

NATIONAL RECONNAISSANCE OFFICE

# Weibull Analysis Method

Presented to the ICEAA Annual Symposium  
Denver, CO  
June 2014



SUPRA ET ULTRA

**Erik Burgess, Burgess Consulting**  
**James Smirnoff, Wyle**  
**Brianne Wong, Booz Allen Hamilton**



# Topics

---

- Analytical Basis
- Accuracy
- Application



# Weibull Analysis Method (WAM)

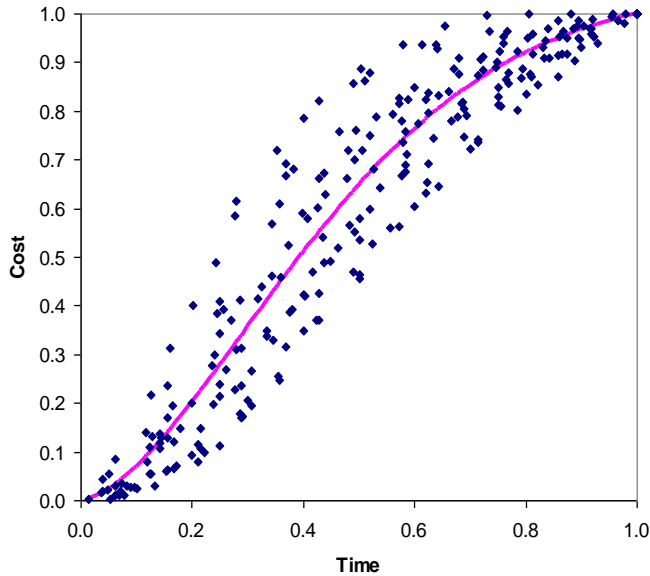
- Uses a program's actual history to estimate future budgets
  - Expenditures
  - Outlay rates
  - Government liability
- Improves accuracy over NRO's baseline parametric phasing model
  - For programs already underway
  - Especially in the near term
    - 2-years out (budget year)
    - FYDP
- Quantifies and reports error bounds based on historical data
  - Annual error vs. historical data
  - Departure from baseline model
- Builds on LMI<sup>1</sup> and CNA<sup>2</sup> research

<sup>1</sup> Dukovich, John et al., "The Rayleigh Analyzer." Logistics Management Institute AT902C1. October, 1999.

<sup>2</sup> Davis, Dan et al. "Using the Rayleigh Model to Assess Future Acquisition Contract Performance and Overall Contract Risk." Center for Naval Analysis CRM D0019289.A2, January 2009.



# NRO's Baseline Parametric Phasing Model<sup>3</sup>



Weibull plus a constant-rate term  
38 NRO & DoD Programs  
387 time-cost pooled data points

$$E(t) = d \left[ Rt + 1 - e^{-\alpha t^\beta} \right]$$

$$d = \frac{\text{total cost}}{R + 1 - e^{-\alpha}}$$

$$0 \leq t \leq 1.0$$

$$R = .002945 \cdot \text{duration (mos.)}$$

$$\alpha = 0.10 + \sum X_i \cdot \text{driver}_i$$

$$\beta = 1.539 + \sum Y_i \cdot \text{driver}_i$$

- Adjusts front/back-loading based on “phasing drivers”
- Starting point for all space-segment estimates
- Phases expenditures, converts to budget authority

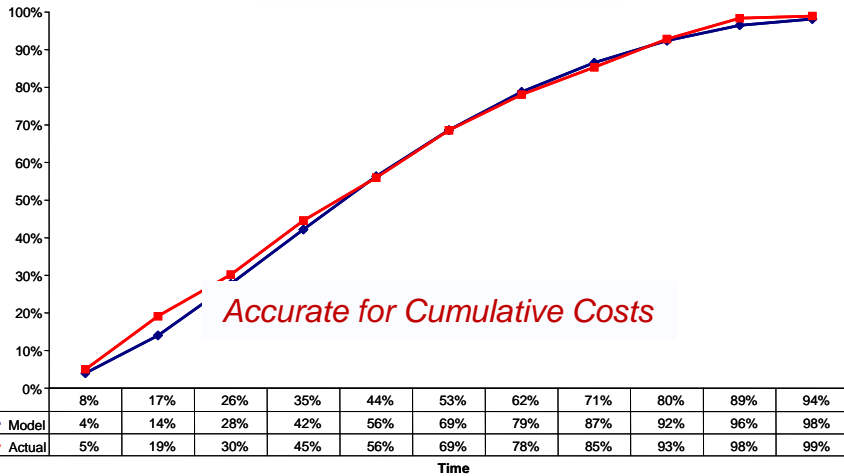
<u>Driver</u>	<u>Coefficient (X)</u>
GFE (1,0)	1.84E+00
% Subs	2.73E-02
BY07\$M	9.57E-04
Duration (mos)	2.79E-02

<u>Driver</u>	<u>Coefficient (Y)</u>
Competitive (1,0)	1.71E-01
GFE (1,0)	3.62E-01
% Subs	4.47E-03
BY07\$M	7.03E-05
Duration (mos)	-1.62E-03

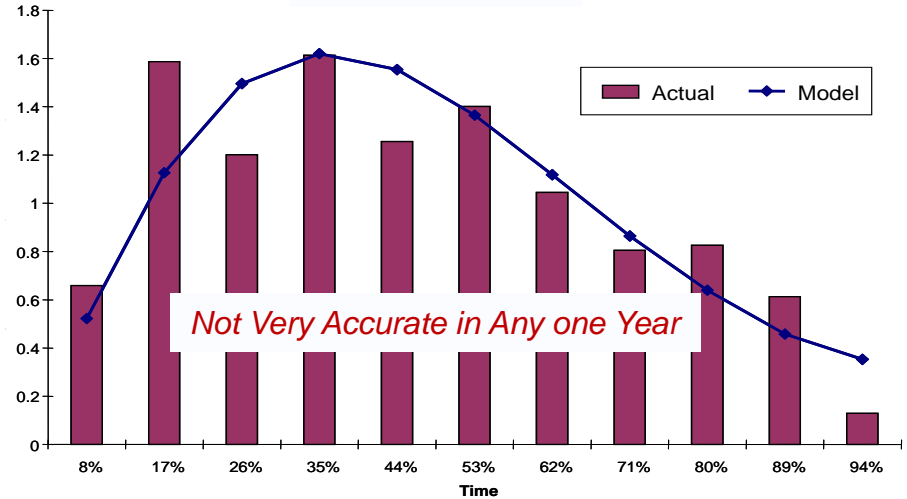


# Baseline Model: Not Very Accurate in Any one Year

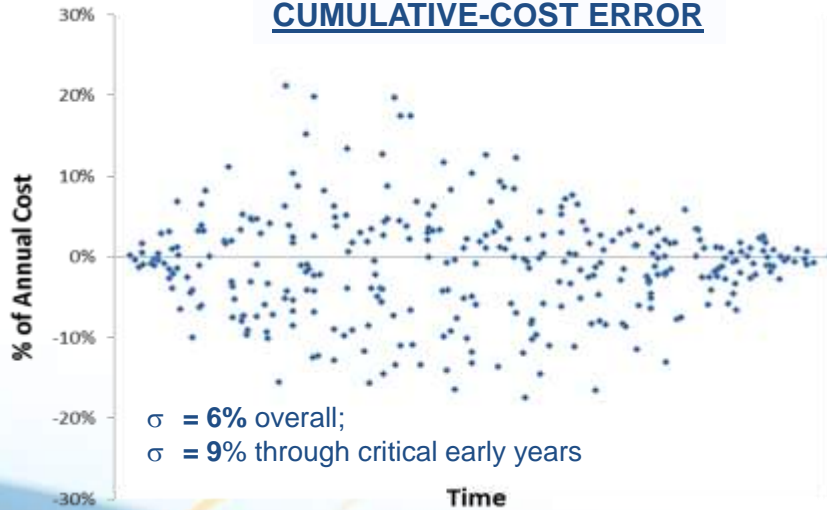
**CUMULATIVE COSTS**



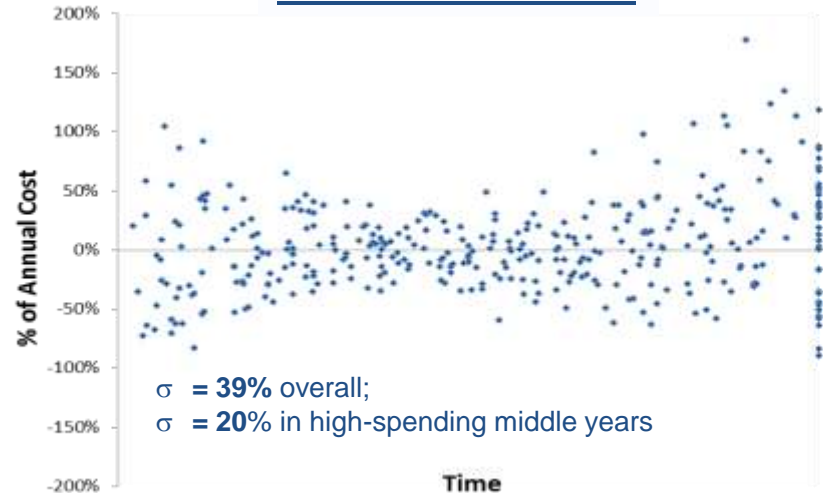
**ANNUAL COSTS**



**CUMULATIVE-COST ERROR**



**ANNUAL-COST ERROR**





# Motivation for WAM

## Baseline model establishes solid historical reference

- Cumulative accuracy through early years is quantified
- Powerful tool to link budget profile to schedule

But ...

- Mid and late-program assessments now occur every year
  - Comparing government estimate at complete (GEAC) to program-office plan
  - Search for margin
  - Re-phasing the ICE
- Better method needed for evaluating annual budgets
  - Baseline model not very accurate for annual costs, especially in later years
  - Unclear how to apply baseline model when prior-year actuals are different
  - Need a method based on actuals, not plans



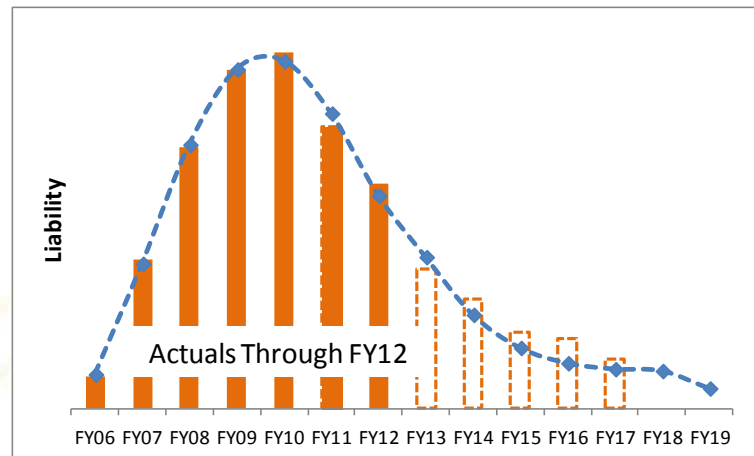
# WAM: Analytical Basis

- Functional form: Weibull plus constant-rate term
  - Same as baseline phasing model
  - Empirical and theoretical basis for satellite acquisitions<sup>4</sup>

$$E(t) = d \left[ R t + 1 - e^{-\alpha t^\beta} \right]$$

$$d = \frac{\text{total cost}}{R + 1 - e^{-\alpha}}, 0 \leq t \leq 1$$

- Use actual program performance to estimate Weibull parameters





# Basic WAM Process

- Input: Actual expenditures for each year to date, BY\$

- Constraints:

- Total cost in BY\$ (set to match ICE)
- Schedule (set to match ICE)
- Cumulative expenditures to date
- Constant-rate term from baseline model:  $R = .002945 * \text{duration}$

Used for re-phasing existing ICE.  
Unconstrained version also can be run.

- Optimization:

- For each year  $i$  of actual data:  $E(t_i) = d \left[ Rt_i + 1 - e^{-\alpha t_i^\beta} \right], d = \frac{\text{total cost}}{R + 1 - e^{-\alpha}}$
- Excel Solver<sup>®</sup> estimates Weibull parameters  $\alpha, \beta$  by minimization:

$$\min \sum_i \left( E(t_i) - \hat{E}(t_i) \right)^2$$

- Forecasting:

- Apply  $\alpha, \beta$  to project expenditures in remaining years
- Convert to TY\$ and compare to funding plans



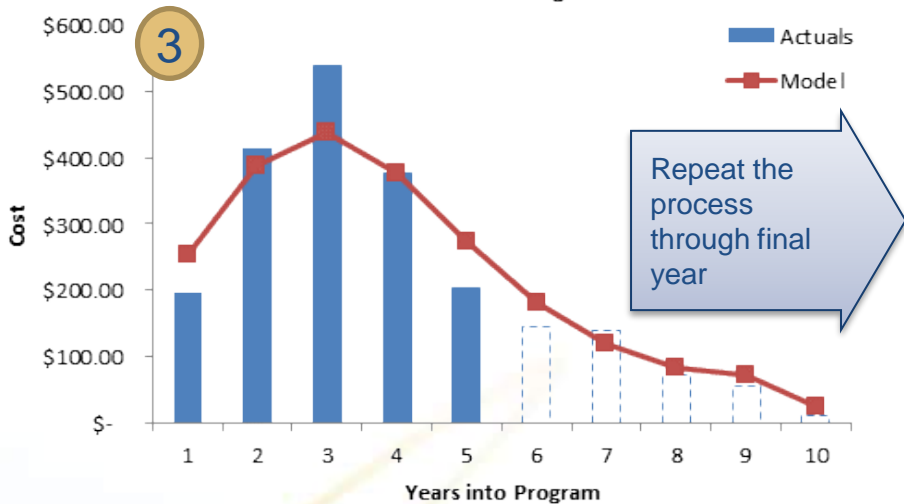
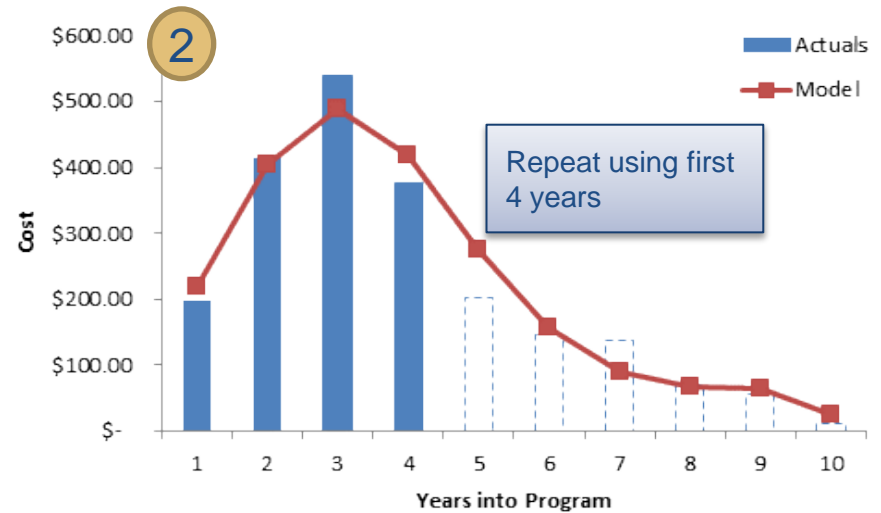
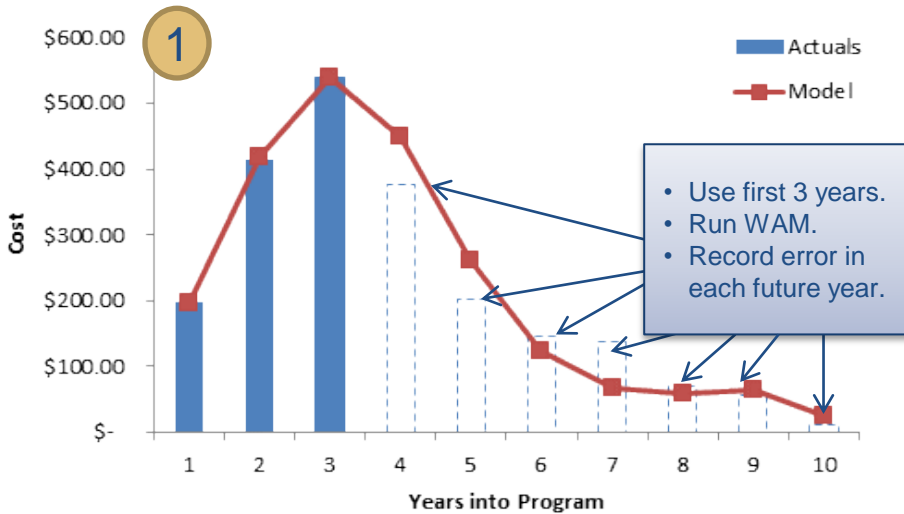


# Measuring the Accuracy of WAM

- Gather and normalize historical phased expenditure data from 38 completed contracts
- Use WAM to generate estimates of “future” time phased program expenditures starting from progressively further points in each program
- Compare the WAM predicted time phased expenditures to the actual time phased expenditures and measure the error of the prediction
- Create a model to characterize WAM accuracy
- Compare the accuracy of WAM to the accuracy of the baseline phasing model



# Generating Error Measurements

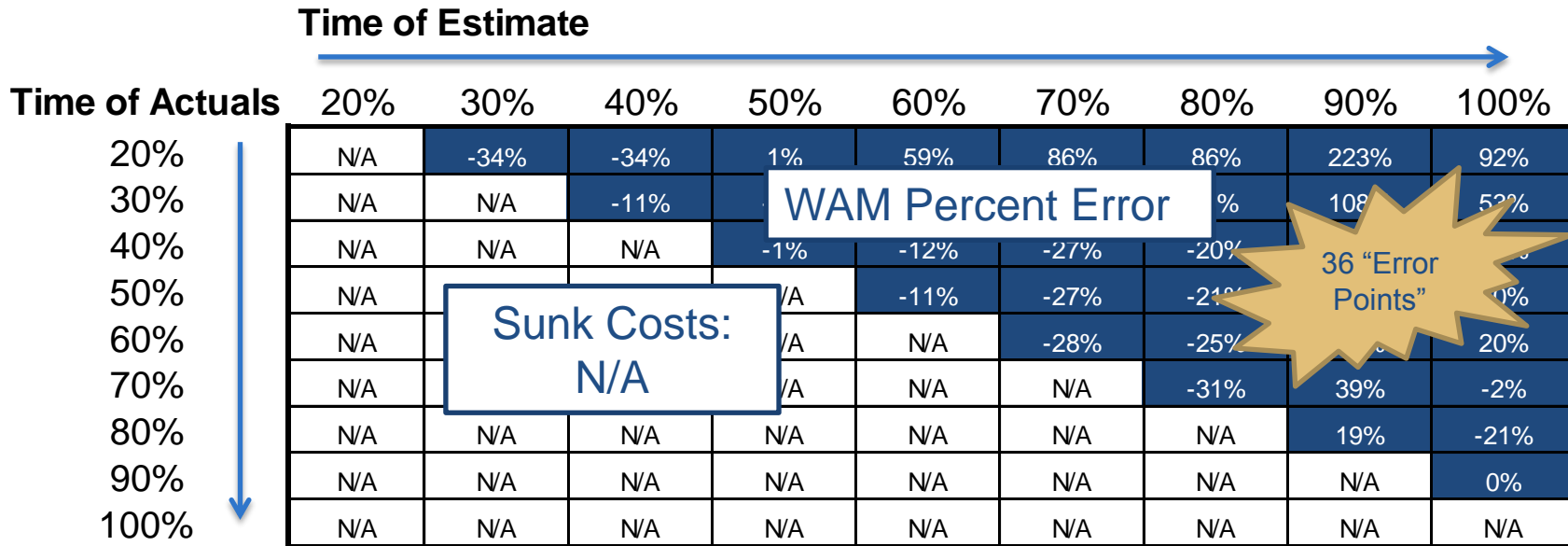


$$\text{Error, } \varepsilon_i = \frac{E(t_i) - \hat{E}(t_i)}{\hat{E}(t_i)}$$

Each program generated 20-40 measurements for a total of 1328 "Error Points"



# Results for One Contract



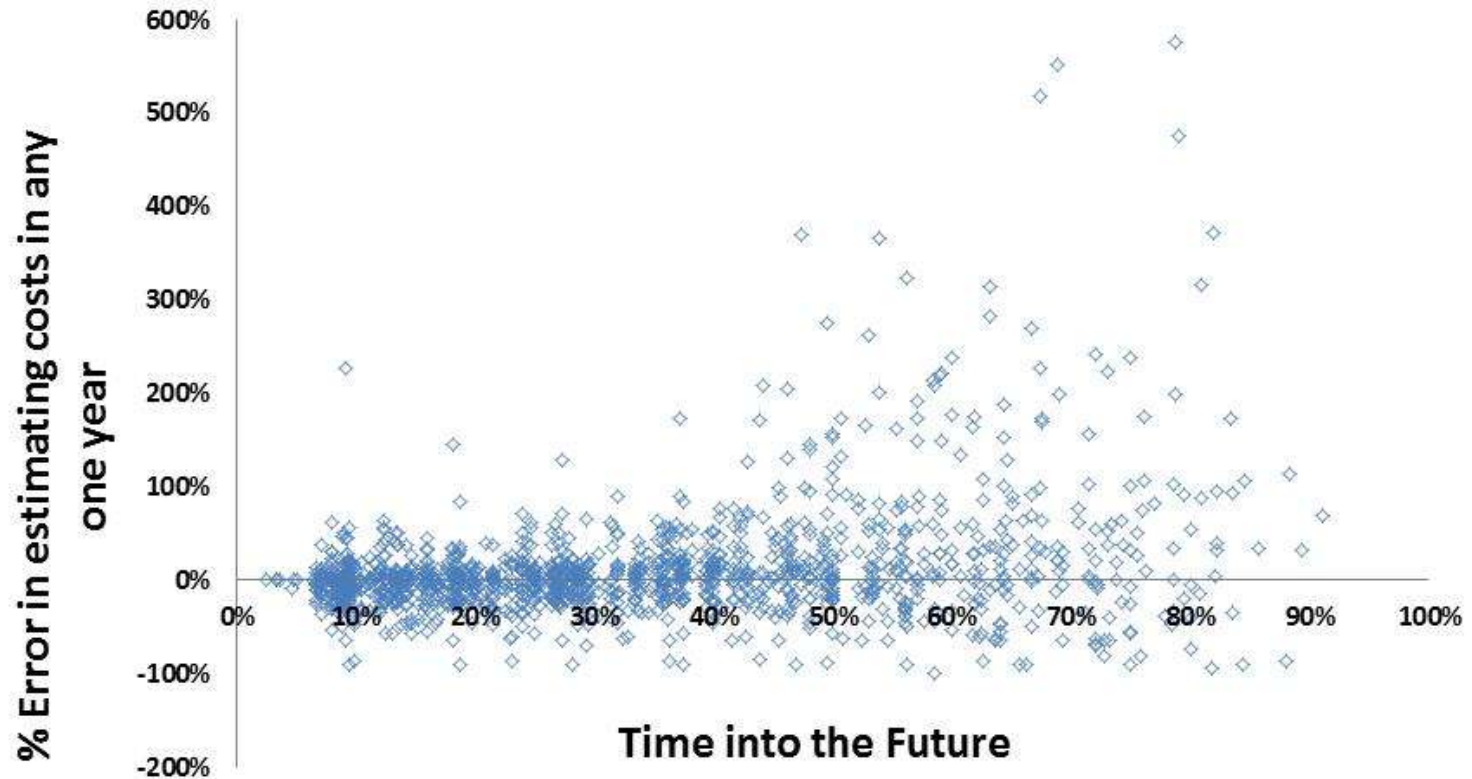
- Time of Actuals: Data at and before this time is used for WAM best fit
- Time of Estimate: Time in contract for which WAM is estimating the expenditure level

Each program generated 20-40 measurements for a total of 1328 "Error Points"



# Error Source 1

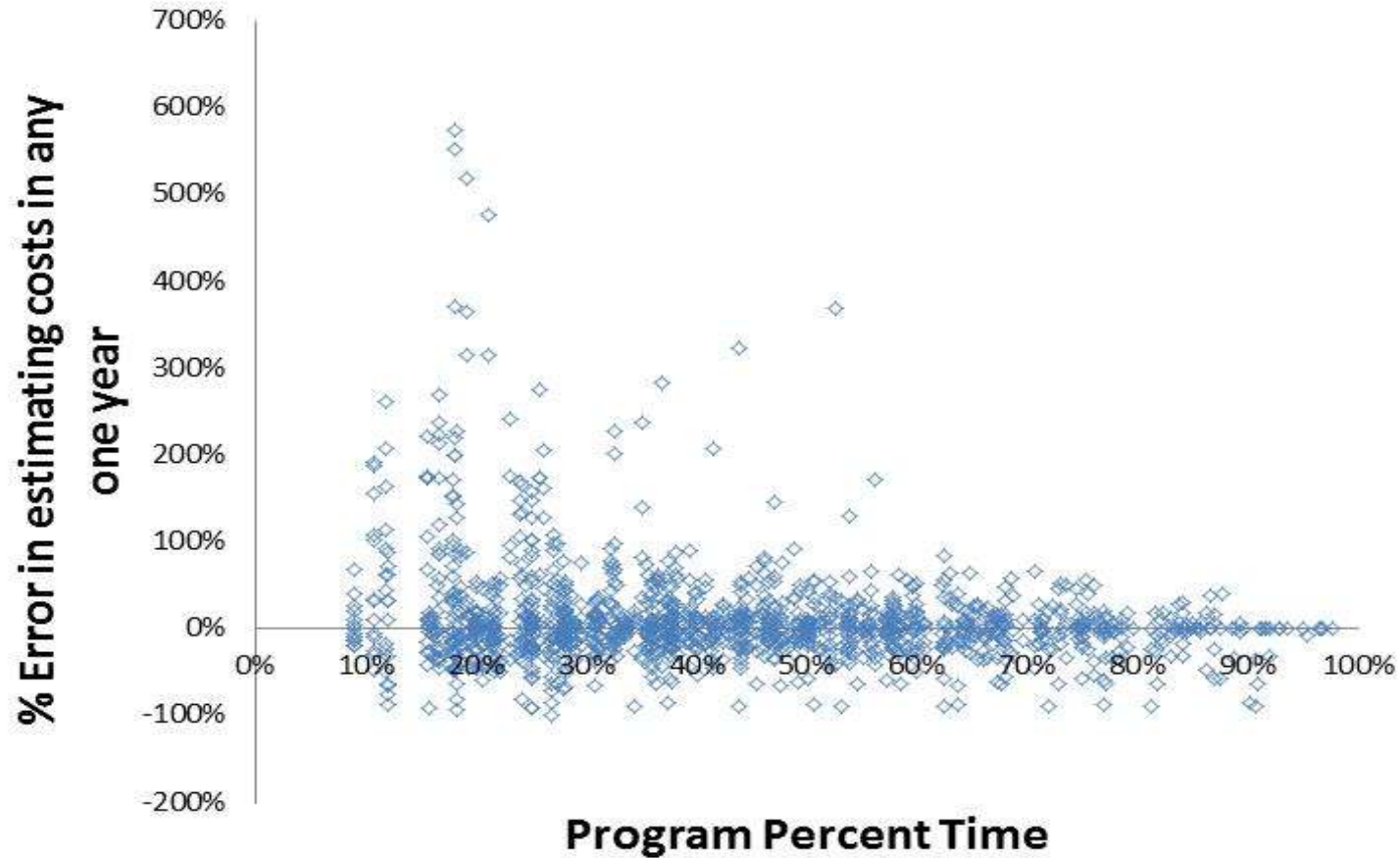
Error increases when projecting farther into the future





## Error Source 2

Error is greater when there are fewer years of data





# WAM Error Model

- All 1328 points used in OLS regression to estimate absolute % error

$$|\%Error| = 0.052 - 0.123 * T_A + 0.968 * T_F$$

$$T_A = \frac{\text{Time of last actual}}{\text{Total time}}$$

$$T_F = \frac{\text{Time into future}}{\text{Total time}}$$

Error increases with  $T_F$  and decreases with  $T_A$

- Both  $T_A$  and  $T_F$  are statistically significantly correlated with  $|\%Error|$

ANOVA					
	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	2	58.60	29.30	129.40	0.000
Residual	1325	300.01	0.23		
Total	1327	358.61			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>
Intercept	0.05	0.05	1.02	0.308
TA	-0.12	0.07	-1.67	0.094
TF	0.97	0.07	12.92	0.000

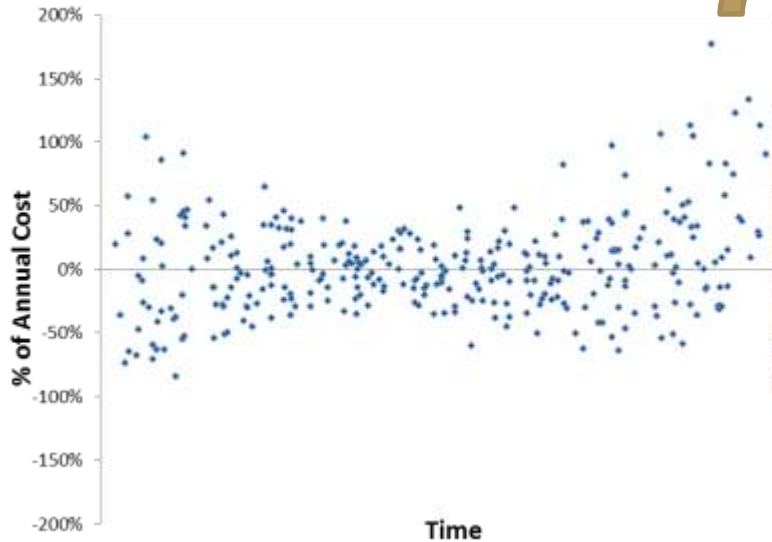
- Since the coefficient modifying  $T_F$  is much larger and the variable ranges are similar,  $T_F$  has much more impact on WAM error

**WAM Error is better when (1) the contract is father along, and (2) projecting near-term spending.**

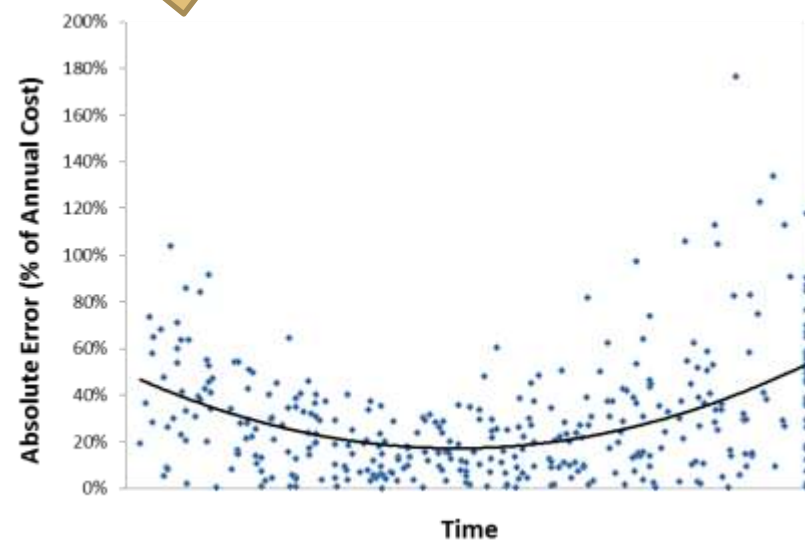


# Absolute Error of Baseline Phasing Model

**BASELINE MODEL:  
Annual-Cost Error**



**BASELINE MODEL:  
Absolute Error**

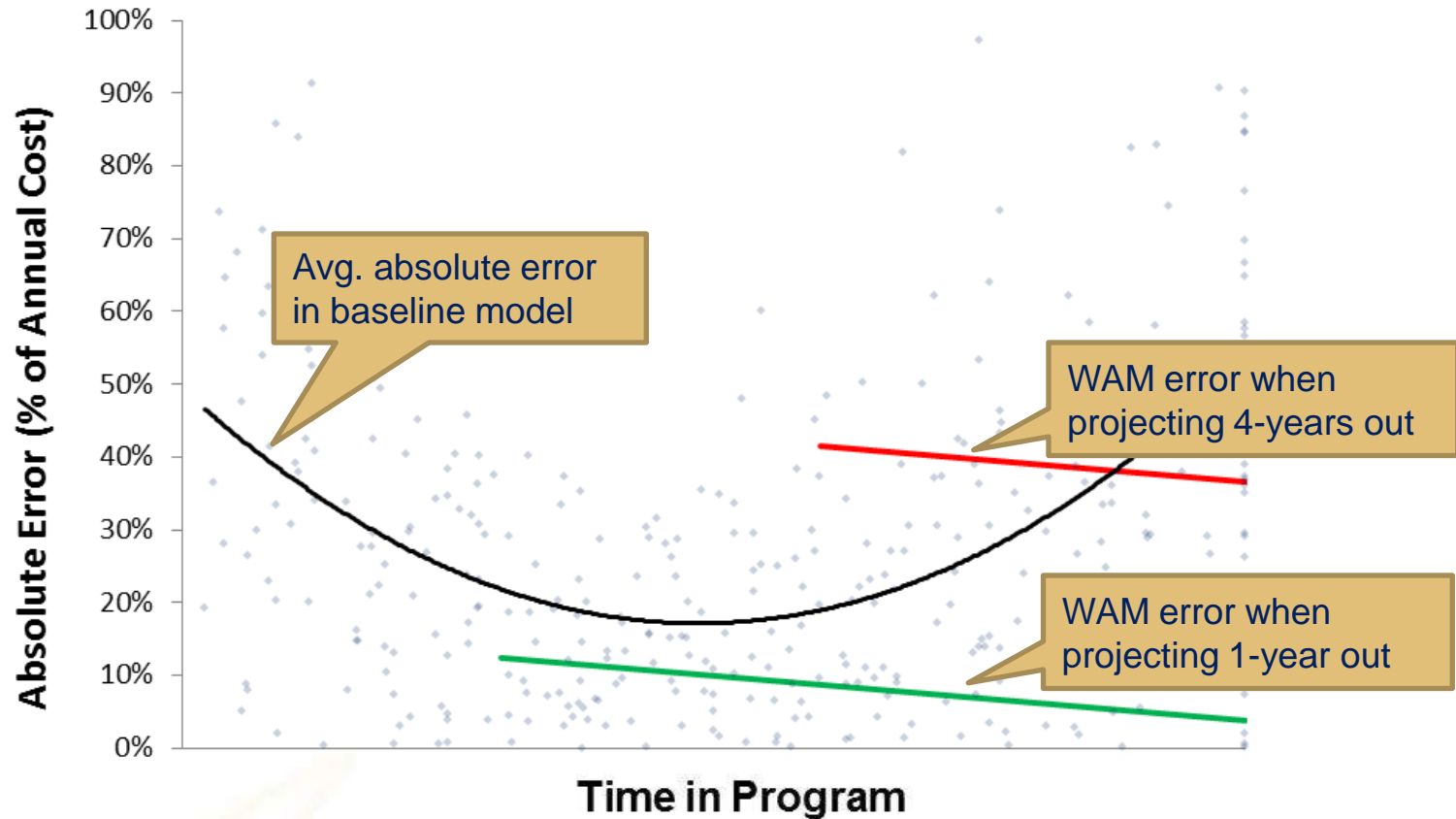


**Absolute Error of Baseline Model is lower in the middle of a program when expenditures are high**





# Comparison to Baseline Model



**WAM is a lot better at estimating program budgets in the near term and not as good at estimating far into the future**





# Weibull Analysis Tool (WAT)

**Weibull Analysis Tool Feb-2013**  
Notional Data

For Analysis Tool  
Date for WAT

01/2008 2017 Date  
619,2416 Last Launch Date  
12K Months Available for Test Search  
1,045 Total Space Segment Cost (BY\$)  
2012 Base Year

7 (Optimistic based (Year1, Year1)  
4 Priority segment in SPE (Year1, Year1)  
8% % Subcontracted

None (Year1, Year1)  
1000 (Year1, Year1)  
1000 (Year1, Year1)

Expenses  
Commitment By  
2006 2007 2008 2009 2010 2011

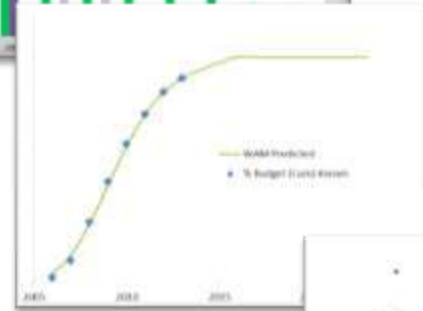
2006 2007 2008 2009 2010 2011  
2006 2007 2008 2009 2010 2011

2006 2007 2008 2009 2010 2011  
2006 2007 2008 2009 2010 2011



Outlay rates

	Year 1	Year 2	Year 3	Year 4	Year 5
Default	68.10%	74.67%	4.07%	1.71%	1.13%
2006	62.00%	40.29%	4.07%	1.71%	1.13%
2007	68.07%	33.22%	4.07%	1.71%	1.13%
2008	63.99%	26.12%	4.07%	1.71%	1.13%
2009	66.98%	26.11%	4.07%	1.71%	1.13%
2010	68.34%	23.73%	4.07%	1.71%	1.13%
2011	71.21%	21.78%	4.07%	1.71%	1.13%
2012	73.08%	20.89%	4.07%	1.71%	1.13%
2013	74.43%	18.60%	4.07%	1.71%	1.13%
2014	75.92%	17.23%	4.07%	1.71%	1.13%



**RUN SOLVER**

The Objective Function (a measure of success) is the sum of annual differences from the known L&C to the predicted. Weibull Solver. This is the target cell that solver will minimize.

Constraint Function: 0.001414042

Constraint: 0%

Solver Output: 0.670, 0.682

Alternatively, you may manually run solver by going to the Data Tables and clicking Solver under Analysis. If done this way, you will be able to track the complete history for the entire model.

	2006	2007	2008	2009	2010	2011	2012	2013	2014
Notional Budget Data	\$ 70	\$ 251	\$ 564	\$ 593	\$ 202	\$ 145	\$ 243	\$ 177	\$ 137
Error vs. WAM (% cum)	-7%	-13%	-12%	-9%	-16%	-22%	-22%	-22%	-22%
Error vs. WAM (% annual)	-24%	-45%	6%	17%	-53%	-56%	0%	0%	0%
WAM Error model (% annual)	N/A	N/A	N/A	N/A	N/A	N/A	7.7%	16.9%	26.2%
Projected BY\$	\$ 79	\$ 275	\$ 599	\$ 617	\$ 207	\$ 146	\$ 241	\$ 173	\$ 130
BY% (Annual)	2%	8%	10%	10%	6%	4%	7%	5%	4%
Baseline Phasing Model	\$ 355	\$ 584	\$ 588	\$ 496	\$ 356	\$ 255	\$ 176	\$ 130	\$ 108
vs. Phasing Model (% cum BY\$)	10%	20%	21%	18%	21%	26%	24%	22%	22%



**Implements WAM for NRO Estimators**



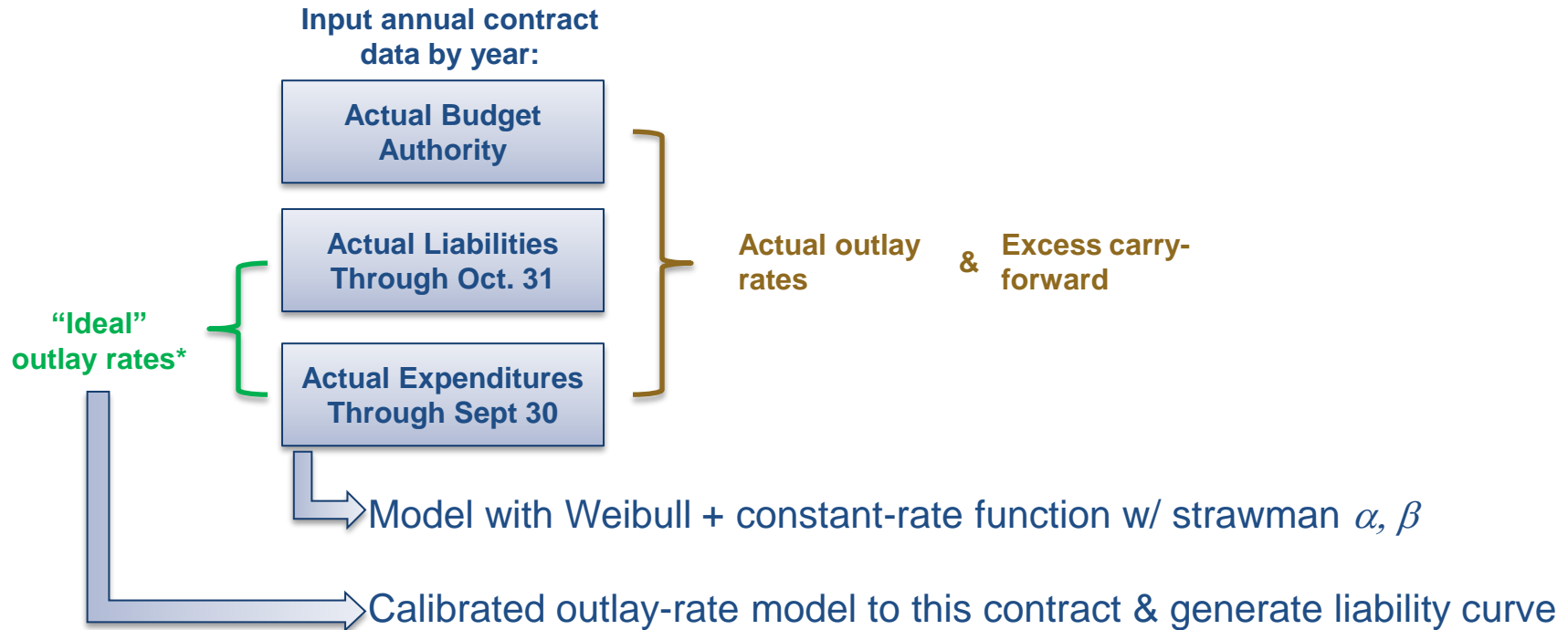
# WAT Goals

- Tool for NRO estimators
  - Apply WAM as repeatable part of estimating process
  - Excel-based, easy to integrate and modify
- Accepts and forecasts all relevant contract data
  - Expenditures
  - Government liability
  - Budget authority
  - Carry-forward
  - Actual program outlay rates
- Compare WAM result to:
  - Existing budget line
  - Program plan (CFSR)
  - Baseline phasing model

**Are they within WAM error bounds?  
Is there excess margin in any year?**



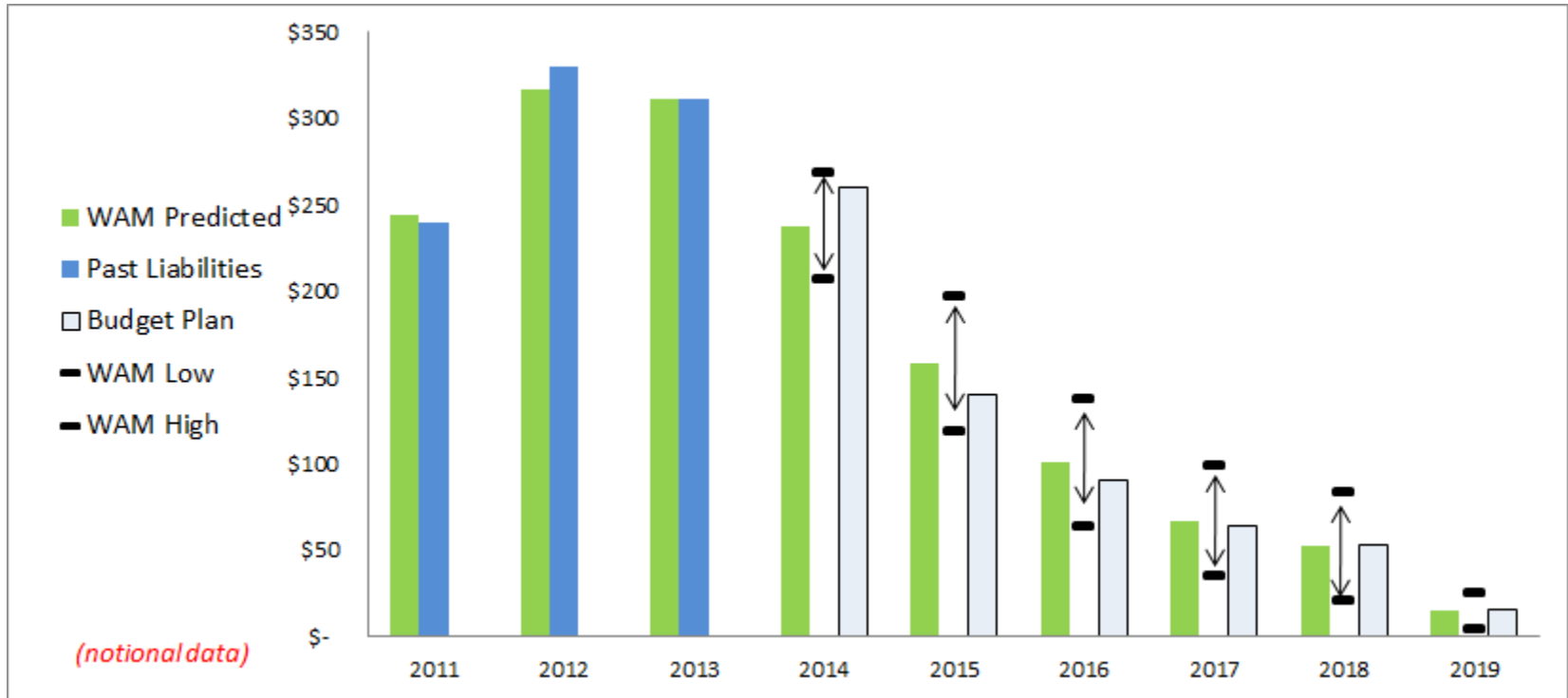
# Overview of WAT Mechanics



- Solve for  $\alpha, \beta$  to fit liability curve through current year
  - Apply constraints, including budget already programmed
- Project future liabilities
  1. Hold excess carry-forward as margin
  2. Assume excess carry-forward eliminated in next budget year



# Application Example



*(notional data)*

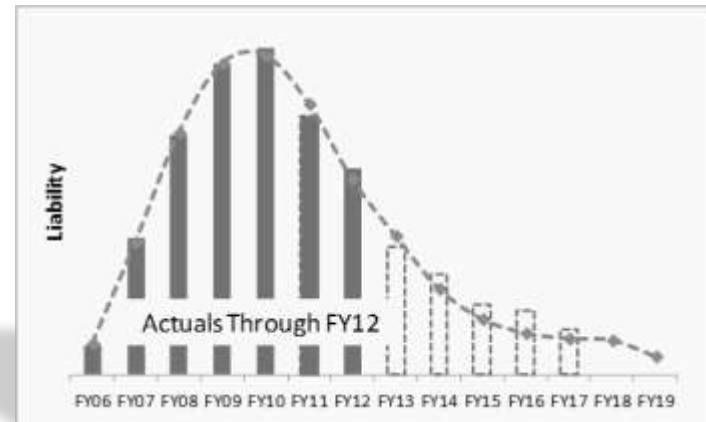
*Notional Data*

	2011	2012	2013	2014	2015	2016	2017	2018	2019
WAM Output (TY\$)	\$ 244	\$ 316	\$ 311	\$ 238	\$ 158	\$ 101	\$ 67	\$ 53	\$ 15
Past and Projected (TY\$)	\$ 239	\$ 330	\$ 311	\$ 260	\$ 140	\$ 90	\$ 64	\$ 53	\$ 15
Error vs. WAM (% annual)	-2%	4%	0%	9%	-12%	-11%	-5%	0%	0%
WAM Error model (% annual)	N/A	N/A	N/A	13%	25%	36%	48%	60%	71%
Baseline Phasing Model (TY\$)	\$ 239	\$ 310	\$ 311	\$ 241	\$ 162	\$ 104	\$ 69	\$ 54	\$ 15
Δ from Phasing Model (% cum)	0%	1%	1%	3%	1%	0%	0%	0%	0%



# Summary

- WAM is a useful addition to NRO's estimating toolkit
- Serves as alternative to baseline phasing model
  - More accurate in near years
  - Calibrated to program-specific outlay patterns
- WAT integrates analysis of several key metrics
  - Expenditures
  - Outlay rates
  - Government liability
  - Budget authority
  - Carry forward



# NATIONAL RECONNAISSANCE OFFICE

SUPRA ET ULTRA





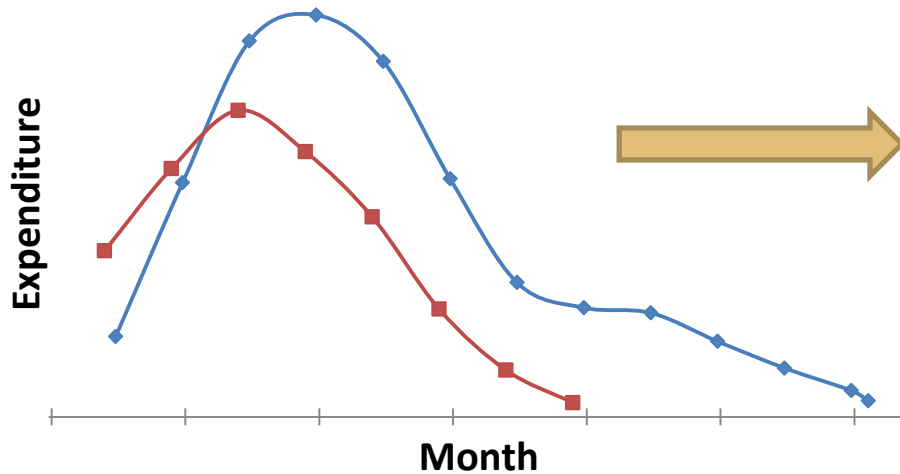
# Backup



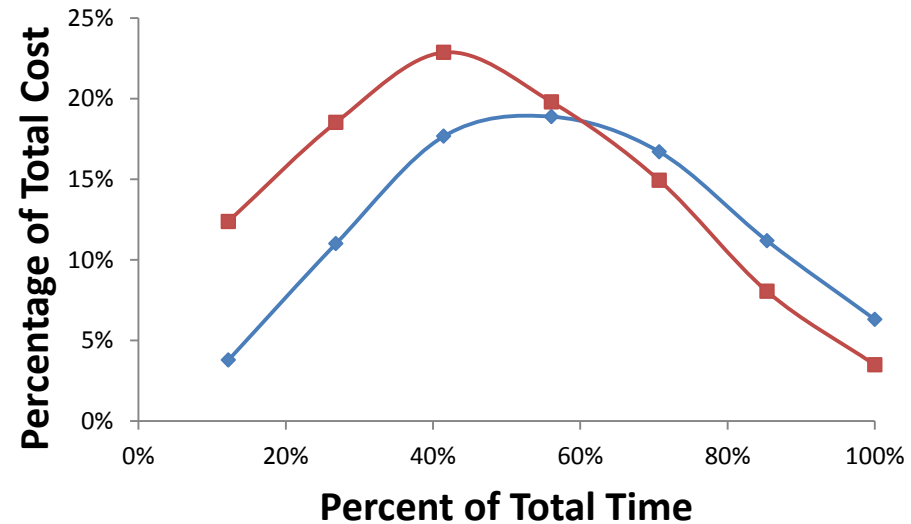
# Normalizing the Data



Cost and time are normalized so profiles can be compared

Not Normalized



Normalized



-  Longer More Expensive Program
-  Shorter and Cheaper Program