

Preliminary Multi-Variable Cost Model for Space Telescopes

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ABSTRACT

Parametric cost models are routinely used to plan missions, compare concepts and justify technology investments. This paper reviews the methodology used to develop space telescope cost models; summarizes recently published single variable models; and presents preliminary results for two and three variable cost models. Some of the findings are that increasing mass reduces cost; it costs less per square meter of collecting aperture to build a large telescope than a small telescope; and technology development as a function of time reduces cost at the rate of 50% per 17 years.

Keywords: Space Telescope Cost Model, Parametric Cost Model

1. INTRODUCTION

Multivariable parametric cost models for space telescopes provide several benefits to designers and space system project managers. They identify major architectural cost drivers and allow high-level design trades. They enable cost-benefit analysis for technology development investment. And, they provide a basis for estimating total project cost. A survey of historical models revealed that there is no definitive space telescope cost model. In fact, the published models vary greatly. [1] Therefore, the opportunity exists to develop a multi-variable parametric cost model for space telescopes that encompasses the latest available data and applies rigorous analytical techniques. The first step in this process was to develop a single variable parametric cost model for space telescopes. [2]

Cost and engineering data has been collected on 59 different parameters for 23 different UV, optical or infrared space telescopes. (Table 1 and Table 2)

Table 1: UV/OIR Cost Model Missions Database	
UV/Optical Telescopes	Infrared Telescopes
EUVE	CALIPSO
FUSE	Herschel
GALEX	ICESat
HiRISE	IRAS
HST	ISO
HUT	JWST
IUE	SOFIA
Kepler	Spitzer (SIRTF)
Copernicus (OAO-3)	TRACE
SOHO/EIT	WIRE
UIT	WISE
WUPPE	

Table 2: Cost Model Variables Study and the completeness of data knowledge	
Parameters	% of Data
OTA Cost	89%
Total Phase A-D Cost w/o LV	84%
Aperture Diameter	100%
Avg. Input Power	95%
Total Mass	89%
OTA Mass	89%
Spectral Range	100%
Wavelength Diffraction Limit	63%
Primary Mirror Focal Length	79%
Design Life	100%
Data Rate	74%
Launch Date	100%
Year of Development	95%
Technology Readiness Level	47%
Operating Temperature	95%
Field of View	79%
Pointing Accuracy	95%
Orbit	89%
Development Period	95%
Average	88%

Statistical correlations have been evaluated between 19 of 59 variables. And, these parameters have been used to develop single and multi-variable cost estimating relationships (CERs) which are evaluated for their 'goodness'. For the purpose of this paper, Optical Telescope Assembly (OTA) is defined as the space observatory subsystem which collects electromagnetic radiation and focuses it (focal) or concentrates it (afocal). An OTA consists of the primary mirror, secondary mirror, auxiliary optics and support structure (such as optical bench or truss structure, primary support structure, secondary support structure or spiders, etc.). An OTA does not include science instruments or spacecraft subsystems. And, cost is defined as prime contract cost without any NASA labor or overhead. Total mission cost is defined as Phase A-D cost, excluding: launch cost; costs associated with NASA labor (civil servant or support contractors) for program management, technical insight/oversight; or any NASA provided ground support equipment, e.g. test facilities. Accounting for NASA overheads would increase the cost by at least 10% and maybe as much as 33%.

Four single variable cost estimating relationships (CERs) have been developed. These CERs estimate OTA cost and total mission cost as a function of OTA diameter, OTA mass and total mission mass. [2] This paper reviews those results including the finding that ‘attached’ OTAs with mass 10X larger than ‘free-flying’ OTAs are 60% less expensive; tests the historical Horak model against the data base; and presents preliminary results regarding the development of a multi-variable cost model.

2. MODEL CREATION

The first step in creating a statistical cost model is to start with the Cross Correlation Matrix (Figure 1) and look for variables which are highly correlated with cost. When using a cross-correlation matrix, there are several things to consider. First, the higher the correlation value, the greater the cost variation explained by that variable. Second, the sign of correlation is important. It must be consistent with known engineering design principals and manufacturing processes. Third, for multi-variable models, we want variables which independently effect cost. Variables which ‘cross-talk’ with each other are multicollinear.

	Total Phase A-D Cost	OTA Cost	Areal Total Cost	Areal OTA Cost	Aperture Diameter	PM F Lens	PM F/N	OTA Volume	FOV	Pointing Accuracy	Total Mass	OTA Mass	Total Areal Density	OTA Areal Density	Spectral Range	Minimum Wavelength	Diffraction Limit	Operating Temperature	Avg. Input Power	Data Rate	Design Life	Technology Readiness Level	Year of Development	Development Period	Launch Date	Orbit
units	(\$MM)	(\$MM)	(\$MM/m²)	(\$MM/m²)	(m)	(m)	unitless	(m³)	(°)	(Arc-Sec)	(kg)	(kg)	(kg/m²)	(kg/m²)	(μm)	(μm)	(μm)	(K)	(Watts)	(Gbps)	(months)	TRL	(year)	(months)	(year)	(km)
Total Phase A-D Cost	1.00	0.70	0.09	-0.36	0.64	0.80	0.38	0.83	0.26	-0.52	0.92	0.72	-0.17	-0.48	-0.02	-0.40	-0.04	0.59	0.44	0.65	-0.41	-0.11	0.78	0.11	0.54	
OTA Cost		1.00	-0.53	-0.30	0.87	0.82	0.39	0.84	0.00	-0.58	0.68	0.82	-0.64	-0.41	0.07	-0.23	0.01	0.14	0.15	0.46	-0.68	-0.31	0.45	-0.16	0.17	
Areal Total Cost			1.00	0.68	0.20	0.22	-0.71	-0.40	0.03	-0.50	0.26	0.34	0.00	0.00	-0.47	0.92	0.38	-0.23	-0.18	-0.07	-0.05	-0.07	-0.20	-0.29	-0.28	
Areal OTA Cost				1.00	-0.74	-0.62	-0.16	-0.71	-0.56	0.30	-0.34	-0.48	0.72	0.59	-0.20	-0.07	-0.03	-0.48	-0.48	-0.41	-0.43	-0.56	-0.22	-0.68	0.04	
Aperture Diameter					1.00	0.88	0.27	0.98	-0.09	-0.58	0.63	0.86	-0.80	-0.60	0.14	-0.11	0.05	0.42	0.38	0.53	-0.29	0.09	0.37	0.26	0.08	
PM F Lens						1.00	0.69	0.96	0.34	-0.66	0.84	0.78	-0.50	-0.44	-0.50	-0.19	0.28	0.49	0.31	0.50	-0.38	-0.07	0.50	0.10	0.28	
PM F/N							1.00	0.45	0.57	-0.41	0.48	0.33	0.09	-0.02	-0.61	-0.43	0.32	0.06	0.20	0.25	-0.37	-0.32	0.21	-0.29	0.08	
OTA Volume								1.00	0.08	-0.65	0.84	0.84	-0.66	-0.54	-0.36	-0.08	0.21	0.65	0.34	0.52	-0.31	0.06	0.54	0.26	0.31	
FOV									1.00	0.12	0.16	-0.05	0.25	0.01	0.05	-0.38	-0.06	-0.02	0.18	0.09	-0.27	0.08	-0.01	0.09	0.09	
Pointing Accuracy										1.00	-0.48	-0.71	0.44	0.14	0.31	0.08	-0.38	-0.37	-0.29	-0.35	-0.15	0.13	-0.55	-0.02	-0.32	
Total Mass											1.00	0.82	-0.04	-0.42	-0.15	-0.49	0.03	0.55	0.17	0.65	-0.56	-0.27	0.64	-0.10	0.33	
OTA Mass												1.00	-0.59	-0.11	-0.06	0.06	-0.03	0.60	0.09	0.40	-0.29	-0.16	0.57	0.02	0.47	
Total Areal Density													1.00	0.62	-0.28	-0.20	-0.01	-0.06	-0.29	-0.19	-0.54	-0.32	-0.07	-0.38	0.15	
OTA Areal Density														1.00	0.05	0.28	-0.31	-0.16	-0.39	-0.55	0.07	-0.36	-0.20	-0.46	-0.09	
Spectral Range															1.00	0.76	-0.79	-0.09	-0.12	-0.25	-0.09	0.21	0.20	0.23	0.01	
Minimum Wavelength																1.00	-0.55	-0.07	-0.25	-0.75	0.51	0.35	0.31	0.28	0.14	
Diffraction Limit																	1.00	0.09	0.31	0.31	0.11	-0.01	-0.30	0.00	-0.30	
Operating Temperature																		1.00	0.34	0.64	0.05	0.35	0.27	0.45	0.04	
Avg. Input Power																			1.00	0.54	0.51	0.49	-0.06	0.52	0.21	
Data Rate																				1.00	0.15	0.12	0.12	0.24	0.14	
Design Life																					1.00	-0.15	0.12	0.12	0.24	0.14
Technology Readiness Level																					1.00	0.68	-0.24	0.64	0.33	
Year of Development																						1.00	-0.23	0.97	-0.05	
Development Period																							1.00	-0.02	0.51	
Launch Date																								1.00	0.04	
Orbit																									1.00	

Figure 1: Cross-Correlation Matrix of data base for 19 Free-Flying Space Telescope Systems. Correlations which are at least 95% significant are **Bolded**, e.g. for 12 data points a correlation of greater than 60% is significant to better than 95%.

The second step is to understand the statistical indicators of ‘Goodness of Fit’ and ‘Significance’. Goodness of Fit is tested via a range of statistical measures, including Pearson’s r^2 coefficient, Student T-Test p-value and standard percent error (SPE). Pearson’s r^2 (typically denoted as just r^2) describes the percentage of agreement between the model and the actual cost. For multi-variable models, we use Adjusted Pearson’s r^2 (or r^2_{adj}) which accounts for the number of data points and the number of variables. In general, the closer r^2 (or r^2_{adj}) is to 1.0 or 100%, the better the model. SPE is a normalized standard deviation of the fit residual (difference between data and fit) to the fit. The closer SPE is to 0, the better the fit. Please note that since SPE is normalized, a small variation divided by a very small parameter coefficient can yield a very large SPE. The p-value is the probability that a fit or correlation would occur if the variables are

independent of each other. The closer the p-value is to 0, the more significant the fit or correlation. The closer it is to 1, the less significant. If the p-value for a given variable is small, then removing it from the model would cause a large change to the model. If it is large, then removing the variable will have a negligible effect. Also, it is important to consider how many data points are included in a given correlation, fit or regression.

Given the complexity and data density of Figure 1, Table 3 digests the cross-correlation between specific key parameters and Total Mission Cost, OTA Cost and OTA Areal Cost (where areal cost is defined as OTA cost divided by OTA collecting area). For each parameter, Table 3 reports its r^2 correlation to cost, the correlation's p-value and the number of data points in the correlation. Diameter appears to be the most significant cost driver. So, in addition to total cost and OTA cost we have examined OTA Areal Cost, i.e. OTA Cost per unit Area of Primary Mirror collecting aperture.

Parameter	Total Cost			OTA Cost			OTA Areal Cost		
	Corr	p	N	Corr	p	N	Corr	p	N
Diameter	.68	.007	14	.87	0	16	-.71	.005	14
Focal Length	.82	.002	11	.82	.001	12	-.42	.194	11
Pointing Accuracy	-.53	.061	14	-.64	.011	15	.47	.087	14
Total Mass	.92	0	15	.68	.005	15	-.0	.997	15
OTA Mass	.72	.002	15	.82	0	15	-.47	.074	15
Spectral Min	-.02	.934	16	.07	.804	17	-.23	.383	16
Operating Temp	-.04	.884	16	0	.975	16	-.07	.802	16
Electrical Power	.59	.021	15	.14	.611	16	-.05	.862	16
Design Life	.65	.007	16	.46	.064	17	-.20	.454	16
TRL	-.41	.307	8	-.68	.061	8	-.29	.481	8
Development Period	.78	.001	15	.45	.083	15	.14	.830	15
Launch Year	.11	.675	16	-.16	.533	17	-.34	.204	16

Mass correlates most significantly with Total Cost while OTA Mass correlates most significantly with OTA Cost. Unexpectedly, Minimum Spectral Range Value and Operating Temperature do not have a significant correlation with any Cost. However, as we will show later, Spectral Minimum does have a role in multi-variable cost models. As expected Electrical Power, Design Life and Development Period have significant correlations (99% confidence) with Total Cost. Also unexpected is that TRL and Launch Year do not have significant correlations. But, as we will discuss later, they both have roles in multi-variable cost models. One problem with TRL is that there are only 8 data points. Also, it is a qualitative and not a quantitative parameter.

3. SUMMARY OF SINGLE VARIABLE COST MODEL RESULTS

Four single variable cost estimating relationships (CERs) have been developed for OTA cost and total mission cost as a function of OTA diameter, OTA mass and total mission mass. [2] These models were developed with and without JWST. The benefit of including JWST is that it is the most current mission. The disadvantage is that its cost is not yet final. For the purpose of this paper, we will include the 2009 JWST C/D final cost estimate. In general, including JWST does affect the model r^2_{adj} but does not increase the noisiness of the fit as represented by the SPE. Additionally, these models are developed only for free-flying missions. Of the 23 missions in the data base, there are 19 free flying telescopes (17 for which we have OTA cost data) and 4 that are attached (3 to the Space Shuttle Orbiter and SOFIA to a Boeing 747 airplane). As will be discussed below with regard to mass models, attached missions have a significantly different cost dependency than free-flying missions. Therefore, we excluded attached missions from the models.

Figure 3 plots OTA Cost for free-flying space telescopes as a function of Primary Mirror Diameter. The regression fit for this data is:

$$\text{OTA Cost} \sim \text{Aperture Diameter}^{1.2} \quad (N = 17; r^2 = 75\%; \text{SPE} = 79\%) \text{ with 2009 JWST}$$

Note that the Chandra data point is included only for reference. It is not included in the regression. And, it is inserted based upon the equivalent normal incidence mirror diameter it would have if all of its x-ray mirrors were unrolled.

Given that the OTA cost might be dominated by the large apertures for HST and JWST, a model was also created for normalized Areal OTA Cost (Figure 4):

$$\text{OTA Areal Cost} \sim \text{Aperture Diameter}^{-0.74} \quad (N = 17; r^2 = 55\%; \text{SPE} = 78\%) \text{ with JWST}$$

A key finding of this analysis is that Areal Cost decreases with aperture size. It is less expensive per photon to build a large aperture telescope than a small aperture telescopes. Large aperture telescopes provide a better ROI.

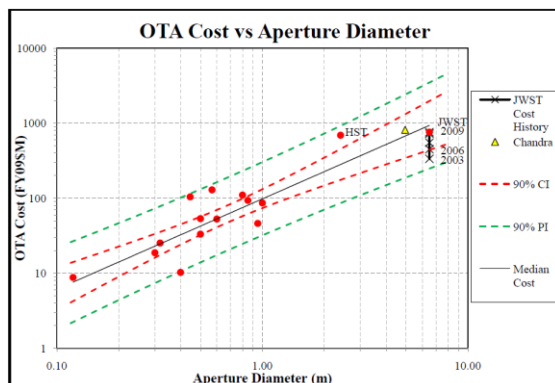


Figure 3: OTA Cost vs Aperture Diameter scaling law for 17 free flying UV/OIR systems (including 2009 JWST). Plot includes 90% confidence and prediction intervals, and data points. Chandra data point is not included in the regression.

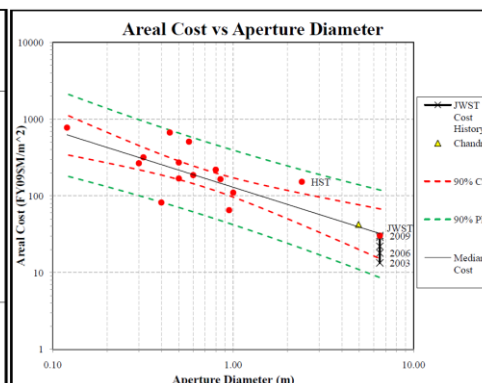


Figure 4: OTA Areal Cost vs Aperture Diameter scaling law for 17 free flying UV/OIR systems (including 2009 JWST). Plot includes 90% confidence and prediction intervals, and data points. Chandra data point is not included in the regression.

From both an engineering and a science perspective, aperture diameter is the best parameter upon which to build a space telescope cost model. Aperture defines the observatory's science performance and determines the payload's size and mass. And, while the results are consistent with some historical cost models, our results invalidate long held 'intuitions' which are often purported to be 'common knowledge'. Space telescope costs vary almost linearly with diameter and not to a power of 1.6X or 2.0X or even 2.8X. But, a model based on diameter alone has only an ~80% agreement with the OTA cost data and ~55% agreement with the OTA areal data. Therefore, other factors must influence cost. The next step is to develop a multi-variable cost model using multi-variable regression techniques.

While aperture diameter is the single most important parameter driving science performance, total system mass determines what vehicle can be used to launch. Significant engineering costs are expended to keep a given payload inside of its allocated mass budget. This includes light-weighting mirrors and structure. Therefore, mass is a potential CER which requires study. Figure 5 plots Total Cost for Free-Flying Missions vs Total Mission Mass. The regression of this data is:

$$\text{Total Cost} \sim \text{Total Mass}^{1.12} \quad (N = 15; r^2 = 86\%; \text{SPE} = 71\%) \text{ with JWST}$$

Figure 6 plots OTA Cost vs OTA Mass for both free-flying and attached missions. The regression for only the free-flying missions is:

$$\text{OTA Cost} \sim \text{OTA Mass}^{0.72} \quad (N = 15; r^2 = 92\%; \text{SPE} = 93\%) \text{ with JWST}$$

While OTA Mass may appear to be a good indicator of OTA Cost because it has the highest Pearson's r^2 , it also has the highest SPE. In general mass should be avoided as a CER because it is a secondary indicator. Mass depends upon the size of the telescope. Bigger telescopes have more mass. And, bigger telescopes typically require bigger spacecraft and bigger science instruments which both require more power – all which require more mass. And, because many missions are designed to a mass-budget defined by launch vehicle constraints, the result can be a very complex, risky, and expensive mission architecture when trying to extend the state-of-the-art in either wavelength or aperture. An indication of this is given in Figure 5 where JWST has nearly half the total mass of HST but still has a higher total mission cost – because JWST is much more complex than HST. But, this does not have to be the case. The key finding of Figure 6 (as indicated by the square data points) is that attached OTAs are ~10X more massive and ~60% less expensive than free-flying missions. This finding actually invalidates the 'common assumption' that the more massive the mission the more expensive the mission. The only reason that more massive missions are more expensive is because they have more 'stuff'. When one compares missions with similar performance properties, it is less expensive to design, build and fly a simple mission with more mass than a lightweight complex mission. Therefore, maybe the best way to reduce the cost

of future large aperture space telescopes is to develop cost effective heavy lift launch vehicles which will enable mission planners to trade complexity for mass.

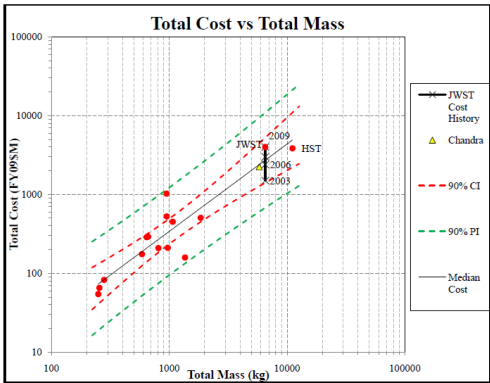


Figure 5: Total Cost vs Total Mass scaling law for free-flying UV/OIR space telescopes (including 2009 JWST). Plot includes 90% confidence and prediction intervals, and data points. Chandra data point is not included in the regression.

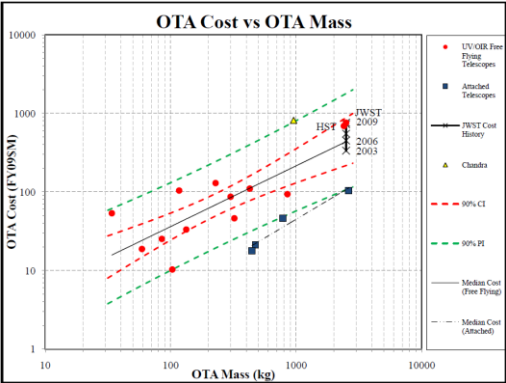


Figure 6: OTA Cost vs OTA Mass scaling law for free-flying UV/OIR space telescopes (including 2009 JWST). Plot includes 90% confidence and prediction intervals, and data points. Chandra data point is not included in the regression.

4. TEST OF HISTORICAL MODELS

One of the goals of the study was to test historical models against our data base. Of all the historical models, the Horak model is the easiest to test. Our database has parameters equivalent to the Horak database. And, Horak published the details of his statistical fit including his Student-T Test Confidence values (Figure 7):

CER:

$$T1 = 0.357 (Matl)(Dsn)(Apr)^{0.705} (\#Elem)^{0.473} (\lambda)^{-0.178} (K^{\circ})^{-0.191} e^{-0.033(Yr-80)}$$

T Statistics:

(1.74)

(8.80)

(2.55)

(-2.04)

(-2.61)

(-2.31)

Statistics

$R^2 = 97.0\%$
 $s = 0.212 \quad (17 \text{ Data Points})$

Figure 7: Horak Cost Estimating Relationship with Statistical Details

For this comparison, we ignore the Material (glass vs metal) and Design (on vs off-axis) factors and concentrate on the parameters with power terms. The first step is to convert the Horak T Statistics for each parameter into p-values:

Parameter	Apr	#Elem	Wave	Temp	Year
T Statistics:	8.80	2.55	-2.04	-2.61	-2.31
p-values:	0.00	0.022	0.059	0.020	0.036

Based on 17 data points, the reported T statistics and p-values indicate that all variables in Horak’s model are significant. And, given that he reported an $R^2 = 97\%$, his model yielded a good fit to his data. Next, we regress the Horak parameters against our data base yielding:

$$OTA\ Cost \sim Diam^{1.39} \#Elem^{-1.11} \lambda^{-0.024} K^{\circ -0.045} e^{-0.0369(Yr-80)}$$

Parameter

Apr

#Elem

Wave

Temp

Year

T Statistics:

9.34

-1.03

-0.22

-0.38

-2.80

p-values:

0.00

0.320

0.829

0.710

0.014

For our data, based on 16 data points, only Diameter and Launch Year are significant and the fit has a good $R^2=90.8\%$ and $R^2_{adj} = 86.2\%$. (Please note: R^2 statistics, which are calculated in log-space, are reported so as to match Horak’s measures of ‘goodness’.) These T statistics and p-values are generated from the t-test for the coefficients which tests whether the coefficient is equal to zero or not. The p-value represents the probability of getting the result we got (the coefficient) if the actual coefficient equals zero. The standard is to reject the coefficient (argue that it is really zero) if

the p-value is greater than 0.05 (corresponding to a 5% chance of getting the coefficient we got if it is really zero). For this model regression against our data base, we cannot claim that the exponents for #Elem, λ , and K° are statistically different than zero. Thus those terms have little effect on the model and are not drivers of telescope cost. Removing those variables yields the following model:

$$OTA\ Cost \sim Diam^{1.33} e^{-0.0434(Yr-80)}$$

T Statistics:	10.61	-4.22
p-values:	0.00	0.001

Based on 17 data points, both Diameter and Launch Year are significant and the fit has a good $R^2=89.2\%$ and $R^2_{adj}=87.6\%$. The R^2 is marginally smaller after removing the three variables and the R^2_{adj} increases from the removal.

The explanation is in the databases. The Horak data base consists mostly of DoD strategic systems most of which were laboratory experiments that were never deployed. Of the systems which were flown, most were airframe or missile systems. Our database consists entirely of space telescope missions.

5. MULTI-VARIABLE PARAMETRIC MODEL

The first step in developing a multi-variable parametric model is to start with a single variable model and evaluate its statistics ‘goodness’. For OTA Cost, we start with Aperture Diameter. Then we perform a two variable regression of Diameter and each of the other parameters under study – one parameter at a time – and evaluate the statistical ‘goodness’ of each regression. Once a good two variable model is selected, the process is repeated to add a third variable. This process is generally called step-wise regression.

There are five specific criteria that must be satisfied when adding parameters to a multi-variable model:

The correlation of each variable must be ‘significant’. An arbitrary criterion is that the parameter’s p-value must be less than 0.10 (for a 90% confidence limit) but in some cases, we will consider variables whose correlations are significant only at the 80% confidence level.

The parameter’s coefficient must be ‘consistent’ with engineering judgment. For example, it violates engineering judgment if the coefficient for TRL is positive. This would indicate that the higher the TRL level the higher the cost of a telescope or mission.

The addition of a variable should ‘increase’ Pearson Adjusted r^2 . The closer r^2_{adj} is to 1.0 the better the model agrees with the data.

The addition of a variable should ‘decrease’ SPE. The closer SPE is to 0.0 the less noise there is in the fit.

The parameters should not be multicollinear. However, this rule may be violated with proper justification. Although at this point we have not identified such a justification.

5.1 Two-Variable Models

Figure 8 summarizes the results of a two variable model regression for OTA Cost as a function of Aperture Diameter and a second variable. Three parameters have significance greater than 98%: TRL, Year of Development (YoD) and Launch Year (LYr). The Diameter + TRL model has a slightly higher r^2_{adj} than the other models, but it also has a high SPE. This may be because of the relatively few TRL data points in our data base. Or, it may be because TRL value is subjective and thus has a natural ‘fuzziness’ to its data values. Based on coefficient significance, other parameters of potential interest are Field of View (82%), OTA Mass (74%), OTA Areal Density (74%), Power (77%) and Data Rate (72%). But, all, except Data Rate, do not simultaneously increase r^2_{adj} and decrease SPE. And, some, such as FOV, are particularly poor. It should also be noted that OTA Mass is multicollinear with Aperture Diameter – which only makes sense, i.e. the larger the telescope, the more mass it should have. Therefore, mass is not a good second variable candidate. For the purpose of future three variable model regressions, we will use YoD or LYr.

Both YoD and LYr have similarly high r^2_{adj} values and significantly lower SPE values. And, if you round significant digits, each model is virtually identical:

$$OTA\ Cost \sim D^{1.34} e^{-0.04(LYr-1960)} \quad (N = 17, r^2_{adj} = 93\%; SPE=39\%)$$

$$OTA\ Cost \sim D^{1.27} e^{-0.04(YoD-1960)} \quad (N = 16, r^2_{adj} = 95\%; SPE=39\%)$$

At this stage of our study, we have not determined which parameter to use. Launch Year has the advantage that it is a definite date, but it also has the disadvantage that a launch can be delayed. And, while a launch delay tends to increase the total mission cost, it may or may not increase the OTA cost. Year of Development yields a slightly better regression, but its exact date is subject to definition. Is it the Start of Phase A or B or C? Regardless, the message of either 'year' model is clear: technology improvements reduce OTA cost as a function of time by approximately 50% every 17 years. Further cost reductions could probably be obtained if a procurement strategy was pursued to maintain an industrial infrastructure and a skilled workforce by making multiple system procurements spread out over many years.

	coef		p		OTA Cost versus Diameter and V2																	
Second Variable	Diameter alone		PM F Len.		PM F/N		OTA Volume		FOV		Pointing Accuracy		OTA Mass		OTA Areal Density		Spectral Range minimum		Wavelength Diffraction Limit		Operating Temperature	
Diameter	1.20	0.00	0.68	0.27	1.05	0.00	-0.02	0.99	1.16	0.01	1.14	0.00	0.76	0.12	1.45	0.00	1.22	0.00	1.19	0.00	1.21	0.00
Second Variable	-	-	0.35	0.45	0.26	0.57	0.35	0.45	-0.26	0.18	-0.05	0.45	0.35	0.26	0.35	0.26	-0.04	0.63	-0.10	0.55	-0.01	0.96
Adjusted r2	73%		71%		71%		71%		14%		73%		83%		83%		73%		75%		71%	
SPE	79%		77%		78%		77%		73%		78%		83%		83%		84%		95%		82%	
n	17		13		13		13		13		16		15		15		17		11		16	
Multicollinearity?	N/A		Yes		No		Yes		No		No		Yes		No		No		No		No	

Second Variable			Avg. Input Power		Data Rate		Design Life		Design Life (exp)		Technology Readiness Level		YoD (exp)		Development Period		Dev Per (exp)		Launch Date (exp)		Orbit	
Diameter			1.41	0.00	1.40	0.00	1.21	0.00	1.13	0.00	1.31	0.00	1.27	0.00	1.19	0.00	1.20	0.00	1.34	0.00	1.23	0.00
Second Variable			-0.15	0.23	-0.08	0.28	-0.01	0.98	0.00	0.51	-0.09	0.02	-0.04	0.00	0.23	0.60	0.00	0.73	-0.04	0.00	0.02	0.62
Adjusted r2			70%		91%		71%		84%		97%		95%		71%		71%		93%		66%	
SPE			58%		59%		83%		81%		83%		39%		77%		78%		39%		85%	
n			16		12		17		17		8		16		16		16		17		15	
Multicollinearity?			No		No		No		No		No		No		No		No		No		No	

Figure 8: Two Variable Model Regression for Optical Telescope Assembly (OTA) Cost vs Aperture Diameter and a 2nd Variable

As with the single variable model, there is utility in looking at OTA Areal Cost. And, the two variable regression confirms the single variable result. It costs less per square meter to build a large aperture space telescope than a small aperture telescope. The regression for OTA Areal Cost is very similar to the regression for OTA Cost, therefore, we will not show a summary chart similar to Figure 8. TRL, YoD and LYr are all significant with a confidence of >98%. TRL is less noisy, but YoD and LYr have higher correlation values.

$$\text{OTA Areal Cost} \sim D^{-0.61} e^{-0.04(\text{LYr}-1960)} \quad (N = 17, r^2_{adj} = 76\%; \text{SPE}=40\%)$$

$$\text{OTA Areal Cost} \sim D^{-0.68} e^{-0.04(\text{YoD}-1960)} \quad (N = 16, r^2_{adj} = 76\%; \text{SPE}=39\%)$$

$$\text{OTA Areal Cost} \sim D^{-0.69} \text{TRL}^{-0.93} \quad (N = 8, r^2_{adj} = 56\%; \text{SPE}=35\%)$$

Like OTA Cost, OTA Areal Cost also reduces as a function of time. It is intuitively obvious that a higher TRL value lowers cost. And the model result should help justify maximizing technology re-use and pre-Phase A development funding. But, given the requirement that all technology should be at TRL-6 before beginning Phase C/D, a TRL based cost model is probably not very useful. Finally, both Mass parameters had terrible r^2_{adj} values of only 15%.

Next, we look at two variable models for Total Mission Cost. First we used Diameter as the primary variable. Based on a plot of Total Cost versus Diameter (Figure 9), it appears that there is a dependency. And, in fact there may be. However, while the regression yields a coefficient with 99% significance, its SPE is very noisy:

$$\text{Total Cost} \sim D^{0.88} \quad (N = 16, r^2_{adj} = 72\%; \text{SPE}=203\%)$$

IR and UV Telescopes

This scatter plot illustrates the relationship between the aperture diameter and the total mission cost for various IR and UV telescopes. The x-axis represents the Aperture Diameter in meters (m) on a logarithmic scale from 0.10 to 10.00. The y-axis represents the Total Mission Cost in units of 10^9 \$ on a logarithmic scale from 1.00 to 10000.00. A power-law fit line is shown, and specific data points are labeled: SOHO, HST, and JWST.

Telescope	Aperture Diameter (m)	Total Mission Cost (10^9 \$)
SOHO	0.10	~500
HST	~2.4	~3000
JWST	~6.5	~4000

Figure 9: Total Mission Cost vs Aperture Diameter.

[illegible]

Figure 10: Two Variable Model Regression for Total Mission Cost vs Aperture Diameter and a 2nd Variable

Now for a disclaimer, we continue to believe that Mass is not an appropriate cost driver (consider that JWST cost more than HST but has half the mass, and that attached OTAs with 10X more mass are 60% lower cost), but we decided to examine it anyway. Figure 11 shows the two variable regression of Total Mission Cost as a function of Total Mass and a 2nd Variable. Potential models from this regression include:

Total Cost $\sim \text{TM}^{0.77} \text{e}^{0.02\text{DevPeriod}}$	$(N = 15, r^2_{adj} = 98\%; \text{SPE}=58\%)$
Total Cost $\sim \text{TM}^{0.85} \text{DevPeriod}^{1.16}$	$(N = 15, r^2_{adj} = 95\%; \text{SPE}=54\%)$
Total Cost $\sim \text{TM}^{1.01} \text{Orbit}^{0.09}$	$(N = 14, r^2_{adj} = 96\%; \text{SPE}=38\%)$
Total Cost $\sim \text{TM}^{0.98} \text{DataRate}^{0.10}$	$(N = 12, r^2_{adj} = 95\%; \text{SPE}=63\%)$
Total Cost $\sim \text{TM}^{1.13} \text{Temp}^{-0.13}$	$(N = 15, r^2_{adj} = 91\%; \text{SPE}=56\%)$

Development Period and Orbit are both 99% significant. TRL, Operating Temperature and Data Rate are all significant at >80% confidence level. But the TRL 'sign' is wrong. Minimum Spectral Wavelength is boarder line significant. One problem with minimum spectral wavelength is that in some cases its coefficient is positive and in other cases it is negative. Neither YoD or LYr are significant.

	coef		p		Total Cost vs Total Mass and V2																			
Second Variable	Total Mass alone		PM F Len.		PM F/N		OTA Volume		FOV		Pointing Accuracy		OTA Mass		Total Areal Density		OTA Areal Density		Spectral Range minimum		Wavelength Diffraction Limit		Operating Temperature	
Total Mass	1.11	0.00	1.07	0.02	1.12	0.00	0.97	0.03	1.07	0.00	1.06	0.00	1.20	0.00	1.10	0.00	1.06	0.00	1.14	0.00	1.03	0.00	1.13	0.00
Second Variable	-	-	0.03	0.93	-0.14	0.75	0.06	0.71	0.04	0.61	-0.03	0.60	-0.08	0.69	-0.06	0.54	-0.16	0.47	0.09	0.20	-1.03	0.85	-0.13	0.16
Adjusted r ²	85%		84%		81%		91%		9%		85%		82%		90%		94%		90%		26%		91%	
SPE	71%		81%		77%		81%		79%		72%		81%		71%		87%		61%		243%		56%	
n	15		11		11		11		12		15		14		15		14		15		9		15	
Multicollinearity?	No		Yes		No		Yes		No		No		Yes		No		No		No		No		No	

Second Variable			Avg. Input Power		Data Rate		Design Life		Design Life (exp)		Technology Readiness Level		YoD (exp)		Development Period		Dev Per (exp)		Date of Launch (exp)		Orbit	
Total Mass			1.06	0.00	0.98	0.00	1.04	0.00	1.12	0.00	1.28	0.00	1.12	0.00	0.85	0.00	0.77	0.00	1.10	0.00	1.01	0.00
Second Variable			-0.02	0.92	0.10	0.17	0.12	0.62	0.00	0.96	0.70	0.19	0.01	0.62	1.16	0.01	0.02	0.01	0.01	0.41	0.09	0.01
Adjusted r ²			94%		95%		81%		85%		96%		90%		95%		98%		93%		96%	
SPE			79%		63%		73%		74%		58%		72%		54%		58%		71%		38%	
n			14		12		15		15		8		15		15		15		15		14	
Multicollinearity?			No		No		No		No		No		No		No		No		No		No	

Figure 11: Two Variable Model Regression for Total Mission Cost vs Total Mass and a 2nd Variable

5.2 Three-Variable Models

The next step after developing a two variable model is to try adding a third parameter. Given that the only satisfactory two variable models were OTA Cost versus Diameter and a ‘year’ parameter, we did two regressions: Diameter and Year of Development, and Diameter and Launch Year with all the other variables. Both regressions gave similar results with the Year of Launch version yielding slightly better results (Figure 12). None of the regressions yielded a satisfactory model. The only 3rd variable with a significant coefficient was TRL. And, the addition of TRL forced the Launch Year (and Year of Development) coefficient to zero. The only other parameter with an even remotely significant coefficient is Orbit. We will revisit Orbit in the future.

Finally, given the failure of the three variable regression, we decided to add some wavelength diversity by including missions with shorter and longer wavelengths. Specifically, we added WMAP, TDRS-1, TDRS-7, EUVE, Chandra and Einstein. The regression yielded to satisfactory models with the Year of Development being slightly better (Figure 13):

$$\text{OTA Cost} \sim D^{1.15} \lambda^{-0.17} e^{-0.03(\text{YoD}-1960)} \quad (N = 20, r_{adj}^2 = 92\%; \text{SPE} = 76\%)$$

$$\text{OTA Cost} \sim D^{1.05} \lambda^{-0.13} e^{-0.03(\text{LY}-1960)} \quad (N = 23, r_{adj}^2 = 63\%; \text{SPE} = 69\%)$$

Interestingly, adding wavelength diversity to the regression yields a wavelength coefficients similar to the Horak model:

$$T1 \sim Apr^{0.705} \#Elem^{0.473} \lambda^{-0.178} K^{-0.191} e^{-0.033(Y-80)}$$

Please note, as indicated by the title of this paper, these results are preliminary and may change over the coming year as we continue our analysis.

	coef	p	OTA Cost vs Diameter, DoL, and V3																	
Third Variable	Diam and Launch Date (exp)		PM F Len.		PM F/N		OTA Volume		FOV		Pointing Accuracy		OTA Mass		OTA Areal Density		Spectral Range minimum		Wavelength Diffraction Limit	
Aperture Diameter	1.34	0.00	1.62	0.01	1.38	0.00	2.10	0.10	1.23	0.00	1.32	0.00	1.44	0.00	1.38	0.00	1.34	0.00	1.53	0.00
Launch Date	-0.04	0.00	-0.04	0.01	-0.04	0.01	-0.04	0.01	-0.06	0.00	-0.04	0.00	-0.04	0.00	-0.04	0.00	-0.04	0.00	-0.06	0.00
Third Variable	-	-	-0.24	0.51	-0.24	0.48	-0.24	0.51	0.03	0.83	-0.02	0.70	-0.03	0.90	-0.03	0.90	0.01	0.85	0.03	0.73
Adjusted r ²	93%		93%		93%		93%		78%		93%		92%		92%		93%		100%	
SPE	39%		43%		43%		43%		42%		41%		44%		44%		41%		50%	
n	17		13		13		13		13		16		15		15		17		11	
Multicollinearity?	No		Yes		No		Yes		No		No		Yes		No		No		No	
Third Variable			Operating Temperature		Avg. Input Power		Data Rate		Design Life		Design Life (exp)		Technology Readiness Level		Development Period		Dev Per (exp)		Orbit	
Aperture Diameter			1.34	0.00	1.27	0.00	1.28	0.00	1.30	0.00	1.28	0.00	1.33	0.00	1.34	0.00	1.35	0.00	1.36	0.00
Launch Date			-0.04	0.00	-0.05	0.00	-0.04	0.02	-0.04	0.00	-0.04	0.00	-0.01	0.28	-0.04	0.00	-0.04	0.00	-0.04	0.00
Third Variable			-0.02	0.79	0.02	0.83	0.02	0.78	0.07	0.57	0.00	0.43	-0.70	0.10	0.07	0.80	0.00	0.92	0.03	0.25
Adjusted r ²			93%		85%		88%		95%		96%		99%		93%		93%		88%	
SPE			40%		43%		38%		40%		40%		32%		40%		40%		35%	
n			16		16		12		17		17		8		16		16		15	
Multicollinearity?			No		No		No		No		No		No		No		No		No	

Figure 12: Three Variable Model Regression for OTA Cost vs Diameter, Launch Year and 3rd Parameter

	coef	p	OTA Cost vs Diameter, YoD, DoL, and Spct min									
	Diam		Diam, spct min		Diam, YoD(exp)		Diam, YoD(exp), spct min		Diam, DoL(exp)		Diam, DoL(exp), spct min	
Aperture Diameter	0.84	0.00	1.03	0.00	0.78	0.00	1.15	0.00	0.85	0.00	1.05	0.00
YoD	-	-	-	-	-0.03	0.12	-0.03	0.04	-	-	-	-
Launch Date	-	-	-	-	-	-	-	-	-0.03	0.08	-0.03	0.03
Spct Min	-	-	-0.13	0.00	-	-	-0.17	0.00	-	-	-0.13	0.00
Adjusted r ²	43%		69%		18%		92%		18%		63%	
SPE	126%		88%		97%		76%		99%		69%	
n	23		23		20		20		23		23	
Multicollinearity?	N/A		No		No		No		No		No	

Figure 13: Three Variable Model Regression for OTA Cost vs Diameter, 'year' and Spectral Minimum Wavelength

6. CONCLUSIONS

Cost models are invaluable for system designers. They identify major architectural cost drivers and allow high-level design trades. They enable cost-benefit analysis for technology development investment. And, they provide a basis for estimating total project cost. A study has begun to develop a multivariable parametric cost model for space telescopes. Cost and engineering parametric data has been collected on 30 different missions and extensively analyzed for 23 normal incidence UV/OIR space telescopes. Statistical correlations have been developed for 19 of the 59 variables sampled.

From an engineering & science perspective, Aperture Diameter is the best parameter for a space telescope cost model. But, the single variable model only predicts 75% of OTA Cost:

$$\text{OTA Cost} \sim D^{1.2} \quad (N = 17; r^2_{\text{adj}} = 75\%; \text{SPE}=79\%) \text{ with 2009 JWST}$$

Two and three variable models provide better estimates:

$$\text{OTA Cost} \sim D^{1.3} e^{-0.04(\text{LYr}-1960)} \quad (N = 17, r^2_{\text{adj}} = 93\%; \text{SPE}=39\%)$$

$$\text{OTA Cost} \sim D^{1.3} e^{-0.04(\text{YoD}-1960)} \quad (N = 16, r^2_{\text{adj}} = 95\%; \text{SPE}=39\%)$$

$$\text{OTA Cost} \sim D^{1.15} \lambda^{-0.17} e^{-0.03(\text{YoD}-1960)} \quad (N = 20, r^2_{\text{adj}} = 92\%; \text{SPE} = 76\%)$$

where: D = Aperture Diameter, LYr = Launch Year, YoD = Year of Development, and λ = Spectral Min Wavelength.

Similar results were obtained for OTA Areal Cost:

$$\text{OTA Areal Cost} \sim D^{-0.6} e^{-0.04(\text{LYr}-1960)} \quad (N = 17, r^2_{\text{adj}} = 76\%; \text{SPE}=40\%)$$

$$\text{OTA Areal Cost} \sim D^{-0.7} e^{-0.04(\text{YoD}-1960)} \quad (N = 16, r^2_{\text{adj}} = 76\%; \text{SPE}=39\%)$$

At present, no satisfactory model has been developed for Total Mission Cost. While total mass does yield a statistically significant result, it contradicts other findings, i.e. that JWST cost more than HST but has half the mass, and that attached OTAs with 10X more mass are 60% lower cost.

The primary conclusions of the cost modeling study to date are:

- The primary cost driver for Space Telescope Assemblies is Aperture Diameter.
- It costs less per collecting area to build a large aperture telescope than a small aperture telescope.
- Technology development as a function of time reduces cost at the rate of 50% per 17 years.
- If all other parameters are held constant, adding mass reduces cost.

REFERENCES

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