1040
C ..... ENG CPS  1050

```

```

SUBROUTINE ONEWTN (MAXIT,STOPCHG,N,XNOW,DELTAX,NOEC,PENCRIT) ONE 10
C ..... ONE 20
C ..... ONE 30
C ..... ONE 40
C ..... THIS OPTIMIZATION EMPLOYS SELF-SCALING, RESTARTING, ONE 50
C ..... GLASS-NEWTON METHOD. ONE 60
C ..... REFERENCE: D.G. LUENBERGER INTRO. TO LINEAR AND NONLINEAR ONE 70
C ..... PROGRAMMING; P.204, SEC.9.5 ONE 80
C ..... MAXIT: MAXIMUM NUMBER OF ITERATIONS ALLOWED ONE 90
C ..... STOPCHG: STOP IF PERCENTAGE CHANGE IN SUCCESSIVE COSTS IS ONE 100
C ..... LESS THAN THIS VALUE ONE 110
C ..... NI: DIMENSION OF THE UNKNOWN X ONE 120
C ..... XNOW: PRESENT OF INITIAL VALUE OF UNKNOWN X ONE 130
C ..... ONE 140
C ..... ONE 150
C ..... ONE 160

```

```

DIMENSION XNOW(NDEC,1), DELTAX(NDEC)
DIMENSION GNOW(30,1), GNEXT(30,1), P(30,1), C(30,1), PC(1,1)
DIMENSION SQS(1,1), PP(30,30), SQQS(30,30), S(30,30)
DIMENSION XTEMP(30,1), PT(1,30), QS(1,30), SO(1,30), QT(1,30)
DIMENSION SQQ(30,30)
DIMENSION D(30,1)
IT = 0
CALL COST1 (G,N,FNCB,XNCW,ANII,PNALTY,CLOSE,THRESH,NDEC)
IF (PNALTY.GE.PENCRIT) GO TO 10
CALL MONIT (IT,XNEW,N,FNCW,ANII,PNALTY,CLOSE,THRESH,NDEC)
IF (MAXIT.EQ.0) STOP
GC TO 20
10 WRITE (6,9020) PNALTY,ANII
CALL CORNER (XNCW,N,NDEC)
IF (MAXIT.EQ.0) STOP
C
C .....
C
C STEP 1: SET S = IDENTITY MATRIX AND CALCULATE GRADIENT G
C
C .....
C
20 DO 30 I = 1,N
   DO 30 J = 1,N
     S(I,J) = 0.
     IF (I.EQ.J) S(I,J) = 1.
30 CONTINUE
CALL FGRAD (N,FNOW,XNCW,GNOW,DELTAX,NDEC)
C
C .....
C
C STEP 2: SET D = -SG
C
C .....
C
40 CALL MPLY (N,N,1,S,GNOW,D,NDEC,NDEC,1)
DO 50 I = 1,N
  D(I,1) = -D(I,1)
50 CONTINUE
C
C .....
C
C STEP 3: LINE SEARCH ALONG D TO FIND AFA THAT SATISFIES PO>0
C
C .....
C
K = 0
60 K = K+1
CALL LINESCH (K,N,FNOW,XNOW,D,FAA2,XTEMP,FSMALL,ANII,PNALTY,CLOSE,
  THRESH,NDEC)
CALL FGRAD (N,FSMALL,XTEMP,GNEXT,DELTAX,NDEC)
DO 70 I = 1,N
  Q(I,1) = GNEXT(I,1)-GNOW(I,1)
  P(I,1) = FAA2*D(I,1)
70 CONTINUE
CALL TRANSPOS (NDEC,1,P,PT)
CALL MPLY (1,N,1,PT,Q,PQ,1,NDEC,1)
IF (PQ(1,1).GT.0.) GO TO 90
IF (K.GT.3) RETURN
FNOW = FSMALL
DO 80 I = 1,N
  XNOW(I,1) = XTEMP(I,1)
80 CONTINUE
GC TO 60

```

```

ONE 170
ONE 180
ONE 190
ONE 200
ONE 210
ONE 220
ONE 230
ONE 240
ONE 250
ONE 260
ONE 270
ONE 280
ONE 290
ONE 300
ONE 310
ONE 320
ONE 330
ONE 340
ONE 350
ONE 360
ONE 370
ONE 380
ONE 390
ONE 400
ONE 410
ONE 420
ONE 430
ONE 440
ONE 450
ONE 460
ONE 470
ONE 480
ONE 490
ONE 500
ONE 510
ONE 520
ONE 530
ONE 540
ONE 550
ONE 560
ONE 570
ONE 580
ONE 590
ONE 600
ONE 610
ONE 620
ONE 630
ONE 640
ONE 650
ONE 660
ONE 670
ONE 680
ONE 690
ONE 700
ONE 710
ONE 720
ONE 730
ONE 740
ONE 750
ONE 760
ONE 770
ONE 780
ONE 790
ONE 800

```

```

90 DC 100 I = 1,N
   XNCW(I,1) = XTEMP(I,1)
100 CONTINUE
   IT = IT+1
   PRCNT = ABS((FSPALL-FNOW)/FNOW)
   IF (PRCNT.GE.STOPCHG) GO TO 120
   WRITE (6,9030) PRCNT,STOPCHG
   IF (PNALTY.GE.PENCRIT) GO TO 110
   CALL MONIT (IT,XNOW,N,FSPALL,ANII,PNALTY,CLOSE,THRESH,NDEC)
C
   RETURN
110 WRITE (6,9010) PNALTY
   CALL CORNER (XNOW,N,NDEC)
   RETURN
120 IF (IT.LT.MAXIT) GO TO 140
   WRITE (6,9040)
   IF (PNALTY.GE.PENCRIT) GO TO 130
   CALL MONIT (IT,XNOW,N,FSPALL,ANII,PNALTY,CLOSE,THRESH,NDEC)
C
   RETURN
130 WRITE (6,9010) PNALTY
   CALL CORNER (XNOW,N,NDEC)
   RETURN
140 IF (PNALTY.GE.PENCRIT) GO TO 150
   CALL MONIT (IT,XNOW,N,FSPALL,ANII,PNALTY,CLOSE,THRESH,NDEC)
   GO TO 160
150 WRITE (6,9020) PNALTY,ANII
   CALL CORNER (XNOW,N,NDEC)
160 CONTINUE
C
C .....
C : STEP 4: IF "IT" IS INTEGER MULTIPLE OF N GO TO STEP 1;
C :       IF NOT, UPDATE S
C :       IS "IT" INTEGER MULTIPLE OF N?
C : .....
C : IF ((FLOAT(IT)/FLOAT(N)).NE.FLOAT(IT/N)) GO TO 170
C : FACW = FSPALL
C : GO TO 20
C
C .....
C : UPDATE MATRIX S; GO TO STEP 2
C : .....
C : .....
170 CALL MPLY (N,N,1,S,G,SO,NDEC,NDEC,1)
   CALL TRANSPOS (NDEC,1,Q,QT)
   CALL MPLY (N,1,N,SC,QT,SCQ,NDEC,1,NDEC)
   CALL MPLY (N,N,N,SCQ,S,SCQS,NDEC,NDEC,NDEC)
   CALL MPLY (1,N,N,CT,S,CS,1,NDEC,NDEC)
   CALL MPLY (1,N,1,CS,G,CSQ,1,NDEC,1)
   CALL MPLY (N,1,N,P,PT,PP,NDEC,1,NDEC)
   DO 180 I = 1,N
     DC 180 J = 1,N
     S(I,J) = (S(I,J)-SCQS(I,J)/CSQ(1,1))*(PQ(1,1)/CSQ(1,1))+PP(I,J)/
$   PQ(1,1)
180 CONTINUE
   FNOW = FSPALL
   DC 190 I = 1,N
     GNCW(I,1) = GNEXT(I,1)
190 CONTINUE

```

```

ONE P10
ONE P20
ONE P30
ONE P40
ONE P50
ONE P60
ONE P70
ONE P80
ONE P90
ONE P100
ONE P110
ONE P120
ONE P130
ONE P140
ONE P150
ONE P160
ONE P170
ONE P180
ONE P190
ONE P200
ONE P210
ONE P220
ONE P230
ONE P240
ONE P250
ONE P260
ONE P270
ONE P280
ONE P290
ONE P300
ONE P310
ONE P320
ONE P330
ONE P340
ONE P350
ONE P360
ONE P370
ONE P380
ONE P390
ONE P400
ONE P410
ONE P420
ONE P430
ONE P440

```

```

      GO TO 40
C
9010 FOPMAT (5X,40HIN RESULT: DYNAMIC CONSTRAINTS VIOLATION,,10X,9HPENCKE
      PNALTY =,1PE16.9)
C
9020 FOPMAT (5X,39HIN MONIT: DYNAMIC CONSTRAINTS VIOLATION,,10X,9HPENCKE
      PNALTY =,1PE16.9,,10X,"ANII = ",E16.9," (IF = 0., THEN ANII HAS",MNDONE
      IT BEEN COMPUTED IN THIS ITERATION, DUE TO LARGE PNALTY")
C
9030 FOPMAT (2X,"RELATIVE CHANGE IN SUCCESSIVE COSTS ",1PE16.3," LESS",ONE
      1.," THAN STOP CRITERION ",1PE10.3)
C
9040 FOPMAT (2X,29HMAXIMUM ITERATION SET REACHED)
      END

```

```

OFF 142C
ONE 1440
ONE 1470
ONE 148C
ONE 1490
ONE 1500
ONE 1510
ONE 1520
ONE 1530
ONE 1540
ONE 1550

```

```

      SUBROUTINE MPLY (L,M,N,A,B,C,LDEC,MDEC,NDEC)
C
C .....
C
C . CALCULATE MATRIX MULTIPLICATION C = AB
C
C .....
C
      DIMENSION A(LDEC,MDEC), B(MDEC,NDEC), C(LDEC,NDEC)
      DO 10 I = 1,L
        DO 10 J = 1,N
          C(I,J) = 0.
          DO 10 K = 1,M
            C(I,J) = C(I,J)+A(I,K)*B(K,J)
          10 CONTINUE
      RETURN
      END

```

```

MPL 10
MPL 20
MPL 30
MPL 40
MPL 50
MPL 60
MPL 70
MPL 80
MPL 90
MPL 100
MPL 110
MPL 120
MPL 130
MPL 140
MPL 150
MPL 160
MPL 170

```

```

      SUBROUTINE TRNSPOS (M,N,A,B)
C
C .....
C
C . TRANSPOSE OF MATRIX A IS RETURNED IN MATRIX B
C
C .....
C
      DIMENSION A(M,N), B(N,M)
      DO 10 I = 1,M
        DO 10 J = 1,N
          B(J,I) = A(I,J)
        10 CONTINUE
      RETURN
      END

```

```

TFN 10
TFN 20
TFN 30
TFN 40
TFN 50
TFN 60
TFN 70
TFN 80
TFN 90
TFN 100
TFN 110
TFN 120
TFN 130
TFN 140
TFN 150

```

```

      SUBROUTINE FGRAD (N,F,X,C,DELTA,X,NDEC)
C
C .....
C
C . CALCULATE GRADIENT OF COST F WITH RESPECT TO UNKNOWN X
C
C .....
C
      DIMENSION X(NDEC,1), G(NDEC,1), DELTA(NDEC,1)

```

```

FCF 10
FCF 20
FCF 30
FCF 40
FCF 50
FCF 60
FCF 70
FCF 80
FCF 90

```

```

DO 10 I = 1,N
  G(I,1) = 0.
  IF (MOD(I,3).EQ.0) GO TO 10
  X(I,1) = X(I,1)+DELTA(I)
  CALL COST1 (1,N,FF,X,ANII,PNALTY,CLOSE,THRESH,NDEC)
  G(I,1) = (FF-F)/DELTA(I)
  X(I,1) = X(I,1)-DELTA(I)
10 CONTINUE
RETURN
END

```

```

FGP 100
FGP 110
FGP 120
FGP 130
FGP 140
FGP 150
FGP 160
FGP 170
FGP 180
FGP 190

```

```

SUBROUTINE GAFA (N,F,DAFA,G,X,D,NDEC)
C .....
C : CALCULATE GRADIENT OF F WITH RESPECT TO AFA
C : .....
C
  DIMENSION X(NDEC,1), D(NDEC,1), XTEMP(30,1)
  DO 10 I = 1,N
    XTEMP(I,1) = X(I,1)+DAFA*D(I,1)
10 CONTINUE
  CALL COST1 (1,N,FF,XTEMP,ANII,PNALTY,CLOSE,THRESH)
  G = (FF-F)/DAFA
  RETURN
END

```

```

GAF 10
GAF 20
GAF 30
GAF 40
GAF 50
GAF 60
GAF 70
GAF 80
GAF 90
GAF 100
GAF 110
GAF 120
GAF 130
GAF 140
GAF 150
GAF 160

```

```

SUBROUTINE ERROR (K,MESSAGE)
  DIMENSION MESSAGE(12), ITEXT(12)
  DATA ITEXT/12*" "
  DO 10 I = 1,K
    ITEXT(I) = MESSAGE(I)
  WRITE (6,9010) (ITEXT(I),I=1,K)
  STOP
9010 FORMAT (///,1X,"ERROR: ",12A10,/)
END

```

```

1P0 10
1P0 20
1P0 30
1P0 40
1P0 50
1P0 60
1P0 70
1P0 80
1P0 90
1P0 100

```

```

SUBROUTINE LINESCH (F,N,FNOW,XNOW,D,AF42,XTEMP,FSMALL,ANII,PNALTY,LIN
  1CLOSE,THRESH,NDEC)
  DIMENSION D(NDEC,1), XTEMP(NDEC,1), XNOW(NDEC,1)
C .....
C : CUBIC FIT BY INITIALLY LETTING AFA = 0 AND DAFA = 5.
C : .....
C
  DNORM = 0.
  DO 10 I = 1,N
    DNORM = DNORM+D(I,1)**2
10 CONTINUE
  DNORM = SQRT(DNORM)
  IF (DNORM.EQ.0.) CALL CHECK (1,N,FNOW,XNOW,D,AF40,FAFA0,GAFA0,AF41LIN

```

```

LIN 10
LIN 20
LIN 30
LIN 40
LIN 50
LIN 60
LIN 70
LIN 80
LIN 90
LIN 100
LIN 110
LIN 120
LIN 130
LIN 140
LIN 150
LIN 160

```

```

      S,FAFA1,GAFAL,U1,LU1,U2,NDEC)                                LIN 170
C .....                                                              LIN 180
C .....                                                              LIN 190
C .....                                                              LIN 200
C ..... NORMALIZE D .....                                          LIN 210
C .....                                                              LIN 220
C .....                                                              LIN 230
C .....                                                              LIN 240
C .....                                                              LIN 250
      DO 20 I = 1,N                                                LIN 260
        D(I,1) = D(I,1)/DNCRH                                       LIN 270
20  CONTINUE                                                       LIN 280
      DAFA = 5.                                                     LIN 290
      AFAO = 0.                                                     LIN 300
      FAFAC = FNQW                                                  LIN 310
      KK = 1                                                         LIN 320
      CALL GAFA (N,FAFAO,DAFA,GAFAL,XNQW,D,NDEC)                  LIN 330
      IF (GAFAL.EQ.0) CALL CHECK (1,N,FNQW,XNQW,D,FAFAO,FAFAO,GAFAL,FA1, LIN 340
      SFAFA1,GAFAL,U1,LU1,U2,NDEC)
      IF (-FAFAO/GAFAL.EQ.0) CALL ERROR (6,"-FAFAC/GAFAL .LE. 0., TPAJ LIN 350
      1. SET PROBABLY NEAR MINIMUM PT.")
      AFA1 = AMIN1(1000.,-FAFAC/GAFAL)                             LIN 360
30  GO 40 I = 1,N                                                  LIN 370
      XTEMP(I,1) = XNQW(I,1)+AFA1*D(I,1)                          LIN 380
40  CONTINUE                                                       LIN 390
      CALL COST1 (1,N,FAFA1,XTEMP,ANII,PNALTY,CLOSE,THRESH,NDEC)   LIN 400
      IF (KK.EQ.4) CALL ERROR (7,"PROBLEM IN LINESCH. CAN'T FIND LOWER LIN 410
      1CCST AFTER 3 CUBIC FITS")
      CALL GAFA (N,FAFA1,DAFA,GAFAL,XTEMP,D,NDEC)                  LIN 420
      IF (ABS((AFAO-AFA1)/AFA1).LT..001) CALL CHECK (2,N,FNQW,XNQW,D,FAFA1, LIN 430
      10,FAFAO,GAFAL,FAFA1,FAFA1,GAFAL,U1,U1,U2,NDEC)
      U1 = GAFAL+GAFAL-3.*(FAFAO-FAFA1)/(AFAC-AFA1)               LIN 440
      LU1 = U1*2-GAFAL+GAFAL                                       LIN 450
      IF (U1.LE.0.) U2 = 0.                                         LIN 460
      IF (LU1.GT.0.) U2 = SORT(LU1)                                 LIN 470
      AQ = AMIN1(AFAO,FAFA1)                                       LIN 480
      A1 = AMAX1(AFAO,FAFA1)                                       LIN 490
      GO = GAFAL                                                    LIN 500
      FO = FAFAO                                                    LIN 510
      G1 = GAFAL                                                    LIN 520
      F1 = FAFA1                                                    LIN 530
      IF (AFAO.LT.AFA1) GO TO 50                                    LIN 540
      GO = GAFAL                                                    LIN 550
      FO = FAFA1                                                    LIN 560
      G1 = GAFAL                                                    LIN 570
      F1 = FAFAO                                                    LIN 580
50  IF (G1-GO*2.*U2.EQ.0.) CALL CHECK (3,N,FNQW,XNQW,D,AQ,FO,GO,A1,F1, LIN 590
      10,G1,U1,U1,U2,NDEC)
      AFA2 = A1-(A1-AQ)*(G1+U2-U1)/(G1-GO*2.*U2)                  LIN 600
      DO 60 I = 1,N                                                LIN 610
        XTEMP(I,1) = XNQW(I,1)+AFA2*D(I,1)                        LIN 620
60  CONTINUE                                                       LIN 630
      CALL COST1 (0,N,FSPALL,XTEMP,ANII,PNALTY,CLOSE,THRESH,NDEC)   LIN 640
      WRITE (6,9010) AQ,A1,AFA2,DNCRH,FO,F1,FSPALL,GO,G1          LIN 650
      WRITE (6,9020) (D(I,1),I=1,N)                                LIN 660
      IF (FSPALL.GE.FNQW) GO TO 70                                  LIN 670
      RETURN                                                         LIN 680
70  AFA1 = AFA1/2.                                                  LIN 690
      KK = KK+1                                                      LIN 700
      GO TO 30                                                       LIN 710
C .....                                                              LIN 720
9010 FORMAT (15X,"AQ = ",1PE13.6,5X,"A1 = ",E13.6,5X,"AFA2 = ",E13.6,5X, LIN 730
      1,"DNCRH = ",E13.6,/,15X,"FO = ",E13.6,5X,"F1 = ",E13.6,5X,"FSPALL LIN 740
      1 = ",E13.6,/,15X,"GO = ",E13.6,5X,"G1 = ",E13.6,/)
9020 FORMAT (5X,"D = ",1C(1X,1PE9.2,7X))                          LIN 750

```



```

COMMON /DYNAMIC/ RADIUS,PWEIT1,PWEIT2,ZWEIT1,GHMIN,GHMAX,NSAMP DYN 50
CCPMCN /CORNER/ ANGLE(3,3),POSIT(3,20C,3),XCENR(3,3),YCENR(3,3) DYN 60
ZCENR(3,3),XT1(3,5),YT1(3,5),ZT1(3,5),XT2(3,5),YT2(3,5),ZT2(3,5) DYN 70
INPOSIT(3),CIS(3) DYN 80
REAL LENGTH DYN 90
LENGTH(A,B,C,D,E,F) = SORT((A-B)**2+(C-D)**2+(E-F)**2) DYN 100
PI = ATAN(1.)*4. DYN 110
C ..... DYN 120
C ..... DYN 130
C ..... DYN 140
C ..... DYN 150
C ..... DYN 160
C ..... DYN 170
C ..... DYN 180
DC 60 I = 1,NTRAJ DYN 190
DC 10 J = 2,NSEG DYN 200
YT2(I,J) = 0. DYN 210
YT1(I,J) = YT2(I,J) DYN 220
XT2(I,J) = YT1(I,J) DYN 230
YT1(I,J) = XT2(I,J) DYN 240
ZT1(I,J) = 0. DYN 250
ZT2(I,J) = 0. DYN 260
ANGLE(I,J-1) = XT1(I,J) DYN 270
YCENR(I,J-1) = ANGLE(I,J-1) DYN 280
XCENR(I,J-1) = YCENR(I,J-1) DYN 290
10 CONTINUE DYN 300
DIS(I) = 0. DYN 310
XT1(I,1) = XM(I,1) DYN 320
XT2(I,1) = XM(I,1) DYN 330
YT1(I,1) = YM(I,1) DYN 340
YT2(I,1) = YM(I,1) DYN 350
ZT1(I,1) = ZM(I,1) DYN 360
ZT2(I,1) = ZM(I,1) DYN 370
ISEG2 = ISEG1+1 DYN 380
DO 40 J = ISEG2,NSEG DYN 390
A = LENGTH(XM(I,J-1),XM(I,J),YM(I,J-1),YM(I,J),ZM(I,J-1),ZM(I,J)) DYN 400
J) DYN 410
B = LENGTH(XM(I,J),XM(I,J+1),YM(I,J),YM(I,J+1),ZM(I,J),ZM(I,J+1)) DYN 420
I) DYN 430
C = LENGTH(XM(I,J-1),XM(I,J+1),YM(I,J-1),YM(I,J+1),ZM(I,J-1),ZM(I,J+1)) DYN 440
J) DYN 450
C ..... DYN 460
C ..... DYN 470
C ..... DYN 480
C ..... DYN 490
C ..... DYN 500
C ..... DYN 510
C ..... DYN 520
IF (ABS(A+B-C).GT.1.) GO TO 20 DYN 530
SEG = LENGTH(XT2(I,J-1),XM(I,J),YT2(I,J-1),YM(I,J),ZT2(I,J-1),ZM(I,J)) DYN 540
ZM(I,J)) DYN 550
XT1(I,J) = XM(I,J) DYN 560
XT2(I,J) = XM(I,J) DYN 570
YT1(I,J) = YM(I,J) DYN 580
YT2(I,J) = YM(I,J) DYN 590
ZT1(I,J) = ZM(I,J) DYN 600
ZT2(I,J) = ZM(I,J) DYN 610
ZCENR(I,J-1) = 0. DYN 620
YCENR(I,J-1) = 0. DYN 630
XCENR(I,J-1) = YCENR(I,J-1) DYN 640
ALONG1 = 0. DYN 650
ANGLE(I,J-1) = C. DYN 660
GO TO 30 DYN 670
20 AXY = LENGTH(XM(I,J-1),XM(I,J),YM(I,J-1),YM(I,J),0.,0.) DYN 680

```

```

      BXY = LENGTH(XM(I,J),XM(I,J+1),YM(I,J),YM(I,J+1),0.,0.)
      CXY = LENGTH(XM(I,J-1),XM(I,J+1),YM(I,J-1),YM(I,J+1),0.,0.)
      BETA = AACOS(AXY,BXY,CXY)
      ALONG2 = RADIUS/TAN(BETA/2.)
      PNALTY = PNALTY+PWEIT1*(AMAX1(0.,(ALONG2-AXY)))*2+AMAX1(0.,
1  ALONG2-BXY))*2
      IF (J.NE.ISEG2) PNALTY = PNALTY+PWEIT1*(AMAX1(0.,(ALONG1+ALONG2-
1  2-AXY)))*2
      .....
      CALCULATE TANGENTIAL POINTS, CENTER OF RADIUS R AT J-TH CORNER
      .....
      XT1(I,J) = XM(I,J)+ALONG2*(XM(I,J-1)-YM(I,J))/AXY
      YT1(I,J) = YM(I,J)+ALONG2*(YM(I,J-1)-XM(I,J))/AXY
      XT2(I,J) = XM(I,J)+ALONG2*(YM(I,J+1)-XM(I,J))/BXY
      YT2(I,J) = YM(I,J)+ALONG2*(YM(I,J+1)-YM(I,J))/BXY
      ZT1(I,J) = XM(I,J)+ALONG2*(ZM(I,J-1)-ZM(I,J))/AXY
      ZT2(I,J) = ZM(I,J)+ALONG2*(ZM(I,J+1)-ZM(I,J))/BXY
      SEG = LENGTH(XT2(I,J-1),XT1(I,J),YT2(I,J-1),YT1(I,J),ZT2(I,J-1),
1  ZT1(I,J))
      DISM2 = ALONG2*COS(BETA/2.)
      DISC2 = RADIUS/SIN(BETA/2.)
      X = (XT1(I,J)+XT2(I,J))/2.
      Y = (YT1(I,J)+YT2(I,J))/2.
      YCENTR(I,J-1) = XM(I,J)+DISC2*(X-XM(I,J))/DISM2
      YCENTR(I,J-1) = YM(I,J)+DISC2*(Y-YM(I,J))/DISM2
      ALONG1 = ALONG2
      D = LENGTH(XT1(I,J),XT2(I,J),YT1(I,J),YT2(I,J),0.,0.)
      ANGLE(I,J-1) = AACOS(RADIUS,RADIUS,D)
      SEG = SEG+ANGLE(I,J-1)*RADIUS
3C  DIS(I) = DIS(I)+SEG
4C  CONTINUE
      DO 50 J = ISEG2,NSEG
      .....
      CALCULATE VERTICAL SLOPE OF EACH SEGMENT AND
      ADD PENALTY FOR EACH SEGMENT WITH VERTICAL SLOPE GREATER
      THAN GAMAX OR LESS THAN GAMIN DEGREES
      .....
      XYL = LENGTH(XM(I,J+1),XM(I,J),YM(I,J+1),YM(I,J),0.,0.)
      VSLOPE = (ZM(I,J+1)-ZM(I,J))/XYL
      PNALTY = PNALTY+ZWEIT1*(AMAX1(0.,VSLOPE-TAND(GMAXX)))*2+ZWEIT
1  1*(AMAX1(0.,TAND(GMINI)-VSLOPE))*2
50  CONTINUE
      AXY = LENGTH(XM(I,NSEG),XM(I,NSEG+1),YM(I,NSEG),YM(I,NSEG+1),0.,
1  0.)
      BXY = LENGTH(XM(I,NSEG+1),XM(I,NSEG+2),YM(I,NSEG+1),YM(I,NSEG+2),
1  0.,0.)
      CXY = LENGTH(XM(I,NSEG),XM(I,NSEG+2),YM(I,NSEG),YM(I,NSEG+2),0.,
1  0.)
      ANG = AACOS(AXY,BXY,CXY)
      PNALTY = PNALTY+PWEIT1*(AMAX1(0.,(4C0.+ALONG1-AXY)))*2+PWEIT2*(
1  AMAX1(0.,(2-3ECS-ANG)))*2
      DIS(I) = DIS(I)+(LENGTH(XT2(I,NSEG),XM(I,NSEG+1),YT2(I,NSEG),YM(I,
1  NSEG+1),ZT2(I,NSEG),ZM(I,NSEG+1))
      FATOFB = FATOFB+FATOFB*DIS(I)
6C  CONTINUE
      RETURN

```

END

DYN 1330

```

SUBROUTINE NII (ANII)
COMMON NTRAJ,NMAP,NSEG,XM(3,6),YM(3,6),ZM(3,6),ARRAY(576,9),SPOSIT,NCI
(3,200,3),XC(3),XF(3),YC(3),YF(3),ZC(3),ZF(3),NPOSITS(3),PHC(576),NCI
ISEG1,IPPOF,MAXIT,PENCRIT,AWEIT,TVICLA,ALWP,FATOF,FC,FATOFW
PEOPLE = 0.
ALWP = C.
DO 10 K = 1,NMAP
  IF (ARRAY(K,3).LT.1.) GO TO 10
  ALCN = ARRAY(K,4)
  W = 3.36E-6*10.**40.103*ALCN
  L = W/(0.2*10.**40.03*ALCN)+1.43E-4*10.**40.08*ALCN
  PEOPLE = PEOPLE+ARRAY(K,3)*W
  ALWP = ALWP+ARRAY(K,5)
10 CONTINUE
ANII = ALWP/PEOPLE*AWEIT
RETURN
END

```

NII	10
NII	20
NII	30
NII	40
NII	50
NII	60
NII	70
NII	80
NII	90
NII	100
NII	110
NII	120
NII	130
NII	140
NII	150
NII	160
NII	170
NII	180

```

SUBROUTINE NCISEL (THRESH)
COMMON NTRAJ,NMAP,NSEG,XM(3,6),YM(3,6),ZM(3,6),ARRAY(576,9),SPOSIT,NCI
(3,200,3),XC(3),XF(3),YC(3),YF(3),ZC(3),ZF(3),NPOSITS(3),PHC(576),NCI
ISEG1,IPPOF,MAXIT,PENCRIT,AWEIT,TVICLA,ALWP,FATOF,FC,FATOFW
COMMON /CORNER/ ANGLE(3,3),POSIT(3,200,3),XCEN(3,3),YCEN(3,3),NCI
ZCENTR(3,3),XT1(3,5),YT1(3,5),ZT1(3,5),XT2(3,5),YT2(3,5),ZT2(3,5),NCI
INPCISIT(3),DIS(3)
COMMON /DYNAMIC/ RADIUS,PWEIT1,PWEIT2,ZWEIT1,GMAHIN,GMAHAX,NSAMP
COMMON /THRESH/ ALMAX,MAXFLIT,TWEIT1,TWEIT2,NPLANE
COMMON /SAMPLE/ SAMPT,PERIOD,SAMPLTH,NAC(3,32),CNCIS(16,2)
DIMENSION ARCLTH(4), SEGLTH(5), AL(16)
REAL LENGTH
COORD(A,B,C,D) = (A/B)*(C-D)*D
LENGTH(A,B,C,D,E,F) = SORT((A-E)**2+(C-D)**2+(E-F)**2)
PI = ATAN(1.)*4.
DO 100 I = 1,NTRAJ
  XT2(I,NSEG+1)=XM(I,NSEG+1)
  YT1(I,NSEG+1)=YT2(I,NSEG+1)
  YT2(I,NSEG+1)=YM(I,NSEG+1)
  YT1(I,NSEG+1)=YT2(I,NSEG+1)
  ZT2(I,NSEG+1)=ZM(I,NSEG+1)
  ZT1(I,NSEG+1)=ZT2(I,NSEG+1)
  OLZPM = (ZC(I)-ZF(I))/DIS(I)
  PCSIT(I,1,1) = XM(I,ISEG1)
  PCSIT(I,1,2) = YM(I,ISEG1)
  POSIT(I,1,3) = ZM(I,ISEG1)
  NSEG1 = NSEG-1
  DO 10 J = ISEG1,NSEG1
    SEGLTH(J) = LENGTH(XT2(I,J),XT1(I,J+1),YT2(I,J),YT1(I,J+1),ZT2(I,J),ZT1(I,J+1))
    ARCLTH(J) = RADIUS*ANGLE(I,J)
10 CONTINUE
  SEGLTH(NSEG) = LENGTH(XT2(I,NSEG),XM(I,NSEG+1),YT2(I,NSEG),YM(I,NSEG+1),ZT2(I,NSEG),ZM(I,NSEG+1))
  II = 1
  DO 70 J = ISEG1,NSEG1
    IF (SEGLTH(J).LT.SAMPLTH) GO TO 30

```

NCI	10
NCI	20
NCI	30
NCI	40
NCI	50
NCI	60
NCI	70
NCI	80
NCI	90
NCI	100
NCI	110
NCI	120
NCI	130
NCI	140
NCI	150
NCI	160
NCI	170
NCI	180
NCI	190
NCI	200
NCI	210
NCI	220
NCI	230
NCI	240
NCI	250
NCI	260
NCI	270
NCI	280
NCI	290
NCI	300
NCI	310
NCI	320
NCI	330
NCI	340
NCI	350
NCI	360
NCI	370

```

      II = II+1
      POSIT(I,II,1) = COORD(SAMLTN,SEGLTH(J),XT1(I,J+1),POSIT(I,II-1,1))
      POSIT(I,II,2) = COORD(SAMLTN,SEGLTH(J),YT1(I,J+1),POSIT(I,II-1,2))
      POSIT(I,II,3) = POSIT(I,II-1,3)-DLZPM*SAMLTN
      IF (IPROF.EQ.1) POSIT(I,II,3) = COORD(SAMLTN,SEGLTH(J),ZT1(I,J+1),POSIT(I,II-1,3))
      SEGLTH(J) = SEGLTH(J)-SAMLTN
      GO TO 20
30  IF (SEGLTH(J)+ARCLTH(J).GE.SAMLTN) GO TO 40
      PIECE = SAMLTN-ARCLTH(J)-SEGLTH(J)
      II = II+1
      POSIT(I,II,1) = COORD(PIECE,SEGLTH(J+1),XT1(I,J+2),XT2(I,J+1))
      POSIT(I,II,2) = COORD(PIECE,SEGLTH(J+1),YT1(I,J+2),YT2(I,J+1))
      POSIT(I,II,3) = POSIT(I,II-1,3)-DLZPM*SAMLTN
      IF (IPROF.EQ.1) POSIT(I,II,3) = COORD(PIECE,SEGLTH(J+1),ZT1(I,J+2),ZT2(I,J+1))
      SEGLTH(J+1) = SEGLTH(J+1)-PIECE
      GO TO 70
40  PIECE = SAMLTN-SEGLTH(J)
      GAMMA = PIECE/RADIUS
      ARCLTH(J) = ARCLTH(J)-PIECE
      II = II+1
      ALFA1 = ATAN2((YT1(I,J+1)-YCENF(I,J)),(XT1(I,J+1)-XCENF(I,J)))
      IPLUS = 1
      AA = ALFA1+ANGLE(I,J)/4.
      X22 = XCENF(I,J)+RADIUS*COS(AA)
      Y22 = YCENF(I,J)+RADIUS*SIN(AA)
      ADIS = SQRT((X22-XM(I,J+1))**2+(Y22-YM(I,J+1))**2)
      BDIS = SQRT((XT1(I,J+1)-XM(I,J+1))**2+(YT1(I,J+1)-YM(I,J+1))**2)
      IF (BDIS.LT.ADIS) IPLUS = -1
      AA = ALFA1+IPLUS*GAMMA
      POSIT(I,II,1) = XCENF(I,J)+RADIUS*COS(AA)
      POSIT(I,II,2) = YCENF(I,J)+RADIUS*SIN(AA)
      POSIT(I,II,3) = POSIT(I,II-1,3)-DLZPM*SAMLTN
      IF (IPROF.EQ.1) POSIT(I,II,3) = COORD(PIECE,ARCLTH(J)+PIECE,ZT1(I,J+1),ZT2(I,J+1))
50  IF (ARCLTH(J).LT.SAMLTN) GO TO 60
      AA = AA+IPLUS*SAMLTN/RADIUS
      ARCLTH(J) = ARCLTH(J)-SAMLTN
      II = II+1
      POSIT(I,II,1) = XCENF(I,J)+RADIUS*COS(AA)
      POSIT(I,II,2) = YCENF(I,J)+RADIUS*SIN(AA)
      POSIT(I,II,3) = POSIT(I,II-1,3)-DLZPM*SAMLTN
      IF (IPROF.EQ.1) POSIT(I,II,3) = COORD(SAMLTN,ARCLTH(J)+SAMLTN,ZT1(I,J+1),ZT2(I,J+1))
      GO TO 50
60  PIECE = SAMLTN-ARCLTH(J)
      II = II+1
      POSIT(I,II,2) = COORD(PIECE,SEGLTH(J+1),YT1(I,J+2),YT2(I,J+1))
      POSIT(I,II,1) = COORD(PIECE,SEGLTH(J+1),XT1(I,J+2),XT2(I,J+1))
      POSIT(I,II,3) = POSIT(I,II-1,3)-DLZPM*SAMLTN
      IF (IPROF.EQ.1) POSIT(I,II,3) = COORD(PIECE,SEGLTH(J+1),ZT1(I,J+2),ZT2(I,J+1))
      SEGLTH(J+1) = SEGLTH(J+1)-PIECE
70  CONTINUE
80  IF (SEGLTH(NSEG).LT.SAMLTN) GO TO 90
      II = II+1
      POSIT(I,II,1) = COORD(SAMLTN,SEGLTH(NSEG),YT1(I,NSEG+1),POSIT(I,II-1,1))
      POSIT(I,II,2) = COORD(SAMLTN,SEGLTH(NSEG),YT1(I,NSEG+1),POSIT(I,II-1,2))

```

ORIGINAL PAGE IS
OF POOR QUALITY

```

PROGRAM MAIN (INPUT,OUTPUT,TAPES=INPUT,TAPE7,TAPE8,TAPE9,TAPE10) MAI 1C
..... MAI 2C
C ..... MAI 3C
C ..... MAI 4C
C ..... MAI 5C
C ..... MAI 6C
C ..... MAI 7C
C ..... MAI 8C
C ..... MAI 9C
C ..... MAI 10C
C ..... MAI 11C
C ..... MAI 12C
C ..... MAI 13C
C ..... MAI 14C
C ..... MAI 15C
C ..... MAI 16C
C ..... MAI 17C
C ..... MAI 18C
C ..... MAI 19C
C ..... MAI 20C
C ..... MAI 21C
C ..... MAI 22C
C ..... MAI 23C
C ..... MAI 24C
C ..... MAI 25C
C ..... MAI 26C
C ..... MAI 27C
C ..... MAI 28C
C ..... MAI 29C
C ..... MAI 30C
C ..... MAI 31C
C ..... MAI 32C
C ..... MAI 33C
C ..... MAI 34C
C ..... MAI 35C
C ..... MAI 36C
C ..... MAI 37C
C ..... MAI 38C
C ..... MAI 39C
C ..... MAI 40C
C ..... MAI 41C
C ..... MAI 42C
C ..... MAI 43C
C ..... MAI 44C
C ..... MAI 45C
C ..... MAI 46C
C ..... MAI 47C
C ..... MAI 48C
C ..... MAI 49C
C ..... MAI 50C
C ..... MAI 51C
C ..... MAI 52C
C ..... MAI 53C
C ..... MAI 54C
C ..... MAI 55C
C ..... MAI 56C
C ..... MAI 57C
C ..... MAI 58C
C ..... MAI 59C
C ..... MAI 60C
C ..... MAI 61C
C ..... MAI 62C
C ..... MAI 63C
C ..... MAI 64C

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***** C Y C L I C *****
 MINIMAL AIRCRAFT NOISE IMPACT PROGRAM
 WITH CYCLIC TRAJECTORY OPTIMIZATION
 AUTHORS: P. MELTON AND P. CHANG
 UNIVERSITY OF VIRGINIA
 CHARLOTTESVILLE, VA
 REVISED: MAY 26, 1981

THIS PROGRAM USES NONLINEAR OPTIMIZATION TECHNIQUES TO
 FIND THE SET OF AIRCRAFT TRAJECTORIES WITH THE MINIMUM
 NOISE IMPACT INDEX (NII) FOR A SPECIFIED MIX OF AIRCRAFT
 AND FOR THE PARTICULAR POPULATION DISTRIBUTION INPUT.
 THE EQUIVALENT NOISE CONCEPT IS USED, ALONG WITH AN APPROACH
 THAT PERFORMS THE OPTIMIZATION ON ONE TRAJECTORY AT A TIME

THE SUBROUTINES USED ARE:

JPARAMS (NMACFTN PACKAGE ON NASA LRC PUBLIC LIB)
 CCST,CCST1,AACCS,MONIT,RESULT,CPCSCVP,CNEWTON,MPLY,TRNSPDS,
 FCRAO,GAFA,ERRCP,LINESCH,CHECK,CORNER,DYNARIC,NII,NCISEL,
 PCVRY,MNORCEN,ADDR

EXTERNAL PCVRY
 INTEGER OPD
 LOGICAL ZERGLCB,NEWTFAJ,ZENABLE
 COMMON NTRAJ,NHAP,NSEC,XM(15,13),YM(15,13),ZF(15,13),SPCSIT(15,250)
 \$,3),XO(15),XF(15),YC(15),YF(15),ZO(15),ZF(15),NPCSITS(15),PHO(573)
 \$,ISEG1,IPPCF,MAXIT,PENCPIT,AWEIT,TVIGLA,ALWP,PEOPLE,NTRCPT,TMPLON
 \$,73),CPD(15),ITG,ICPT,IT,ZERGLCB,NEWTFAJ,ARRAY(573,10)
 COMMON /DYNAMIC/ RADIUS,PWEIT1,PWEIT2,ZWEIT1,GHMIN(15),GHMAX(15)
 \$,NSAPP
 COMMON /TRESH/ ALFAY,MAXFLIT,TWEIT1,TWEIT2,NPLANE
 COMMON /CROSS/ XBEGIN,XFINAL,YOIS,CWEIT1,CWEIT2
 COMMON /PRINT/ SXCENTR(15,12),SYCENTR(15,12),SXT1(15,12),SYT1(15,12)
 \$,SZT1(15,12),SXT2(15,12),SYT2(15,12),SZT2(15,12),SANGLE(15,12)
 COMMON /SAMPLE/ SAPPT,PERIOD,SAMPLTH,NAC(15,32),CNCIS(16,2)
 COMMON /ENABLE/ ZENABLE
 DIMENSION XNOW(30,1), DELTAX(30), LABEL(F)
 DIMENSION XPCRT(15), YPORT(15), JN(30)
 DIMENSION NDC30(15,2), NDC9SAM(15,2), NDC10(15,2), N707SAM(15,2),
 \$,N720(15,2), N72720C(15,2), N727SAM(15,2), N737100(15,2), N737SAM(15,2),
 \$,N747200(15,2), NL1011(15,2), NA300(15,2), NBAC111(15,2), NVMAI
 \$,C10(15,2), NCV990(15,2), NSST(15,2), NAMAC(32)
 EQUIVALENCE (NAC(1),NDC30(1)), (NAC(31),NDC9SAM(1)), (NAC(61),NDCMAI
 \$,10(1)), (NAC(91),N707SAM(1)), (NAC(121),N720(1)), (NAC(151),N72720C
 \$,10(1)), (NAC(181),N727SAM(1)), (NAC(211),N737100(1)), (NAC(241),N737SAM
 \$,7SAM(1)), (NAC(271),N747200(1)), (NAC(301),NL1011(1)), (NAC(331),NVMAI
 \$,A300(1)), (NAC(361),NBAC111(1)), (NAC(391),NCV990(1)), (NAC(421),NCHMAI
 \$,V990(1)), (NAC(451),NSST(1))
 DATA NAMAC/"DC-P-30", " ", " ", "DC-P W/SAM", "ENG", " ", "DCMAI
 \$,10-10", " ", " ", "707 W/SAM", "ENG", " ", "720", " ", "MAI
 \$, " ", "727-200", " ", " ", "727 W/SAM", "ENG", " ", "737-MAI
 \$,100/20", "0", " ", "737 W/SAM", "ENG", " ", "747-200", " ", "MAI

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      S      "N",NL1011      "N",N      "N",NA3CO      "N",N      "N",NAC-11PAI 650
      S1      "N",N      "N",NVC-1C      "N",N      "N",NVC-990      "N",N      PAI 660
      S      "N",NCONCORDE S"N",N51      "N",N      PAI 670
      DATA NAC/48000/      PAI 680
      NAMELIST /ACHIX/ NDC93C,NDC95AM,NDC10,N707SAM,N72C,N727200,N7275AMPAI 690
      S,N737100,N7375AM,N747200,NL1011,NA3CO,NBAC111,NVC10,NVC990,N55T      PAI 700
      NAMELIST /OPCEP/ CPD,MAXITG,ZENABLE      PAI 710
      C      PAI 720
      C      PAI 730
      C      PAI 740
      C      PAI 750
      C      PAI 760
      C      PAI 770
      C      PAI 780
      C      PAI 790
      C      PAI 800
      C      PAI 810
      C      PAI 820
      C      PAI 830
      C      PAI 840
      C      PAI 850
      C      PAI 860
      C      PAI 870
      C      PAI 880
      C      PAI 890
      C      PAI 900
      C      PAI 910
      C      PAI 920
      C      PAI 930
      C      PAI 940
      C      PAI 950
      C      PAI 960
      C      PAI 970
      C      PAI 980
      C      PAI 990
      C      PAI 1000
      C      PAI 1010
      C      PAI 1020
      C      PAI 1030
      C      PAI 1040
      C      PAI 1050
      C      PAI 1060
      C      PAI 1070
      C      PAI 1080
      C      PAI 1090
      C      PAI 1100
      C      PAI 1110
      C      PAI 1120
      C      PAI 1130
      C      PAI 1140
      C      PAI 1150
      C      PAI 1160
      C      PAI 1170
      C      PAI 1180
      C      PAI 1190
      C      PAI 1200
      C      PAI 1210
      C      PAI 1220
      C      PAI 1230
      C      PAI 1240
      C      PAI 1250
      C      PAI 1260
      C      PAI 1270
      C      PAI 1280

```

INPUT PARAMETERS:
 ORDER = NAMELIST: CPO(I) = TRAJ. NO. WITH ORDINAL I
 MAXITG = TOTAL NO. OF GLOBAL ITERATIONS
 ZENABLE = .TRUE., Z-COORDS OF CORNER PTS.
 ALLOWED TO VARY
 .FALSE., Z-COORDS HELD CONSTANT
 IENTRY = .FIXED FOR FIXED ENTRY PTS.
 .VARIABLE FOR VARIABLE ENTRY PTS.
 ICHINU = 0, INITIAL TRAJ. ARE STRAIGHT LINES
 .1, INITIAL TRAJ. ARE USER INPUT
 SAMPT = TIME IN SECONDS BETWEEN NOISE LEVEL SAMPLES
 PERIOD = TIME IN HRS. OVER WHICH THE SIMULATION OCCURS
 ACSPD = A/C SPEED IN METERS/SEC.
 NTPAJ = NUMBER OF FLIGHT TRAJECTORIES
 NMAP = NUMBER OF POPULATION POINTS ON MAP
 NSEG = NUMBER OF LINE SEGMENTS ON EACH TRAJECTORY
 MAXIT = MAXIMUM NO. OF ITERATIONS ALLOWED
 IF ONLY RESULT OF INITIAL CONDITION IS NEEDED, MAXIT = 0
 DELH = PERTURBATION (METERS) IN X,Y DIRECTIONS AT CORNER
 PCINTS FOR GRADIENT CALCULATION
 STOPCHG = STOP CRITERION FOR SUCCESSIVE COST CHANGE
 XC,YC,ZC,XF,YF,ZF = STARTING AND FINAL POINTS OF TRAJECTORIES
 XPORT,YPCRY = AIRPORT LOCATION (METERS)
 XP,YP,ZM (I,J) = COORDS OF J-TH CORNER PT. ON I-TH TRAJ.
 PWEIT1 = PENALTY WEIGHT ON DYN. CONSTR. ON TRAJ.
 PWEIT2 = PENALTY WEIGHT ON CONSTR. ON FINAL ANGLE
 RADIUS = MIN. ALLOWED TURNING RADIUS (METERS)
 XMIN,XMAX,YMIN,YMAX = BOUNDARIES OF AREA EXCLUDED
 FROM MAX. NOISE (THRESHOLD) CONSTR.
 ALMAX = MAX. ALLOWED A-LEVEL NOISE
 PATIC = FRACTION OF FLIGHTS ALLOWED TO VIOLATE MAX. A-LEVEL
 TWFIT1 = PENALTY WEIGHT ON MAX. NOISE AND NUMBER CONSTR.
 TWFIT2 = NOT USED NOW, BUT SOME VALUE MUST BE INPUT
 XBEGIN,XFINAL = BOUNDARIES ON X-AXIS WHERE SEPARATING
 CONSTR. IS IMPOSED
 YDIS = MIN. SEPARATING DISTANCE (Y-AXIS) BETWEEN TRAJ.
 CWEIT1,CWEIT2 = PENALTY WEIGHTS ON SEP. AND CROSSOVER
 CONSTR. RESPECTIVELY
 IPROF = 0, NO PROFILE OPTIMIZATION
 .1, PROFILE OPTIMIZATION OCCURS
 NOTE: IPROF IS ALWAYS SET = 1 IN THIS PROGRAM
 ZWFIT1 = PENALTY WEIGHT ON VERTICAL SLOPE CONSTR.
 GMAMAX = MAX. ALLOWED VERTICAL ANGLE ON EACH SEG.
 GMAMIN = MIN. ALLOWED VERTICAL ANGLE ON EACH SEG.
 PENCRIT = MIN. SIGNIFICANT DYN. PENALTY LEVEL
 AWEIT = SCALING ON AII
 ACHIX = NAMELIST: ACTYPE (I,1) = NO. OF ACTYPE ON TRAJ. I
 IN DAYTIME
 ACTYPE (I,2) = NO. OF ACTYPE ON TRAJ. I
 AT NIGHT
 GMAMIN (I) = MIN. FLIGHT PATH ANGLE (DEG) ON
 TRAJ. I
 GMAMAX (I) = MAX. FLIGHT PATH ANGLE (DEG) ON
 TRAJ. I


```

C : ..... MAY 1290
C ..... MAY 1300
C ..... MAY 1310
CALL KECQVR (RCVFY,778,0) MAY 1320
CALL DATE (DAT) MAY 1330
ENCODE (10,9010,IDAT) DAT MAY 1340
CALL REMARK (IDAT) MAY 1350
CALL JPARAMS (JN) MAY 1360
ENCODE (7,9020,JCBNAME) JN(1) MAY 1370
CALL REMARK (JCBNAME) MAY 1380
WRITE (6,9130) MAY 1390
WRITE (6,9030) MAY 1400
WRITE (6,9040) JCBNAME,IDAT MAY 1410
NPLANE = 0 MAY 1420
C ..... MAY 1430
C ..... MAY 1440
C ..... MAY 1450
C : INPUT ALL THE INFORMATION MAY 1460
C ..... MAY 1470
C ..... MAY 1480
C ..... MAY 1490
C ..... MAY 1500
READ (7,9150) (LABEL(I),I=1,8) MAY 1510
READ (7,9000) MAY 1520
READ (7,9050) IENTRY MAY 1530
IF (IENTRY.EQ."VARIABLE".AND.IENTRY.NE."FIXED") CALL ERROR (4,"ILLEGAL MAY 1540
  ENTRY POINT TYPE, CHECK DATA") MAY 1550
C ..... MAY 1560
C ..... MAY 1570
C : ISEG1 = 1 : FIXED ENTRY PTS., 1ST SEG. INCLUDED IN ALL CALCS MAY 1580
C : ISEG1 = 2 : VARIABLE ENTRY PTS., 1ST SEG. EXCLUDED MAY 1590
C ..... MAY 1600
C ..... MAY 1610
C ..... MAY 1620
C ..... MAY 1630
ISEG1 = 1 MAY 1640
IF (IENTRY.EQ."VARIABLE") ISEG1 = 2 MAY 1650
READ (7,*) ICNTINU,SAMPT,PERIOD,ACSPD MAY 1660
READ (7,*) NTRAJ,NTRCPT,NTRCNST,NMAP,NSEG,MAXIT MAY 1670
READ (7,*) DELH,STEPCHG MAY 1680
SAMPTH = SAMPT*ACSPD MAY 1690
DO 30 I = 1,NTRAJ MAY 1700
  READ (7,*) XC(CRD(I)),YC(CRD(I)),ZC(CRD(I)) MAY 1710
  READ (7,*) XF(CRD(I)),YF(CRD(I)),ZF(CRD(I)) MAY 1720
  READ (7,*) XPCPT(CPD(I)),YPCPT(CPD(I)) MAY 1730
C ..... MAY 1740
C ..... MAY 1750
C : CALCULATE INITIAL CORNEW POINTS MAY 1760
C ..... MAY 1770
C ..... MAY 1780
C ..... MAY 1790
C ..... MAY 1800
C ..... MAY 1810
C ..... MAY 1820
C ..... MAY 1830
C ..... MAY 1840
C ..... MAY 1850
C ..... MAY 1860
C ..... MAY 1870
C ..... MAY 1880
C ..... MAY 1890
C ..... MAY 1900
C ..... MAY 1910
C ..... MAY 1920
C ..... MAY 1930
C ..... MAY 1940
C ..... MAY 1950
C ..... MAY 1960
C ..... MAY 1970
C ..... MAY 1980
C ..... MAY 1990
C ..... MAY 2000

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      ZM(I,J) = ZO(I) + (FLOAT(J-1)/FLOAT(NSEG)) * (ZF(I) - ZO(I))
10  CONTINUE
      GC TC 30
20  READ (7,*) (XM(ORD(I),J),YM(ORD(I),J),ZM(ORD(I),J),J=2,NSEG)
30  CONTINUE
      DO 40 I = 1,NMAP
      DO 40 J = 4,6
      ARRAY(I,J) = 0.
40  CONTINUE
      READ (7,*) PWEIT1,PWEIT2,RADIUS
      READ (7,*) XMIN,XMAX,YMIN,YMAX
      READ (7,*) ALMAX,RATIO,TWEIT1,TWEIT2
      READ (7,*) XBEGIN,XFINAL,YDIS,CWEIT1,CWEIT2
      READ (7,*) IPRUF,ZWEIT1,PENCRIT,AWEIT
      READ (7,*) ((CNOIS(CRD(I),J),J=1,2),I=1,NTRAJ)
      READ (10,*) ((ARRAY(I,J),J=1,3),(ARRAY(I,J),J=6,9),R+O(I),I=1,NMAP
5)
      READ (7,ACMIX)
      READ (7,*) (GMAMIN(CRD(I)),I=1,NTRAJ)
      READ (7,*) (GMAMAX(CRD(I)),I=1,NTRAJ)
      IPRCF = 1
      DO 50 I = 1,NTRAJ
      DO 50 L = 1,32
      NPLANE = NPLANE + NAC(I,L)
50  CONTINUE
      DO 60 I = 1,NMAP
      ARRAY(I,10) = 1.
      IF (XMIN.LE.ARRAY(I,1).AND.ARRAY(I,1).LE.XMAX.AND.YMIN.LE.ARRAY(I,2).AND.ARRAY(I,2).LE.YMAX.AND.ARRAY(I,3).NE.0.) ARRAY(I,10) = 0.
60  CONTINUE
      MAXFLIT = IFIX(NPLANE*RATIO*0.5)
C .....
C . PRINT INFORMATION INPUT
C .
C .....
C
      WRITE (6,9160) (LABEL(I),I=1,R)
      WRITE (6,9140) (CRD(I),I=1,NTRAJ),MAXITG
      WRITE (6,9170) MAXIT,NTRAJ,NSEG,NMAP,DELH,STOPCHG
      WRITE (6,9180) NTRAJ,NTRAPT,NTPCNST
      WRITE (6,9060) SAMPT,PERIOD,ACSPD
      IF (IENTRY.EQ."FIXED") WRITE (6,9070)
      IF (IENTRY.EQ."VARIABLE") WRITE (6,9080)
      IF (IPROF.EQ.1) WRITE (6,9090)
      IF (IPROF.EQ.0) WRITE (6,9100)
      IF (ZENABLE) WRITE (6,9110)
      IF (.NOT.ZENABLE) WRITE (6,9120)
      WRITE (6,9190) ALMAX,MAXFLIT,XMIN,XMAX,YMIN,YMAX
      WRITE (6,9200) XBEGIN,XFINAL,YDIS
      WRITE (6,9210) RADIUS,PWEIT1,PWEIT2,ZWEIT1,TWEIT1,TWEIT2,CWEIT1,CWEIT2,PENCRIT,AWEIT
      WRITE (6,9220)
      WRITE (6,9230) (CRD(I),GMAMIN(CRD(I)),GMAMAX(CRD(I)),I=1,NTRAJ)
      DO 90 I = 1,NTRAJ
      WRITE (6,9240) CRD(I),XO(CRD(I)),YO(CRD(I)),ZO(CRD(I)),XF(CRD(I)),YF(CRD(I)),ZF(CRD(I))
90  CONTINUE
      NSEG1 = NSEG+1
      NSEG2 = NSEG+2
      DO 70 J = 1,NSEG2
      WRITE (6,9260) J,XM(ORD(I),J),YM(ORD(I),J),ZM(ORD(I),J)
70  CONTINUE

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9160 FORMAT (/,5X,"TOTAL NUMBER OF TRAJECTORIES = ",I2,/,5X,"NUMBER OF MAI 321C
      TRAJECTORIES TO BE OPTIMIZED = ",I2,/,5X,"NUMBER OF TRAJECTORIES PMAI 3220
      SELD CONSTANT = ",I2,/) MAI 3230
9190 FORMAT (5X,"NO BLOCK MAY RECEIVE ",F5.1," DB (OR HIGHER) A-LEVEL",MAI 3240
      " NOISE MORE THAN ",I4," TIMES A DAY",/,3X,". THIS CONSTRAINT ",MAI 3250
      "EXCLUDES BLOCKS WITH CENTROIDS INSIDE THE RANGES: ",F8.1," TO ",F8MAI 3260
      ".1," (X) AND ",F8.1," TO ",F8.1," (Y)",/) MAI 3270
9200 FORMAT (5X,"WITHIN THE AREA OF X-COORDINATES BETWEEN,F8.1,5H ANDMAI 3280
      ",F8.1,4H, THE SEPARATING DISTANCE BETWEEN TRAJECT,5HORIES,/,5X,1MAI 3290
      "SHOULD BE AT LEAST,F6.1,7H METERS,/) MAI 3300
9210 FCPMAT (/,8X,"OTHER-CONSTRAINT PARAMETERS",/,4X,"MINIMUM RADIUS",MAI 3310
      " OF CURVATURE (IN X-Y PLANE PROJECTION) = ",1PE9.2,/,4X,"DYNAMIC PMAT 3320
      "PENALTY WEIGHTS: PWEIT1 = ",E9.2," PWEIT2 = ",E9.2," ZWEIT1 = ",MAI 3330
      "E9.2,/,4X,"MAXIMUM NOISE LEVEL (THRESHOLD) ",PENALTY WEIGHTS: TMAI 3340
      "EIT1 = ",E9.2,"TWEIT2 = ",E9.2,/,4X,"SEPARATION PENALTY WEIGHTS: MAI 3350
      "CWEIT1 = ",E9.2," CWEIT2 = ",E9.2,/,4X," PENCRT = ",1PE9.2," ANEMAI 3360
      "IT = ",E9.2,/) MAI 3370
9220 FOPMAT (/,8X,"VERTICAL SLOPE CONSTRAINTS",/,8X,"TRAJ NO.",2X,"GHAMMAI 3380
      "IN GAMMAX",/) MAI 3390
9230 FCPMAT (11X,I2,5X,F5.1,3X,F5.1) MAI 3400
9240 FOPMAT (10X,15HFLIGHTPATH NO: ,I2,/,12X,13HINITIAL X,Y,Z,1X,23HCOCMAI 3410
      "RDINATES IN METERS: ,3(F9.1,3X),/,12X,11HFINAL X,Y,Z,1X,23HCOCPOINMAI 3420
      "ATES IN METERS: ,3(F9.1,3X),/,12X,14HINITIAL CORNER,12H POINT POSIMAT 3430
      "TIONS: ,/,14X,10HCCORNER NO.,6X,1HX,9X,1HY,9X,1HZ,/) MAI 3440
9250 FCPMAT (10X,"RUNWAY LOCATION: ",F8.1,10X,F8.1,/) MAI 3450
9260 FOPMAT (15X,I2,6X,F9.1,3X,F9.1,3X,F9.1) MAI 3460
9270 FCPMAT (15X,"A/C MIX",T45,"NC. OF A/C",/,T40,"DAYTIME",3X,"NIGHTTMAI 3470
      "IME",/) MAI 3480
9280 FCPMAT (T20,ZA10,T42,I3,T52,I3/) MAI 3490
      END MAI 3500

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SUBROUTINE COST (IGRAD,TOTAL,ANII,PALTY,CLOSE,THRESH) CCS 10
COMMON NTRAJ,NMAP,NSEG,XM(15,13),YM(15,13),ZM(15,13),SPOSIT(15,25)CCCS 20
$,3),XD(15),XF(15),YC(15),YF(15),ZD(15),ZF(15),NPOSITS(15),RHO(573)CCS 30
$,ISEG1,IPRCF,MAXIT,PENCRT,AWEIT,TVIOLA,ALWP,PEOPLE,NTROPT,TMPLDN(CCS 40
$573),ORDI(15),ITG,IQPT,IT,ZERGLDB,NEWTRAJ,ARRAY(573,10) CCS 50
INTEGER ORD CCS 60
LOGICAL ZERGLDB,NEWTRAJ CCS 70
COMMON /CORNER/ ANGLE(15,12),POSIT(15,250,3),XCENTR(15,12),YCENTR(CCS 80
$15,12),ZCENTR(15,12),XT1(15,12),YT1(15,12),ZT1(15,12),XT2(15,12),YCCS 90
$T2(15,12),ZT2(15,12),NPOSIT(15),DIS(15) CCS 100
COMMON /PPINT/ SXCENTR(15,12),SYCENTR(15,12),SXT1(15,12),SYT1(15,10CCS 110
$2),SZT1(15,12),SXT2(15,12),SYT2(15,12),SZT2(15,12),SANGLE(15,12) CCS 120
TOTAL=0. CCS 130
ANII=TOTAL CCS 140
CLOSE=ANII CCS 150
THRESH=CLOSE CCS 160
PALTY=THRESH CCS 170
CALL DYNARID (PALTY) CCS 180
C ..... CCS 190
C ..... CCS 200
C ..... CCS 210
C ..... SUBTP. NOISE EFFECTS OF TRAJ. BEING OPT. CCS 220
C ..... DURING PREV. CALL TO COST CCS 230
C ..... CCS 240
C ..... CCS 250
C ..... CCS 260
IF (.NOT. ZERGLDB) CALL ADDR (-1) CCS 270
IF (PALTY.LT.PENCRT) GO TO 10 CCS 280
TOTAL = PALTY CCS 290
IF (MAXIT.NE.0) RETURN CCS 300

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10 CALL NOISEL (THRESH)
CALL NII (ANII)
IF (NTRAJ.EQ.1) GO TO 20
CALL CRCSCVR (NPCSIT,POSIT,CLOSE)
20 TCTAL = ANII*THRESH*CLOSE*PNALTY
IF (IGRAD.EQ.1) PETLPN
DO 50 II = 1,NTRAJ
  I = ORD(II)
  SXT1(I,1) = XT1(I,1)
  SXT2(I,1) = XT2(I,1)
  SYT1(I,1) = YT1(I,1)
  SYT2(I,1) = YT2(I,1)
  SZT1(I,1) = ZT1(I,1)
  SZT2(I,1) = ZT2(I,1)
DO 30 J = 2,NSEG
  SXCENTR(I,J-1) = XCENTR(I,J-1)
  SYCENTR(I,J-1) = YCENTR(I,J-1)
  SXT1(I,J) = XT1(I,J)
  SYT1(I,J) = YT1(I,J)
  SXT2(I,J) = XT2(I,J)
  SYT2(I,J) = YT2(I,J)
  SZT1(I,J) = ZT1(I,J)
  SZT2(I,J) = ZT2(I,J)
  SANGLE(I,J-1) = ANGLE(I,J-1)
30 CONTINUE
NPCSITS(I) = NPCSIT(I)
IPOST = NPOSITS(I)
DO 40 J = 1,IPOST
  SPOSIT(I,J,1) = PCSIT(I,J,1)
  SPOSIT(I,J,2) = POSIT(I,J,2)
  SPCSIT(I,J,3) = PCSIT(I,J,3)
40 CONTINUE
50 CONTINUE
RETURN
END

```

```

COS 310
COS 320
COS 330
COS 340
COS 350
COS 360
COS 370
COS 380
COS 390
COS 400
COS 410
COS 420
COS 430
COS 440
COS 450
COS 460
COS 470
COS 480
COS 490
COS 500
COS 510
COS 520
COS 530
COS 540
COS 550
COS 560
COS 570
COS 580
COS 590
COS 600
COS 610
COS 620
COS 630
COS 640
COS 650

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SUBROUTINE COST1 (ICRAD,N,F,X,ANII,PNALTY,CLOSE,THRESH,NDEC)
COMMON NTRAJ,NMAP,NSEG,XM(15,13),YM(15,13),ZM(15,13),SPLSIT(15,25000)
S,3),XO(15),XF(15),YC(15),YF(15),ZO(15),ZF(15),NPOSITS(15),PHO(573)
S,ISEG1,IPOF,MAXIT,PENCRIT,AWEIT,TVICLA,ALWP,PEOPLE,NTRCPT,TMPLDN(CO
$573),ORD(15),ITG,ICPT,IT,ZERGLCB,NWTRAJ,APRAY(573,10)
INTEGER ORD
LOGICAL ZERGLCB,NWTRAJ
DIMENSION X(NDEC,1)
NVAR = 2
IF (IPOF.EQ.1) NVAR = 3
I = 1
NSEG1 = NSEG-1
DO 10 J = 1,NSEG1
  DO 10 K = 1,NVAR
    L = (I-1)*NVAR+NSEG1+(J-1)*NVAR+K
    IF (K.EQ.1) XM(ORD(I),J+1) = X(L,1)
    IF (K.EQ.2) YM(ORD(I),J+1) = X(L,1)
    IF (K.EQ.3) ZM(ORD(I),J+1) = X(L,1)
10 CONTINUE
CALL COST (IGRAD,F,ANII,PNALTY,CLOSE,THRESH)
RETURN
END

```

```

CO1 10
CO1 20
CO1 30
CO1 40
CO1 50
CO1 60
CO1 70
CO1 80
CO1 90
CO1 100
CO1 110
CO1 120
CO1 130
CO1 140
CO1 150
CO1 160
CO1 170
CO1 180
CO1 190
CO1 200
CO1 210
CO1 220

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FUNCTION AACOS (A,B,C)
X = (A**2+B**2-C**2)/(2.*A*B)
IF (ABS(X).LT.1.1) GO TO 10
WRITE (6,901C) X,A,B,C
STOP
10 IF (X.GT.1.) X = 1.
IF (X.LT.-1.) X = -1.
AACOS = ACOS(X)
RETURN
C
901C FORMAT (29H TROUBLE IN AACOS, X,A,B,C = ,4(1PE16.9,3X))
END
AAC 10
AAC 20
AAC 30
AAC 40
AAC 50
AAC 60
AAC 70
AAC 80
AAC 90
AAC 100
AAC 110
AAC 120

SUBROUTINE MCNIT (X,N,F,ANII,PNALTY,CLOSE,THRESH,NDEC) MCN 10
COMMON NTRAJ,NMAP,NSEG,XM(15,13),YM(15,13),ZM(15,13),SPOSIT(15,25) MCN 20
,3),XG(15),XF(15),YO(15),YF(15),ZC(15),ZF(15),NPOSITS(15),RHO(573) MCN 30
,ISEG1,IPPOF,MAXIT,PENCRIT,AWEIT,TVIOLA,ALWP,PEOPLE,NTRCPT,TMPLON(MCN 40
5573),CRD(15),ITG,ICPT,IT,ZERGLOB,NEWTRAJ,ARRAY(573,10) MCN 50
INTEGER ORD MCN 60
LOGICAL ZERGLOB,NEWTRAJ MCN 70
COMMON /PPINT/ SXCENTR(15,12),SYCENTP(15,12),SXT1(15,12),SYT1(15,1 MCN 80
2),SZT1(15,12),SXT2(15,12),SYT2(15,12),SZT2(15,12),SANGLE(15,12) MCN 90
DIMENSION X(NDEC,1) MCN 100
PI = ATAN(1.)*4. MCN 110
NVAR = 2 MCN 120
IF (IPRCF.EQ.1) NVAR = 3 MCN 130
I = 1 MCN 140
NSEG1 = NSEG-1 MCN 150
DO 10 J = 1,NSEG1 MCN 160
DO 10 K = 1,NVAR MCN 170
L = (I-1)*NVAR+NSEG1+(J-1)*NVAR+K MCN 180
IF (K.EQ.1) XM(CRD(I),J+1) = X(L,1) MCN 190
IF (K.EQ.2) YM(CRD(I),J+1) = X(L,1) MCN 200
IF (K.EQ.3) ZM(CRD(I),J+1) = X(L,1) MCN 210
10 CONTINUE MCN 220
WRITE (6,901C) IT,F,ANII,PNALTY,CLOSE,THRESH,TVIOLA,ALWP MCN 230
MAX = 1 MCN 240
IF (ZERGLOB) MAX = NTRAJ MCN 250
DO 30 I = 1,MAX MCN 260
WRITE (6,902C) CRD(I),XM(CRD(I),1),YM(CRD(I),1),ZM(CRD(I),1) MCN 270
DO 20 J = 2,NSEG MCN 280
WRITE (6,903C) XM(CRD(I),J),YM(CRD(I),J),ZM(CRD(I),J),SXCENTR(MCN 290
, CRD(I),J-1),SYCENTP(CRD(I),J-1),SXT1(CRD(I),J),SYT1(CRD(I),J),MCN 300
, SZT1(CRD(I),J),SXT2(CRD(I),J),SYT2(CRD(I),J),SZT2(CRD(I),J),SAMCN 310
, NGLE(CRD(I),J-1)*180./PI MCN 320
20 CONTINUE MCN 330
WRITE (6,904C) XM(CRD(I),NSEG+1),YM(CRD(I),NSEG+1),ZM(CRD(I),NSEG MCN 340
, G+1) MCN 350
NPCSITI = NPOSITS(CRD(I)) MCN 360
30 CONTINUE MCN 370
WRITE (8,905C) (IFI)(ARRAY(K,3)),ARRAY(K,4),K=1,NMAP) MCN 380
ZERGLOB = .FALSE. MCN 390
RETURN MCN 400
C
901C FORMAT (1H1,9X,11HITERATION: ,12,/,12X,12HTOTAL COST: ,1PE16.9,/,1MCN 420
2X,17HANNYANCE (NII): ,1PE16.9,/,12X,31HPENALTY ON DYNAMIC CONSTR MCN 430
1AINT: ,1PE16.9,/,12X,34HPENALTY ON SEPARATING CONSTRAINT: ,1PE16.9, MCN 440
,/,12X,2FHPENALTY ON THRESHOLD NOISE: ,1PE16.9,/,12X,"THRESH- VIOLAMCN 450
STION = ",E16.9,/,12X,"LWP = ",E16.9,/) MCN 460
902C FORMAT (12X,"FLIGHT PATH NO: ",I2,/,T20,"CORNER PTS.",T40,"CENTER MCN 470

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$OF CIRCLE",T74,"TANGENTIAL PTS.",T11C,"ANGLE (DEGREE)",/,14X,"(",FMCN 480
$7.0,"",F7.0,"",F7.0,"")" MON 490
9030 FORMAT (14X,"(",F7.0,"",F7.0,"",F7.0,"")",",",F7.0,"",F7.0,"")",2MCN 500
$((",F7.0,"",F7.0,"",F7.0,"")",3X,F6.1) MCN 510
9040 FORMAT (14X,"(",F7.0,"",F7.0,"",F7.0,"")" MCN 520
9050 FORMAT (5X,I8,5X,F9.5) MCN 530
END MON 540

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SUBROUTINE RESULT (X,N,F,ANII,PALTY,CLOSE,THRESH,NOEC) PFS 10
COMMON NTAJ,NMAP,NSEG,XM(15,13),YM(15,13),ZM(15,13),SPOSIT(15,25)RES 20
$,3),XO(15),XF(15),YC(15),YF(15),ZO(15),ZF(15),NPOSITS(15),RHO(573)RES 30
$,ISEC1,IPROF,MAXIT,FENCRT,ALWIT,TVIOLA,ALWP,PEOPLE,NTRDPT,TMPLDN(PFS 40
$573),ORD(15),ITG,ICPT,IT,ZERGLOR,NEWTRAJ,ARRAY(573,10) RES 50
INTEGER ORD RES 60
LOGICAL ZERGLOR,NEWTRAJ RES 70
COMMON /PRINT/ SXCENTR(15,12),SYCENTR(15,12),SXT1(15,12),SYT1(15,12)RES 80
$,SZT1(15,12),SXT2(15,12),SYT2(15,12),SZT2(15,12),SANGLE(15,12) RES 90
DIMENSION X(NDEC,1) RES 100
PI = ATAN(1.)*%. RES 110
NVAR = 2 RES 120
IF (IPROF.EQ.1) NVAR = 3 RES 130
I = 1 RES 140
NSEG1 = NSEG-1 RES 150
DO 10 J = 1,NSEG1 RES 160
  DO 10 K = 1,NVAR RES 170
    L = (I-1)*NVAR+NSEG1+(J-1)*NVAR+K RES 180
    IF (K.EQ.1) XM(CRD(I),J+1) = X(L,1) RES 190
    IF (K.EQ.2) YM(CRD(I),J+1) = X(L,1) RES 200
    IF (K.EQ.3) ZM(CRD(I),J+1) = X(L,1) RES 210
  10 CONTINUE RES 220
  WRITE (6,9010) IT,F,ANII,PALTY,CLOSE,THRESH,TVIOLA,ALWP RES 230
  DO 30 I = 1,NTRAJ RES 240
    WRITE (6,9020) ORD(I),XM(CRD(I),1),YM(CRD(I),1),ZM(CRD(I),1) RES 250
    DO 20 J = 2,NSEG RES 260
      WRITE (6,9030) XM(CRD(I),J),YM(CRD(I),J),ZM(CRD(I),J),SXCENTR(RES 270
      CRD(I),J-1),SYCENTR(CRD(I),J-1),SXT1(CRD(I),J),SYT1(CRD(I),J),RES 280
      SZT1(CRD(I),J),SXT2(CRD(I),J),SYT2(CRD(I),J),SZT2(CRD(I),J),SAPES RES 290
      $ NGLE(CRD(I),J-1)*180./PI RES 300
    20 CONTINUE RES 310
    WRITE (6,9040) XM(CRD(I),NSEG+1),YM(CRD(I),NSEG+1),ZM(CRD(I),NSEG+1)RES 320
    $ G+1) RES 330
    WRITE (6,9050) RES 340
    NPOSITI = NPOSITS(CRD(I)) RES 350
    WRITE (6,9060) ((SPOSITI(CRD(I),J,K),K=1,3),J=1,NPOSITI) RES 360
  30 CONTINUE RES 370
  RETURN RES 380
C RES 390
9010 FORMAT (1H,19X,11HITERATION: ,I2,/,12X,12HTOTAL COST: ,1PE16.9,/,1PES 400
$2X,17HANNOYANCE (NII): ,1PE16.9,/,12X,31HPENALTY ON DYNAMIC CCNSTRES 410
$AINT: ,1PE16.9,/,12X,24HPENALTY ON SEPARATING CONSTRAINT: ,1PE16.9PES 420
$,/,12X,28HPENALTY CH THRESHOLD NOISE: ,1PE16.9,/,12X,"THRESH VIOLARES 430
$TION = ",E16.9,/,12X,"LWP = ",E16.9,/) RES 440
9020 FORMAT (12X,"FLIGHT PATH NO: ",I2,/,T20,"CORNER PTS.",T40,"CENTER RES 450
$OF CIRCLE",T74,"TANGENTIAL PTS.",T11C,"ANGLE (DEGREE)",/,14X,"(",FRES 460
$7.0,"",F7.0,"",F7.0,"")" PFS 470
9030 FORMAT (14X,"(",F7.0,"",F7.0,"",F7.0,"")",",",F7.0,"",F7.0,"")",2PES 480
$((",F7.0,"",F7.0,"",F7.0,"")",3X,F6.1) RES 490
9040 FORMAT (14X,"(",F7.0,"",F7.0,"",F7.0,"")" PFS 500
9050 FORMAT (/,12X,36HFLIGHT TRAJECTORY X, Y, Z, IN METERS,/,19X,1HX,1PES 510
$4X,1HY,14X,1HZ,/) RES 520
9060 FORMAT (14X,1PE10.3,5X,E10.3,5X,E10.3,5X) PFS 530

```

END

PES 340

```

C      SUBROUTINE CROSCVR (I,A,P)                                CRD  10
C      .....                                                    CRD  20
C      .....                                                    CRD  30
C      .....                                                    CRD  40
C      . THIS IS A DUMMY ROUTINE, REPLACING THE ORIGINAL CROSOVR, . CRD  50
C      . WHICH DOES NOT WORK FOR TRAJS THAT DO NOT SHARE A COMMON . CRD  60
C      . SEGMENT OF THE X-AXIS . CRD  70
C      . . CRD  80
C      .....                                                    CRD  90
C      .....                                                    CRD 100
C      RETURN                                                    CRD 110
C      END                                                        CRD 120

```

```

C      SUBROUTINE ONEWTN (N,XNOW,DELTAX,STOPCHG,NDEC)           ONE  10
C      COMMON NTPAJ,NMAP,NSEG,XM(15,13),YM(15,13),ZM(15,13),SPOSIT(15,25) ONE  20
C      ,XO(15),XF(15),YC(15),YF(15),ZO(15),ZF(15),NPOSITS(15),RHO(573) ONE  30
C      ,ISEG1,IPRCF,MAXIT,PENCRIT,AWEIT,TVICLA,ALWP,PEOPLE,NTROPT,TMPLON(ONE  40
C      ,573),ORD(15),ITG,ICPT,IT,ZERFLOB,NEWTRAJ,ARRAY(573,10) ONE  50
C      .....                                                    ONE  60
C      .....                                                    ONE  70
C      . THIS OPTIMIZATION EMPLOYS SELF-SCALING, RESTARTING, . ONE  80
C      . QUASI-NEWTON METHOD. . ONE  90
C      . REFERENCE: D.G. LUENBERGER INTRO. TO LINEAR AND NONLINEAR . ONE 100
C      . PROGRAMMING; P.204, SEC.9.5 . ONE 110
C      . MAXIT: MAXIMUM NUMBER OF ITERATIONS ALLOWED . ONE 120
C      . STOPCHG: STOP IF PERCENTAGE CHANGE IN SUCCESSIVE COSTS IS . ONE 130
C      . LESS THAN THIS VALUE . ONE 140
C      . N: DIMENSION OF THE UNKNOWN X . ONE 150
C      . XNOW: PRESENT OR INITIAL VALUE OF UNKNOWN X . ONE 160
C      .....                                                    ONE 170
C      .....                                                    ONE 180
C      .....                                                    ONE 190
C      .....                                                    ONE 200
C      DIMENSION XNOW(NDEC,1), DELTAX(NDEC) ONE 210
C      DIMENSION GACW(30,1), GNEXT(30,1), P(30,1), Q(30,1), PO(1,1) ONE 220
C      DIMENSION OSO(1,1), PP(30,30), SQQS(30,30), S(30,30) ONE 230
C      DIMENSION XTEMP(30,1), PT(1,30), QS(1,30), SO(1,30), QT(1,30) ONE 240
C      DIMENSION SCC(30,30) ONE 250
C      DIMENSION D(30,1) ONE 260
C      IT = 0 ONE 270
C      CALL COST1 (0,N,FNCW,XNOW,ANII,PNALTY,CLOSE,THRESH,NDEC) ONE 280
C      IF (PNALTY.GE.PENCRIT) GO TO 10 ONE 290
C      CALL MONIT (XNOW,N,FNCW,ANII,PNALTY,CLOSE,THRESH,NDEC) ONE 300
C      IF (MAXIT.EQ.0) RETURN ONE 310
C      GO TO 20 ONE 320
C      10 WRITE (6,9C30) PNALTY,ANII ONE 330
C      CALL CORNRP (XNOW,N,NDEC) ONE 340
C      IF (MAXIT.EQ.0) PETFN ONE 350
C      .....                                                    ONE 360
C      .....                                                    ONE 370
C      .....                                                    ONE 380
C      . STEP 1: SET S = IDENTITY MATRIX AND CALCULATE GRADIENT G . ONE 390
C      . . ONE 400
C      .....                                                    ONE 410
C      .....                                                    ONE 420
C      20 DO 30 I = 1,N ONE 430

```



```

      * CC 30 J = 1,N
      S(I,J) = 0.
      IF (I.EQ.J) S(I,J) = 1.
30  CONTINUE
      CALL FGRAD (N,FNCW,XNOW,GNCW,DELTAX,NDEC)
C
C .....
C
C STEP 2: SET D = -SG
C
C .....
C
40  CALL MPLY (N,N,1,S,GNCW,D,NDEC,NDEC,1)
    DO 50 I = 1,N
      D(I,1) = -D(I,1)
50  CONTINUE
C
C .....
C
C STEP 3: MOST IMPORTANT!
C LINE SEARCH ALONG D TO FIND AFA THAT SATISFIES PQ>C
C IF COST FUNCTION IS ANALYTIC, SLIGHT MODIFICATION IS
C NEEDED (SIMPLIFICATION)(OPTIONAL)
C
C .....
C
      K = C
      K = K+1
40  CALL LINESCH (K,N,FNCW,XNOW,D,AFA2,XTEMP,FSMALL,ANII,FNALTY,CLOSE,
$THRESH,NDEC), RETURN$ (220)
      IF (IT+1.LT.MAXIT) GO TO 70
      IT = IT+1
      WRITE (6,9050)
      IF (PNALTY.GE.PENCRIT) GO TO 150
      IF (FSMALL.LE.FNCW) CALL MONIT (XNCW,N,FSMALL,ANII,PNALTY,CLOSE,THRESH,NDEC)
      RETURN
70  CALL FGRAD (N,FSMALL,XTEMP,GNEXT,DELTAX,NDEC)
      DO 80 I = 1,N
        D(I,1) = GNEXT(I,1)-GNCW(I,1)
        P(I,1) = AFA2*D(I,1)
80  CONTINUE
      CALL TRNSPOS (NDEC,1,P,PT)
      CALL MPLY (1,N,1,PT,C,PQ,1,NDEC,1)
      IF (PQ(1,1).GT.0.) GO TO 110
      IF (K.LE.1) GO TO 50
      WRITE (6,9010)
      RETURN
90  FNCW = FSMALL
      DO 100 I = 1,N
        XNOW(I,1) = XTEMP(I,1)
100  CONTINUE
      GO TO 60
110  DO 120 I = 1,N
        XNCW(I,1) = XTEMP(I,1)
120  CONTINUE
      IT = IT+1
      PRCNT = ABS((FSMALL-FNCW)/FNCW)
      IF (PRCNT.GE.STGPCHG) GO TO 140
      WRITE (6,9040) PRCNT,STGPCHG
      IF (PNALTY.GE.PENCRIT) GO TO 130
      CALL MONIT (XNOW,N,FSMALL,ANII,PNALTY,CLOSE,THRESH,NDEC)
      RETURN
      C
      RETURN

```

```

130 WRITE (6,9020) PNALTY
    CALL CORNER (XNOW,N,NDEC)
    RETURN
140 IF (IT.LT.MAXIT) GO TO 160
    WRITE (6,9050)
    IF (PNALTY.GE.PENCRIT) GO TO 150
    CALL MONIT (XNOW,N,FSMALL,ANII,PNALTY,CLOSE,THRESH,NDEC)
C
    RETURN
150 WRITE (6,9020) PNALTY
    CALL CORNER (XNOW,N,NDEC)
    RETURN
160 IF (PNALTY.GE.PENCRIT) GO TO 170
    CALL MONIT (XNOW,N,FSMALL,ANII,PNALTY,CLOSE,THRESH,NDEC)
    GO TO 180
170 WRITE (6,9030) PNALTY,ANII
    CALL CORNER (XNOW,N,NDEC)
180 CONTINUE
C
C .....
C STEP 4: IF "IT" IS INTEGER MULTIPLE OF N GO TO STEP 1;
C IF NOT, UPDATE S
C IS "IT" INTEGER MULTIPLE OF N?
C .....
C IF ((FLOAT(IT)/FLCAT(N)).NE.FLOAT(IT/N)) GO TO 190
C FNOW = FSMALL
C GO TO 20
C
C .....
C UPDATE MATRIX S; GO TO STEP 2
C .....
C
190 CALL MPLY (N,N,1,S,Q,SQ,NDEC,NDEC,1)
    CALL TRNSPOS (NDEC,1,Q,QT)
    CALL MPLY (N,1,N,SC,QT,SQ,NDEC,1,NDEC)
    CALL MPLY (N,N,N,SGO,S,SGO,NDEC,NDEC,NDEC)
    CALL MPLY (1,N,N,QT,S,QS,1,NDEC,NDEC)
    CALL MPLY (1,N,1,GS,G,GSQ,1,NDEC,1)
    CALL MPLY (N,1,N,P,PT,PP,NDEC,1,NDEC)
    DO 200 I = 1,N
        DO 200 J = 1,N
            S(I,J) = (S(I,J)-SCOS(I,J)/GSQ(1,1))*(PG(1,1)/GSQ(1,1))+PP(I,J)/
1 PC(1,1)
200 CONTINUE
    FNOW = FSMALL
    DO 210 I = 1,N
        GNCW(I,1) = GNEXT(I,1)
210 CONTINUE
    GO TO 40
220 WRITE (6,9060) CFD(1)
    RETURN
C
9010 FCPAT (/,10X,"AFTER 2 CALLS TO LINESCH, CANT FIND TRAJ. WITH",N LGNE 1650
    $OWER COST. RETURN TO PARPOBN.")
9020 FORMAT (5X,40HIN RESULT: DYNAMIC CONSTRAINTS VIOLATION,,10X,9HPENCNF 1670
    $ALTY =,1PE16.9)
9030 FCPAT (5X,30HIN MONIT: DYNAMIC CONSTRAINTS VIOLATION,,10X,9HPENAO 1690
    $LTY =,1PE16.9,,10X,"ANII = ",E16.9," (IF = 0., THEN ANII HAS", "NOONE 1700
    $T BEEN COMPUTED IN THIS ITERATION, DUE TO LARGE PNALTY")

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9040 FCFMAT (2X,"RELATIVE CHANGE IN SUCCESSIVE COSTS ",1PE10.3," LESS" ONE 1720
      1," THAN STOP CRITERION ",1PE10.3) ONE 1730
9050 FCFMAT (2X,29HMAXIMUM ITERATION SET REACHED) ONE 1740
9060 FCFMAT (//," ..... OPTIMIZATION FAILS FOR TRAJ. ",12," ON THIS", ONE 1750
      $GLOBAL ITERATION",//) ONE 1760
      END ONE 1770

```

```

      SUBROUTINE MPLY (L,M,N,A,B,C,LDEC,MDEC,NDEC) MPL 10
C ..... MPL 20
C ..... MPL 30
C : ..... MPL 40
C : CALCULATE MATRIX MULTIPLICATION C = AB ..... MPL 50
C : ..... MPL 60
C ..... MPL 70
C ..... MPL 80
C ..... MPL 90
      DIMENSION A(LDEC,MDEC), B(MDEC,NDEC), C(LDEC,NDEC) MPL 100
      DO 10 I = 1,L MPL 110
        DO 10 J = 1,N MPL 120
          C(I,J) = 0. MPL 130
          DO 10 K = 1,M MPL 140
            C(I,J) = C(I,J)+A(I,K)*B(K,J) MPL 150
          10 CONTINUE MPL 160
        RETURN MPL 170
      END

```

```

      SUBROUTINE TRANSPC (M,N,A,B) TPN 10
C ..... TPN 20
C ..... TPN 30
C : ..... TPN 40
C : TRANSPOSE OF MATRIX A IS RETURNED IN MATRIX B ..... TPN 50
C : ..... TPN 60
C ..... TPN 70
C ..... TPN 80
C ..... TPN 90
      DIMENSION A(M,N), B(N,M) TPN 100
      DO 10 I = 1,M TPN 110
        DO 10 J = 1,N TPN 120
          B(J,I) = A(I,J) TPN 130
        10 CONTINUE TPN 140
      RETURN TPN 150
      END

```

```

      SUBROUTINE FGRAD (N,F,X,G,DELTA,NDEC) FGR 10
C ..... FGR 20
C ..... FGR 30
C : ..... FGR 40
C : CALCULATE GRADIENT OF COST F WITH RESPECT TO UNKNOWN X ..... FGR 50
C : ..... FGR 60
C ..... FGR 70
C ..... FGR 80
C ..... FGR 90
      COMMON /ENABLE/ ZENABLE FGR 100
      LOGICAL ZENABLE FGR 110
      DIMENSION X(NDEC,1), G(NDEC,1), DELTA(NDEC,1) FGR 120
      DO 10 I = 1,N FGR 130
        G(I,1) = 0. FGR 140
        IF (.NOT.ZENABLE.AND.MOD(I,3).EQ.0) GO TO 10

```



```

C : NORMALIZE D . LIN 200
C : . LIN 210
C : . LIN 220
C : ..... LIN 230
C : ..... LIN 240
C : ..... LIN 250
      DO 20 I = 1,N
        D(I,1) = D(I,1)/DACRM . LIN 260
20  CONTINUE . LIN 270
      DAFA = 5. . LIN 280
      AFAO = 0. . LIN 290
      FAFAO = FNOW . LIN 300
      KK = 1 . LIN 310
      CALL GAFA (N,FAFAO,DAFA,GAFAO,XNOW,D,NDEC) . LIN 320
      IF (GAFAO.EQ.0) CALL CHECK (1,N,FNOW,XNOW,D,AFAO,FAFAO,GAFAC,AF1, . LIN 330
1FAFA1,GAF1,U1,LL1,L2,NDEC), RETURNS (90) . LIN 340
      AFA1 = AMIN1(1000.,-FAFAO/GAFAO) . LIN 350
      IF (-FAFAO/GAFAO.LT.0) AFA1 = AMAX1(-1000.,-FAFAO/GAFAO) . LIN 360
30  DO 40 I = 1,N . LIN 370
      XTEMP(I,1) = XNOW(I,1)+AFA1*D(I,1) . LIN 380
40  CONTINUE . LIN 390
      CALL COST1 (1,N,FAFA1,XTEMP,ANTI,PNALTY,CLOSE,THRESH,NDEC) . LIN 400
      IF (MK.LT.4) GO TO 50 . LIN 410
      WRITE (6,9040) . LIN 420
      RETURN . LIN 430
50  CALL GAFA (N,FAFA1,DAFA,GAF1,XTEMP,D,NDEC) . LIN 440
      IF (ABS((AFAC-AFA1)/AFA1).LT..001) CALL CHECK (2,N,FNOW,XNOW,D,AFA1, . LIN 450
10,FAFAO,GAFAO,AF1,FAFA1,GAF1,U1,UU1,U2,NDEC), RETURNS (90) . LIN 460
      U1 = GAFAC+GAF1=3.*(FAFAO-AFA1)/(AFAO-AFA1) . LIN 470
      UU1 = U1*2-GAFAC+GAF1 . LIN 480
      IF (UU1.LE.0.) U2 = 0. . LIN 490
      IF (UU1.GT.0.) U2 = SQRT(UU1) . LIN 500
      AO = AMIN1(AFAO,AFA1) . LIN 510
      A1 = AMAX1(AFAC,AFA1) . LIN 520
      GC = GAFAO . LIN 530
      FO = FAFAO . LIN 540
      G1 = GAF1 . LIN 550
      F1 = FAFA1 . LIN 560
      IF (AFAO.LT.AFA1) GC TO 60 . LIN 570
      GO = GAF1 . LIN 580
      FC = FAFA1 . LIN 590
      G1 = GAFAO . LIN 600
      F1 = FAFAO . LIN 610
60  IF (G1-GO+2.*U2.EQ.0.) CALL CHECK (3,N,FNOW,XNOW,D,AO,FO,GO,A1,F1, . LIN 620
1G1,U1,LL1,U2,NDEC), RETURNS (90) . LIN 630
      AFA2 = A1-(A1-AO)*(G1+L2-U1)/(G1-GC+2.*U2) . LIN 640
C : ..... LIN 650
C : ..... LIN 660
C : ..... LIN 670
C : RESTRICT MAXIMUM AFA2 TO 1.E4 . LIN 680
C : ..... LIN 690
C : ..... LIN 700
C : ..... LIN 710
      AFA2 = AMIN1(AFA2,1.E4) . LIN 720
      IF (AFA2.GT.0.) AFA2 = AMIN1(AFA2,1.E4) . LIN 730
      IF (AFA2.LT.0.) AFA2 = AMAX1(AFA2,-1.E4) . LIN 740
      DO 70 I = 1,N . LIN 750
      XTEMP(I,1) = XNOW(I,1)+AFA2*D(I,1) . LIN 760
70  CONTINUE . LIN 770
      WRITE (6,9010) AC,A1,AFA2,DNORM,FC,F1,GC,G1 . LIN 780
      WRITE (6,9030) (D(I,1),I=1,N) . LIN 790
      CALL COST1 (10,N,FSMALL,XTEMP,ANTI,PNALTY,CLOSE,THRESH,NDEC) . LIN 800
      WRITE (6,9020) FSMALL . LIN 810
      IF (FSMALL.GE.FNOW) GO TO AC . LIN 820
      RETURN . LIN 830

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80 BFA1 = BFA1/2.                                LIN 84C
   KK = KK+1                                       LIN 85C
   GO TO 30                                         LIN 86C
90 RETURN KABCRT                                   LIN 87C
C                                                    LIN 88C
901C FCFMAT (15X,"AO = ",1PE13.6,5X,"AI = ",E13.6,5X,"AFA2 = ",E13.6,5X)LIN 89C
   1,"DACRM = ",E13.6,/,15X,"FO = ",E13.6,5X,"FI = ",E13.6,/,15X,"GO = LIN 90C
   1"E13.6,5X,"GI = ",E13.6,/)                  LIN 91C
902C FCFMAT (15X,"FMSALL = ",1PE13.6,/)         LIN 92C
903C FCFMAT (5X,"D = ",1C(1X,1PE9.2,2X))         LIN 93C
904C FCFMAT (/,10X,"LINESCH CANT FIND LOWER COST AFTER 3 CUBIC FITS") LIN 94C
   END                                             LIN 95C

```

SLRPGUTINE CHECK (ICHECK,N,FNOW,XNOW,D,AO,FO,GO,A1,F1,G1,U1,UU1,U2CHE	10
1,NDEC), RETURNS (KABORT)	CHE 2C
DIMENSION XNCW(INDEC,1), D(INDEC,1)	CHE 3C
WRITE (6,10) FNOW,N,(1,XNCW(1,1),D(1,1),I=1,N)	CHE 40
IF (ICHECK.EQ.1) RETURN KABORT	CHE 50
WRITE (6,20) AO,FO,GO,A1,F1,G1	CHE 60
IF (ICHECK.EQ.2) RETURN KABORT	CHE 7C
WRITE (6,30) U1,UU1,U2	CHE 80
RETURN KABORT	CHE 90
	CHE 100
10 FORMAT (20X,44HTHIS IS SUBROUTINE CHECK WHICH GIVES ALL THE,35H INCH	110
1FCRPATION IN SLRROUTINE LINESCH,/,/,30X,7HFNOW = ,1PE16.9,/,/,37X,4CH	120
1HXNOW,17Y,1HO,/,/,=(25X,12,3X,1PE16.9,4Y,1PE16.9,/))	CHE 130
20 FORMAT (/,/,20X,MAC = ",1PE16.9,5X,NFO = ",E16.9,5X,MGO = ",E16.9,CH	140
12CX,"A1 = ",E16.9,5X,"F1 = ",E16.9,5X,"G1 = ",E16.9)	CHE 150
30 FORMAT (/,/,20X,5HU1 = ,1E16.9,5X,6HUU1 = ,1PE16.9,5X,5HU2 = ,1PE16CHE	160
1.9)	CHE 170
END	CHE 180

	SLBPCUTINE CORNER (X,N,NDEC)	CCP	10
	CCMCHN NTRAJ,NMAP,NSEG,XM(15,13),YM(15,13),ZM(15,13),SPOSIT(15,25)CCP	20	
	1,31,XG(15),XF(15),YF(15),ZG(15),ZF(15),NPOSITS(15),RMO(573)CCP	30	
	1,ISEG1,IPPCF,MAXIT,PENCFIT,AWEIT,TVIOLA,ALWP,PEOPLE,NTRCPT,TMPLON(CCP	40	
	5573),OPD(15),ITG,ICPT,IT,ZEPGLQD,NEWTRAJ,ARRAY(573,10)	50	
	INTEGER OMO	CCP	60
	LOGICAL ZEPGLQD,NEWTRAJ	CCP	70
	DIMENSION X(NDEC,1)	CCP	80
	NVAR = 2	CCP	90
	IF (IPPCF.EC.1) NVAR = 3	CCP	100
	DC 10 I = 1,NTRCPT	CCP	110
	NSEG1 = NSEG-1	CCP	120
	DC 10 J = 1,NSEG1	CCP	130
	DO 10 K = 1,NVAR	CCP	140
	L = (I-1)*NVAR+NSEG1+(J-1)*NVAR+K	CCP	150
	IF (K.EQ.1) XM(OPD(I),J+1) = X(L,1)	CCP	160
	IF (K.EQ.2) YM(OPD(I),J+1) = X(L,1)	CCP	170
	IF (K.EQ.3) ZM(OPD(I),J+1) = X(L,1)	CCP	180
10	CONTINUE	CCP	190
	DC 30 I = 1,NTRAJ	CCP	200
	WRITE (6,9010) I,XM(OPD(I),1),YM(OPD(I),1),ZM(OPD(I),1)	CCP	210
	DC 20 J = 2,NSEG	CCP	220
	WRITE (6,9020) XM(OPD(I),J),YM(OPD(I),J),ZM(OPD(I),J)	CCP	230
20	CONTINUE	CCP	240
	WRITE (6,9020) XM(OPD(I),NSEG+1),YM(OPD(I),NSEG+1),ZM(OPD(I),NSEG+1)	CCP	250
	1 C+1)	CCP	260


```

C . * DO THE TWO SEGMENTS FORM A STRAIGHT LINE? . DYN 540
C . . DYN 550
C . . . . . DYN 560
C . . . . . DYN 570
C . . . . . DYN 580
      IF (ABS(A+B-C).GT.1.) GO TO 20
      SEG = LENGTH(XT2(I,J-1),XM(I,J),YT2(I,J-1),YM(I,J),ZT2(I,J-1),DYN 590
1      ZM(I,J))
      YT1(I,J) = YM(I,J)
      XT2(I,J) = XM(I,J)
      YT1(I,J) = YM(I,J)
      YT2(I,J) = YM(I,J)
      ZT1(I,J) = ZM(I,J)
      ZT2(I,J) = ZM(I,J)
      ZCENTR(I,J-1) = C.
      YCENTR(I,J-1) = 0.
      XCENTR(I,J-1) = YCENTR(I,J-1)
      ALONG1 = 0.
      ANGLE(I,J-1) = 0.
      GO TO 30
20  AXI = LENGTH(XM(I,J-1),XM(I,J),YM(I,J-1),YM(I,J),0.,0.)
      BXY = LENGTH(XM(I,J),XM(I,J+1),YM(I,J),YM(I,J+1),0.,0.)
      CXY = LENGTH(XM(I,J-1),XM(I,J+1),YM(I,J-1),YM(I,J+1),0.,0.)
      BETA = AACOS(AXI,BXY,CXY)
      ALONG2 = RADIUS/TAN(BETA/2.)
      PNALTY = PNALTY+PWEIT1*((AMAX1(0.,(ALONG2-AXI)))*2+AMAX1(0.,(DYN 770
1      ALONG2-BXY)))*2
      IF (J.NE.ISEG2) PNALTY = PNALTY+PWEIT1*(AMAX1(0.,(ALONG1+ALONG2DYN 800
1      2-AXI)))*2
C . . . . . DYN 810
C . . . . . DYN 820
C . . . . . DYN 830
C . . . . . DYN 840
C . . . . . DYN 850
C . . . . . DYN 860
C . . . . . DYN 870
C . . . . . DYN 880
      XT1(I,J) = XM(I,J)+ALONG2*(XM(I,J-1)-XM(I,J))/AXI
      YT1(I,J) = YM(I,J)+ALONG2*(YM(I,J-1)-YM(I,J))/AXI
      XT2(I,J) = XM(I,J)+ALONG2*(XM(I,J+1)-XM(I,J))/BXY
      YT2(I,J) = YM(I,J)+ALONG2*(YM(I,J+1)-YM(I,J))/BXY
      ZT1(I,J) = ZM(I,J)+ALONG2*(ZM(I,J-1)-ZM(I,J))/AXI
      ZT2(I,J) = ZM(I,J)+ALONG2*(ZM(I,J+1)-ZM(I,J))/BXY
      SEG = LENGTH(XT2(I,J-1),XT1(I,J),YT2(I,J-1),YT1(I,J),ZT2(I,J-1)DYN 950
1      ),ZT1(I,J))
      DISM2 = ALONG2*COS(BETA/2.)
      DISC2 = RADIUS/SIN(BETA/2.)
      X = (XT1(I,J)+YT2(I,J))/2.
      Y = (YT1(I,J)+YT2(I,J))/2.
      XCENTR(I,J-1) = XM(I,J)+DISC2*(X-XM(I,J))/DISM2
      YCENTR(I,J-1) = YM(I,J)+DISC2*(Y-YM(I,J))/DISM2
      ALONG1 = ALONG2
      D = LENGTH(XT1(I,J),XT2(I,J),YT1(I,J),YT2(I,J),0.,0.)
      ANGLE(I,J-1) = AACOS(RADIUS,RADIUS,D)
      SEG = SEG+ANGLE(I,J-1)*RADIUS
30  DIS(I) = DIS(I)+SEG
40  CONTINUE
      DO 50 J = ISEG2,NSEG
C . . . . . DYN 1090
C . . . . . DYN 1100
C . . . . . DYN 1110
C . . . . . DYN 1120
C . . . . . DYN 1130
C . . . . . DYN 1140
C . . . . . DYN 1150
C . . . . . DYN 1160
C . . . . . DYN 1170

```

C-3


```

C
    XYL = LENGTH(XM(I,J+1),XM(I,J),YM(I,J+1),YM(I,J),0.,0.)      DYN 1190
    VSLCPE = (ZM(I,J+1)-ZM(I,J))/XYL                                DYN 1190
    PNALTY = PNALTY+ZWEIT1*(AMAX1(0.,VSLOPE-TAND(GHAMAX(I))))**2+ZDYN 1200
    WEIT1*(AMAX1(0.,TAND(GHAMIN(I))-VSLOPE))**2                      DYN 1210
    CONTINUE                                                         DYN 1220
5C    CONTINUE                                                         DYN 1230
    AXI = LENGTH(XM(I,NSEG),XM(I,NSEG+1),YM(I,NSEG),YM(I,NSEG+1),0.,0.) DYN 1240
    BXY = LENGTH(XM(I,NSEG+1),XM(I,NSEG+2),YM(I,NSEG+1),YM(I,NSEG+2),0.,0.) DYN 1250
    CXY = LENGTH(XM(I,NSEG),XM(I,NSEG+2),YM(I,NSEG),YM(I,NSEG+2),0.,0.) DYN 1260
    ANG = AACOS(AXI,BXY,CXY)                                          DYN 1270
    PNALTY = PNALTY+PWEIT1*(AMAX1(0.,(BXY*ALONG1-AXI))**2+PWEIT2*(ADYN 1280
    MAX1(0.,(Z-3805-ANG))**2                                          DYN 1290
    DIS(I) = DIS(I)+LENGTH(XT2(I,NSEG),XM(I,NSEG+1),YT2(I,NSEG),YM(I,NSEG+1), DYN 1300
    ZT2(I,NSEG),ZM(I,NSEG+1))                                          DYN 1310
5C    CONTINUE                                                         DYN 1320
    RETURN                                                            DYN 1330
END                                                                    DYN 1340

```

```

SUBROUTINE NII (ANII)
COMMON NTRAJ,NMAP,NSEG,XM(15,13),YM(15,13),ZM(15,13),SPOSIT(15,25)
1,3),XO(15),XF(15),YC(15),YF(15),ZO(15),ZF(15),NPOSITS(15),PHO(573)
1,ISEG1,IPROF,MAXIT,PENCRIT,AWEIT,TVICLA,ALWP,PEOPLE,NTRCPT,TMPLON
573),ORD(15),ITG,ICPT,IT,ZERGLCB,NEWTRAJ,ARRAY(573,10)
INTEGER ORD
LOGICAL ZERGLCB,NEWTRAJ
PEOPLE = 0.
ALWP = 0.
DO 10 K = 1,NMAP
    IF (ARRAY(K,3).LT.1.) GO TO 1C
    ALON = TMPLON(K)
    IF (ALON.LT.55.) GO TO 10
    W = 3.36E-6*10.**((0.103*ALON)
    W = W/(0.2*10.**((0.03*ALON)+1.43E-4*10.**((0.08*ALON)))
    PEOPLE = PEOPLE+ARRAY(K,3)
    ARRAY(K,5) = ARRAY(K,3)*W
    ALWP = ALWP+ARRAY(K,5)
10    CONTINUE
    ANII = ALWP/PEOPLE*AWEIT
    RETURN
END

```

```

SUBROUTINE NOISEL (THRESH)
COMMON NTRAJ,NMAP,NSEG,XM(15,13),YM(15,13),ZM(15,13),SPOSIT(15,25)
1,3),XO(15),XF(15),YC(15),YF(15),ZO(15),ZF(15),NPOSITS(15),PHO(573)
1,ISEG1,IPROF,MAXIT,PENCRIT,AWEIT,TVICLA,ALWP,PEOPLE,NTRCPT,TMPLON
573),ORD(15),ITG,ICPT,IT,ZERGLCB,NEWTRAJ,ARRAY(573,10)
INTEGER ORD
LOGICAL ZERGLCB,NEWTRAJ
COMMON /COMMON/ ANGLE(15,12),POSIT(15,25),XCEN(15,12),YCEN(15,12),
15,12),ZCEN(15,12),XT1(15,12),YT1(15,12),ZT1(15,12),XT2(15,12),
YT2(15,12),ZT2(15,12),NPOSIT(15),DIS(15)
COMMON /DYNAMIC/ RADIUS,PWEIT1,PWEIT2,ZWEIT1,GHAMIN(15),GHAMAX(15)
15,SAMP
COMMON /THRESH/ ALMAX,MAXFLIT,TWEIT1,TWEIT2,NPLANE
COMMON /SAMPLE/ SAMPT,PERIOD,SAMLEN,NAC(15,32),CNOIS(16,2)

```

```

      DIMENSION ARCLTH(8), SEGLTH(9)
      REAL LENGTH
      CCCPD(A,B,C,D) = (A/E)*(C-D)+D
      LENGTH(A,B,C,D,E,F) = SQRT((A-B)**2+(C-D)**2+(E-F)**2)
      PI = ATAN(1.)*4.
      MAX = 1
C
C .....
C : INCLUDE ALL TRAJS FOR THE ZEROth GLOBAL ITERATION
C :
C .....
C
      IF (ZEPGLOB) MAX = NTRAJ
      DO 100 KK = 1,MAX
      I = OPON(KK)
      XT2(I,NSEG+1)=XM(I,NSEG+1)
      YT1(I,NSEG+1)=YT2(I,NSEG+1)
      YT2(I,NSEG+1)=YM(I,NSEG+1)
      YT1(I,NSEG+1)=YT2(I,NSEG+1)
      ZT2(I,NSEG+1)=ZM(I,NSEG+1)
      ZT1(I,NSEG+1)=ZT2(I,NSEG+1)
      DLZPM = (ZQ(I)-ZF(I))/DIS(I)
      POSIT(I,1,1) = XM(I,ISEG1)
      POSIT(I,1,2) = YM(I,ISEG1)
      PCSIT(I,1,3) = ZM(I,ISEG1)
      NSEG1 = NSEG-1
      DO 10 J = ISEG1,NSEG1
      SEGLTH(J) = LENGTH(XT2(I,J),XT1(I,J+1),YT2(I,J),YT1(I,J+1),ZT2(I,J),ZT1(I,J+1))
      ARCLTH(J) = RADILS*ANGLE(I,J)
      CONTINUE
10  SEGLTH(NSEG) = LENGTH(XT2(I,NSEG),XM(I,NSEG+1),YT2(I,NSEG),YM(I,NSEG+1),ZT2(I,NSEG),ZM(I,NSEG+1))
      II = 1
      DO 70 J = ISEG1,NSEG1
      IF (SEGLTH(J).LT.SAMLTH) GO TO 30
      II = II+1
      IF (II.GT.250) WRITE (6,9010) I,J,II
      IF (II.GT.250) STOP
      POSIT(I,II,1) = CCCPD(SAMLTH,SEGLTH(J),XT1(I,J+1),PCSIT(I,II-1,1))
      POSIT(I,II,2) = CCCPD(SAMLTH,SEGLTH(J),YT1(I,J+1),POSIT(I,II-1,2))
      POSIT(I,II,3) = PCSIT(I,II-1,3)-DLZPM*SAMLTH
      IF (IPROF.EC.1) POSIT(I,II,3) = CCCPD(SAMLTH,SEGLTH(J),ZT1(I,J+1),POSIT(I,II-1,3))
      SEGLTH(J) = SEGLTH(J)-SAMLTH
      GO TO 20
30  IF (SEGLTH(J)+ARCLTH(J).GE.SAMLTH) GO TO 40
      PIECE = SAMLTH-ARCLTH(J)-SEGLTH(J)
      II = II+1
      IF (II.GT.250) WRITE (6,9010) I,J,II
      IF (II.GT.250) STOP
      PCSIT(I,II,1) = CCCPD(PIECE,SEGLTH(J+1),XT1(I,J+2),XT2(I,J+1))
      POSIT(I,II,2) = CCCPD(PIECE,SEGLTH(J+1),YT1(I,J+2),YT2(I,J+1))
      POSIT(I,II,3) = POSIT(I,II-1,3)-DLZPM*SAMLTH
      IF (IPROF.EC.1) POSIT(I,II,3) = CCCPD(PIECE,SEGLTH(J+1),ZT1(I,J+2),ZT2(I,J+1))
      SEGLTH(J+1) = SEGLTH(J+1)-PIECE
      GO TO 70
40  PIECE = SAMLTH-SEGLTH(J)
      GAMMA = PIECE/RADILS
      ARCLTH(J) = ARCLTH(J)-PIECE

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      II = II+1                                NOI 790
      IF (II.GT.250) WRITE (6,9010) I,J,II    NOI 800
      IF (II.GT.250) STOP                      NOI 810
      ALFA1 = ATAN2((YT1(I,J+1)-YCENTR(I,J)),(XT1(I,J+1)-XCENTR(I,J)) NOI 820
      )                                         NOI 830
      IPLUS = 1                                NOI 840
      AA = ALFA1*ANGLE(I,J)/4.                NOI 850
      X22 = XCENTR(I,J)+RADIUS*COS(AA)        NOI 860
      Y22 = YCENTR(I,J)+RADIUS*SIN(AA)        NOI 870
      ADIS = SQRT((Y22-YM(I,J+1))**2+(Y22-YM(I,J+1))**2) NOI 880
      BDIS = SQRT((XT1(I,J+1)-XM(I,J+1))**2+(YT1(I,J+1)-YM(I,J+1))**2) NOI 890
      2)                                       NOI 900
      IF (BDIS.LT.ADIS) IPLUS = -1            NOI 910
      AA = ALFA1+IPLUS*GAMMA                  NOI 920
      POSIT(I,II,1) = XCENTR(I,J)+RADIUS*COS(AA) NOI 930
      POSIT(I,II,2) = YCENTR(I,J)+RADIUS*SIN(AA) NOI 940
      POSIT(I,II,3) = POSIT(I,II-1,3)-DLZPM*SAMPLTH NOI 950
      IF (IPROF.EQ.1) POSIT(I,II,3) = COORD(PIECE,ARCLTH(J)+PIECE,ZTN) NOI 960
      2(I,J+1),ZT1(I,J+1))                  NOI 970
50    IF (ARCLTH(J).LT.SAMPLTH) GO TO 60      NOI 980
      AA = AA+IPLUS*SAMPLTH/RADIUS            NOI 990
      ARCLTH(J) = ARCLTH(J)-SAMPLTH           NOI 1000
      II = II+1                                NOI 1010
      IF (II.GT.250) WRITE (6,9010) I,J,II    NOI 1020
      IF (II.GT.250) STOP                      NOI 1030
      POSIT(I,II,1) = XCENTR(I,J)+RADIUS*COS(AA) NOI 1040
      POSIT(I,II,2) = YCENTR(I,J)+RADIUS*SIN(AA) NOI 1050
      POSIT(I,II,3) = POSIT(I,II-1,3)-DLZPM*SAMPLTH NOI 1060
      IF (IPROF.EQ.1) POSIT(I,II,3) = COORD(SAMPLTH,ARCLTH(J)+SAMPLTH,NOI 1070
      ZT2(I,J+1),POSIT(I,II-1,3))           NOI 1080
      GO TO 50                                NOI 1090
60    PIECE = SAMPLTH-ARCLTH(J)              NOI 1100
      II = II+1                                NOI 1110
      IF (II.GT.250) WRITE (6,9010) I,J,II    NOI 1120
      IF (II.GT.250) STOP                      NOI 1130
      POSIT(I,II,2) = CCCRD(PIECE,SEGLTH(J+1),YT1(I,J+2),YT2(I,J+1)) NOI 1140
      POSIT(I,II,1) = CCCRD(PIECE,SEGLTH(J+1),XT1(I,J+2),XT2(I,J+1)) NOI 1150
      POSIT(I,II,3) = POSIT(I,II-1,3)-DLZPM*SAMPLTH NOI 1160
      IF (IPROF.EQ.1) POSIT(I,II,3) = COORD(PIECE,SEGLTH(J+1),ZT1(I,NOI 1170
      J+2),ZT2(I,J+1))                      NOI 1180
      SEGLTH(J+1) = SEGLTH(J+1)-PIECE        NOI 1190
70    CONTINUE                                NOI 1200
PC    IF (SEGLTH(NSEG).LT.SAMPLTH) GO TO 90   NOI 1210
      II = II+1                                NOI 1220
      IF (II.GT.250) WRITE (6,9010) I,J,II    NOI 1230
      IF (II.GT.250) STOP                      NOI 1240
      POSIT(I,II,1) = CCCRD(SAMPLTH,SEGLTH(NSEG),XT1(I,NSEG+1),POSIT(I,NOI 1250
      II-1,1))                               NOI 1260
      POSIT(I,II,2) = CCCRD(SAMPLTH,SEGLTH(NSEG),YT1(I,NSEG+1),POSIT(I,NOI 1270
      II-1,2))                               NOI 1280
      POSIT(I,II,3) = POSIT(I,II-1,3)-DLZPM*SAMPLTH NOI 1290
      IF (IPROF.EQ.1) POSIT(I,II,3) = CCCRD(SAMPLTH,SEGLTH(NSEG),ZT1(I,NOI 1300
      NSEG+1),POSIT(I,II-1,3))              NOI 1310
      SEGLTH(NSEG) = SEGLTH(NSEG)-SAMPLTH    NOI 1320
      GO TO 60                                NOI 1330
90    APOSIT(I) = II                          NOI 1340
      IF (II.GT.250) CALL ERROR (8,"NO. OF TRAJ. SAMPLES .GT. 250, NOI 1350
      RE-DIMENSION APPAYS POSIT AND SPOSIT") NOI 1360
100   CONTINUE                                NOI 1370
      NSAMP = IFIX(PERIOD*3600/SAMPT)        NOI 1380
      TVIOLA = 0.                             NOI 1390
      DO 110 K = 1,NSAMP                      NOI 1400
      AMEN = 0.                               NOI 1410
      C                                         NOI 1420

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```

C ..... NCI 1430
C ..... NCI 1440
C : CAL. ENERGY CONTRIB. BY AMBIENT NOISE, NCI 1450
C : CURRENTLY AMBIENT LEVEL = 0. DB NCI 1460
C : NCI 1470
C ..... NCI 1480
C : NCI 1490
C : ARRAY(K,4) = NSAMP*10.** (AMBN/10.) NCI 1500
110 CONTINUE NCI 1510
C : NCI 1520
C ..... NCI 1530
C : NCI 1540
C : IF THIS IS THE ZEROETH GLOBAL ITERATION, INCLUDE NCI 1550
C : NOISE EFFECTS OF ALL TRAJS IN TMPLDN NCI 1560
C : ELSE, CALC. EFFECT OF TRAJ. BEING OPT. AND NCI 1570
C : STOPE IN ARRAY(*,4) NCI 1580
C : NCI 1590
C ..... NCI 1600
C : NCI 1610
C : IF (.NOT. ZERGLGB. OR. MAX.EQ.1) GO TO 160 NCI 1620
C : NCI 1630
C : DO 150 K = 1,NMAP NCI 1640
C : BLKVIOL = 0. NCI 1650
C : IF (ARRAY(K,3).LT.1) GO TO 140 NCI 1660
C : DO 130 II = 2,MAX NCI 1670
C : I = ORD(II) NCI 1680
C : NPCSITI = NPOSIT(I) NCI 1690
C : ENERGY = 0. NCI 1700
C : DO 120 J = 1,NPOSITI NCI 1710
C : RANGE = SQRT((PCOSIT(I,J,1)-ARRAY(K,1))**2+(PCOSIT(I,J,2)-ARRAY(K,2))**2+(PCOSIT(I,J,3))**2) NCI 1720
C : Y(K,2)**2+(PCOSIT(I,J,3))**2) NCI 1730
C : AL = CNOIS(I,1)-CNOIS(I,2)*ALOG10(RANGE) NCI 1740
C : ENERGY = ENERGY+10.** (AL/10.) NCI 1750
120 CONTINUE NCI 1760
C : TMPLDN(K) = ENERGY+TMPLDN(K) NCI 1770
130 CONTINUE NCI 1780
140 TMPLDN(K) = 10.*ALOG10(TMPLDN(K)/(PERIOD*3600./SAMP)) NCI 1790
C : BLKVIOL = AMAX1(C.,BLKVIOL-ALMAX*MAXFLIT) NCI 1800
C : IF (ARRAY(K,10).EQ.1.) TVICLA = TVICLA+BLKVIOL NCI 1810
150 CONTINUE NCI 1820
160 DO 160 K = 1,NMAP NCI 1830
C : IF (ARRAY(K,3).LT.1) GO TO 160 NCI 1840
C : I = ORD(1) NCI 1850
C : NPCSITI = NPOSIT(I) NCI 1860
C : ENERGY = 0. NCI 1870
C : DO 170 J = 1,NPOSITI NCI 1880
C : RANGE = SQRT((PCOSIT(I,J,1)-ARRAY(K,1))**2+(PCOSIT(I,J,2)-ARRAY(K,2))**2+(PCOSIT(I,J,3))**2) NCI 1890
C : Y(K,2)**2+(PCOSIT(I,J,3))**2) NCI 1900
C : AL = CNOIS(I,1)-CNOIS(I,2)*ALOG10(RANGE) NCI 1910
C : ENERGY = ENERGY+10.** (AL/10.) NCI 1920
170 CONTINUE NCI 1930
C : ARRAY(K,4) = 10.*ALOG10(ENERGY/(PERIOD*3600./SAMP)) NCI 1940
180 CONTINUE NCI 1950
C : NCI 1960
C : NCI 1970
C : ADD THE NOISE FROM TRAJ. BEING OPT. TO THE OTHER NOISE NCI 1980
C : NCI 1990
C ..... NCI 2000
C : NCI 2010
C : IF (ZERGLGB. OR. .NOT. NEWTRAJ) CALL ACOR (1) NCI 2020
C : THRESH = TWEIT1*TVICLA**2 NCI 2030
C : NEWTRAJ = .FALSE. NCI 2040
C : RETURN NCI 2050
C : NCI 2060

```

9010 FORMAT ('1 = ",I2," J = ",I2," II = ",I10,/')
END

NOI 2070
NOI 2080

SUBROUTINE RCVR (I,J,K)
STOP
END

PCV 10
PCV 20
PCV 30

```

SUBROUTINE RNDROBN (MAXITG,STOPCHG,N,XNOW,DELTA,NDEC)      RND 10
COMMON NTRAJ,NHAF,NSEG,XM(15,13),YM(15,13),ZM(15,13),SPOSIT(15,25) RND 20
,3),XCI(15),XF(15),YD(15),YF(15),ZD(15),ZF(15),NPOSITS(15),RHO(573) RND 30
,ISEG1,IPOF,MAXIT,PENCRIT,AWEIT,TVIOLA,ALWP,PEOPLE,NTRPT,TMPLON RND 40
$573),ORD(15),ITG,ICPT,IT,ZEPGLOB,NEWTRAJ,ARPA(573,10) RND 50
INTEGER DPO RND 60
LOGICAL ZEPGLOB,NEWTRAJ RND 70
DIMENSION XNOW(NDEC,1), DELTA(NDEC) RND 80
NSEG1 = NSEG-1 RND 90
DO 30 KK = 1,MAXITG RND 100
  ITG = KK RND 110
  DO 30 I = 1,NTRPT RND 120
    IOPT = I RND 130
    NEWTRAJ = .TRUE. RND 140
    WRITE (6,9010) (ORD(J),J=1,NTRAJ) RND 150
    CALL ONEWTN (N,XNOW,DELTA,STOPCHG,NDEC) RND 160
C ..... RND 170
C ..... RND 180
C ..... RND 190
C ..... TMPLON NOW CONTAINS NOISE EFFECTS FROM THE CONSTANT TRAJ, RND 200
C ..... SC ADD IN THE NOISE FROM THE TRAJ. BEING OPT. RND 210
C ..... RND 220
C ..... RND 230
C ..... RND 240
C ..... CALL ADDR (1) RND 250
C ..... IF (NTRCPT.EQ.1) RETURN RND 260
C ..... RND 270
C ..... RND 280
C ..... RND 290
C ..... CYCLE THE ORDER OF THE TRAJ RND 300
C ..... RND 310
C ..... RND 320
C ..... RND 330
C ..... RND 340
C ..... RND 350
C ..... RND 360
C ..... RND 370
10 CONTINUE RND 380
ORD(NTRCPT) = ORDTMP RND 390
DO 20 J = 1,NSEG1 RND 400
  DO 20 K = 1,3 RND 410
    L = (J-1)*3+K RND 420
    IF (K.EQ.1) YNCW(L,1) = XM(ORD(1),J+1) RND 430
    IF (K.EQ.2) YNCW(L,1) = YM(CPD(J),J+1) RND 440
    IF (K.EQ.3) YNCW(L,1) = ZM(ORD(1),J+1) RND 450
20 CONTINUE RND 460
30 CONTINUE RND 470
WRITE (6,9020) NTRCPT,MAXITG RND 480
RETURN RND 490
C ..... RND 500
9010 FORMAT (/,5X,"ORDER OF TRAJ: ",15(I2,2Y)) RND 510

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9020 FORMAT (/,5X,"RGUNC FGBIN PROCESS COMPLETE. EACH OF THE ",I2," TRPNC 520
          SAJS OPTIMIZED ",I2," TIME(S).") RND 530
          END RNC 540

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SUBROUTINE ADDR (ISIGN) ACD 10
COMMON NTRAJ,NMA-,NSEC,XM(15,13),YM(15,13),ZM(15,13),SPOSIT(15,25) ACD 20
1,3),XO(15),XP(15),YO(15),YF(15),ZC(15),ZF(15),NPOSITS(15),PHO(573) ACD 30
1,ISEG1,IPOF,MAXIT,PENGTIT,AWETT,TVIOLA,ALWP,PEOPLE,NTROPT,THPLDN( ACD 40
573),OFO(15),ITG,ICPT,IT,ZERGLOB,NEWTRAJ,ARRAY(573,10) ACD 50
INTEGER ORD ACD 60
LOGICAL ZERGLOB,NEWTRAJ ACD 70
CALL SECOND (TO) ACD 80
DO 10 K = 1,NMAP ACD 90
    E1 = 10.**(ARRAY(K,4)/10.) ACD 100
    E2 = 10.**(THPLDN(K)/10.) ACD 110
    ENEW = E2*ISIGN*E1 ACD 120
    THPLDN(K) = 10.*ALOG10(ENEW) ACD 130
10 CONTINUE ACD 140
RETURN ACD 150
END ACD 160

```

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