



**TECOLOTE
RESEARCH, INC.**
Bridging Engineering and Economics
Since 1973

Real Data, Real Uncertainty

Alfred Smith Tecolote Research
Jeff McDowell Tecolote Research
Dr. Lew Fichter Tecolote Research
27 June 2012

Outline

- **CRUAMM Purpose**
- **CRUAMM Process**
- **Illustrative Result**
- **CRUAMM Final Report**



The Purpose of CRUAMM

- **If you are provided a method without uncertainty information, rather than guess or using one of the AFCAA CRUH defaults, locate the method and its input in CRUAMM for a representative uncertainty distribution**
- **You can also use CRUAMM to cross check the uncertainty you derive from your data or your other sources**
- **CRUAMM does not supplant existing best practices**
 - If you have the data to build a CER, use it to develop the prediction interval for use in your model
 - If you are provided a CER complete with uncertainty information, use them in your model



Expressing Uncertainty

Given:

Cost Element Point Estimate = **Your Methodology**

Its uncertainty can then be expressed:

Cost Element Uncertainty = $f(\text{Your Methodology, Distribution Shape, Distribution Parameters})$

With CRUAMM, its uncertainty can then be expressed:

Cost Element Uncertainty = **Your Methodology * Unitized Distribution**



Orders of Dispersion

CRUAMM

■ First Order

- Amount of spread in a single parameter of a dataset
- Example: cost elements, cost drivers

■ Second Order

- Amount of spread in a factor relationship
- Example: Non Recurring as a factor of T1

■ Third Order

- Amount of spread in a the error term of a best fit equation
- Example: Confidence interval of a CER derived from regression

■ Fourth Order

- Prediction interval from a statistically derived CER
- Example: CER prediction interval for a given independent variable



Distribution Fit Process

■ Collect and normalize the data

- Sort sample data in ascending order
- Assign a cumulative percentile using the NIST¹ formula (different than Excel) but apply a “correction for continuity”² $(0.5 * \text{ObsFreq} + \text{NumObsBelow}) / \text{ObsCount}$
- Use the sample descriptive statistics to provide a starting point for fit parameters

■ Assess the difference between the sample and fit using either:

- Sum squared error (SSE) - highly influenced by very large sample points (as compared to mean)
- Sum squared percent error (SSPE) - highly influenced by fitted points close to zero (divide by zero)

■ Use *Excel Solver or ACEIT CO\$TAT to find the fit parameters that minimize SSE or SSPE

- Set optional constraints such as: $\text{Low} > 0$, $\text{Low} < \text{LowestSamplePoint}$, $\text{High} > \text{HighestSamplePoint}$

■ Rank the fits using standard error of the estimate (SEE) or standard percent error (SPE)

- This is a preferred method to rank the fits (rather than SSE or SSPE directly) because it accounts for the degrees of freedom
- Accounts for number of parameters required to uniquely define a distribution
 - 2 for Normal, lognormal, 3 for Triangular, 4 for Beta
- Many tools do not account for the number of parameters as they are designed to address fits on hundreds or thousands of data points. The difference between 1000, 997 and 995 degrees of freedom (on 1000 data points) has no impact on SEE but for 10 data points, the difference between 10, 7 and 5 degrees of freedom can be dramatic.

■ Use a Goodness-of-Fit test (Chi²) to determine significance of the fit

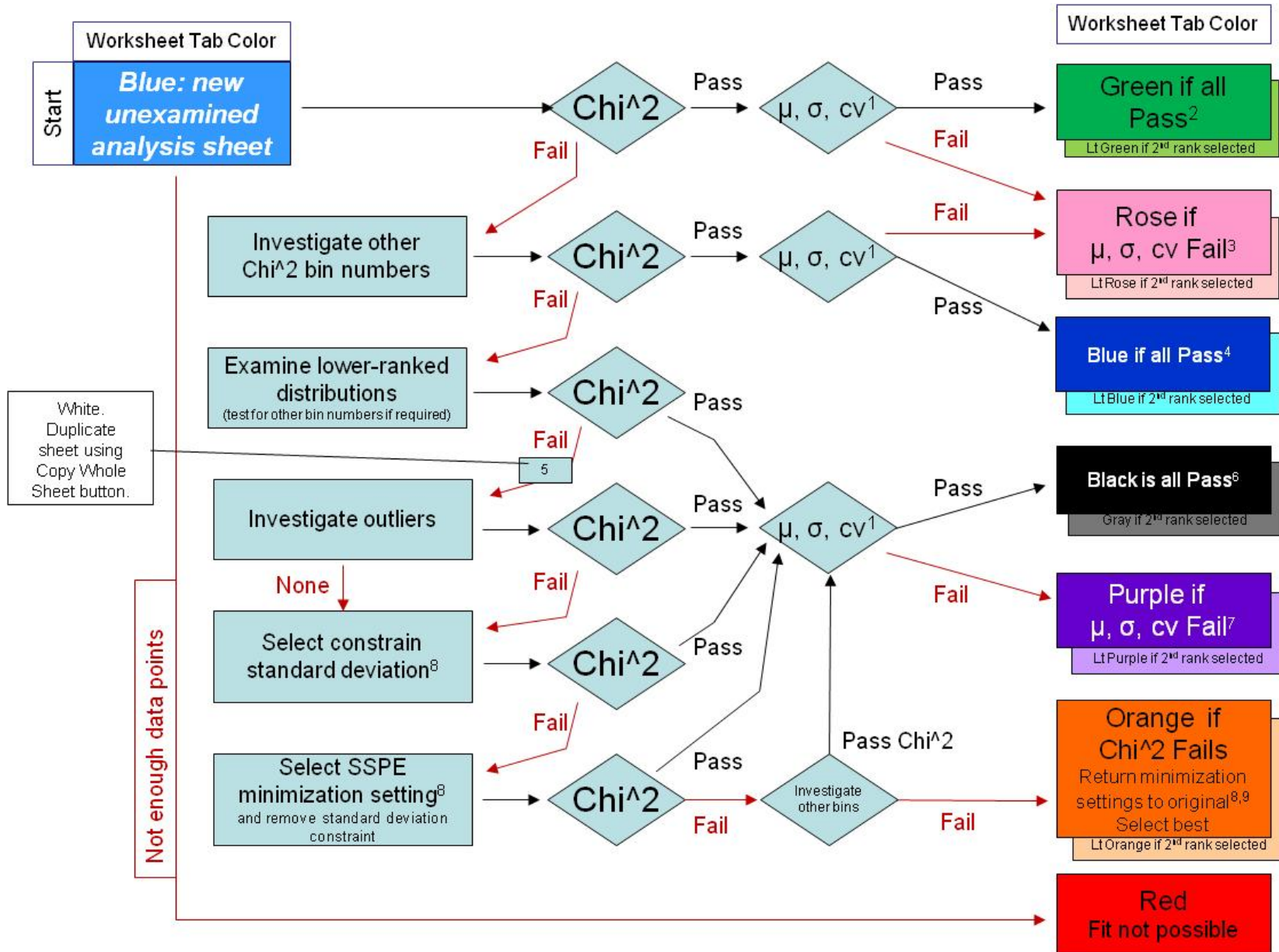
- Minimum SSE or SSPE alone does not necessarily mean the fit is meaningful
- Other tests (K-S, AD) cannot be applied to every distribution. Only Chi² provides a p-value for any distribution

1. NIST= National Institute of Standards and Technology

2. From "Reliability and Information Functions for Percentile Ranks" Kim May and W. Alan Nicewander, Journal of Educational Measurement, Vol. 31, No. 4 (Winter, 1994), pp. 313-325

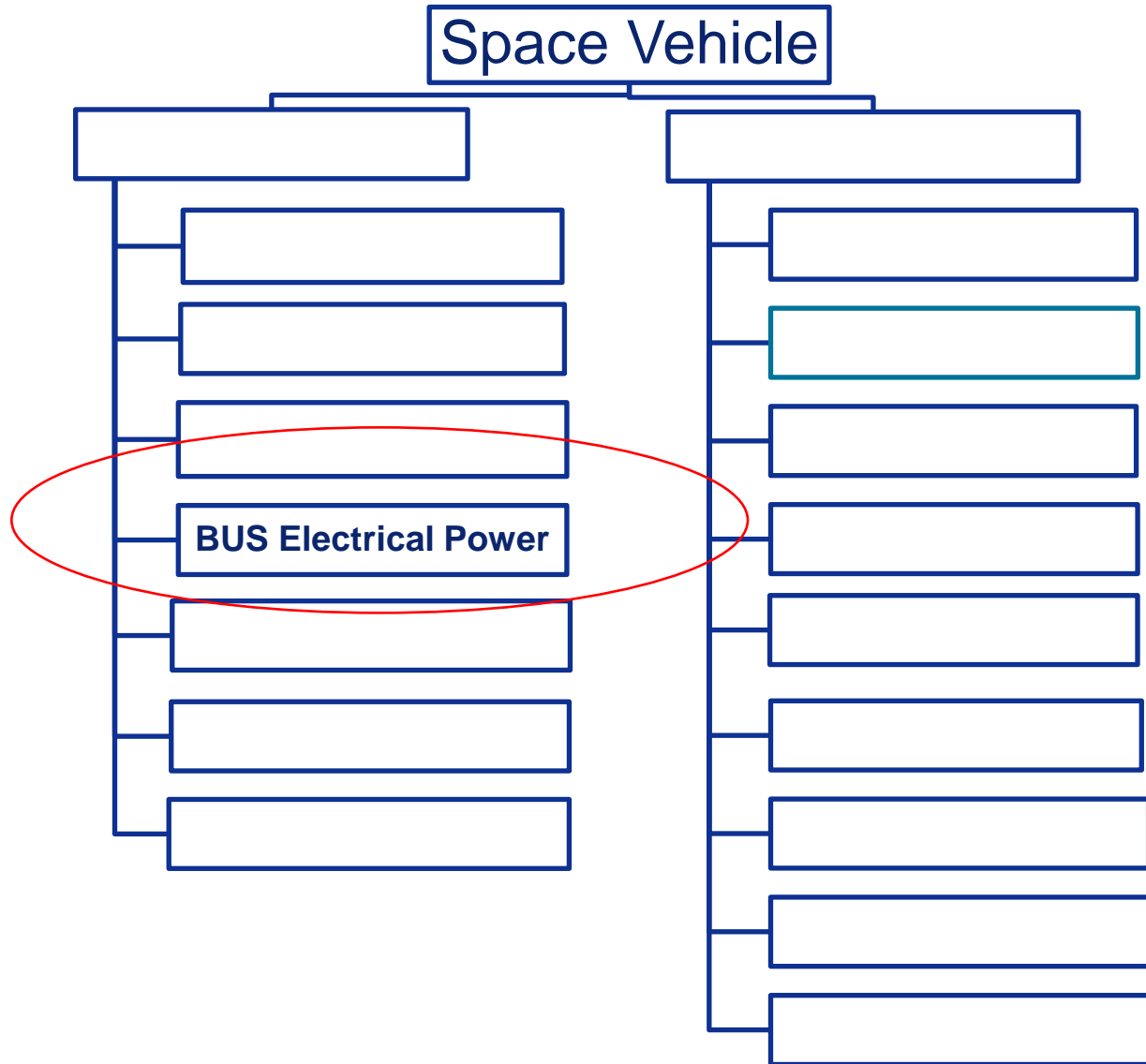


CRUAMM Results are Based on a Consistent Process for Evaluating Distribution Fits





Illustrative Results for One Subsystem



Analyst Wants to Use the FireSat CER for EPS, but needs Uncertainty Guidance

- Use CRUAMM to identify representative uncertainty for:
 - the Electrical Power Subsystem CER
 - the EPS CER independent variable (EPS Weight)

WBS Description	Unique ID	CER	Point Estimate	Forecast (PE * Uncert)	Distribution	Uncertainty	Min	Max or Std Dev
Satellite			\$298,647	\$298,647		Mean or Mode		
Payload	RDPayload		\$168,462	\$168,462				
IR Sensor		$356851 * \text{PayloadDiam}^{0.562}$	\$168,462	\$168,462	Triangular	1	0.071	1.9640
Spacecraft Bus	RDSpBus		\$15,096	\$15,096				
Structure		$157 * \text{StrWgt}^{0.83}$	\$2,787	\$2,787	Triangular	1	0	1.9528
Thermal		$394 * \text{ThermWgt}^{0.635}$	\$1,331	\$1,331	Triangular	1	0	1.9528
Electrical Power System (EPS)		$62.7 * \text{EPSWgt}$	\$2,865	\$2,865		1		
Telemetry Tracking & Command (TT&C) and Command & Data Handling (C&DH)		$545 * \text{TTCWgt}^{0.761}$	\$2,344	\$2,344	Lognormal	1		0.2941
Attitude Determination & Control Sys		$464 * \text{ADCSWgt}^{0.867}$	\$5,769	\$5,769	Lognormal	1		0.4231
Int & Assy		$989 + 0.215 * (\text{RDPayload\$} + \text{RDSpBus})$	\$40,454	\$40,454	Normal	1		0.3544
Program Level Costs		$1.963 * (\text{RDPayload\$} + \text{RDSpBus})^{0.841}$	\$52,451	\$52,451	Lognormal	1		0.4609
Ground Support Equip (GSE)		$9.262 * (\text{RDPayload\$} + \text{RDSpBus})^{0.642}$	\$22,184	\$22,184	Triangular	1	0	3.1204

Input Variable	Unique ID	CER	Point Estimate	Forecast	Distribution	Uncertainty	Min	Max or Std Dev
Payload Sensor diameter (m)	PayloadDiam		0.263	0.263	Lognormal	1		0.3000
Structure weight (kg)	StrWgt		32.000	32.000	Lognormal	1		0.5019
Thermal weight (kg)	ThermWgt		6.800	6.800	Triangular	1	0.261	2.7028
EPS weight (kg)	EPSWgt		45.700	45.700		1		
TT&D/DH weight (kg)	TTCWgt		6.800	6.800	Normal	1		0.1893
ADCS weight (kg)	ADCSWgt		18.300	18.300	Normal	1		0.3103
Spacecraft + Payload Weight (kg)	SpPyWgt		140.000	140.000	Lognormal	1		0.2307

All examples use fictitious data



Sample Extract From CRUAMM Public Domain Table

- The “unitized” results are intended for public domain and are intended for use in “your” model
 - uncertainty scales with “your” point estimate

Parameter	Filter	The Point Estimate is:		
		Mean	Median	Mode
T1 CER ¹	None	Lognormal (1.0000, 0.4512)	Lognormal (1.0971, 0.4950)	Lognormal (1.3204, 0.5958)
T1 CER ³	None	Triangular (0.1552, 0.7265, 2.1183)	Triangular (0.1635, 0.7651, 2.2308)	Triangular (0.2137, 1.0000, 2.9158)
NR CER ³	None	Triangular (0.5206, 0.7584, 1.7210)	Triangular (0.5418, 0.7892, 1.7911)	Triangular (0.6865, 1.0000, 2.2694)
Weight Growth ²	None	Triangular (0.5568, 0.6370, 1.8062)	Triangular (0.5850, 0.6692, 1.8975)	Triangular (0.8742, 1.0000, 2.8357)
	>50% New	Triangular (0.5853, 0.7180, 1.6967)	Triangular (0.6092, 0.7474, 1.7662)	Triangular (0.8151, 1.0000, 2.3631)
	<50% New	Triangular (0.5043, 0.6168, 1.8789)	Triangular (0.5316, 0.6502, 1.9805)	Triangular (0.8176, 1.0000, 3.0461)
T1 / Weight ³	Commercial	Triangular (0.3442, 0.6256, 2.0302)	Triangular (0.3649, 0.6631, 2.1520)	Triangular (0.5503, 1.0000, 3.2454)
	Government	Lognormal (1.0000, 0.7077)	Lognormal (1.2251, 0.8670)	Lognormal (1.8387, 1.3013)
NRE / Weight ³	New	Triangular (0.0283, 0.1925, 2.7792)	Triangular (0.0316, 0.2155, 3.1103)	Triangular (0.1468, 1.0000, 14.4348)
	Modified	Lognormal (1.0000, 0.9282)	Lognormal (1.3644, 1.2663)	Lognormal (2.5397, 2.3573)
NRE / T1 ³	New	Triangular (0.0000, 0.1905, 2.8095)	Triangular (0.0000, 0.2136, 3.1495)	Triangular (0.0000, 1.0000, 14.7445)
	Modified	Lognormal (1.0000, 1.3844)	Lognormal (1.7078, 2.3642)	Lognormal (4.9806, 6.8950)

See Previous Chart for Filter Information

These footnotes identify the data source for each fit on this particular chart:

- 1 – CRUAMM – AFCAA
- 2 – CRUAMM – NRO
- 3 – CRUAMM - USCM

For this example, the CER produces the mean and the weight input is assumed to be the mode.

All examples use fictitious data



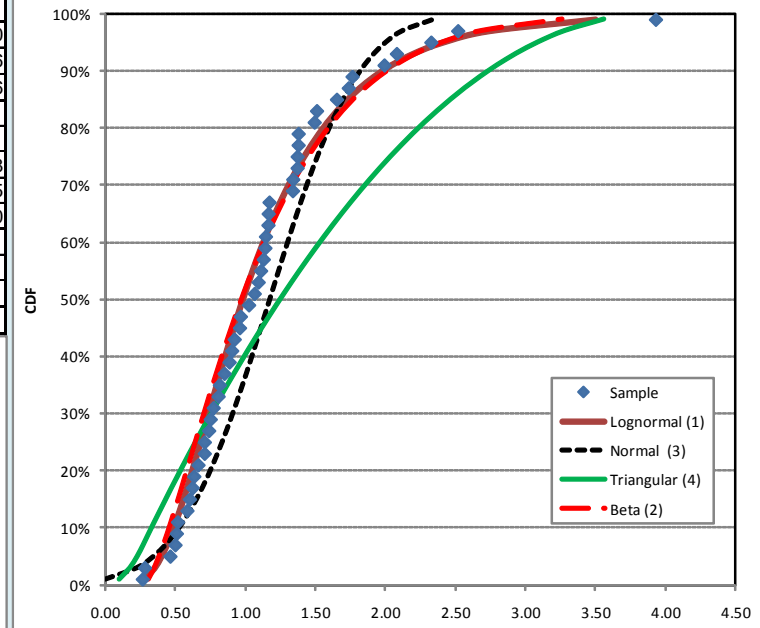
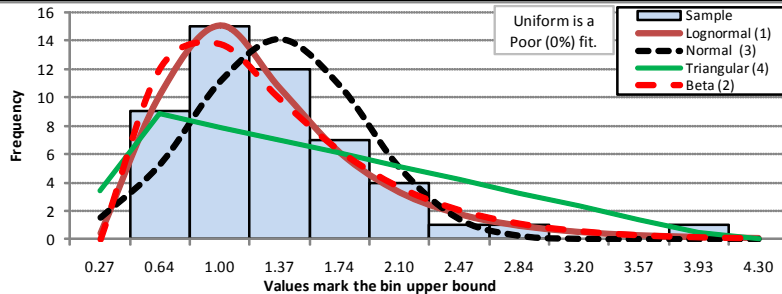
WBS:	Electrical Power	Cost Driver:	T1 CER (\$K)
CER:	220.0 + Bus Electrical Power Weight*13.0 + CCL * 6000.0		CCL = 0; If Commercial CCL= 1; If Government
Dist Form:	Actual/Predicted		
Best Fit	Lognormal (Mean =1.1388, Std Dev = 0.6722)		
Dependent	753 – 4087 6\$K	Independent	94 – 1969 lbs

Cost Risk and Uncertainty Analysis Data Report

Commodity	Space	Dispersion	Third
WBS	Residuals	Units	
Element	EPS	Data Comment	EPS T1 = 220 + EPS Weight(lbs) * 13 + CCL * 9000
Sub-Element	T1 Rec	Phase	
Category	EPSRecT1Residuals	Comment	
Filter and Fit Settings	Filter= NONE : , Fit = SSE on Value, Percentile = CoC, Using: CO\$TAT Engine		
Fit Constraints	Force Min => 0: Nor TRI Beta Force Include Sample Min/Max: TRI Beta		
Recommended Fit	Lognormal (Mean = 1.14, Std Dev = 0.67)		

Dataset Statistics and Fitted Distributions

Sample		Sample	Lognormal	Normal	Triangular	Beta	
Can be used to define a CDF directly							
Percentile	Value	Mean	1.1408	1.1388	1.1760	1.4015	1.1320
0%	0.2623	StdDev	0.6413	0.6722	0.5055	0.8972	0.6526
5%	0.4677	CV	0.562	0.590	0.430	0.640	0.576
10%	0.5174	Min	0.2701			0.0000	0.2701
15%	0.6067	Mode		0.7272	1.1760	0.2701	0.6171
20%	0.6529	Max	3.9345			3.9345	17.2398
25%	0.7129	Data Count	50			Alpha	1.605
30%	0.7658					Beta	30.000
35%	0.8204	Rank	1	3	4	2	
40%	0.8993	Standard Error of Estimate	0.0945	0.2744	0.4729	0.1341	
45%	0.9642	SEE / Fit Mean	8.3%	23.3%	33.7%	11.8%	
50%	1.0493	Chi^2 Fit test 9 Bins, Sig 0.05	Good (63%)	Good (27%)	Poor (0%)	Good (21%)	
55%	1.1147						
60%	1.1475						
65%	1.1692						
70%	1.3436						
75%	1.3760						
80%	1.4403						
85%	1.6569						
90%	1.8824						
95%	2.3306						
100%	3.9345						



CRUAMM Commodity Appendix

Not intended for public domain.

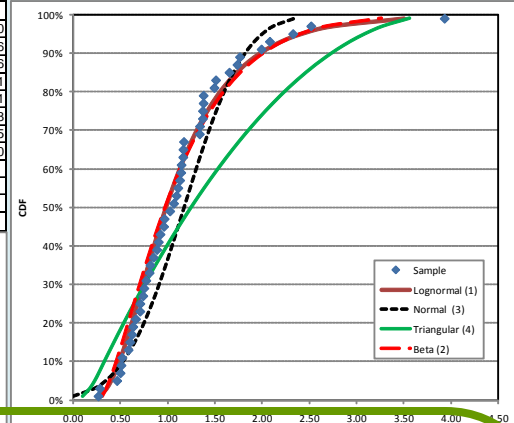
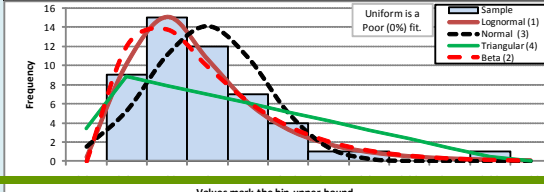
CRUAMM appendix contains data points.

Cost Risk and Uncertainty Analysis Data Report

Commodity	Space	Dispersion	Third
WBS	Residuals	Units	
Element	EPS	Data Comment	EPS T1 = 220 + EPS Weight(bs) * 13 + CCL * 9000
Sub-Element	T1 Rec	Phase	
Category	EPSRecT1Residuals	Comment	
Filter and Fit Settings	Filter= NONE; Fit = SSE on Value, Percentile = CoC, Using: COSTAT Engine		
Fit Constraints	Force Min => 0; Nor TRI Beta Force Include Sample Min/Max: TRI Beta		
Recommended Fit	Lognormal (Mean = 1.14, Std Dev = 0.67)		

Dataset Statistics and Fitted Distributions

Sample	Sample	Lognormal	Normal	Triangular	Beta	
Can be used to define a CDF directly	Mean	1.1408	1.1388	1.1760	1.4015	1.1320
Percentile	StdDev	0.6413	0.6722	0.5055	0.8972	0.6526
0%	CV	0.562	0.590	0.430	0.640	0.576
5%	Min	0.2701			0.0000	0.2701
10%	Mode		0.7272	1.1760	0.2701	0.6171
15%	0.6067				3.9345	17.2398
20%	0.6529				Alpha	1.605
25%	0.7129				Beta	30.000
30%	0.7658					
35%	0.8204					
40%	0.8993					
45%	0.9642					
50%	1.0493					
55%	1.1147					
60%	1.1475					
65%	1.1692					
70%	1.3436					
75%	1.3760					
80%	1.4403					
85%	1.6569					
90%	1.8824					
95%	2.3306					



Dataset

Source	Space CERs Faux Data
Dataset Comments	Update

CER 1st Var	EPST1	Min	753.000	Max	40664.000	CV	0.838	Units	0.000
CER 2nd Var	Wt		94.154		1969.846		0.622		0.000

Observation	EPSRecT1Residual	EPST1	Wt	Observation	ResT1Residu	EPST1	Wt	Observation	PSRecT1Residu	EPST1	Wt
Project Name 1	0.2701	1,719.0000	472.6923	Project Name 26	1.6569	4,752.0000	203.6923				
Project Name 2	0.5035	1,809.0000	259.4615	Project Name 27	1.1752	3,710.0000	225.9231				
Project Name 3	1.1339	31,849.0000	1,451.3077	Project Name 28	2.3306	6,563.0000	199.6923				
Project Name 4	1.7411	40,864.0000	1,096.1538	Project Name 29	0.8902	2,911.0000	234.6154				
Project Name 5	1.4974	25,500.0000	1,293.0000	Project Name 30	0.9642	23,086.0000	1,132.6154				
Project Name 6	0.9265	4,259.0000	336.6923	Project Name 31	1.0694	10,534.0000	740.7692				
Project Name 7	0.4677	4,273.0000	685.9231	Project Name 32	1.0976	9,269.0000	632.6923				
Project Name 8	0.2855	1,324.0000	339.7692	Project Name 33	1.0292	13,398.0000	984.4615				
Project Name 9	0.9704	6,532.0000	500.8462	Project Name 34	0.7764	8,886.0000	863.4615				
Project Name 10	0.6067	8,253.0000	1,029.4615	Project Name 35	0.8204	9,394.0000	863.9231				
Project Name 11	0.7123	5,901.0000	620.3846	Project Name 36	1.1495	12,276.0000	804.5385				
Project Name 12	0.9084	6,538.0000	536.6923	Project Name 37	0.8113	15,444.0000	1,447.3077				
Project Name 13	0.6269	5,626.0000	673.3846	Project Name 38	1.1454	11,492.0000	754.8462				
Project Name 14	1.3755	22,891.0000	1,263.2308	Project Name 39	0.6379	4,682.0000	547.6923				
Project Name 15	1.3447	24,238.0000	1,369.6154	Project Name 40	2.5223	4,124.0000	108.8462				
Project Name 16	1.5143	14,674.0000	728.4615	Project Name 41	0.5215	753.0000	94.1538				
Project Name 17	1.1147	13,079.0000	885.6154	Project Name 42	1.9965	4,001.0000	137.2308				
Project Name 18	2.0847	11,937.0000	423.5385	Project Name 43	1.3829	35,011.0000	1,238.3077				
Project Name 19	0.7551	4,536.0000	445.1538	Project Name 44	0.8539	14,477.0000	1,287.1538				
Project Name 20	1.1692	6,421.0000	405.5385	Project Name 45	0.6678	9,281.0000	1,052.0769				
Project Name 21	1.3760	11,349.0000	617.5385	Project Name 46	1.3832	35,724.0000	1,969.8462				
Project Name 22	1.1673	11,031.0000	710.0000	Project Name 47	0.5888	11,283.0000	1,457.2308				
Project Name 23	3.9345	8,656.0000	152.3077	Project Name 48	0.5134	7,164.0000	1,056.4615				
Project Name 24	1.7682	35,147.0000	1,512.0769	Project Name 49	0.7439	10,302.0000	1,048.3077				
Project Name 25	1.3429	1,500.0000	33.1338	Project Name 50	0.7123	25,620.0000	1,208.6923				

All examples use fictitious data

Using CRUAMM To Define Uncertainty

- Notional model, no correlation applied (functional only), not representative of a properly correlated result
- CRUAMM used to identify representative uncertainty for:
 - the Electrical Power Subsystem CER
 - the EPS CER independent variable (EPS Weight)

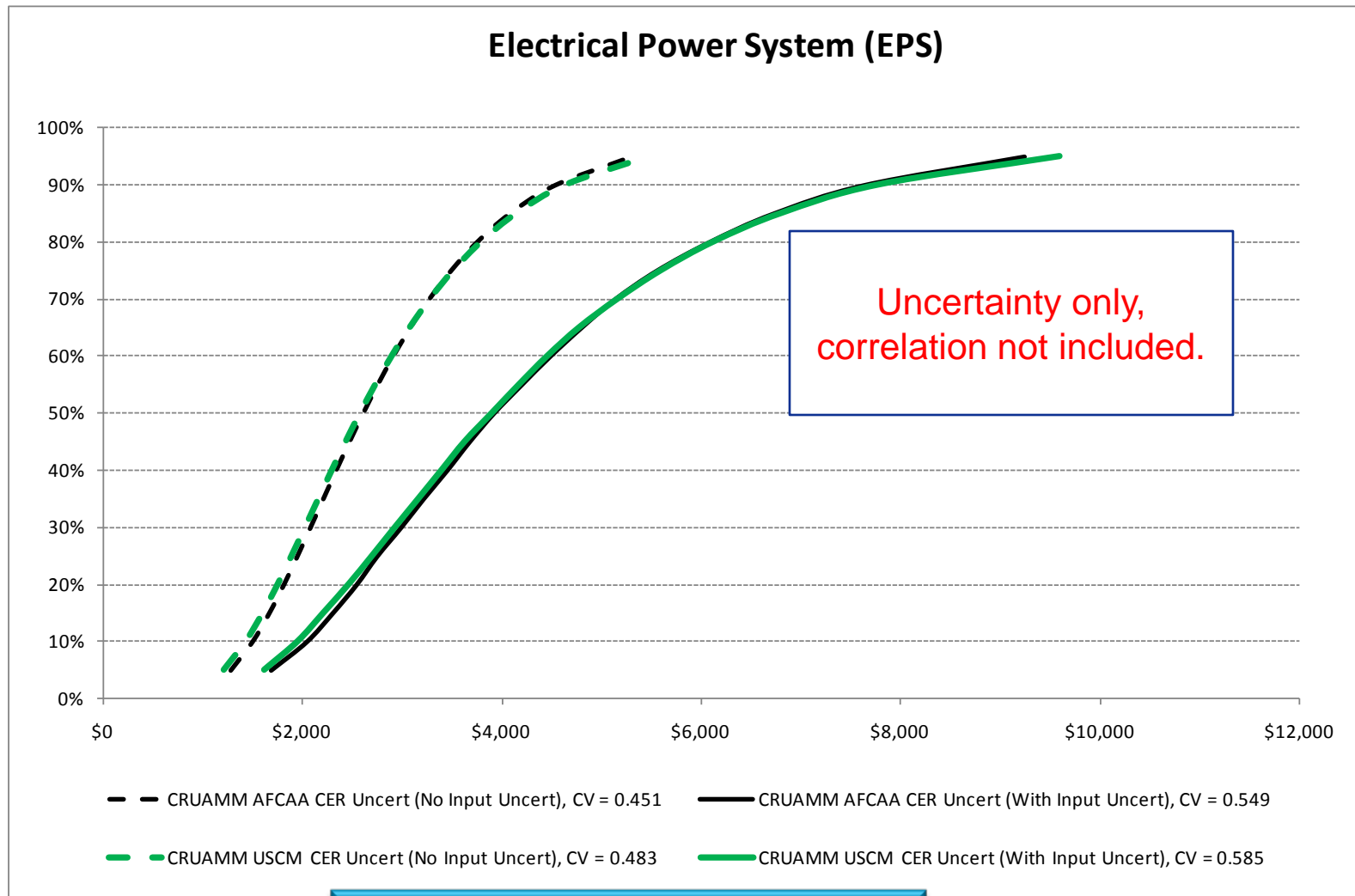
WBS Description	Unique ID	CER	Point Estimate	Forecast (PE * Uncert)	Distribution	Uncertainty	Min	Max or Std Dev
Satellite			\$298,647	\$298,647				
Payload	RDPayload		\$168,462	\$168,462				
IR Sensor		$356851 * \text{PayloadDiam}^{\wedge}0.562$	\$168,462	\$168,462	Triangular	1	0.071	1.9640
Spacecraft Bus	RDSpBus		\$15,096	\$15,096				
Structure		$157 * \text{StrWgt}^{\wedge}0.83$	\$2,787	\$2,787	Triangular	1	0	1.9528
Thermal		$394 * \text{ThermWgt}^{\wedge}0.635$	\$1,331	\$1,331	Triangular	1	0	1.9528
Electrical Power System (EPS)		$62.7 * \text{EPSWgt}$	\$2,865	\$2,865	Lognormal	1		0.4512
Telemetry Tracking & Command (TT&C) and Command & Data Handling (C&DH)		$545 * \text{TTCWgt}^{\wedge}0.761$	\$2,344	\$2,344	Lognormal	1		0.2941
Attitude Determination & Control Sys		$464 * \text{ADCSWgt}^{\wedge}0.867$	\$5,769	\$5,769	Lognormal	1		0.4231
Int & Assy		$989 + 0.215 * (\text{RDPayload\$} + \text{RDSpBus})$	\$40,454	\$40,454	Normal	1		0.3544
Program Level Costs		$1.963 * (\text{RDPayload\$} + \text{RDSpBus})^{\wedge}0.841$	\$52,451	\$52,451	Lognormal	1		0.4609
Ground Support Equip (GSE)		$9.262 * (\text{RDPayload\$} + \text{RDSpBus})^{\wedge}0.642$	\$22,184	\$22,184	Triangular	1	0	3.1204

Input Variable	Unique ID	CER	Point Estimate	Forecast	Distribution	Uncertainty	Min	Max or Std Dev
Payload Sensor diameter (m)	PayloadDiam		0.263	0.263	Lognormal	1		0.3000
Structure weight (kg)	StrWgt		32.000	32.000	Lognormal	1		0.5019
Thermal weight (kg)	ThermWgt		6.800	6.800	Triangular	1	0.264	2.7028
EPS weight (kg)	EPSWgt		45.700	45.700	Triangular	1	0.8742	2.8357
TT&D/DH weight (kg)	TTCWgt		6.800	6.800	Normal	1		0.1893
ADCS weight (kg)	ADCSWgt		18.300	18.300	Normal	1		0.3103
Spacecraft + Payload Weight (kg)	SpPvWgt		140.000	140.000	Lognormal	1		0.2307

All examples use fictitious data

Presented at the 2012 SCEA/ISPA Joint Annual Conference and Training Workshop - www.iceaaonline.com
Compare Spacecraft Bus EPS FireSat CER Uncertainty
Derived from Fitting AFCAA or USCM CER Residuals

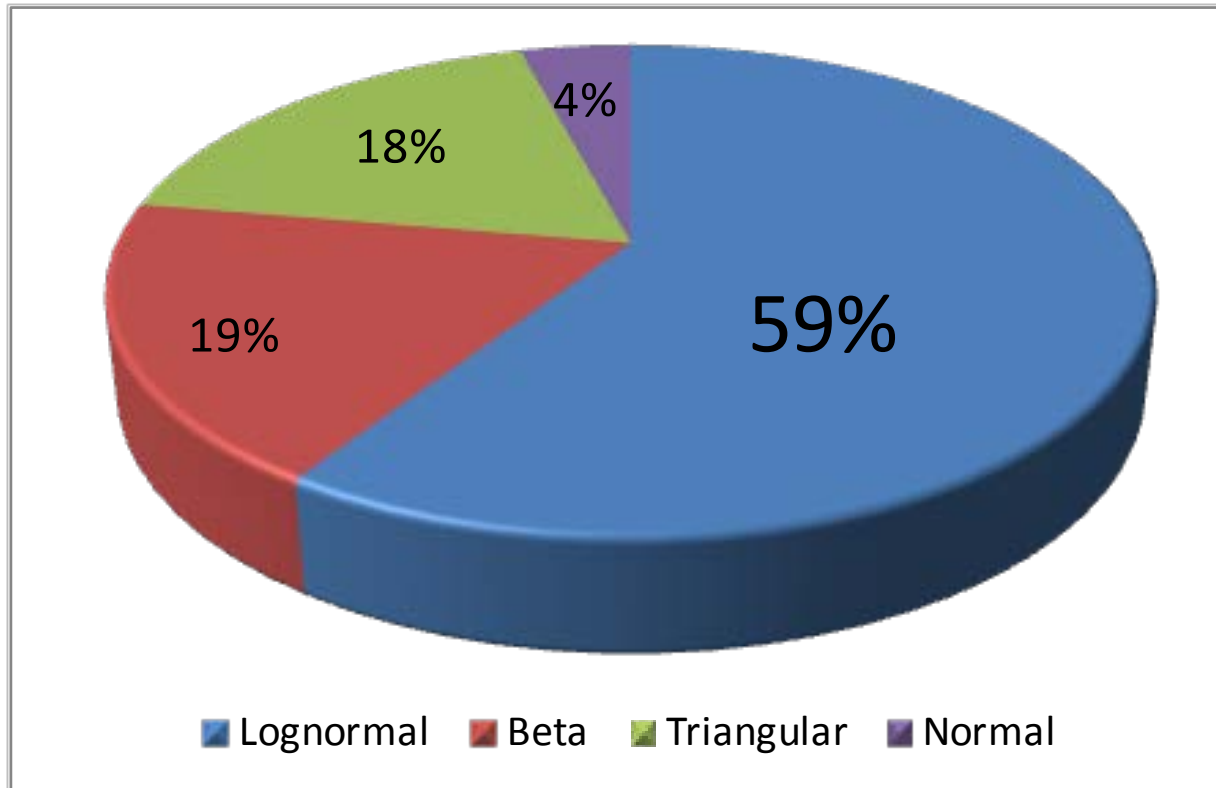
- Dotted lines are result using CER uncertainty only
- Solid lines are result using CER plus Input uncertainty



All examples use fictitious data

Relative Frequency of Distribution Shapes

- **Over 1400 fitted distributions**
 - Lognormal Dominates
 - Normal is Notably Rare
- **“Regardless ... the uncertainty ... may be modeled with a lognormal distribution.”** Air Force Handbook for Cost Uncertainty and Risk Analysis July 2007



CRUAMM Final Report

For Official Use Only



CR-1501/1

Cost Risk and Uncertainty Analysis Metrics Manual (CRUAMM)

Alf Smith, Jeff McDowell, Dr. Lew Fichter, Bryan Blevins, Nick DeTore

PREPARED FOR:
AIR FORCE COST ANALYSIS AGENCY (AFCAA)

Prime Contract Number: GS-35F-5115H S132-51

BPA Number: FA7014-06-A-0015

Delivery Order 0006

DISTRIBUTION LIMITATION: This document was prepared for and submitted to the Air Force Cost Analysis Agency (Wilson Ross) under Contract No. GS-35F-5115H FA7014-06-A-0015-0006. Further distribution of this document is subject to any distribution restrictions set forth in the Contract and consent of the Government Official.

TECOLOTE RESEARCH, INC.
420 S. Fairview Avenue, Suite 201
Goleta, CA 93117-3626
(805) 571-6366

TECOLOTE RESEARCH
101 Quality Circle NW, S
Huntsville, AL 35806
(256) 895-0373

For Official Use Only

TABLE OF CONTENTS

1: Introduction.....	1-1
1.1 Purpose.....	1-1
2: Analysis Approach.....	2-1
2.1 Orders of Dispersion.....	2-1
2.1.1 First Order Dispersion.....	2-2
2.1.2 Second Order Dispersion.....	2-2
2.1.3 Third Order Dispersion.....	2-3
2.1.4 Fourth Order Dispersion.....	2-3
2.2 Distribution Fitting.....	2-3
3: Usage Guidelines.....	3-1
3.1 General.....	3-1
3.2 How to Read the Results.....	3-2
3.3 Additional Examples Using Unitized Results.....	3-4
3.3.1 Example 1: Lognormal.....	3-4
3.3.2 Example 2: Triangular.....	3-5
3.3.3 Example 3: Beta.....	3-6
4: CRUAMM Space System (USCM).....	4-1
4.1 CRUAMM Space System (USCM) Work Breakdown Structure.....	4-1
4.2 CRUAMM Space Systems (USCM) Empirically Based Uncertainty Parameters.....	4-4
4.3 CRUAMM Space Systems (FMS) Empirically Based Uncertainty Parameters.....	4-52
4.4 CRUAMM Space Systems (Growth) Empirically Based Uncertainty Parameters.....	4-53
5: CRUAMM Missiles.....	5-1
5.1 CRUAMM Missiles Work Breakdown Structure.....	5-1
5.2 CRUAMM Missiles Empirically Based Uncertainty Parameters.....	5-2
6: CRUAMM Software.....	6-1
6.1 CRUAMM Software Work Breakdown Structure.....	6-1
6.2 CRUAMM Software Empirically Based Uncertainty Parameters.....	6-1
7: CRUAMM Airborne Radar.....	7-1
7.1 CRUAMM Airborne Radar Work Breakdown Structure.....	7-1
7.2 CRUAMM Airborne Radar Empirically Based Uncertainty Parameters.....	7-1
8: CRUAMM Aircraft.....	8-1
8.1 CRUAMM Aircraft Work Breakdown Structure.....	8-1
8.2 CRUAMM Aircraft Empirically Based Uncertainty Parameters.....	8-1
9: CRUAMM Communications Radio.....	9-1
9.1 CRUAMM Comm Radio Work Breakdown Structure.....	9-1
9.2 CRUAMM Comm Radio Empirically Based Uncertainty Parameters.....	9-2
10: Flow Chart Used to Guide the Fit Process.....	10-1



Sample CRUAMM Public Domain Document Table (Listed by Commodity and WBS Element)

- Results for a single WBS element from a specific commodity
- Colors identify different orders of dispersion (Orange= 1st, Yellow= 2nd, Green= 3rd)
- Green typically has the lowest CV and are associated with CER residuals
- Orange typically has the highest CV and are associated with raw data

Dataset	Count	Numerator		Denominator		Sample CV	Fitted CV	My Point Estimate is the:		
		Label	Range	Label	Range			Mean	Median	Mode
Metrics from CER Dataset										
NR CER, All	11	NR	2477.38 - 11424.26	Weight	86.96 - 621.13	0.271	0.259	Triangular (0.5206, 0.7584, 1.7210)	Triangular (0.5418, 0.7892, 1.7911)	Triangular (0.6865, 1.0000, 2.2694)
NR Cost per Lb, NR CER Datapoints, All	12	NR	2477.38 - 11424.26	Weight	86.96 - 621.13	0.463	0.445	Triangular (0.0709, 0.7294, 2.1997)	Triangular (0.0746, 0.7680, 2.3160)	Triangular (0.0972, 1.0000, 3.0157)
NR Typical Cost, NR CER Datapoints, All	12					0.486	0.463	Triangular (0.2965, 0.3942, 2.3093)	Triangular (0.3215, 0.4275, 2.5040)	Triangular (0.7521, 1.0000, 5.8576)
T1 CER, All	44	NR	0 - 20222.18	Weight	10.92 - 621.13	0.421	0.412	Triangular (0.1552, 0.7265, 2.1183)	Triangular (0.1635, 0.7651, 2.2308)	Triangular (0.2137, 1.0000, 2.9158)
T1 Cost per Lb, Commercial	10	T1	1839.7 - 13567.79	Weight	66.9 - 525.5	0.387	0.369	Triangular (0.3442, 0.6256, 2.0302)	Triangular (0.3649, 0.6631, 2.1520)	Triangular (0.5503, 1.0000, 3.2454)
T1 Cost per Lb, Government	34	T1	319.1 - 17861.82	Weight	10.92 - 621.13	0.673	0.708	Lognormal (1.0000, 0.7077)	Lognormal (1.2251, 0.8670)	Lognormal (1.8387, 1.3013)
T1 Typical Cost, Commercial	10					0.642	0.717	Lognormal (1.0000, 0.7168)	Lognormal (1.2304, 0.8819)	Lognormal (1.8625, 1.3350)
T1 Typical Cost, Government	34					1.065	1.011	Lognormal (1.0000, 1.0110)	Lognormal (1.4220, 1.4376)	Lognormal (2.8754, 2.9070)
Metrics from Entire Dataset										
NR to T1 Ratio, New Design	20	NR	184.22 - 20222.18	T1	319.1 - 17450.68	0.707	0.641	Triangular (0.0000, 0.1905, 2.8095)	Triangular (0.0000, 0.2136, 3.1495)	Triangular (0.0000, 1.0000, 14.7445)
NR to T1 Ratio, Modified Design	17	NR	138.54 - 7416.68	T1	765.54 - 17861.82	1.060	1.384	Lognormal (1.0000, 1.3844)	Lognormal (1.7078, 2.3642)	Lognormal (4.9806, 6.8950)
NR Cost per Lb, New Design	20	NR	184.22 - 20222.18	Weight	10.92 - 621.13	0.647	0.630	Triangular (0.0283, 0.1925, 2.7792)	Triangular (0.0316, 0.2155, 3.1103)	Triangular (0.1468, 1.0000, 14.4348)
NR Cost per Lb, Modified Design	17	NR	138.54 - 7416.68	Weight	33.35 - 596.76	0.849	0.928	Lognormal (1.0000, 0.9282)	Lognormal (1.3644, 1.2663)	Lognormal (2.5397, 2.3573)
T1 Cost per Lb, New Design	20	T1	319.1 - 17450.68	Weight	10.92 - 621.13	0.752	0.818	Lognormal (1.0000, 0.8182)	Lognormal (1.2921, 1.0572)	Lognormal (2.1571, 1.7650)
T1 Cost per Lb, Modified Design	17	T1	765.54 - 17861.82	Weight	33.35 - 596.76	0.464	0.452	Triangular (0.3180, 0.4049, 2.2772)	Triangular (0.3445, 0.4387, 2.4674)	Triangular (0.7853, 1.0000, 5.6243)
NR Cost per Lb, All	45	NR	1.24 - 20222.18	Weight	10.92 - 621.13	1.062	1.055	Beta (0.0003, 5.4745, 0.5514, 2.4678)	Beta (0.0004, 8.8015, 0.5514, 2.4678)	Beta (, , 0.5514, 2.4678)
T1 Cost per Lb, All	50	T1	34.88 - 17861.82	Weight	10.92 - 621.13	0.622	0.638	Lognormal (1.0000, 0.6376)	Lognormal (1.1860, 0.7562)	Lognormal (1.6681, 1.0636)
NR Typical Cost, New Design	20					0.985	0.995	Lognormal (1.0000, 0.9950)	Lognormal (1.4107, 1.4036)	Lognormal (2.8072, 2.7931)
NR Typical Cost, Modified Design	17					1.089	1.264	Lognormal (1.0000, 1.2639)	Lognormal (1.6117, 2.0370)	Lognormal (4.1863, 5.2911)
Recurring Typical Cost, New Design	20					1.003	0.937	Beta (0.0527, 3.5826, 0.4797, 1.3079)	Beta (0.0776, 5.2727, 0.4797, 1.3079)	Beta (, , 0.4797, 1.3079)
Recurring Typical Cost, Modified Design	17					1.023	1.213	Lognormal (1.0000, 1.2127)	Lognormal (1.5719, 1.9063)	Lognormal (3.8836, 4.7098)
T1 Typical Cost, Modified Design	20					1.121	1.085	Lognormal (1.0000, 1.0855)	Lognormal (1.4759, 1.6021)	Lognormal (3.2150, 3.4899)
T1 Typical Cost, Modified Design	17					1.002	1.183	Lognormal (1.0000, 1.1832)	Lognormal (1.5491, 1.8329)	Lognormal (3.7177, 4.3986)
NR Typical Cost, All	43					1.189	1.205	Beta (0.0005, 4.5342, 0.3160, 1.1173)	Beta (0.0011, 10.2913, 0.3160, 1.1173)	Beta (, , 0.3160, 1.1173)
Recurring Typical Cost, All	51					1.017	1.028	Beta (0.0029, 4.2807, 0.4890, 1.6092)	Beta (0.0047, 6.8080, 0.4890, 1.6092)	Beta (, , 0.4890, 1.6092)
T1 Typical Cost, All	51					1.080	1.104	Lognormal (1.0000, 1.1044)	Lognormal (1.4899, 1.6454)	Lognormal (3.3071, 3.6524)
Cost Driver Metrics from CER Dataset										
Weight, NR CER Datapoints, All	12					0.735	0.732	Beta (0.3092, 2.2084, 0.2024, 0.3540)	Beta (0.4636, 3.3113, 0.2024, 0.3540)	Beta (, , 0.2024, 0.3540)
Weight, T1 CER Datapoints, Commercial	10					0.375	0.358	Normal (1.0000, 0.3585)	Normal (1.0000, 0.3585)	Normal (1.0000, 0.3585)
Weight, T1 CER Datapoints, Government	34					0.780	0.692	Triangular (0.0000, 0.0429, 2.9571)	Triangular (0.0000, 0.0487, 3.3544)	Triangular (0.0000, 1.0000, 68.9146)
Cost Driver Metrics from Entire Dataset										
Weight, All	50					0.684	0.670	Triangular (0.0000, 0.1062, 2.8938)	Triangular (0.0000, 0.1199, 3.2677)	Triangular (0.0000, 1.0000, 27.2524)



Benefits of the CRUAMM Project

- **Tools to systematically store normalized data and perform fits**
 - Underlying math and assumptions are “open” for scrutiny and validation
 - Every algorithm is supported by authoritative references
 - Validated to generate results very close to other commercial tools
- **Standard process to elicit consistent fit analysis across a variety of analysts**
 - In particular, the analysis flow chart is a critical and compelling product from this effort
- **Over 2000 fits provides a rich source of data to analyze for trends**
 - One example was the dominance of lognormal across all fits
 - Should examine subsets of the data for trends (e.g., T1 or weight)

CRUAMM Provides a Source of Real Uncertainty Derived from Real Data



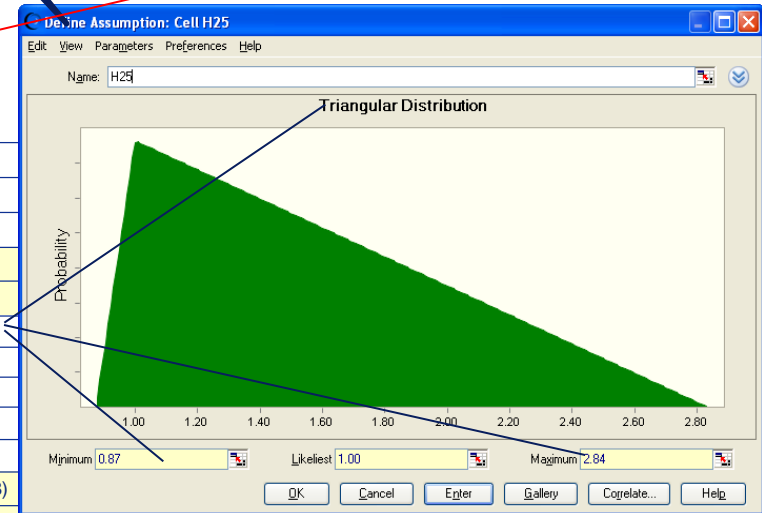
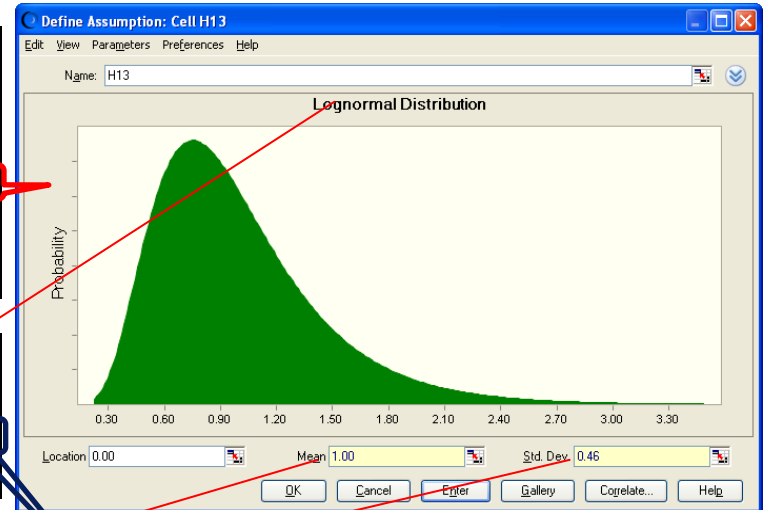


FireSat Spacecraft Bus Backup

Detail of Assigning CRUAMM Unitized Results Into a Crystal Ball Model

WBS Description	Unique ID	CER	Point Estimate	Forecast (PE + Uncert)	Distribution	Uncertainty	Min	Max or Std Dev
Satellite			\$298,647	\$298,647				
Payload	RDPayload		\$168,462	\$168,462				
IR Sensor		356851 * PayloadDiam*0.562	\$168,462	\$168,462	Triangular	1	0.071	1.9640
Spacecraft Bus	RDSpBus		\$15,096	\$15,096				
Structure		157 * StrWgt*0.83	\$2,787	\$2,787	Triangular	1	0	1.9528
Thermal		394 * ThermWgt*0.635	\$1,331	\$1,331	Triangular	1	0	1.0623
Electrical Power System (EPS)		62.7 * EPSWgt	\$2,865	\$2,865	Lognormal	1		0.4512
Telemetry Tracking & Command (TT&C) and Command & Data Handling (C&DH)		545 * TTCWgt*0.761	\$2,344	\$2,344	Lognormal	1		0.2941
Attitude Determination & Control Sys		464 * ADCSWgt*0.867	\$5,769	\$5,769	Lognormal	1		0.4231
Int & Assy		989 + 0.215 * (RDPayload\$ + RDSpBus)	\$40,454	\$40,454	Normal	1		0.3544
Program Level Costs		1.963 * (RDPayload\$ + RDSpBus) *0.841	\$52,451	\$52,451	Lognormal	1		0.4609
Ground Support Equip (GSE)		9.262 * (RDPayload\$ + RDSpBus) *0.642	\$22,184	\$22,184	Triangular	1	0	3.1204

Input Variable	Unique ID	CER	Point Estimate	Forecast	Distribution	Mean	Std Dev
Payload Sensor diameter (m)	PayloadDiam		0.263	0.263	Lognormal	1	0.3000
Structure weight (kg)	StrWgt		32.000	32.000	Lognormal	1	0.5019
Thermal weight (kg)	ThermWgt		6.800	6.800	Triangular	1	0.261
EPS weight (kg)	EPSWgt		45.700	45.700	Triangular	1	0.8742
TT&D/DH weight (kg)	TTCWgt		6.800	6.800	Normal	1	2.8357
ADCS weight (kg)	ADCSWgt		18.300	18.300	Normal	1	0.3100
Spacecraft + Payload Weight (kg)	SpPyWgt		140.000	140.000	Lognormal	1	0.2307

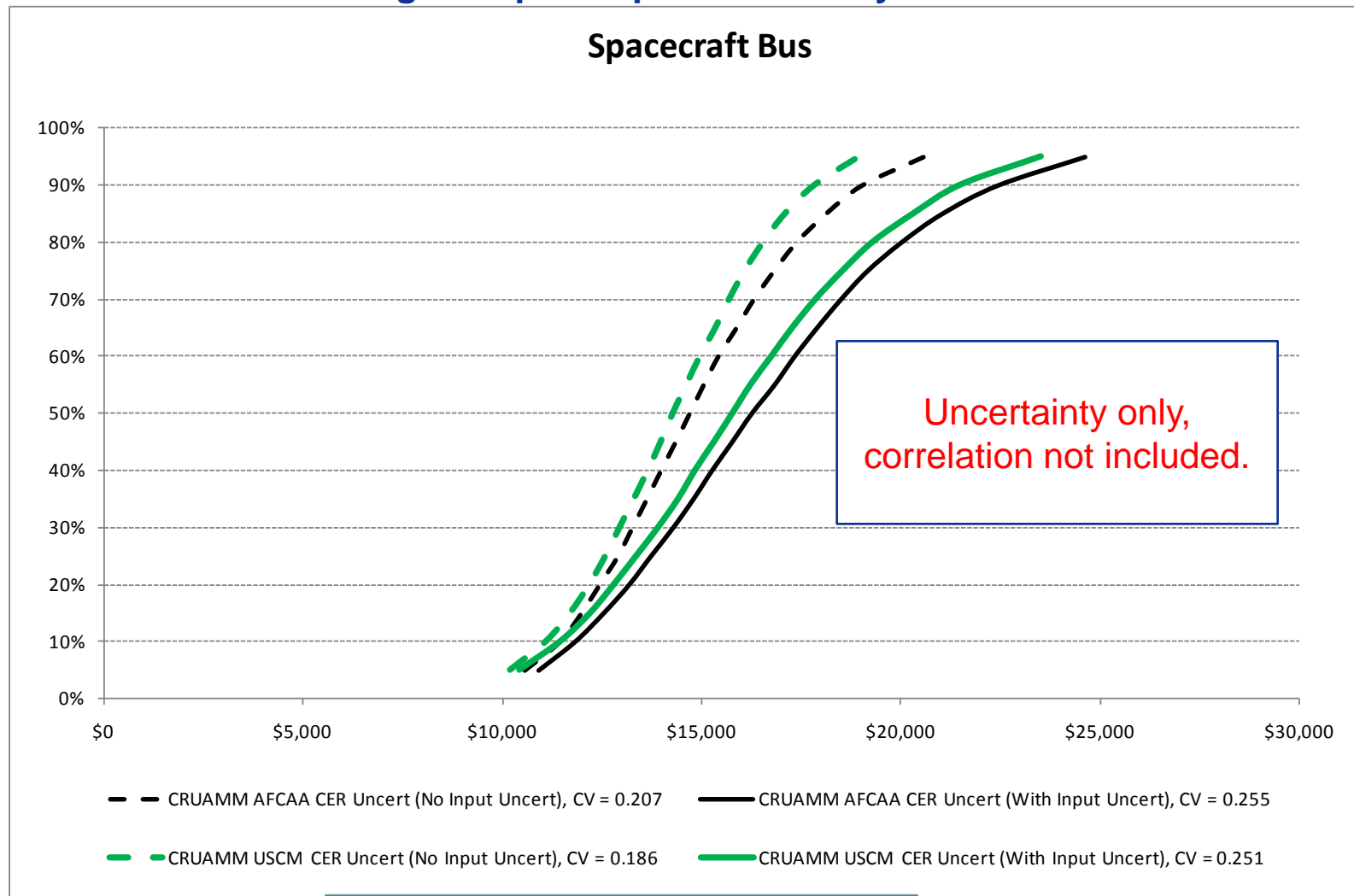


Parameter	Mean	Median	Mode
T1 CER ¹	Lognormal (1.0000, 0.4512)	Lognormal (1.0971, 0.4950)	Lognormal (1.3204, 0.5958)
NR CER ¹			
T1 CER ³	Triangular (0.1552, 0.7265, 2.1183)	Triangular (0.1635, 0.7651, 2.2308)	Triangular (0.2137, 1.0000, 2.9158)
NR CER ³	Triangular (0.5206, 0.7584, 1.7210)	Triangular (0.5418, 0.7892, 1.7911)	Triangular (0.6865, 1.0000, 2.2694)
Weight Growth ²	Triangular (0.5568, 0.6370, 1.8062)	Triangular (0.5850, 0.6692, 1.8975)	Triangular (0.8742, 1.0000, 2.8357)
	Triangular (0.5853, 0.7180, 1.6967)	Triangular (0.6092, 0.7474, 1.7662)	Triangular (0.8151, 1.0000, 2.3631)
	Triangular (0.5043, 0.6168, 1.8789)	Triangular (0.5316, 0.6502, 1.9805)	Triangular (0.8176, 1.0000, 3.0461)
T1 / Weight ³	Triangular (0.3442, 0.6256, 2.0302)	Triangular (0.3649, 0.6631, 2.1520)	Triangular (0.5503, 1.0000, 3.2454)
	Lognormal (1.0000, 0.7077)	Lognormal (1.2251, 0.8670)	Lognormal (1.8387, 1.3013)
NRE / Weight ³	Triangular (0.0283, 0.1925, 2.7792)	Triangular (0.0316, 0.2155, 3.1103)	Triangular (0.1468, 1.0000, 14.4348)
	Lognormal (1.0000, 0.9282)	Lognormal (1.3644, 1.2663)	Lognormal (2.5397, 2.3573)
NRE / T1 ³	Triangular (0.0000, 0.1905, 2.8095)	Triangular (0.0000, 0.2136, 3.1495)	Triangular (0.0000, 1.0000, 14.7445)
	Lognormal (1.0000, 1.3844)	Lognormal (1.7078, 2.3642)	Lognormal (4.9806, 6.8950)



Compare Spacecraft Bus AFCAA and USCM Uncertainty Results

- Dotted lines are result using CER uncertainty only
- Solid lines are result using CER plus Input uncertainty

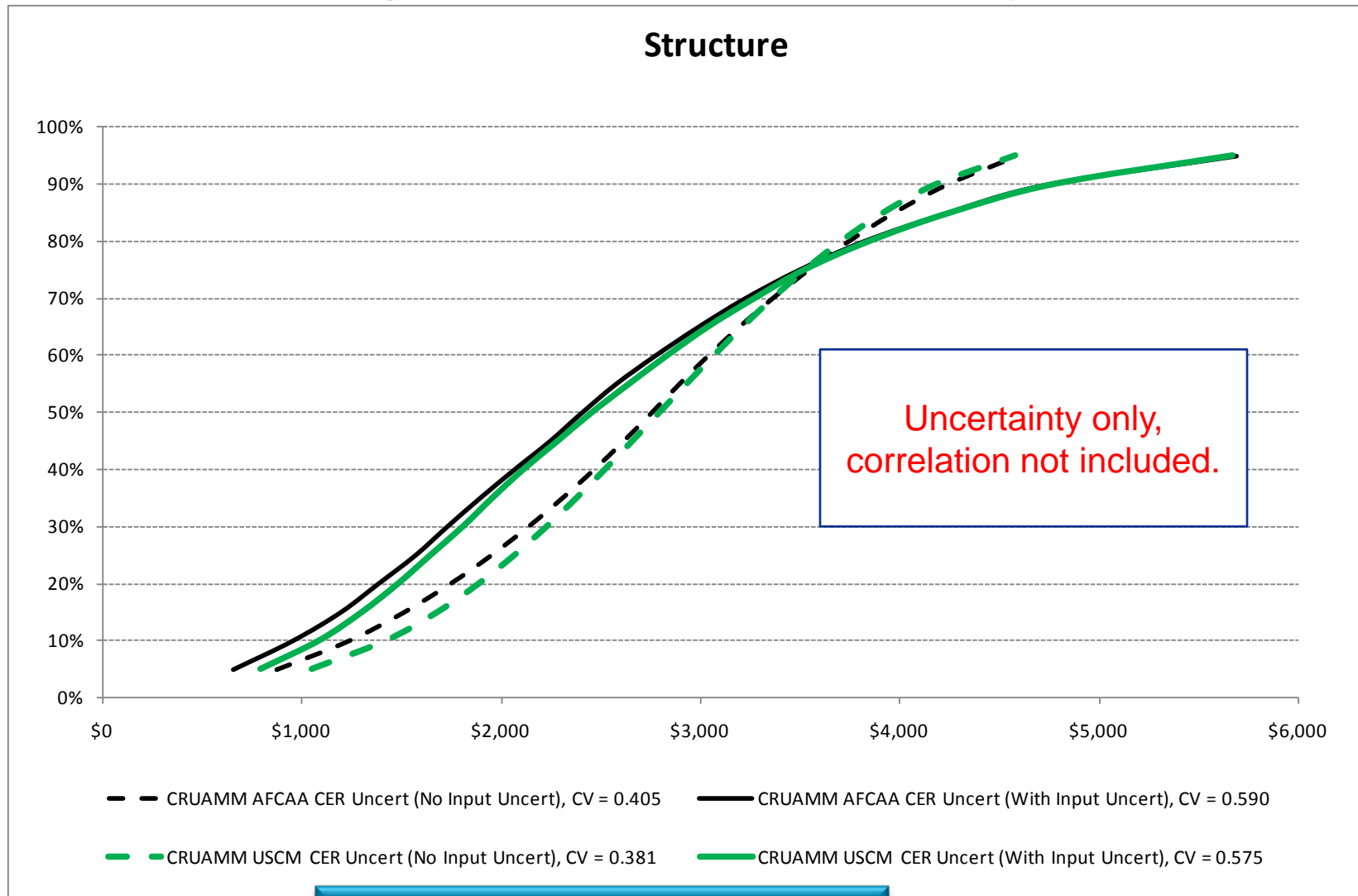


All examples use fictitious data



Compare Spacecraft Bus Structure AFCAA and USCM Uncertainty Results

- Dotted lines are result using CER uncertainty only
- Solid lines are result using CER plus the same Input uncertainty

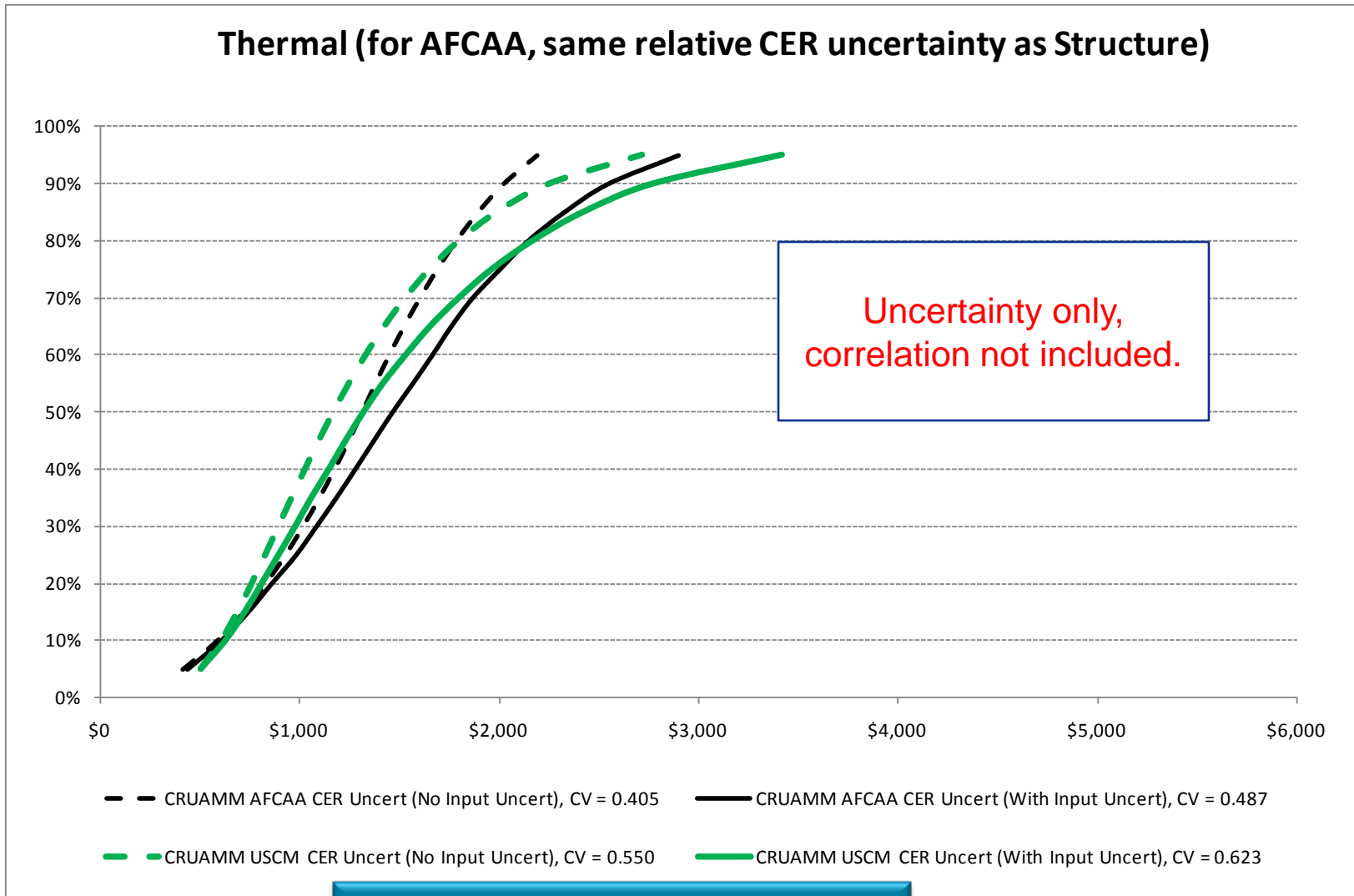


All examples use fictitious data



Compare Spacecraft Bus Thermal AFCAA and USCM Uncertainty Results

- Dotted lines are result using CER uncertainty only
- Solid lines are result using CER plus the same Input uncertainty

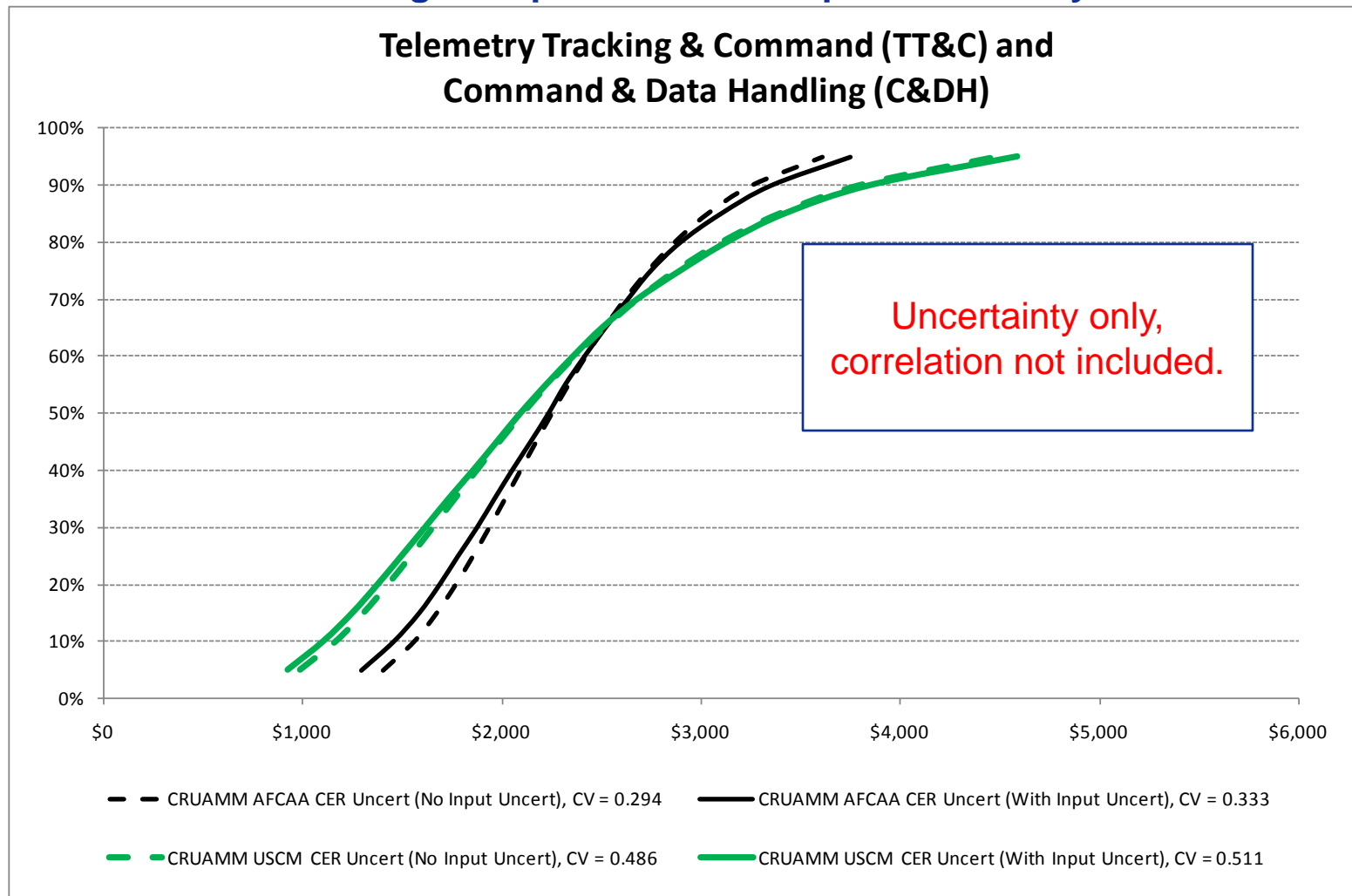


All examples use fictitious data



Compare Spacecraft Bus TT&C AFCAA and USCM Uncertainty Results

- Dotted lines are result using CER uncertainty only
- Solid lines are result using CER plus the same Input uncertainty

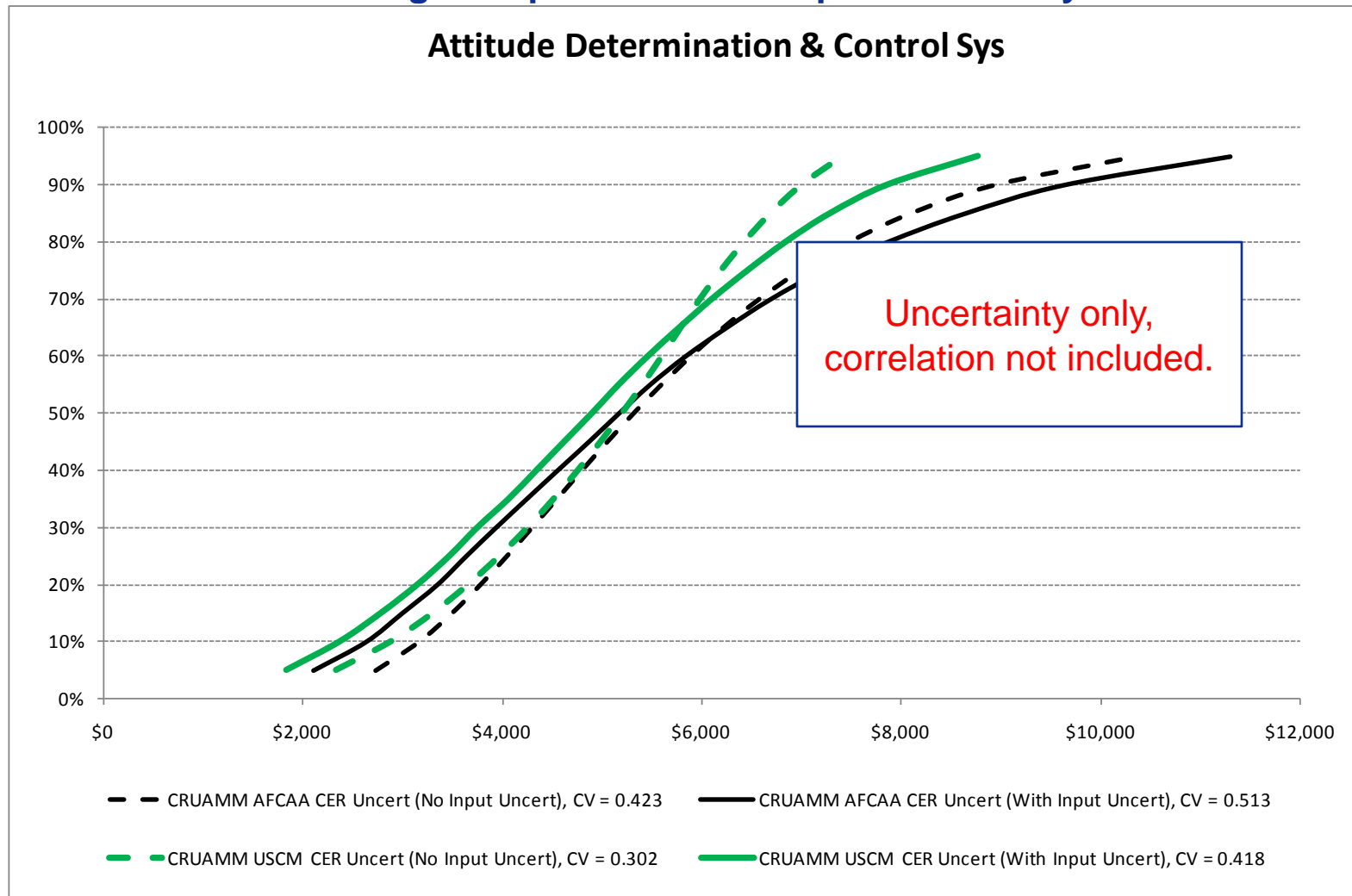


All examples use fictitious data



Compare Spacecraft Bus ADC AFCAA and USCM Uncertainty Results

- Dotted lines are result using CER uncertainty only
- Solid lines are result using CER plus the same Input uncertainty



All examples use fictitious data

References

- AFCAA Cost Risk and Uncertainty Handbook, 2007

- GAO Cost Estimating and Assessment Guide, 2009

- Build Your Own Distribution Finder, 2010 ISPA/SCEA Joint Conference & Training Workshop, 8 - 11 June 2010, Alfred Smith CCEA

- Using Unitized Uncertainty Distributions in ACEIT, ACEIT Users Workshop, February 1-2, 2010, Jeff McDowell

- Real Data, Real Uncertainty, 45th Annual Department of Defense Cost Analysis Symposium, February 14-17, Alfred Smith, Jeff McDowell, Dr. Lew Fichter

