Robust maneuvering envelope estimation based on reachability analysis in an optimal control formulation*

Thomas Lombaerts¹ and Stefan Schuet² and Kevin Wheeler³ and Diana Acosta³ and John Kaneshige³

Abstract—This paper discusses an algorithm for estimating the safe maneuvering envelope of damaged aircraft. The algorithm performs a robust reachability analysis through an optimal control formulation while making use of time scale separation and taking into account uncertainties in the aerodynamic derivatives. Starting with an optimal control formulation, the optimization problem can be rewritten as a Hamilton-Jacobi-Bellman equation. This equation can be solved by level set methods. This approach has been applied on an aircraft example involving structural airframe damage. Monte Carlo validation tests have confirmed that this approach is successful in estimating the safe maneuvering envelope for damaged aircraft.

I. INTRODUCTION

All transportation systems need to focus on safety, but this applies especially to civil aviation. Therefore, in civil aviation, many developments focus on improving safety levels and reducing the risks that critical failures occur. In a recent study by CAST/ICAO, it can be observed that “loss of control in flight” (LOC-I) is the most frequent primary accident cause. This study is based on a statistical analysis of aircraft accidents between 2002 and 2011, and indicates that this category accounts for as much as 23% of all fatal aircraft accidents and involves most fatalities[1]. Benefit can be gained by developing technology which prevents these LOC-I accidents. From a flight dynamics point of view, with the technology and computing power available on this moment, it might have been possible to recover some of the aircraft in the accident category described above on the condition that non-conventional control strategies would have been applied. These non-conventional control strategies involve the so-called concept of fault tolerant flight control (FTFC), where the control system is capable of detecting and adapting to changes in the aircraft behaviour.

One FTFC strategy option is using a model based control routine. Previous research focused on a physical model approach[2]. In this setup, experiments have shown that not only a reconfiguring controller is needed, but also some form of flight envelope protection, which prevents the airplane from leaving the safe flight envelope and losing control in flight[2]. The main challenge in this context is determining the new bounds of the safe flight envelope after failure, which are then used by the envelope protection algorithm[3].

Alternative and complementary research approaches for the purpose of loss of control prevention and prediction are among others passive adaptive control[4], data-based predictive control[5] and real-time optimal envelope limit estimation[6].

Determination of the flight envelope has been done in the literature through various methods. The most straightforward methods include wind tunnel testing, flight test experiments and high-fidelity model-based computation of attainable equilibrium sets or achievable trim points[7], [8], [9]. More complex methods include formulating flight envelope estimation as a reachability problem and solving this with level set methods and Hamilton-Jacobi equations [10], [11], [13], [14], [15], possibly with time scale separation [16] or semi-Lagrangian level sets [18]. Alternative methods rely on linearization and region of attraction analysis [19], determining controllability/maneuverability limits in a quaternion-based control architecture[20] or robustness analysis for determination of reliable flight regimes [21]. An approach suggested by Boeing, as part of the NASA program Dynamic Flight Envelope Assessment and Prediction (DFEAP), uses Control-Centric Modeling, dynamic flexible structure and load models [22]. In the frequency domain, stability margins can be estimated in real time via nonparametric system identification [23]. More focused techniques inspired by flight dynamics exist as well, such as determining the minimum lateral control speed[24].

From the perspective of the physical approach, the preferred interpretation of the safe maneuvering envelope considers reachability from the trim envelope. The stable and controllable trim envelope is considered an a-priori safe set. The backwards reachable set is defined as the set of states from where the trim envelope can be reached. The forwards reachable set is the set of states which can be reached from the trim envelope. Then the safe maneuvering flight envelope is the cross section of the forwards and backwards reachable sets. This interpretation is illustrated in Fig. 1. In addition, the backwards reachable set is the survivable flight envelope. After an upset due to damage, turbulence, a wake encounter etc., one can bring the aircraft back to a safe trim condition if the current flight condition is situated inside the backwards reachable set.

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The aim is to perform a combined forward and backward reachability analysis from the trim envelope as efficiently as possible, for on-line implementations. Based on previous research[6], level set methods are an excellent candidate. Two of the major challenges are the computational load and how to cope with nonlinear systems with higher dimensions. In general, an increase in technology readiness level (TRL) makes it more difficult to control nonlinear systems with higher dimensions.

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The structure of time scale separation is analogous as applied by considering the principle of time scale separation[16]. The overview can be found in Fig. 2, which illustrates that a nine dimensional nonlinear problem is decoupled in three consecutive three dimensional optimization problems.

As done elsewhere in the literature[10], the characterization of this set can be done according to the principle of duality:

\[ \mathcal{R}(t, K) = (\mathcal{I}(t, K^c))^\circ \]

where \( \cdot^\circ \) stands for the complement of \( \cdot \). Through this principle, it can be characterized as an INFMIN problem[10].

The crux is to include the \( \Delta \)'s as disturbances in the optimization function, they oppose the optimization over \( \Delta \). It is assumed that \( u \) is trying to bring or keep the state in the set \( K \), whereas \( \Delta \) is trying to drive it out of \( K \). To ensure the game is well-posed, \( u \) is restricted to play non-anticipative strategies with respect to the unknown uncertainties \( \Delta \).

For the types of safety problems considered here, a set of initial states has to be established such that \( u \) can win the game, in other words the set \( \mathcal{R} \) can be characterized:

\[ \mathcal{R}_\Delta(t, K) = \{ x \in \mathbb{R}^n | \forall \Delta \in D, \exists u \in \mathcal{U}[0,T], \exists \tau \in [t, T], \phi(\tau, t, x, u(\cdot), \Delta) \in K \} \]

where \( \cdot^c \) stands for the complement of \( \cdot \). Through this principle, it can be characterized as an INFMIN problem[10].

The crux is to include the \( \Delta \)'s as disturbances in the optimization function, they oppose the optimization over \( u \). Consider a closed set \( K \), that can be written as the level set of a continuous function \( l : \mathbb{R}^n \rightarrow \mathbb{R} \), i.e. \( K = \{ x \in \mathbb{R}^n | l(x) \geq 0 \} \). As a consequence, the Invariance optimization formulation becomes[16]:

\[ \mathcal{I}(t, K) = \{ x \in \mathbb{R}^n | V_1(x, t) \geq 0 \} \]

with:

\[ V_1(x, t) = \inf_{u(\cdot) \in \mathcal{U}[0,T]} \sup_{\Delta \in D} \min \{ l(\phi(\tau, t, x, u(\cdot), \Delta)) \} \]

This can be reformulated into an Hamilton-Jacobi-Bellmann Partial Differential Equation (HJB PDE)[10], [17]:

\[ \frac{\partial V_1}{\partial t} (x, t) + \min_{\tau \in [t, T]} \left\{ \inf_{u(\cdot) \in \mathcal{U}[0,T]} \sup_{\Delta \in D} \frac{\partial V_1}{\partial x}(x, t) f(x, u, \Delta) \right\} = 0 \]

where \( V_1(x, T) = l(x) \) holds for backward integration and \( V_1(x, t) = l(x) \) applies to forward integration. These HJB PDE's can be solved by level sets, for which a toolbox is available in Matlab®[11].
III. APPROACH
The approach to calculate the safe maneuvering envelope after damage is based on the following steps:

- Identify the updated aircraft parameters after damage (not discussed here, see [12]). This concerns primarily estimating new post-damage values for the aerodynamic derivatives such as $C_{L_{0}}, C_{L_{u}}, C_{D_{0},}, C_{D_{a}}, C_{D_{a_{2}}}$, and $C_{Y_{p}}$ together with their uncertainty bounds.
- Calculate the post-damage trim envelope based on the updated aircraft parameters (not discussed here, see [12]).
- Based on the previous step, define reference trim boundaries for airspeed $V$, flight path angle $\gamma$ and bank angle $\phi$ as well as grid step size $\Delta V$, $\Delta \gamma$ and $\Delta \phi$.
- Define an implicit function accordingly over $V$ and $\gamma$. This needs to be done for every value of $\phi$ in case speed and flight path angle are bank angle dependent, i.e. $V = V(\phi)$ and $\gamma = \gamma(\phi)$.
- The Level sets method toolbox[11] relies on the Hamiltonian in Eq. (6) as a gradient to evolve the implicit function and thus reference boundaries over time.
- The cost function $V_{I}$ becomes a three dimensional functions, where cross sections reflect the situation for a specific time instant $t_{0}$.
- A dissipation function is needed to guarantee numerical stability during these calculations. As a consequence, slightly conservative results are obtained for the boundaries, but analysis has shown that this dissipation has a minor effect on the results. In this specific context, the chosen dissipation function is a Lax Friedrichs dissipation function[11].

IV. APPLICATION EXAMPLE
To illustrate how the envelope estimating algorithm works, a nonlinear 3D aircraft example is considered. At this point, only the slow dynamics as specified in Fig. 2 are considered. Future work will extend to the faster dynamics. The data used in this example are based on the RCAM (Research Civil Aircraft Model) simulation model[25]. The acting forces on the aircraft are illustrated in Fig. 3 for a symmetric flight condition.

$$
\begin{align*}
F_{Ax} - W \sin \gamma &= m \dot{V} \\
F_{Ay} \cos \phi + F_{Az} \sin \phi + W \cos \gamma &= -mV \dot{\gamma}
\end{align*}
$$

(7) (8)

Where the aerodynamic force components can be simplified assuming small aerodynamic angles $\alpha$ and $\beta$:

$$
\begin{align*}
F_{Ax} &= T \cos \beta \cos \alpha - D(V, \alpha) \approx T - D(V, \alpha) \\
F_{Ay} &= -T \sin \alpha - L(V, \alpha) \approx -L(V, \alpha) \\
F_{Az} &= -T \sin \beta \cos \alpha + Y_{aero}(V, \beta) \approx Y_{aero}(V, \beta)
\end{align*}
$$

(9) (10) (11)

with the following expansions for lift $L$, drag $D$ and sideforce $Y_{aero}$:

$$
\begin{align*}
D(V, \alpha) &= q \delta \left(C_{D_{0}} + C_{D_{u}} \alpha + C_{D_{a_{2}}} \alpha^{2}\right) \\
L(V, \alpha) &= q \delta \left(C_{L_{0}} + C_{L_{u}} \alpha \right) \\
Y_{aero}(V, \beta) &= q \delta \left(C_{Y_{p}} \beta \right)
\end{align*}
$$

(12) (13) (14)

where the dynamic pressure $q = 1/2 \rho V^{2}$. Currently, these conventional expansions are used in this approach. However, for future work, more elaborate expansions can be relied on, e.g. where drag $D$ also depends on the absolute value of the sideslip angle $|\beta|$.

The corresponding numerical values are: $m = 120 \cdot 10^{3} \text{ kg}$, $g = 9.81 \text{ m/s}^{2}$, $W = mg$, $\rho = 1.225 \text{ kg/m}^{3}$ (sea level), $S = 260 \text{ m}^{2}$, $C_{L_{0}} = 1.0656$, $C_{L_{u}} = 6.0723$, $C_{D_{0}} = 0.1599$, $C_{D_{u}} = 0.5035$, $C_{D_{a_{2}}} = 2.1175$, $C_{Y_{p}} = -1.6$.

In the perspective of reachability from stable and controllable trim conditions, the primary states of interest are airspeed $V$ and flight path angle $\gamma$. Considering time scale separation as presented in Fig. 2, the virtual inputs for the slow dynamics are roll angle $\phi$, angle of attack $\alpha$, sideslip angle $\beta$ and thrust $T$. This framework and combining Eqs. (7)–(14) allows to define the general dynamics $f$ in Eq. (1) by the following differential equation:

$$
\begin{align*}
\begin{bmatrix} \dot{V} \\ \dot{\gamma} \end{bmatrix} &= \begin{bmatrix} \frac{\rho \delta}{2m} V^{2} C_{D_{0}} - g \sin \gamma \\ -\frac{\delta}{m} \cos \gamma \end{bmatrix} + \begin{bmatrix} \cos \alpha \cos \beta \\ \cos \phi \sin \alpha \cos \beta - \sin \phi \sin \beta \end{bmatrix} \frac{T}{m} + \frac{\rho \delta}{2m} V^{2} \left(C_{D_{u}} \alpha + C_{D_{a_{2}}} \alpha^{2}\right) + \frac{\rho \delta}{2m} V \left(C_{L_{0}} + C_{L_{u}} \alpha \right) \cos \phi - \frac{\rho \delta}{2m} V C_{Y_{p}} \beta \sin \phi
\end{align*}
$$

(15)

Assuming small aerodynamic angles, as earlier, simplifies the differential equation:

$$
\begin{align*}
\begin{bmatrix} \dot{V} \\ \dot{\gamma} \end{bmatrix} &= \begin{bmatrix} \frac{\rho \delta}{2m} V^{2} C_{D_{0}} - g \sin \gamma \\ -\frac{\delta}{m} \cos \gamma \end{bmatrix} + \begin{bmatrix} 0 \\ \frac{\delta}{m} T \end{bmatrix} + \begin{bmatrix} 0 \\ \frac{\rho \delta}{2m} V \left(C_{D_{u}} \alpha + C_{D_{a_{2}}} \alpha^{2}\right) \end{bmatrix} + \begin{bmatrix} 0 \\ -\frac{\rho \delta}{2m} V C_{Y_{p}} \beta \sin \phi \end{bmatrix}
\end{align*}
$$

(16)

Note that the allowed ranges in this specific example are set at $\alpha \in \left[0^\circ; 14.5^\circ\right], \beta \in \left[-5^\circ; 5^\circ\right]$, as defined later. The small angle assumption holds up to $\pm 30^\circ$.
This results in a Hamiltonian function with decoupled virtual inputs \( T, \alpha \) and \( \beta \). Roll angle \( \varphi \) is not decoupled, but will be treated as a discretely gridded input. This decoupling significantly promotes computational efficiency, which is crucial for on-line applications. The Hamiltonian function becomes:

\[
H(\mathbf{p}, \mathbf{x}, \mathbf{u}) = \frac{p_1}{m} T - p_1 \frac{\rho S}{2m} V^2 C_D_{\alpha}, \alpha^2 + p_2 \frac{\rho S}{2m} V (p_2 C_{L_{\alpha}} \cos \varphi - p_1 V C_{D_{\alpha}}) \alpha + p_2 \frac{\rho S}{2m} V C_{\alpha} \sin \varphi \beta
\]

(17)

where \( \mathbf{p} \) are the co-states of the value function: \( p_1 = \frac{\partial \gamma}{\partial \alpha} \) and \( p_2 = \frac{\partial \gamma}{\partial \beta} \). This Hamiltonian is linear in thrust \( T \) and sideslip angle \( \beta \) and quadratic in angle of attack \( \alpha \). This structure allows for an efficient optimization routine over the inputs. The trim envelope boundaries are: \( V_{\text{min}} = 60m/s, V_{\text{max}} = 100m/s \) and \( \gamma_{\text{min/max}} = \pm 10^\circ \). The allowed ranges of the virtual inputs are: \( T_{\text{min}} = 20546N, V_{\text{max}} = 410920N, \alpha_{\text{min}} = 0^\circ, \alpha_{\text{max}} = 14.5^\circ \) (no stall), \( \phi_{\text{min/max}} = \pm 60^\circ \), \( \beta_{\text{min/max}} = \pm 5^\circ \). The maximizers \( \hat{T}, \hat{\alpha} \) and \( \hat{\beta} \) depend on the sign of the costates \( p_1 \) and \( p_2 \). Recall that \( V > 0, C_{D_{\alpha}} > 0, C_{D_{\beta}} > 0, C_{L_{\alpha}} > 0, C_{\alpha} < 0 \). Due to the underlying physics, no sign changes for these parameters are to be expected in case of structural changes. Define \( \hat{p} = \frac{p_2 C_{L_{\alpha}} \cos \varphi - p_1 V C_{D_{\alpha}}}{2p_1 V C_{\alpha}} \) and \( \hat{\alpha} = \frac{\alpha_{\text{min}} - \alpha_{\text{max}}}{2} \). Then the optimizing control inputs can be defined for invariance:

- If \( p_1 > 0 \) then \( \hat{T} = T_{\text{min}} \) and
  - \( \hat{\varphi} > \alpha \) then \( \hat{\alpha} = \alpha_{\text{min}} \)
  - \( \hat{\varphi} = \alpha \) then \( \hat{\alpha} = \alpha_{\text{min}} \) or \( \alpha_{\text{max}} \)
  - \( \hat{\varphi} < \alpha \) then \( \hat{\alpha} = \alpha_{\text{max}} \)
- If \( p_1 = 0 \) then \( \hat{T} \in [T_{\text{min}}; T_{\text{max}}] \)
  - if \( p_2 > 0 \) then \( \hat{\alpha} = \alpha_{\text{min}} \)
  - if \( p_2 = 0 \) then \( \hat{\alpha} \in [\alpha_{\text{min}}; \alpha_{\text{max}}] \)
  - if \( p_2 < 0 \) then \( \hat{\alpha} = \alpha_{\text{max}} \)
- If \( p_1 < 0 \) then \( \hat{T} = T_{\text{max}} \) and
  - if \( \hat{\varphi} < \alpha_{\text{min}} \) then \( \hat{\alpha} = \alpha_{\text{min}} \)
  - if \( \alpha_{\text{min}} \leq \hat{\varphi} \leq \alpha_{\text{max}} \) then \( \hat{\alpha} = \hat{\varphi} \)
  - if \( \hat{\varphi} \geq \alpha_{\text{max}} \) then \( \hat{\alpha} = \alpha_{\text{max}} \)
  - if \( p_2 \sin \varphi > 0 \) then \( \hat{\beta} = \beta_{\text{min}} \)
  - if \( p_2 \sin \varphi = 0 \) then \( \hat{\beta} \in [\beta_{\text{min}}; \beta_{\text{max}}] \)
  - if \( p_2 \sin \varphi < 0 \) then \( \hat{\beta} = \beta_{\text{max}} \)

For the purpose of maximizing the cost function with respect to the uncertainties \( \Delta \), the Hamiltonian from Eq. (17) can be rewritten, this time including parts independent of the inputs \( T, \alpha \) or \( \beta \) but with some aerodynamic derivative(s):

\[
H(\mathbf{p}, \mathbf{x}, \mathbf{u}) = -p_1 \frac{\rho S}{2m} V^2 C_{D_{\alpha}} + p_2 \frac{\rho S}{2m} V C_{L_{\alpha}} \cos \varphi + p_1 \frac{\rho S}{2m} V^2 \left( C_{D_{\alpha}} \alpha + C_{D_{\beta}} \alpha^2 \right) + p_2 \frac{\rho S}{2m} V C_{\alpha} \cos \varphi \alpha - p_2 \frac{\rho S}{2m} V C_{\beta} \sin \varphi \beta
\]

(18)

It can be observed that the aerodynamic derivatives all appear linearly in an uncoupled way, which allows a similar procedure to solve the optimization as previously. By rewriting the Hamiltonian as a summation of terms, where each term is a multiplication of a variable involving a costate, a constant factor and a derivative, one can determine the sign of this factor, which consists of the predefined physical parameters:

\[
H(\mathbf{p}, \mathbf{x}, \mathbf{u}) = -p_1 \frac{\rho S}{2m} V^2 C_{D_{0}} + p_2 \frac{\rho S}{2m} V \cos \varphi C_{L_{0}} + p_1 \frac{\rho S}{2m} V^2 C_{D_{\alpha}} + p_2 \frac{\rho S}{2m} V \alpha \cos \varphi C_{L_{\alpha}} + p_1 \frac{\rho S}{2m} V^2 \alpha C_{D_{\alpha}} - p_2 \sin \varphi \beta \frac{\rho S}{2m} V C_{\beta}
\]

(19)

where it should be noted that \( \varphi \in [-60^\circ; 60^\circ], \alpha \in [0^\circ; 14.5^\circ], \beta \in [-5^\circ; 5^\circ]. \) Furthermore, airspeed \( V > 0 \) and for the aerodynamic derivatives, it is known that \( C_{D_{\alpha}} > 0, C_{D_{\beta}} > 0, C_{L_{\alpha}} > 0, C_{\alpha} < 0 \). Due to the underlying physics, no sign changes for these parameters are to be expected in case of uncertainty or structural changes.

Based on this formulation, optimal control inputs for the aerodynamic derivatives can be defined as given in Table I where \( C_{\alpha} = C_{\alpha_{\text{max}}} \) and \( C_{\beta} = C_{\alpha_{\text{min}}} \).

With this information, it is possible to create an entire “uncertainty band” around the envelope, however, here focus will be placed on the “worst-case” minimal size envelope.

Fig. 4 compares the 3D envelopes with and without uncertainty, where two levels of uncertainty have been considered here, namely 10% and 20% of uncertainty on all aerodynamic derivatives. For the purpose of this example, identical ratios of standard deviations over nominal values have been defined for all derivatives, but the algorithm is capable to deal with individual standard deviations which can vary between the different aerodynamic derivatives. It can be clearly seen in Fig. 4 that larger degrees of uncertainty result in more significant shrinking of the envelope, since this is a “worst-case” minimal size envelope.

Fig. 5 analyzes the \( V, \gamma \) maneuvering envelope for different values of bank angle \( \varphi \), including robustness for uncertainties of 10% and 20%. By comparing Fig. 5(a), 5(b) and 5(c), it can be seen that larger bank angles have an influence on the climb capability of the aircraft. This is due to the physical principle that climb capability of lift force is provided through \( L \cos \varphi \), which confirms a smaller decrease for smaller bank angles (up to \( \varphi = 25^\circ \) as shown in Fig. 5(b)) but a much more significant change for larger bank angles as can be seen in Fig. 5(c).

V. VALIDATION OF RESULTS

The aforementioned results have been validated by means of Monte Carlo analyses. For this purpose, different bang-bang input signals have been inserted in the system. The extreme values of these signals correspond with the range limits. The time instant for the step change and the initial value for the input (maximum or minimum) vary over the Monte Carlo analysis. Running a nonlinear simulation of the
TABLE I

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<tr>
<th>sign of costate</th>
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<td>$p_2 \sin \phi \beta \geq 0$</td>
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<td>$p_2 \sin \phi \beta &lt; 0$</td>
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Fig. 5. $V, \gamma$ maneuvering envelope of RCAM model for time horizon $T = 2s$ for different bank angles and different uncertainty levels ($0\%$, $10\%$ and $20\%$ uncertainty). Smaller envelope areas correspond to larger uncertainty bounds.

Fig. 4. Comparison of 3D envelopes with and without uncertainty: nominal (green), $10\%$ uncertainty (blue), $20\%$ uncertainty (yellow)

In this paper, a computationally efficient algorithm for estimating the safe maneuvering envelope of damaged aircraft has been discussed. The algorithm performs a robust reachability analysis through an optimal control formulation while making use of time scale separation and taking into account uncertainties in the aerodynamic derivatives. The safe maneuvering envelope is defined as the cross section...
between the forwards reachable and backwards reachable sets, which have been calculated starting from the stable trim envelope. Moreover, the backwards reachable set can be considered as the survivable maneuvering envelope, from where it is possible to bring the aircraft back to a safe trim condition after an upset due to damage, turbulence, a wake encounter etc. Results were found to be consistent with the underlying physical principles. This approach differs from others since it is physically inspired. This more transparent approach allows interpreting data in each step, and it is assumed that the physics based approach will therefore facilitate certification for future real life applications.

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