

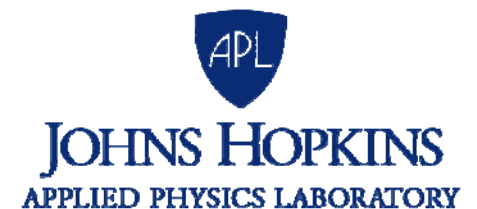
Technology Development Parametric Cost and Schedule Modeling

ICEAA 2017 Professional Development & Training Workshop

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***Chuck Alexander, PE, CPA
National Security Analysis Department
Technology, Programs and Cost***

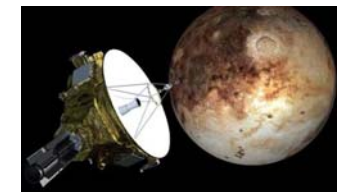


Overview

- **Background**
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 - **Challenges of Technology Development Estimating**
 - **Industry Literature, Tools, and Data Search**
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 - **Technology Development Project Dataset**
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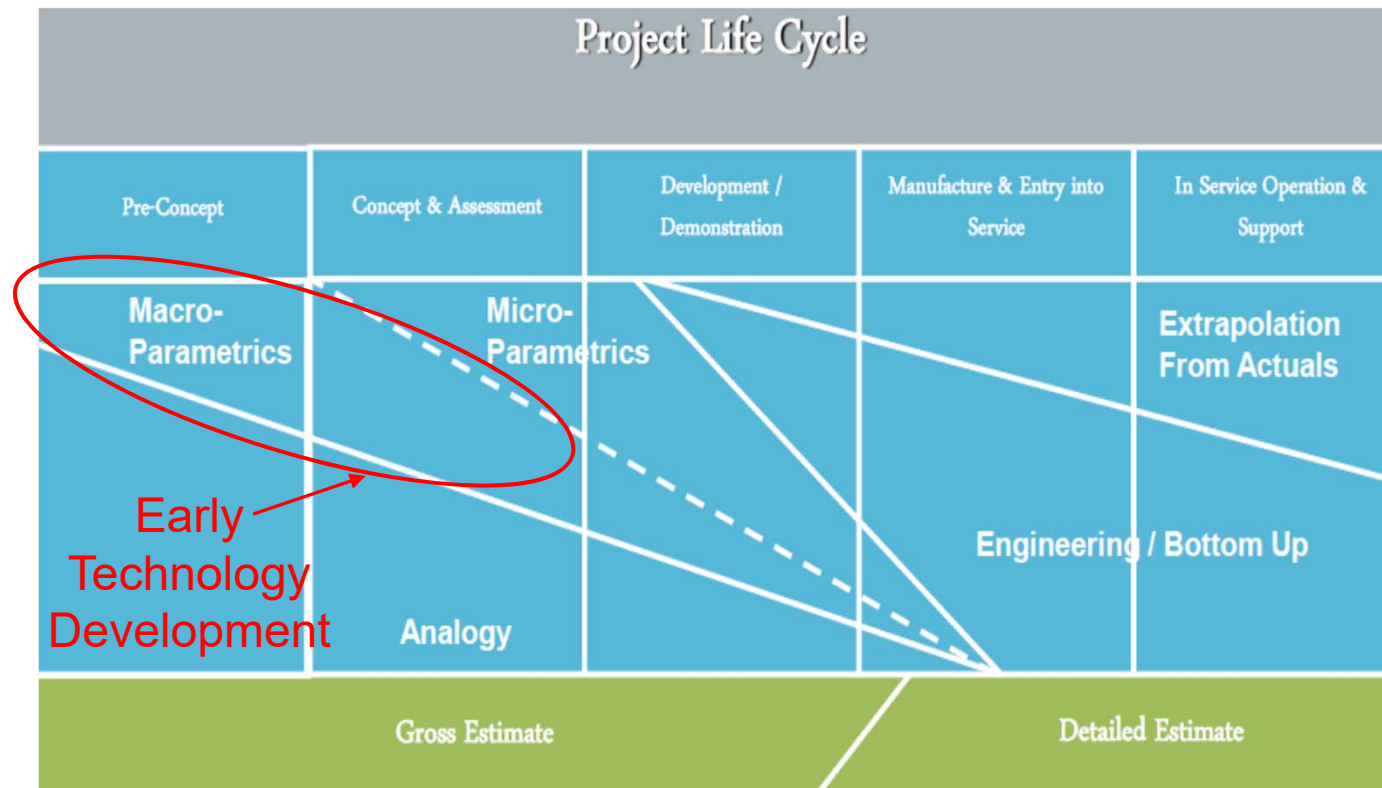
Conceptual & Early Life Cycle Technology Development

- Numerous applications across DoD, Intel, Space and Civil sectors
- Breadth of focus areas and platforms including
 - Sea – Ships / Submarines / Unmanned
 - Air – Aircraft / Airships / Unmanned
 - Space – Satellites / Spacecraft / Probes
 - Weapon Systems – Strategic / Tactical
 - Networks – Ground, Space and Marine Data / Communications / Sensors
 - Robotics / Automation / Nanotechnology
 - Information Technology / Electronics / Cyber
 - Military Strategy and Force Structure
 - Energy and Infrastructure
 - Warrior Armament
 - Healthcare



Challenges of Early Technology Development Estimating

- Little or no analogous or comparable systems / applications
- High uncertainty and level of unknowns
- Lack of conceptual technical, engineering, design or performance parameters available to drive traditional micro-parametric models



Source: “Macro-parameters and the applications of multi-collinearity and Bayesian to enhance early cost modeling” - QinetiQ, Shermom & Barnaby, ICEAA 2015 Professional Development & Training Workshop

Industry Literature, Tool, and Data Search

- **Literature Search for Technology Development Cost and Schedule Estimating Methods / Models**
 - Various frameworks, analysis and modeling approaches have been proposed or developed
 - Research papers offer insightful analysis, methods and considerations for use of “macro level” parameters (e.g., Technology Readiness Level)
 - Deliver varying results but most are based upon limited data sets or focus on select technology areas / applications
- **Leading Technology Development Estimating Tools and Databases**
 - Available tools generally driven by detailed design, performance, and complexity “micro parameters”, not available in early stages
 - General lack of available macro level parametric tools
 - Government sector repositories, databases and models primarily focused on Procurement or O&S Phases
 - Databases generally proprietary / protected or limited access

Technology Development Project Dataset

- Project dataset search conducted to develop broad-based technology development models
- NASA Technology Cost and Schedule Estimating (TCASE) tool identified and selected as resource for model development
 - Contained historical project cost, schedule and technical data with macro variables and project record quantities sought
 - Extensive core technology database containing over 2,900 project records with 164 available data fields across 14 broad-based technology areas (TA)

14+1 TECHNOLOGY AREAS Table 2	
TA #	Description
TA01	Launch Propulsion Systems
TA02	In-Space Propulsion Technologies
TA03	Space Power and Energy Storage
TA04	Robotics, Telerobotics, Autonomous Systems
TA05	Communication and Navigation
TA06	Human Health, Life Support, Habitation Systems
TA07	Human Exploration Destination Systems
TA08	Science Instruments, Observatories, Sensor Systems
TA09	Entry, Descent, and Landing Systems
TA10	Nanotechnology
TA11	Modeling, Simulation, Information Tech
TA12	Materials, Structures, Mechanical Systems, Manufacturing
TA13	Ground and Launch Systems Processing
TA14	Thermal Management Systems
(+) 1	Aeronautics

Model Development Methodology / Approach - Variable Selection and Data Modeling

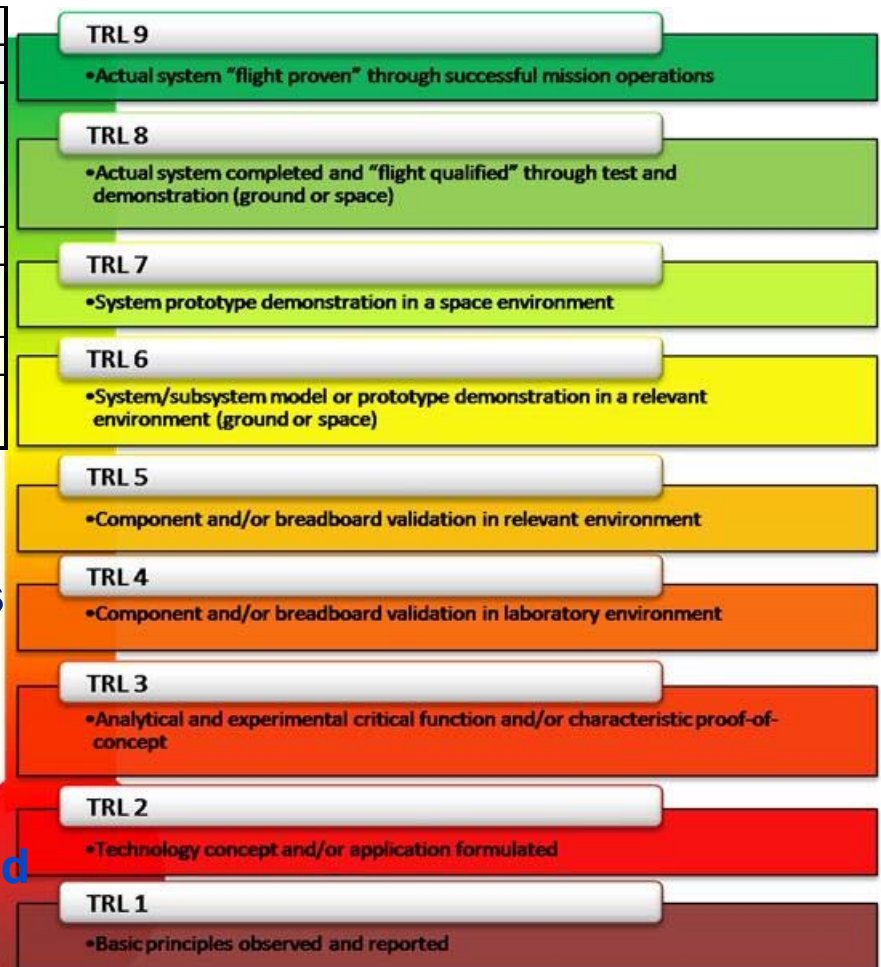
- **Selection of Cost and Schedule dependent response variables**
 - **Total Development Cost (\$) and Project Duration (months)**
 - **Continuous quantitative variables (i.e., cardinal numbers)**
- **Database fields with greatest potential as independent predictor variables for cost and schedule**
 - **System Hierarchy (SH) Level* (1 to 5);**
 - **Project Start / End Technology Readiness Level (TRL) (1 to 9);**
 - **Research and Development Degree of Difficulty (RD3) (Levels I to V);**
 - **Technology Area (TA1 to TA14);**
 - **System Characteristics;**
 - **Key Performance Parameters (KPPs);**
 - **Total FTEs (i.e., Full time Equivalents of project labor);**
 - **Capability Demonstrations**

* *For modeling, the term **Hierarchy Rank** was used to represent the **SH Level***

Model Development Methodology / Approach - Variable Selection and Data Modeling

After careful assessment two viable predictor candidates emerged
System Hierarchy Level* Technology Readiness Level*

System Hierarchy Table		
No.	Tier	Definition
5	System	An integrated set of constituent elements that are combined in an operational or support environment to accomplish a defined objective
4	Subsystem	A portion of a system
3	Assembly	A set of components (as a unit) before they are installed to make a final product
2	Component / Part	A portion of an assembly
1	Hardware / Material	An item or substance used to form a component



- Both ordinal categorical variables
- Other predictor variable candidates were eliminated based upon:
 - Insufficient project records with key predictor or response variables
 - Data relationship screening produced poor fit or overlap with other better suited variables

*Source: NASA TCASE Training Guide and User Manual



Data Modeling – Cost vs. TRL Transition Data

- Investigation determined inadequate no. of observations in most TRL Start - End transition (TRL X-Y) categories for sufficient sample sizes
 - Only 5 of 36 TRL Start-End (TRL X-Y) categories contained “large” sample sizes (>30)
 - Cost curve fits developed for 14 of the 36 TRL X-Y categories (with >7 obs.)¹ produced erratic results
- TRL Improvement (TI) Level² was therefore examined and selected as viable alternative
 - Provided causal relationship & needed sample sizes
 - Resulted in consistent range across starting TRLs with adequate sample sizes for TI Levels 1 to 5
 - TI level 6 or greater appears to be extremely rare

1. Small sample sizes < 8 observations demonstrated substantial volatility produced by limited inputs and considered too small to demonstrate statistical significance

2. TRL Start to TRL End difference, sometimes referred to as “TRL Transition Order” (e.g., TRL 3-5 is of Transition Order 2)

Start TRL	End TRL	TRL X-Y	No. Obs.
TRL Improvement Level 1			177
1	2	1-2	20
2	3	2-3	45
3	4	3-4	66
4	5	4-5	17
5	6	5-6	20
6	7	6-7	8
7	8	7-8	1
8	9	8-9	0
TRL Improvement Level 2			133
1	3	1-3	10
2	4	2-4	51
3	5	3-5	24
4	6	4-6	45
5	7	5-7	3
6	8	6-8	0
7	9	7-9	0
TRL Improvement Level 3			63
1	4	1-4	11
2	5	2-5	18
3	6	3-6	33
4	7	4-7	1
5	8	5-8	0
6	9	6-9	0
TRL Improvement Level 4			22
1	5	1-5	3
2	6	2-6	16
3	7	3-7	1
4	8	4-8	1
5	9	5-9	1
TRL Improvement Level 5			10
1	6	1-6	5
2	7	2-7	3
3	8	3-8	1
4	9	4-9	1
TRL Improvement Level 6			0
1	7	1-7	0
2	8	2-8	0
3	9	3-9	0
TRL Improvement Level 7			0
1	8	1-8	0
2	9	2-9	0
TRL Improvement Level 8			0
1	9	1-9	0
36		Total	405

Model Development – Modeling Forms Investigated

Hundreds of model variants under 4 primary forms

- **Tailored curve fit function models**
 - **Over 20 functions* evaluated for ea. Cost and Schedule, TI and SH level**
- **Simple linear regression models**
 - **Single (univariate) and Composite (multivariate) predictor variables**
 - **Transformed predictor and/or response variables (up to 11 transformation types were evaluated for each variable combination)**
- **Multiple linear regression models**
 - **Multiple predictor - TI and SH Level**
 - **Transformed predictor and/or response variables (up to 11 types each)**
- **A range of nonlinear (NL) models**
 - **21 forms for each predictor variable evaluated including polynomial, sigmoid & logistic curves, exponential & peak models, et.al.**

** Beta, Chi-square, Erlang, Exponential, Gamma, Inverse Gaussian, LaPlace, Levy, Logistic, LogLogistic, Lognorm1/2, Pareto1/2, Pearson5/6, PERT, Raleigh, Triangular, Uniform, Weibull*

Model Evaluation / Selection Criteria

- 1. Statistical Key Performance Measures (KPMs)**
 - **Error Variability and Dispersion Measures:**
 - Coefficient of Determination - R^2 and Adjusted R^2
 - Root Mean Square Error (RMSE)
 - Coefficient of Variation (CV)
 - **Statistical Significance Measures:**
 - F-ratio and t-stat (% of model terms with probability $> |t|$)
 - **Autocorrelation Measure:** Durbin-Watson test
 - **Data Reduction Measure:** Percent (%) of original data sample set unused
- 2. Assessment of prediction model fit to actual sample data**
 - **Various statistical measures and graphic data fit plots / charts**
- 3. Specific measures relevant to the particular model form**
 - **Optimization methods for curve fits and measures applicable to linear and nonlinear models**
 - **VIF to measure multicollinearity for multiple regression models**

Cost Model Performance Results - KPM

Cost Model KPM Results by model type

NOTIONAL PERFORMANCE RATING			
Good	Fair	Marginal	Poor

Mdl. No.	Model Form / Method	Predictor Variable Form	Key Performance Measures (KPM)								
			R-Sq	Adj R-Sq	RMSE (000's)	Coef. of Variation (CV)	F-ratio	Prob. > F	t-stat: % of terms w/ Prob. > t	Durbin-Watson Stat	Data Reduction (%)
1	Tailored Curve Fits	TI Level	N/A	N/A	40,929	0.736	N/A	N/A	N/A	N/A	2.5%
2	Tailored Curve Fits	SH Level	N/A	N/A	26,724	0.711	N/A	N/A	N/A	N/A	3.2%
5	Simple Linear Regression	SH Level	0.935	0.934	2,590	1.249	1893.2	<.0001*	75%	0.896	11.8%
6	Simple Linear Regression	SH Level	0.659	0.657	29,132	2.486	280.8	<.0001*	50%	1.275	3.5%
7	Composite Linear Regression	[TI x SH] ²	0.772	0.771	38,324	1.526	719.5	<.0001*	100%	1.433	3.6%
8	Multiple Linear Regression	TI + SH	0.823	0.816	33,397	1.226	116.7	<.0001*	100%	1.757	5.0%
9	Multiple Linear Regression	[TI + SH] ²	0.788	0.780	2,621	0.617	90.4	<.0001*	50%	1.208	8.1%
10	Nonlinear - Quadratic	NL TI Level	0.610	0.609	32,685	1.606	N/A	N/A	N/A	N/A	15.3%
12	Nonlinear - Exponential 3P	NL SH Level	0.744	0.743	24,966	2.070	N/A	N/A	N/A	N/A	11.3%
13	Nonlinear - Gompertz 4P	NL SH Level	0.742	0.742	25,061	2.078	N/A	N/A	N/A	N/A	11.3%

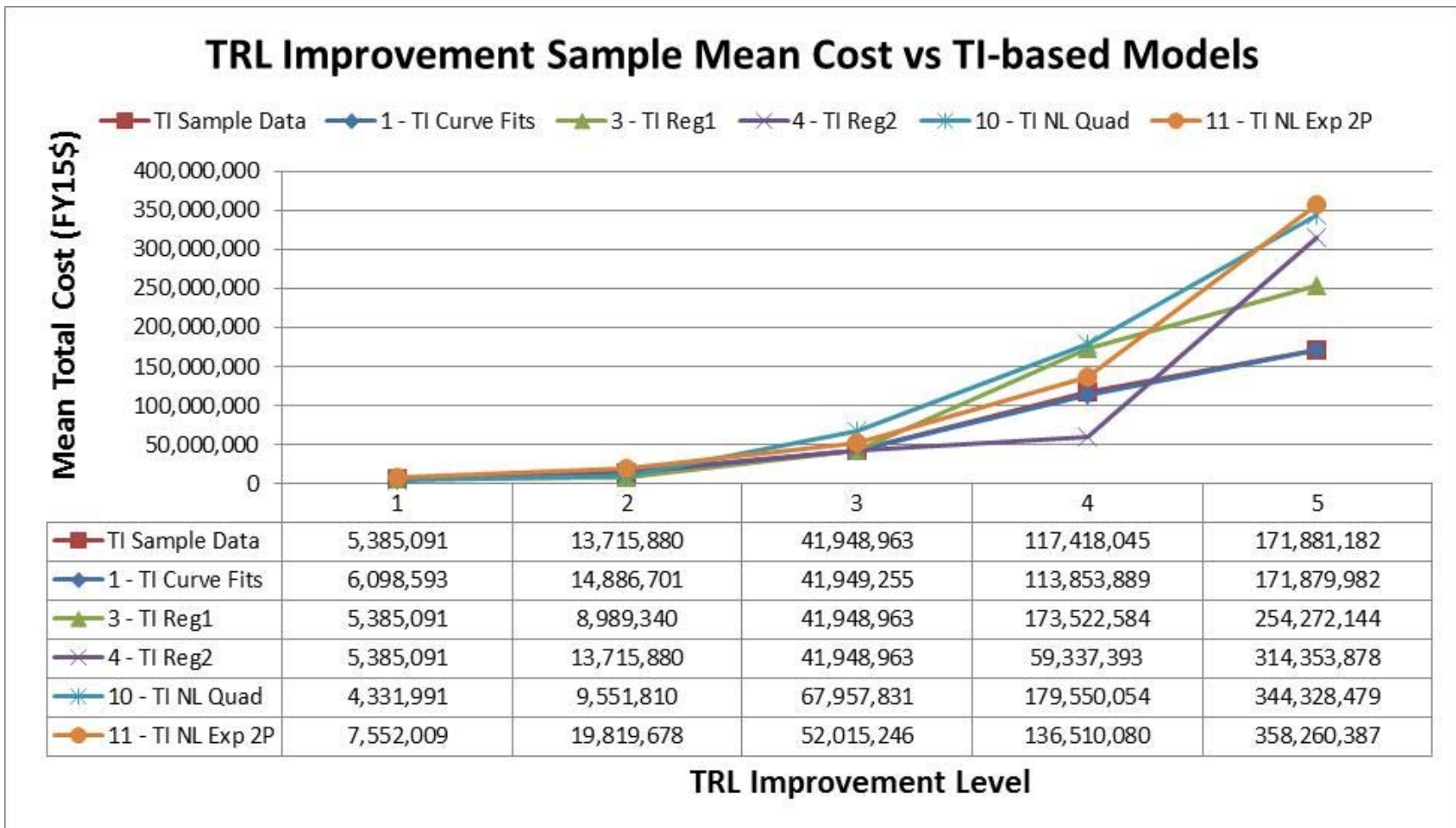
Note - TI level regression models 3 & 4 and TI NL model 11 were eliminated due to poor KPM results.

- **Multiple linear regression models (8 & 9) performed well for KPMs alone**
- **Curve fit models (1 & 2) best replicated the underlying sample data central values and distribution shapes (see following slides)**

*Note: KPM categories that do not apply, cannot be generated, or are not available to a particular model form are shown as **N/A** for not applicable.*

Cost Model Output - TI Level Cost Models vs Actual Data

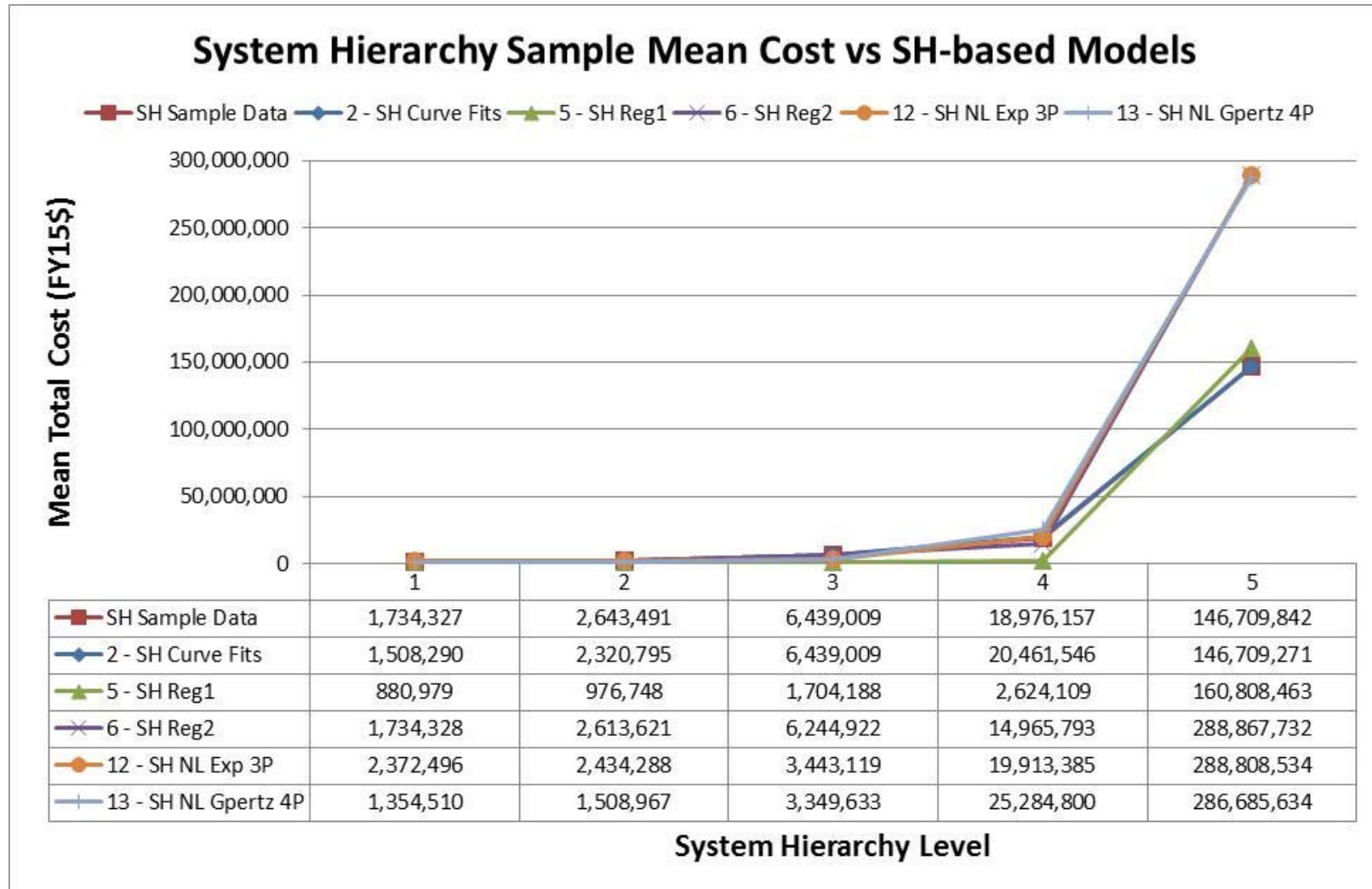
TI Model nos. 1, 3, 4, 10 and 11 vs TI Sample Project Data (means)



Note: Lines do not represent continuous functions but rather demonstrate the progression of model discrete ordinal values.

Cost Model Output - SH Level Cost Models vs Actual Data

SH Model nos. 2, 5, 6, 12 and 13 vs SH Sample Project Data (means)



Cost Model Output - Summary Cost Curve Fit Model Statistics (Model nos. 1 & 2)

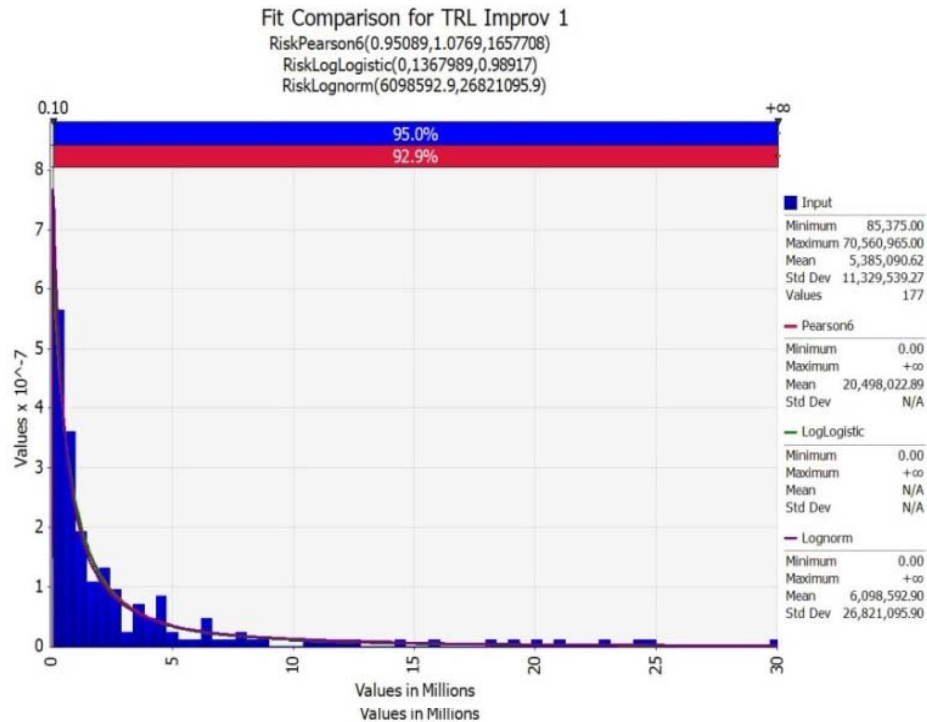
- Lognormal, Gamma and LogLogistic functions produced best curve fit results across Models 1 (TI-level) and 2 (SH-level)
- Lognormal function selected for modeling uncertainty with regression and nonlinear models

Predictor Level / Tier	Mean	Median	60th %ile	80st %ile	Curve Function Type
TRL Improvement Level					
TRL Improvement 1	6,098,593	1,352,186	2,098,994	5,827,153	Lognorm
TRL Improvement 2	14,886,701	2,937,018	4,636,000	13,379,843	Lognorm
TRL Improvement 3	41,949,255	17,585,237	28,194,724	68,557,068	Gamma
TRL Improvement 4	113,853,889	30,765,241	49,013,531	144,529,122	Lognorm
TRL Improvement 5	171,879,982	87,024,759	130,289,167	283,256,614	Gamma
Hierarchy Level					
Hardware / Software / Mat'l.	1,508,290	356,516	492,737	1,077,888	LogLogistic
Component / Part	2,320,795	427,230	600,295	1,366,661	LogLogistic
Assembly	6,439,009	855,392	1,308,794	3,661,668	LogLogistic
Subsystem	20,461,546	2,327,053	3,946,668	13,457,236	Lognorm
System	146,709,271	42,205,134	77,094,954	230,367,198	Gamma

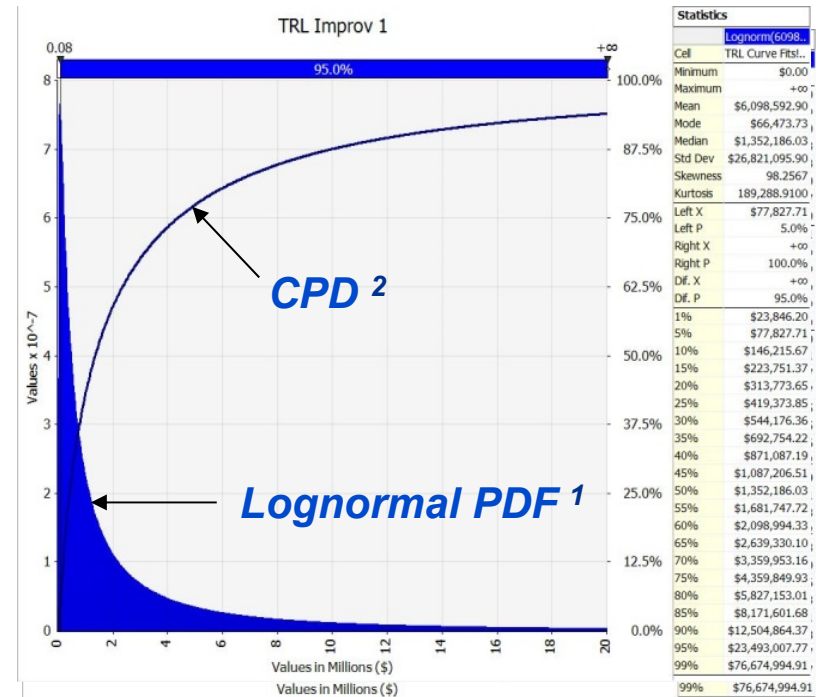
Curve fit model predicted costs in FY15\$K

Cost Model Output - Sample Curve Fit Cost Model Uncertainty (Model no. 1)

Cost Model No. 1: TI Level 1 Sample Data with Higher Performing Curve Fits



Cost Model No. 1: TI Level 1 Selected Curve Fit Model



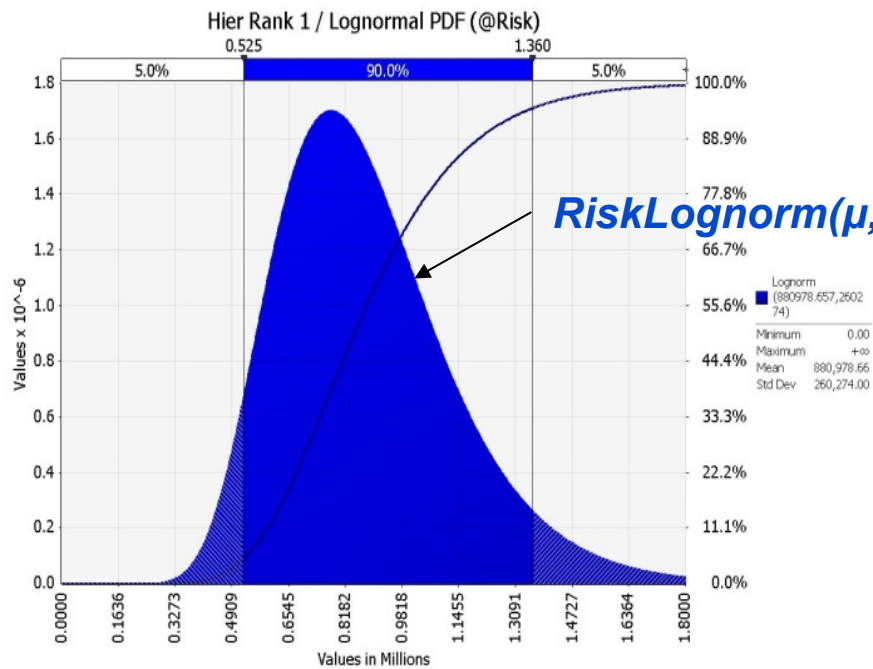
- **Uncertainty PDFs were also developed for the *other* Curve Fit Model ordinal levels (see backup slides)**
 - **Cost Model no. 1 - TI Levels 2 to 5**
 - **Cost Model no. 2 - SH Levels 1 to 5**
1. PDF - Probability Density Function
 2. CPD - Cumulative Probability Distribution (a.k.a. ogive or “S-curve”)



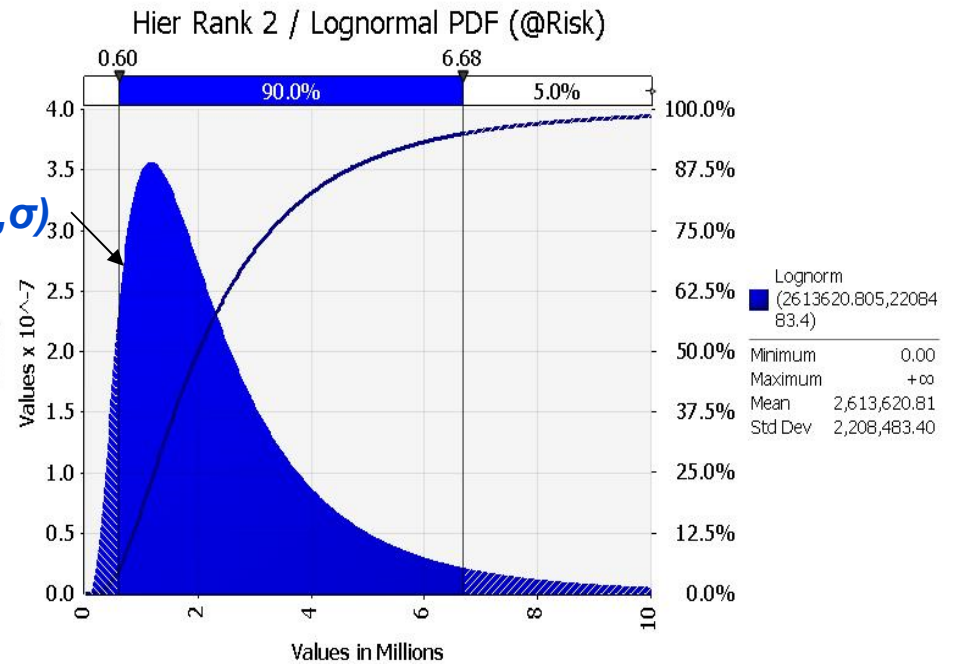
Cost Model Output - Sample Linear Regression Cost Model Uncertainty (Model nos. 5 & 6)

- Sample Uncertainty PDF for Cost Model 5 at SH Level 1 and Cost Model 6 at SH Level 2 (FY\$15)
- Uncertainty PDFs also developed for other SH Levels:
 - Cost Model 5 SH Levels 2, 3, 4, 5 and Cost Model 6 SH levels 1, 3, 4, 5

**Cost Model No. 5:
SH level 1**



**Cost Model No. 6:
SH level 2**

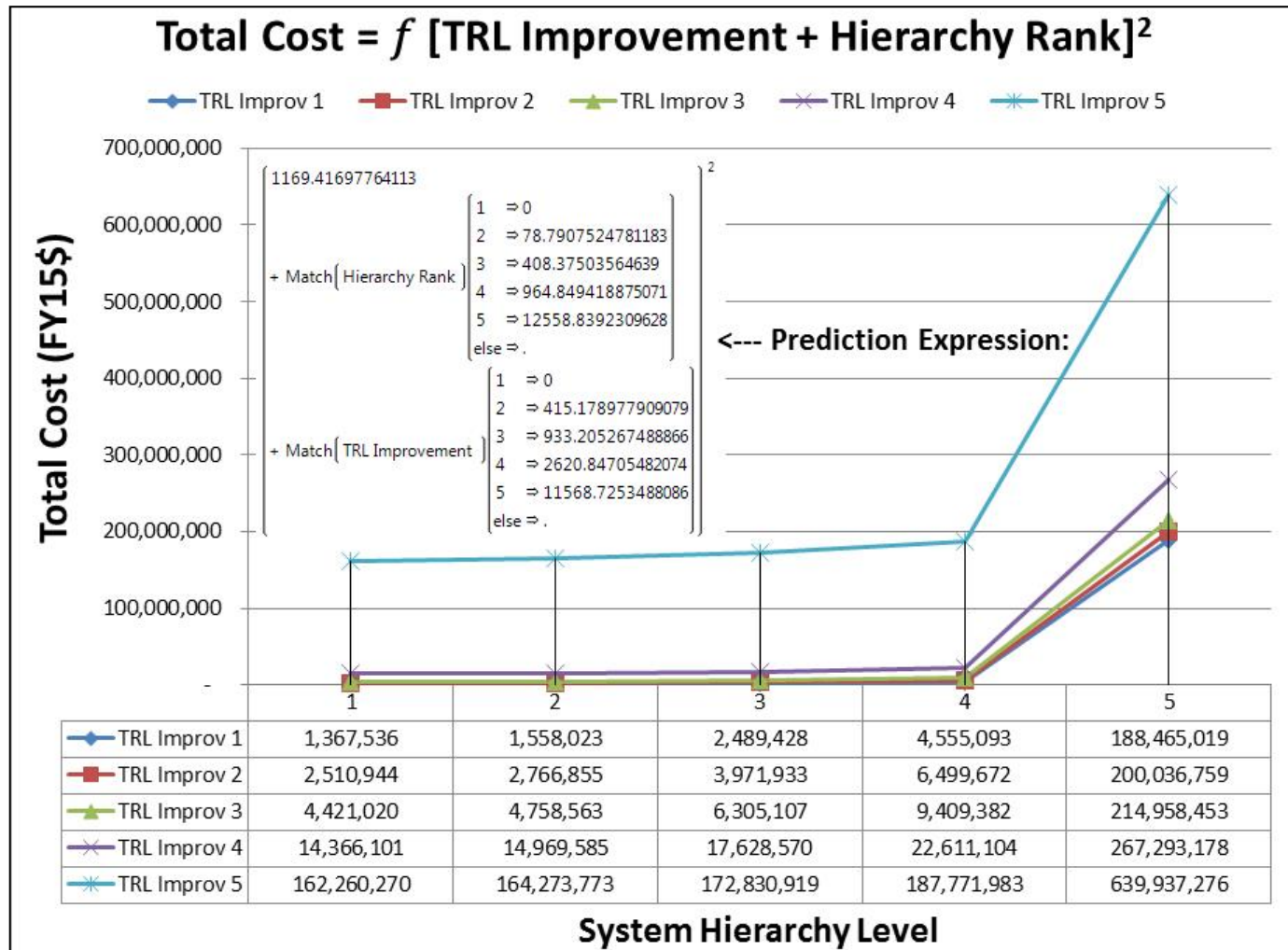


Cost Model Output - Sample Multivariate Regression Cost Model (Model no. 9)*

- Transformation – f [squared \sum of predictor variables]
 - Cost = $c + (a \cdot \text{TI level} + b \cdot \text{SH level})^2$ (expression below)

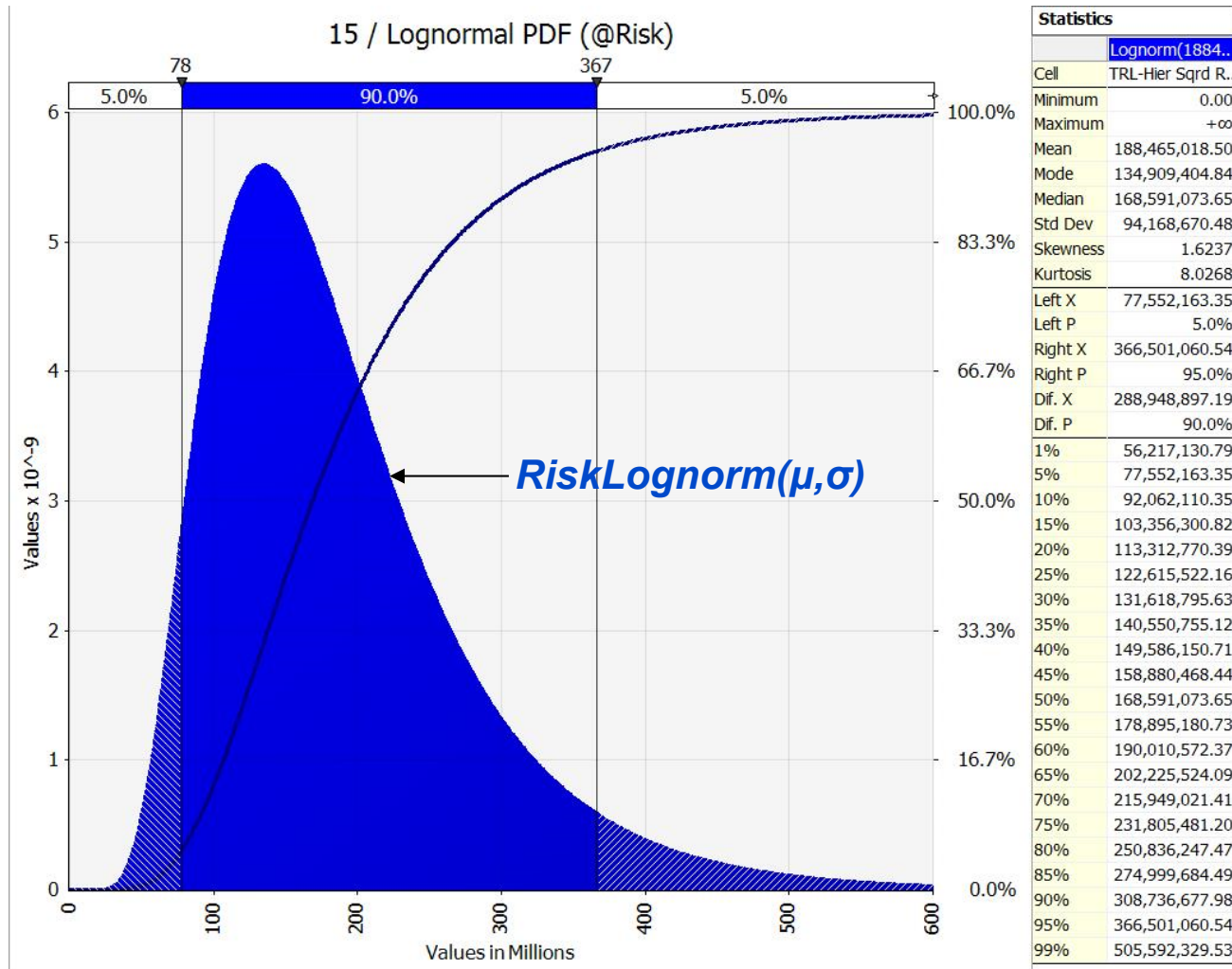
TI level more dominant than SH level based upon coefficients and total response at same ordinal levels up through level 4.

* Other Multivariate Regression Models 7 & 8 provided in backup slides



Cost Model Output - Sample Multivariate Regression Cost Model Uncertainty (Model No. 9)

- Composite Linear Regression (Model No. 9) – Sample PDF Uncertainty for TI Level 1 and SH Level 5 (24 other TI / SH Level PDFs also created)



Cost Model Output – Nonlinear Cost Models (nos. 10, 12 & 13)

■ Nonlinear TI Models 1, 2

➤ Cost Model no. 10 (TRL NL-Quadratic)

○ $Cost = a + b \times TI \text{ level} + c \times TI \text{ Level}^2$

■ Nonlinear SH Models 2

➤ Cost Model no. 12 (Hier NL-Exponential 3P)

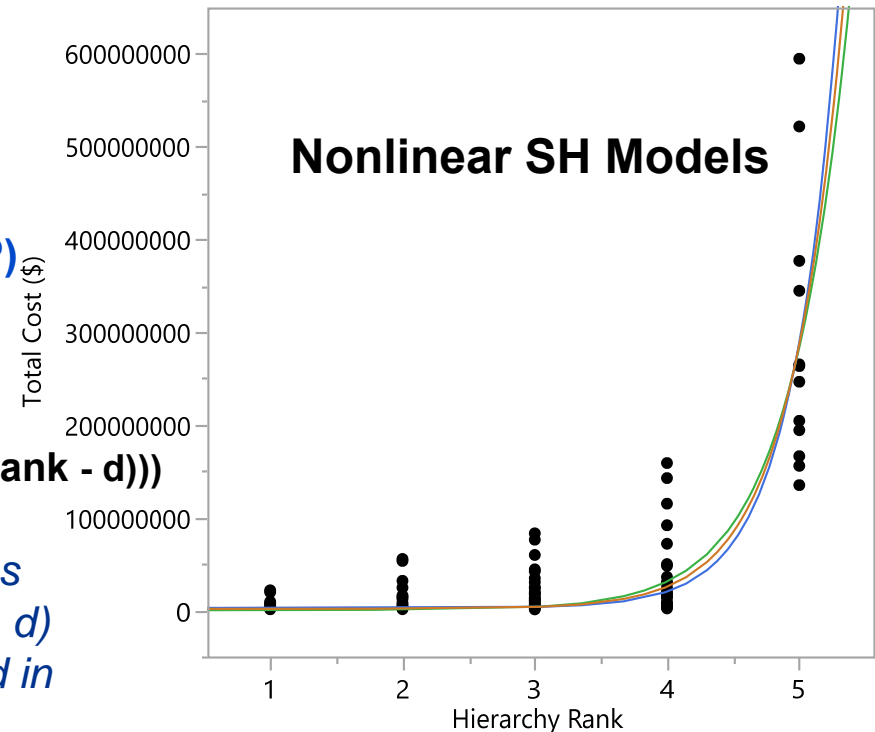
○ $Cost = a + b \times \exp(c \times \text{Hierarchy Rank})$

➤ Cost Model no. 13 (Hier NL-Gompertz 4P)

○ $Cost = a + (b - a) \times \exp(-\exp(-c \times (\text{Hierarchy Rank} - d)))$

1. Plot of TI models provided in backup slides
2. Values for expression parameters (a, b, c, d) and graph for all NL Cost models provided in backup slides

Plot



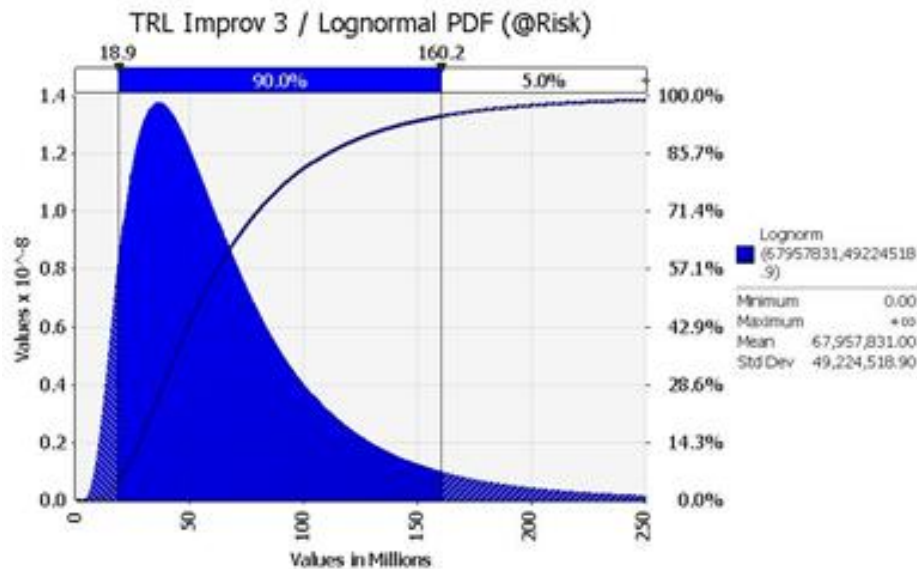
Nonlinear SH Cost Models

Model Comparison

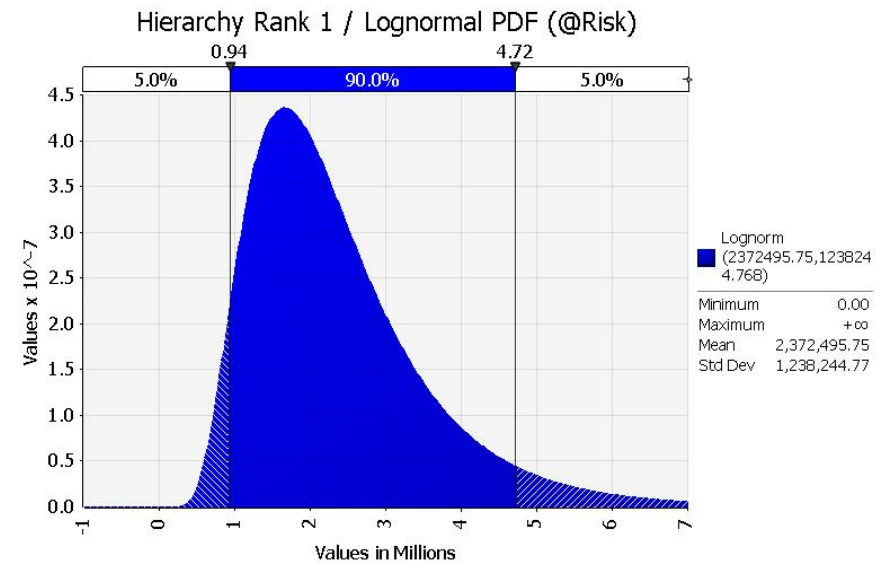
Model	AICc	AICc Weight	BIC	SSE	MSE	RMSE	R-Square
Exponential 3P	19748.669	0.9262116	19765.723	3.316e+17	6.233e+14	24966011	0.7436299
Gompertz 4P	19753.763	0.0725263	19775.061	3.335e+17	6.281e+14	25060990	0.7421611
Logistic 3P	19761.866	0.001262	19778.919	3.399e+17	6.389e+14	25275833	0.7372274

Cost Model Output – Nonlinear Cost Model Uncertainty (nos. 10 & 12)

NL Cost Model No. 10: TI Level 3 Example PDF



NL Cost Model No. 12: SH Level 1 Example PDF



- All other Nonlinear Cost Model Ordinal Level Uncertainty PDFs (15 in total) also developed

- Cost Model no. 10 (Quadratic) - TI levels 1, 2, 4, 5
- Cost Model no. 12 (Exponential 3P) - SH levels 2 to 5
- Cost Model no. 13 (Gompertz 4P) - SH levels 1 to 5

Cost Model General Applicability

Mdl. No.	Model Form / Method	Predictor Variable Form	Model Performance and Technology Development Attributes				
			Best Project Sample Data Fit	Generally Higher KPM Performance	System Level Development (SH level 5)	Below System Level Development (SH Level 1-4)	Generally Higher Cost or Uncertainty Levels*
1	Tailored Curve Fits	TI Level	✓			✓	
2	Tailored Curve Fits	SH Level	✓		✓		
5	Simple Linear Regression	SH Level			✓		
6	Simple Linear Regression	SH Level			✓		✓
7	Composite Linear Regression	$[TI \times SH]^2$		✓	✓	✓	
8	Multiple Linear Regression	TI + SH		✓	✓	✓	✓
9	Multiple Linear Regression	$[TI + SH]^2$	✓	✓	✓	✓	✓
10	Nonlinear - Quadratic	NL TI Level				✓	✓
12	Nonlinear - Exponential 3P	NL SH Level			✓		✓
13	Nonlinear - Gompertz 4P	NL SH Level			✓		✓

* *May be more applicable for higher risk or technology volatility developments*

Schedule Model Performance

- The same model forms were developed and assessed for schedule-based modeling
 - Ordinal Curve Fits, Linear Regression (Univariate and Multivariate including a range of transformations) and Non-linear
 - Dependent variable - Development Project Duration (months)
 - Independent Predictor variables - TI Level, SH Level and Project Spend Rate (investment \$/mo.) added to augment analysis
- Results did not produce the same strength of relationship with the independent predictor variables as experienced with Cost
 - Exception was SH Level Curve Fit model (available KPM below)

Model No.	Fit Model Type	Single / Multiple Predictor Variable(s)	Predictor Type	Reference Model Name	Predictor Variable(s)	Key Performance Measures (KPM)				
						RMSE (months)	Coef. of Variation (CV)	No. of Available Obs.	No. of Applied Obs.	Data Reduction (%)
1	Tailored Curve Fits	Single	System Hierarchy Level	Hier Curve Fits	Hierarchy Rank	20	0.755	551	551	0.0%

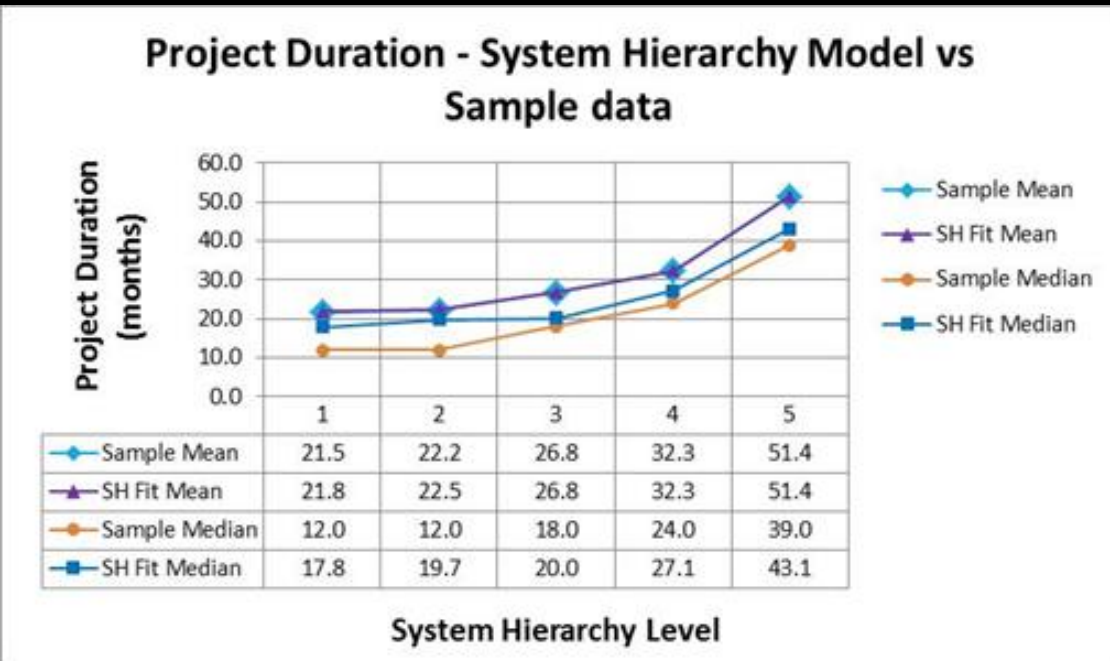
PERFORMANCE RATING			
Good	Fair	Marginal	Poor

Schedule Model Output - SH Level Schedule Model

Schedule Duration (months) vs. System Hierarchy Level Curve Fit

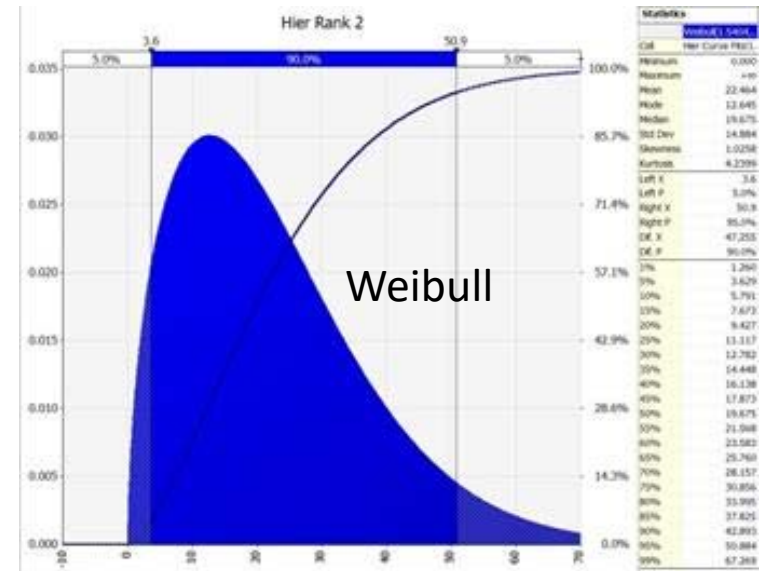
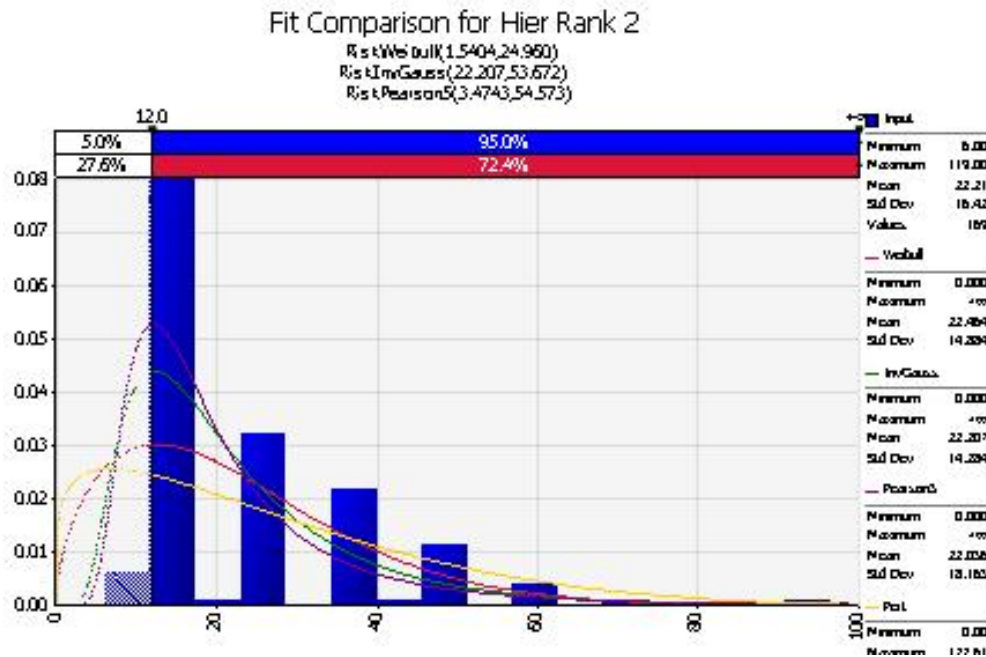
Predictor Level / Tier	Number of Observations	Mean	Median	60th %ile	80th %ile	Curve Function Type
System Hierarchy Level						
Hardware / Software / Mat'l.	98	21.8	17.8	20.4	28.5	Pearson5
Component / Part	169	22.5	19.7	23.6	34.0	Weibull
Assembly	173	26.8	20.0	24.4	38.6	InvGauss
Subsystem	86	32.3	27.1	32.7	48.3	Erlang
System	25	51.4	43.1	52.0	77.0	Erlang
	551	0.0% Data Reduction				

- Consistent cost growth across key benchmark levels with best results from Pearson5, Weibull, Inverse Gaussian and Erlang distributions
- Summary chart demonstrates the closeness of fit to sample mean



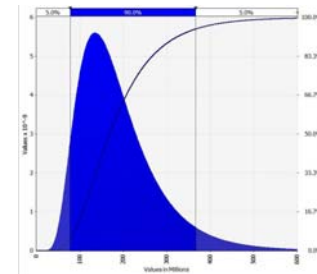
Schedule Model Output - Sample Curve Fit Model Uncertainty

Example Schedule Curve Fits & Selected PDF – Project Duration (months) for System Hierarchy Level 2



Cost and Schedule Model Uncertainty Drivers

- **Range of Technologies in Project Data**
 - Diverse TAs found in the database may contain varying considerations for R&D activities that can drive both cost and schedule
- **TRL and SH Level Assessment Variability**
 - TRL and SH level assessments are subjective qualitative valuations that can vary by source
- **Cost Data Variability / Normalization**
 - Scope and tracking of budgeting, cost accounting methods / categories, contractual CLINs, and indirect costs captured can vary across projects
- **Source Data Characteristics**
 - Data sample sizes are good but unexpected overabundance of smaller projects across higher predictor variable levels
- **Model Forms**
 - Output variability between or across model forms can be related to the nature of particular model relationship characteristics or constraints



Conclusions and Future Work

- **TI and SH macro variable models produced good statistical KPM and goodness-of-fit characteristics but w/ significant variability**
- **Deliver forecasting value above very ROM estimates and SME opinion often applied in early technology development**
- **Other “macro-level” cost & schedule parameters to consider for early stage technology development estimating:**
 - **Research and Development Degree of Difficulty (RD3)**
 - **Capability Demonstrations**
 - **Advanced Degree of Difficulty (AD2)**
 - **System Readiness Level (SRL)**
 - **Integration Readiness Levels (IRL)**
 - **Implementation Readiness Level (ImpRL)**
 - **Manufacturing Readiness Level (MRL)**
 - **Macro-level technology performance or complexity factors**

Questions?

email: chuck.alexander@jhuapl.edu

Additional Information

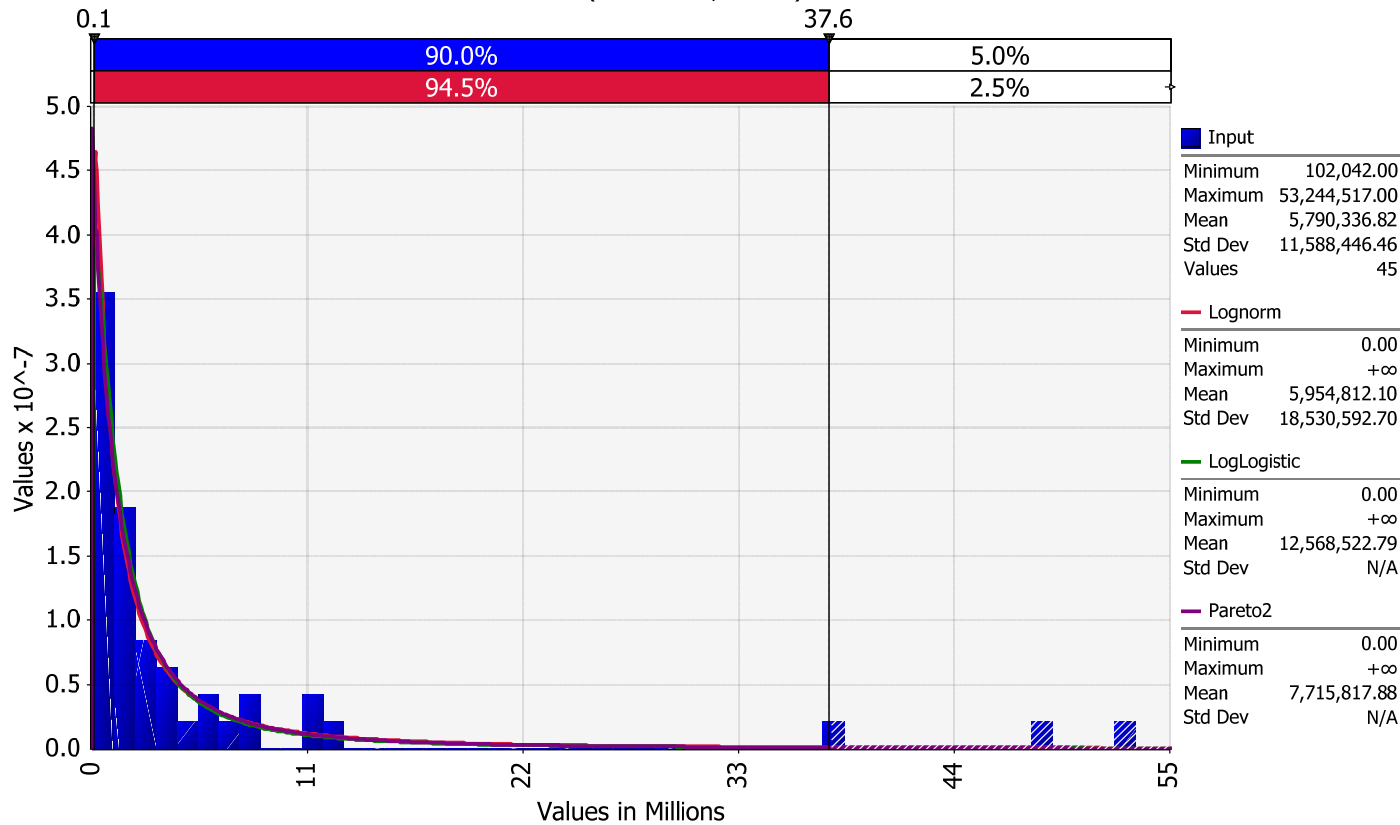
Sample TRL X-Y Transition Data for TRL 2-3

Fit Comparison for TRL 2-3 Cost Curve Fit

RiskLognorm(5954812.1,18530592.7)

RiskLogLogistic(0,1823661,1.1492)

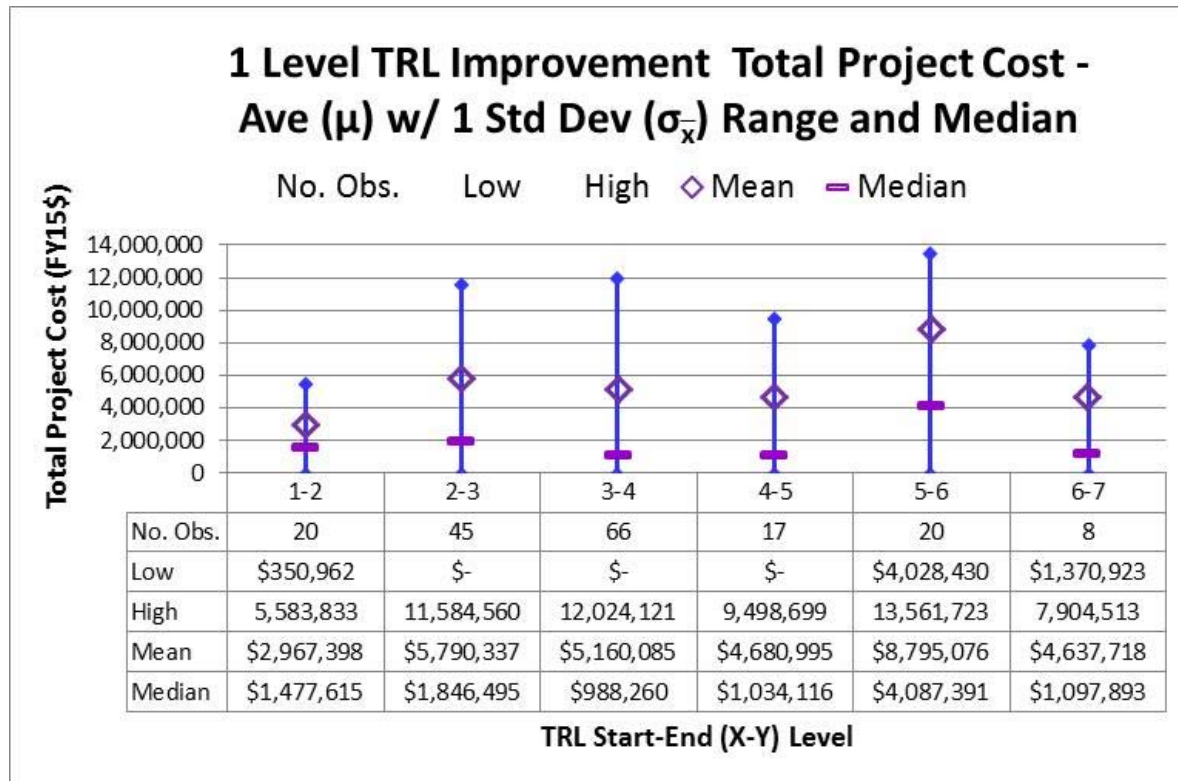
RiskPareto2(2830162.0,1.3668)



Input	
Minimum	102,042
Maximum	53,244,517
Mean	5,790,337
Mode	≈102,093.00
Median	1,846,495
Std Dev	11,588,446
Skewness	3.3134
Kurtosis	13.6324
Left X	102,148
Left P	5.00%
Right X	37,599,597
Right P	95.00%
Dif. X	37,497,449
Dif. P	90.00%
	1%
	5%
	10%
	15%
	20%
	25%
	30%
	35%
	40%
	45%
	50%
	55%
	60%
	65%
	70%
	75%
	80%
	85%
	90%
	95%
	99%

Data Modeling – Cost vs. TI Level Relationship Screening

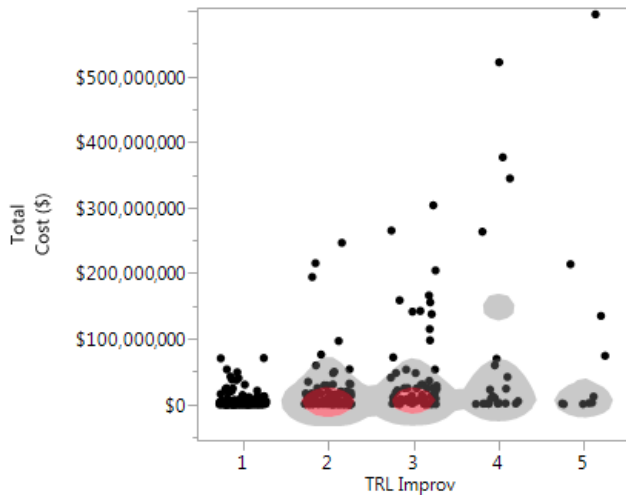
- Data relationship screening for Cost vs. TI level showed stability of a relevant range across TRL Start-End (TRL X-Y) levels
- Representative example plot for for TI Level 1 is shown below (TI Levels 2 through 5 also assessed with similar results)
- This analysis plus other screening techniques supported the use of TI level as an independent predictor variable



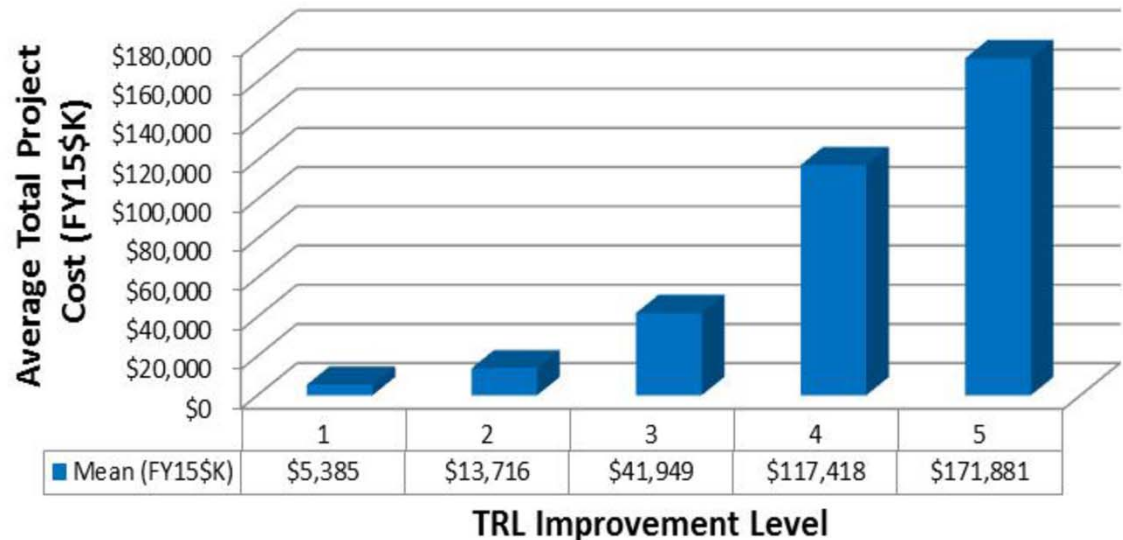
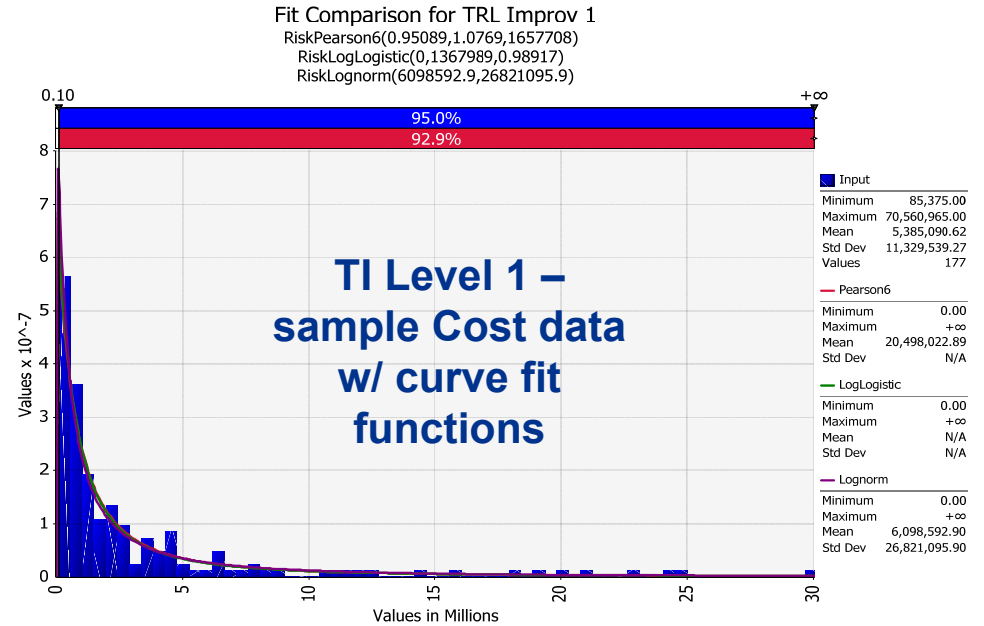
Data Modeling – Cost vs. TI Level Relationship

Project Cost (mean) vs. TI Level

- 405 project record dataset
- Direct relationship of Cost to TI level evident
- Geometric cost growth up through level 4, tapering off at level 5



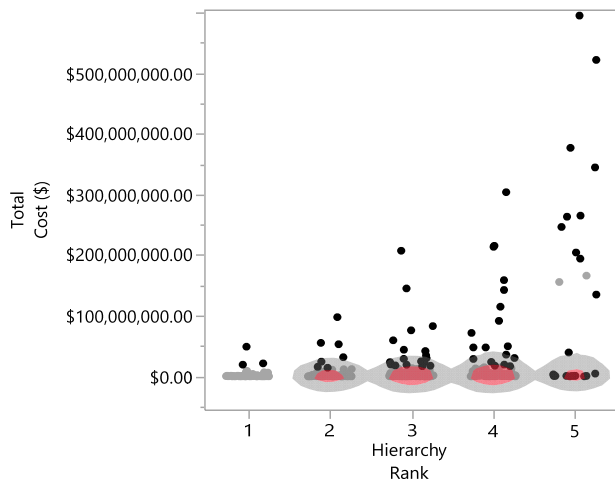
Scatterplot : Total Project Cost vs TI Level



Data Modeling – Cost vs. SH Level Relationship

Project Cost (mean) vs. SH Level

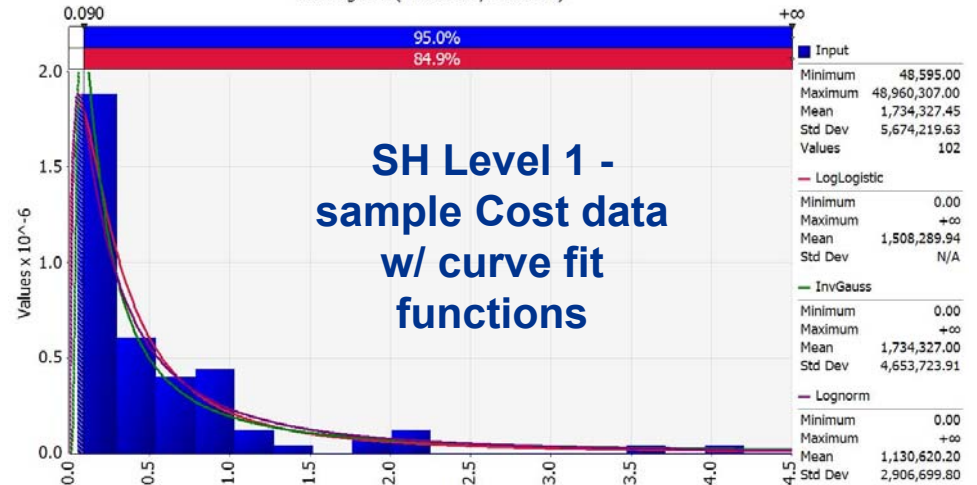
- 603 project record dataset
- Gradual moderate growth up to Subsystem level (4)
- Dramatic increase at the System level (5) suggests possible exponential relationship



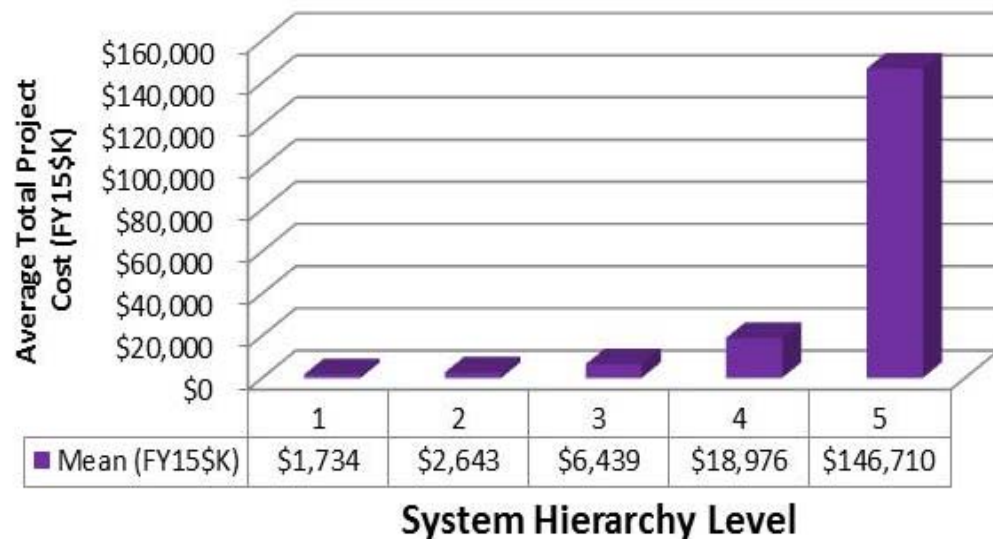
Scatterplot : Total Project Cost vs SH Level

Fit Comparison for Hier Rank 1

RiskLogLogistic(0,356516,1.2530)
 RiskInvGauss(1734327,240875)
 RiskLognorm(1130620.2,2906699.8)



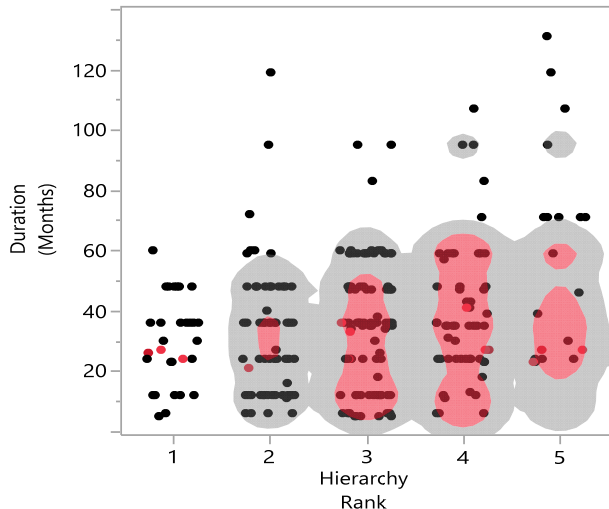
SH Level 1 - sample Cost data w/ curve fit functions



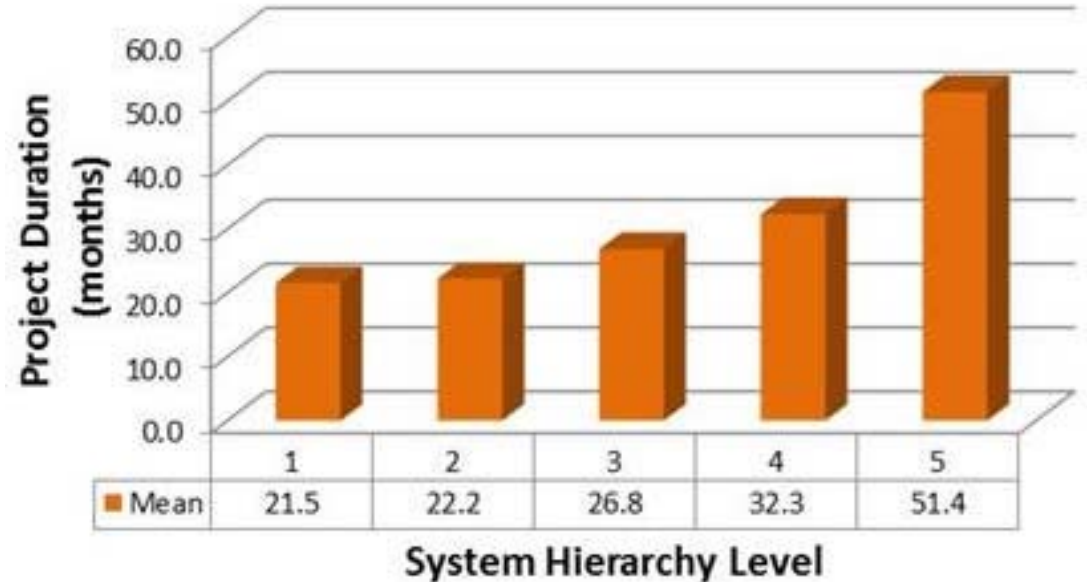
Data Modeling – Schedule vs. SH Level Relationship

Project Duration (mean) vs. SH Level

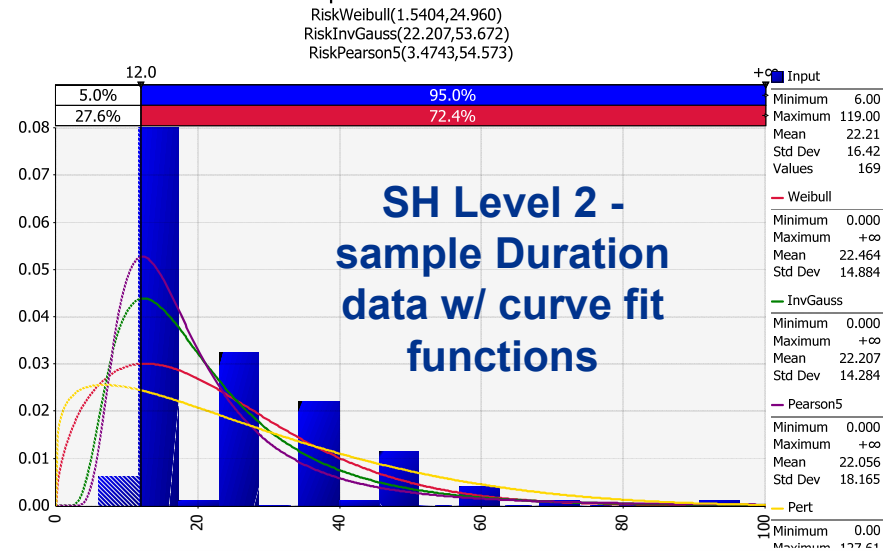
- 551 project record dataset
- Direct relationship between SH level and Project Duration indicated



Scatterplot : Project Duration vs SH Level



Fit Comparison for Hier Rank 2



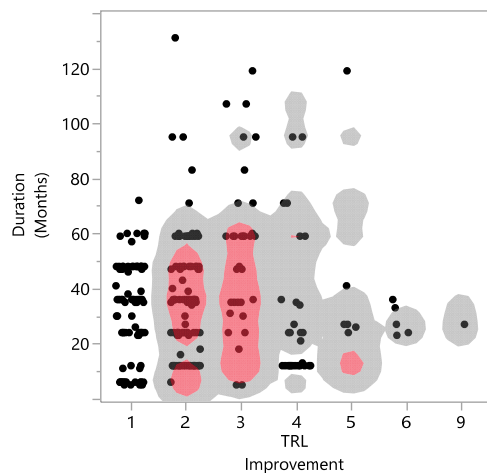
**SH Level 2 -
sample Duration
data w/ curve fit
functions**



Data Modeling – Schedule vs. TI Level Relationship

- A poor affiliation between Project Duration and TI Level is indicated
 - Random data distribution, lack of obvious visual patterns, substantial nonparametric density areas
 - Moderate data correlation ($r = 0.3238$)
 - Columnar chart suggesting the mean project duration does not possess a continuous functional association with the TI level, peaking and then tailing off at level 3

Scatterplot :Project Duration vs TI Level



Data Modeling – Cost and Schedule Model Datasets

- **Data set size limited by projects with valid¹ corresponding predictor and response variable values**
 - **TRL Start and End levels**
 - **System Hierarchy Level**
 - **Total Project Development Cost**
 - **Project Start and End Dates**
- **Resulting data sets available for modeling²**
 - **Total Project Cost vs. TI Level (405 / 395 available project records for cost / schedule models)**
 - **Total Project Cost vs. SH Level (603 / 551 available project records for cost / schedule models)**
 - **Total Project Cost vs. TI Level and SH Level (221 available project records for both cost and schedule models)**

1. Project records with zero, blank or erroneous values removed

2. Not all records for each data set had available project start or end dates so total number of records for schedule duration modeling was slightly less

Key Performance Measure (KPM) Descriptions

▪ Error Variability and Dispersion Measures:

- **Coefficient of Determination - R^2 and Adjusted R^2 .** Most commonly used measure of “goodness of fit”. Relative measure of fit equal to the percent of the variation in the dependent variable (Y) explained by the independent variable (X) = SSR^1 / SST .
- **Root Mean Square Error (RMSE)** – absolute measure of fit or accuracy based upon the differences between sample and population values predicted by a model.
- **Coefficient of Variation (CV)** – RMSE for models, as applied here (vs. Standard Deviation used for individual variables), divided by mean of the Y-data, a unitless relative measure of estimating error (CV < 1 is considered low-variance and CV > 1 considered high)

▪ Statistical Significance Measures:

- **F-ratio** - tests if the entire regression equation is valid (i.e., how well the statistical model is fitted to a sample data set).
- **t-stat** - tests if the individual hypothesized predictor (X-variables) values are valid. t-stat represents the calculated difference represented in units of standard error. The % of expression terms with probability > |t| was applied as an overall measure.

▪ Autocorrelation Measure:

- **Durbin-Watson test** - measures independence of regression residuals.

▪ Data Reduction Measure:

- **Percent (%) of original data sample set unused.** The extent of selectivity in actual data set applied, measured as the % of available sample observations filtered out due to outliers, large residuals or non-core data, etc.

1. SSR represents the sum of squares due to the regression and SST represents the sum of squares total.

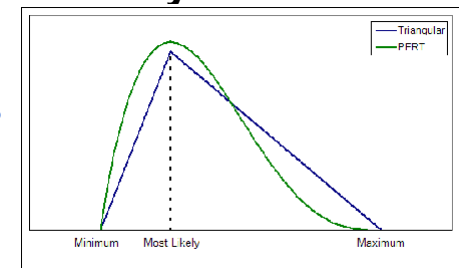
Modeling Uncertainty

■ Uncertainty function evaluation

- Lognormal, PERT, normal and triangular uncertainty PDFs evaluated
 - Inputs necessary to drive these functions (e.g., sample mean, min, max, mode, standard deviation, etc.) were available in most cases
- Significant right-skewed PDFs found for actual Cost and Schedule TI and SH ordinal level sample data
- Lognormal, Gamma and LogLogistic functions were generally highest performing across curve fits within the relevant data range

➤ Uncertainty function selection

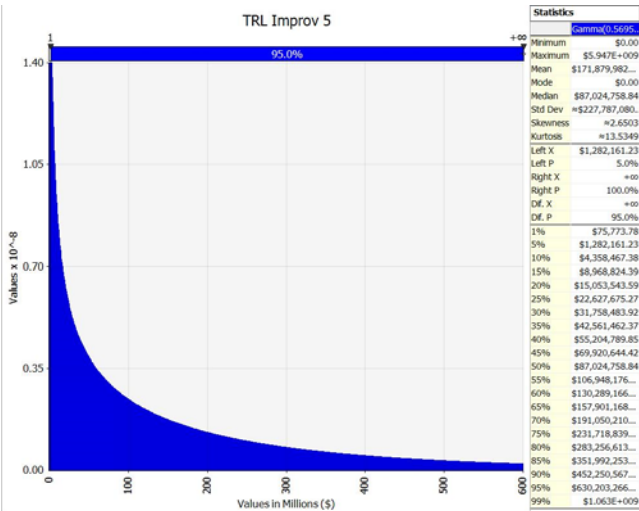
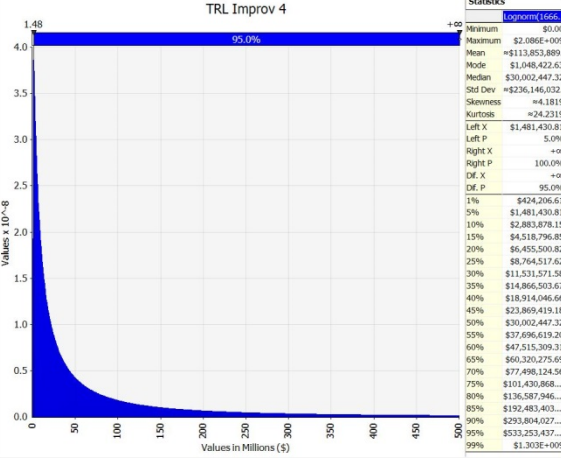
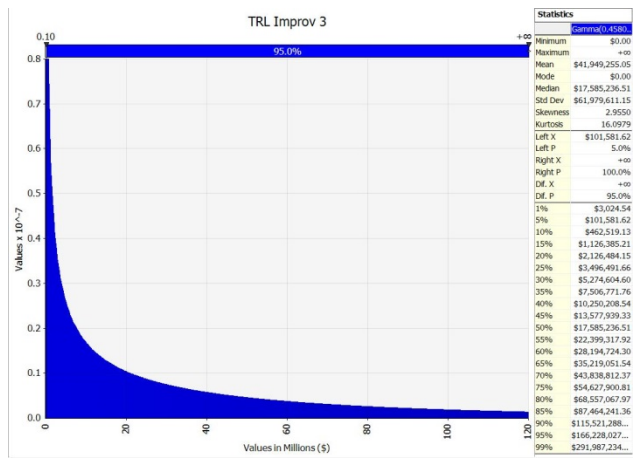
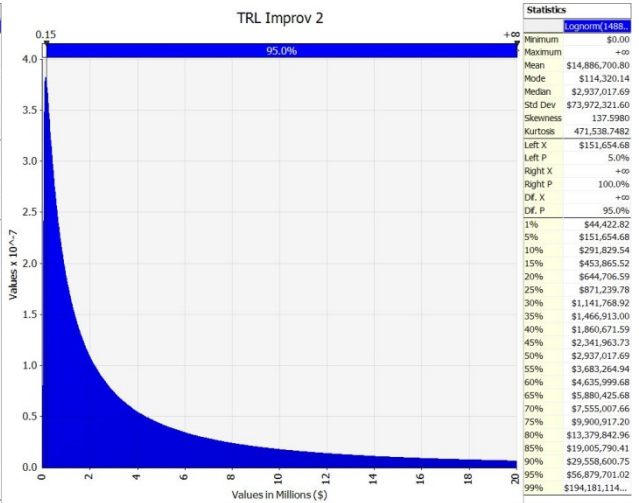
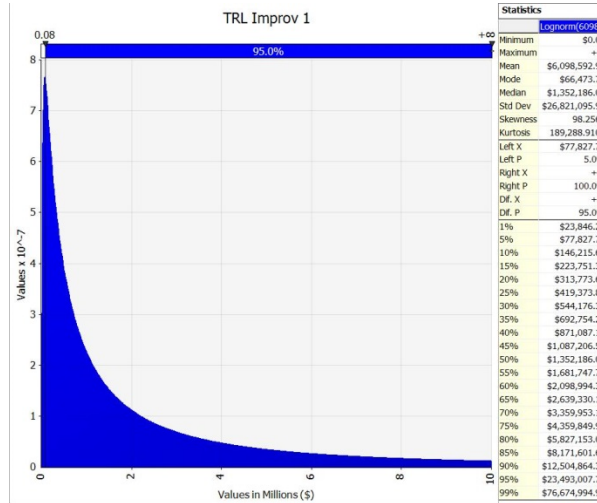
- PERT and Lognormal functions are generally considered superior to the triangular and normal distributions for modeling cost uncertainty
 - Symmetrical normal function poor fit for right skewed data
 - Lognormal and PERT functions deliver natural, continuous distributions with less tendency to overemphasize direction of skew within normal planning range (50th to 80th %ile)
- Lognormal function was generally high performing across curve fits and also closely resembled other high performing Gamma and LogLogistic functions within the planning range



Cost Model No. 1 Selected Curve Fits – Total Project Cost (FY15\$k) vs TRL Improvement Level

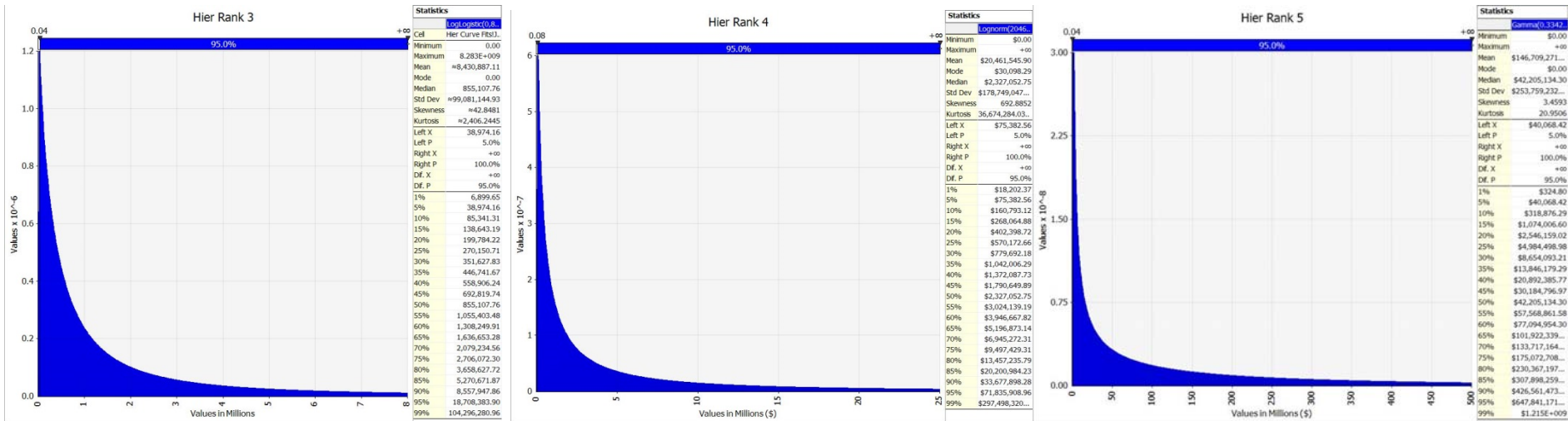
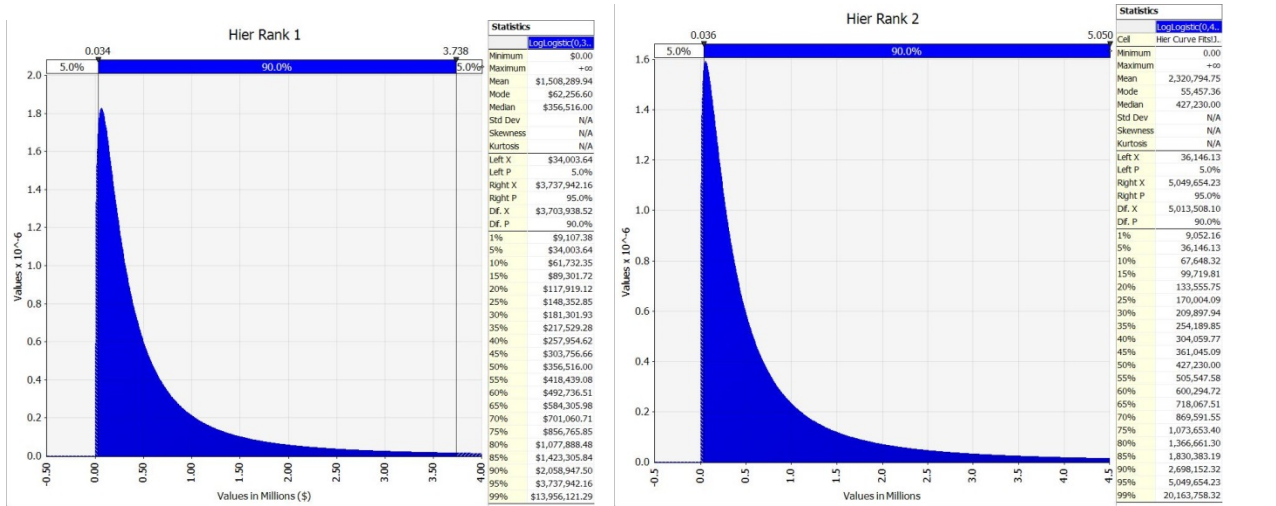
TI Level Curve Fit model PDFs with ventiles by ordinal level

- TI levels 1, 2, 4 - LogLogistic
- TI levels 3 & 5 - Gamma



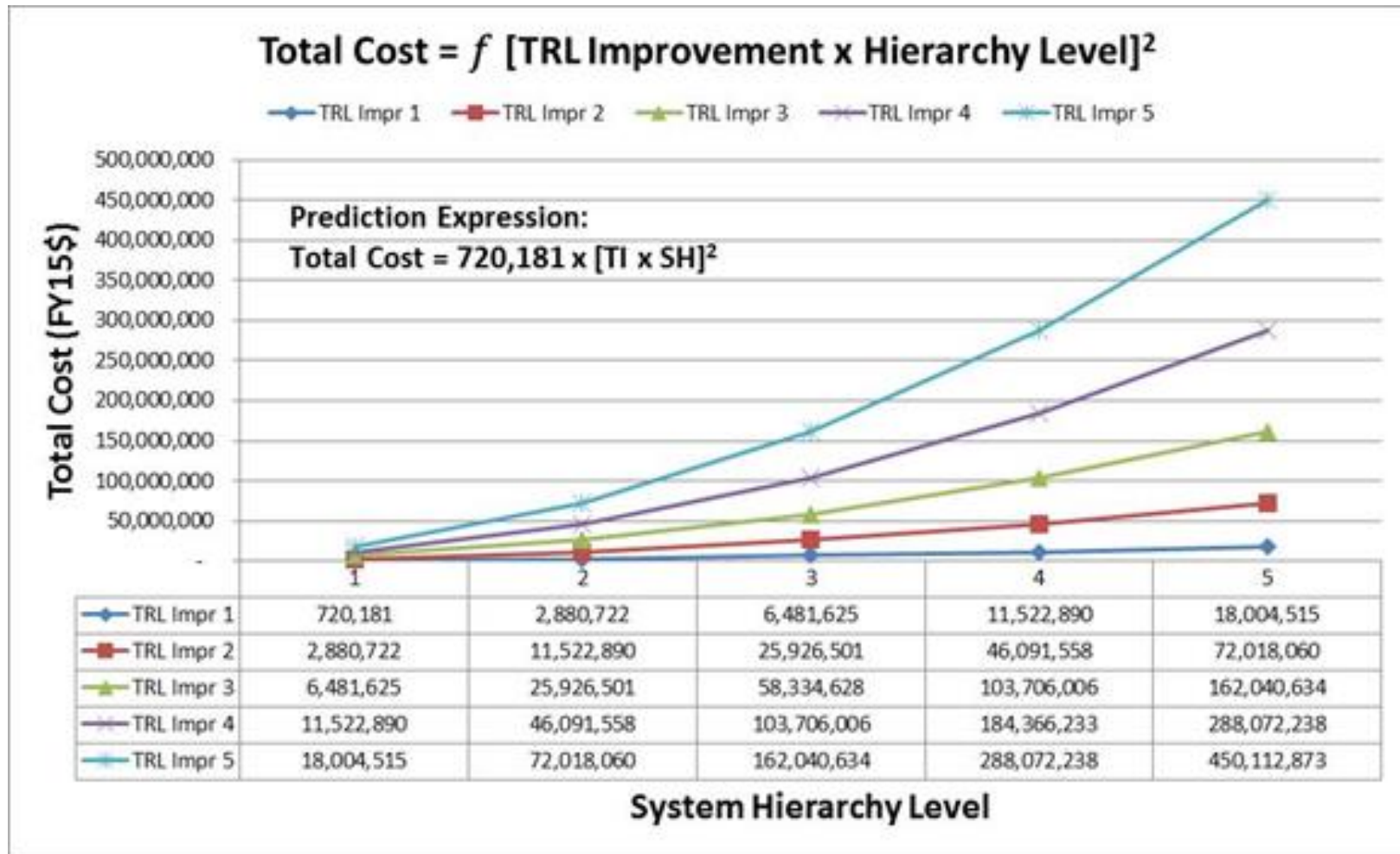
Cost Model No. 2 Selected Curve Fits – Total Project Cost (FY15\$k) vs System Hierarchy Level

- SH Level Curve Fit model PDFs with ventiles by ordinal level
 - SH levels 1, 2, 3 - LogLogistic
 - SH level 4 - Lognormal
 - SH level 5 - Gamma



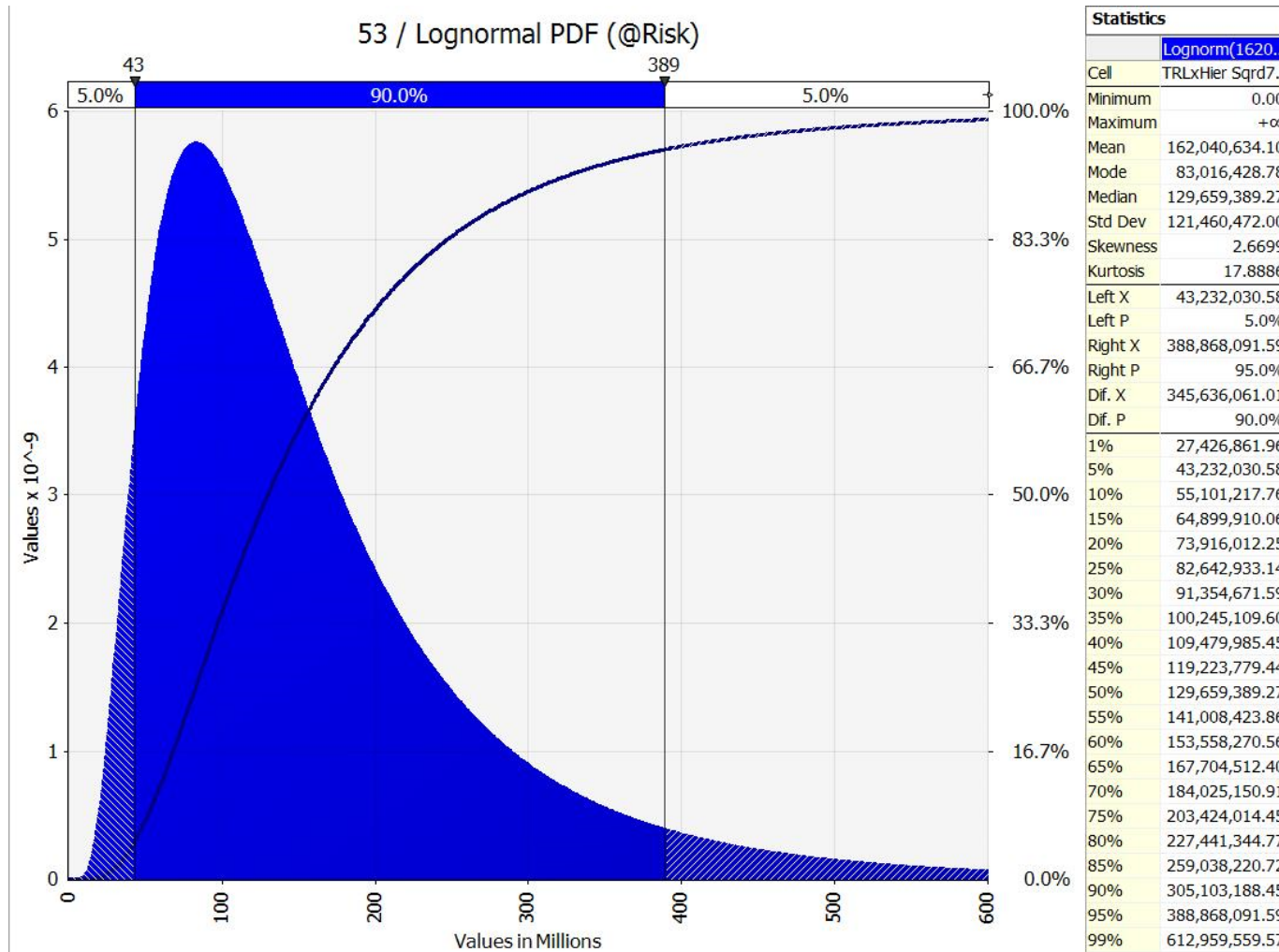
Multivariate Models – Multiple Regression Model no. 7

- Product of predictors squared transformation (expression below)
 - Total Cost = f [TRL Improvement x Hierarchy Level]²



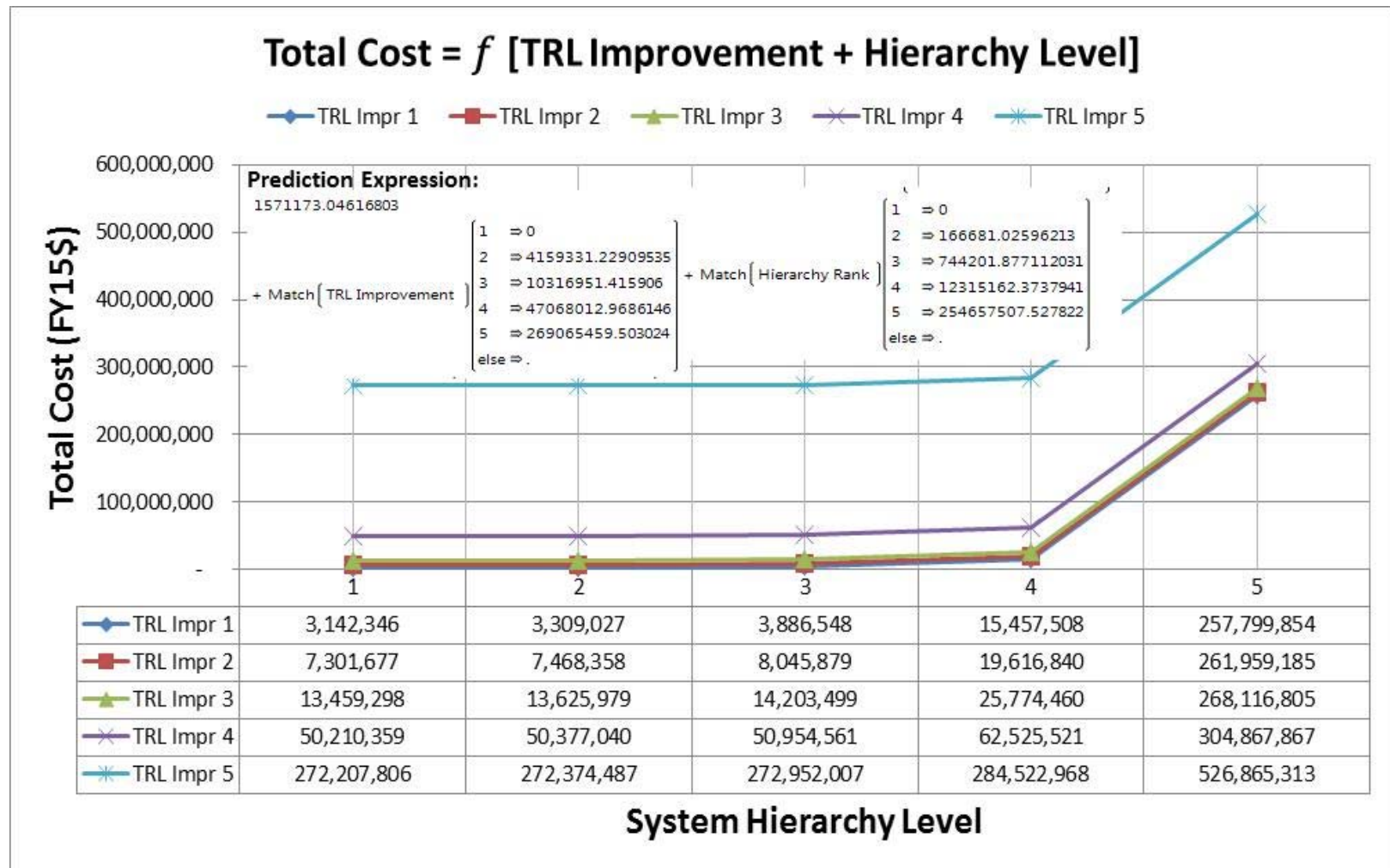
Multivariate Regression Cost Model Output - Sample Cost Model Uncertainty (Model No. 7)

- Composite Linear Regression (Model No. 7) – Sample PDF Uncertainty for TI Level 5 and SH Level 3 (24 other TI/SH Level PDFs also developed)



Multivariate Models – Multiple Regression Model no. 8

- Linear first order TI + SH predictor variable expression below
 - Constant intercept plus graduated TI and SH coefficients by level



Cost Model Output – Nonlinear TI Cost Models

Nonlinear TI Level Cost Model

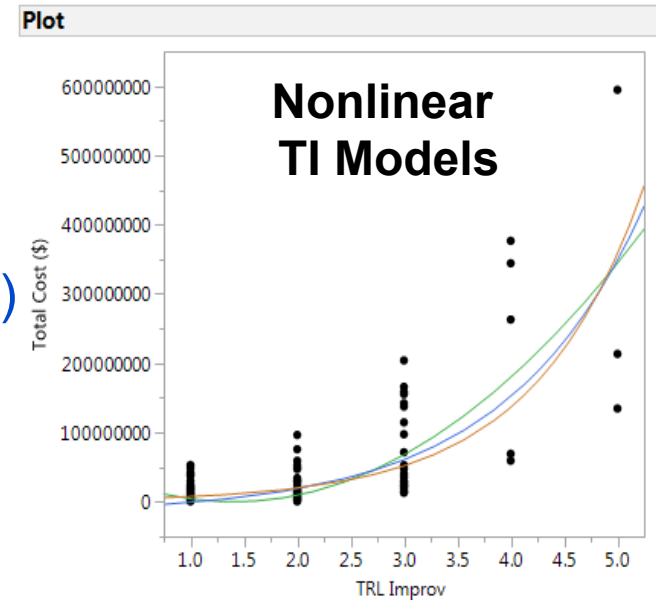
■ Nonlinear TI Models:

➤ Cost Model no. 10 (TI NL-Quadratic)

○ $Cost = a + b \times TI \text{ level} + c \times TI \text{ Level}^2$

➤ Cost Model no. 11 (TI NL-Exponential 2P)

○ $Cost = a \times (1 - b \times \exp(-c \times TI \text{ level}))$



Nonlinear TI Models

Model Comparison

Model	AICc	AICc Weight	BIC	SSE	MSE	RMSE	R-Square
Quadratic	12885.399	0.9999998	12900.644	3.643e+17	1.068e+15	32684768	0.6097699
Mechanistic Growth	12916.095	2.16e-7	12931.34	3.983e+17	1.168e+15	34176060	0.5733479
Exponential 2P	12929.287	2.95e-10	12940.738	4.163e+17	1.217e+15	34890388	0.5540222

Model Output – Nonlinear Cost Models

Model No.	Fit Model Type	Predictor Type	Single / Multiple Predictor Variable(s)	Predictor Variable / Parameter	Predictor Level / Parameter Name	Predictor Level / Parameter	Parameter Est. / No. Obs.	Prediction Estimate	Median	60th %ile	80st %ile	Mode	
10	Nonlinear - Quadratic	TRL Improvement Level	Single	TRL Improvement	TRL Improvement 1	1		4,331,991	1,006,162	1,551,226	4,238,667	54,279	
					Summary of Fit	Ref Model Name: TRL NL - Quadratic		TRL Improvement 2	2		9,551,810	2,743,392	4,093,405
		AICc	12,885		TRL Improvement 3	3		67,957,831	55,036,648	64,879,635	95,067,227	36,097,482	
		BIC	12,901		TRL Improvement 4	4		179,550,054	167,208,302	183,984,663	229,720,738	145,011,495	
		SSE	3.643E+17		TRL Improvement 5	5		344,328,479	331,945,687	355,504,827	416,861,098	308,500,024	
		MSE	1.068E+15		Function	Form	Quadratic						
		RMSE	32,684,768			Equation	Cost = a + b x TRL Improv + c x TRL Improv2			Lower 95%	Upper 95%		
		R-Square	0.6097699		Parameters	Intercept	a	52,298,374		36,460,718	68,136,030	N/A	
		Coef. of Variation (CV) =	1.606			Slope	b	(74,559,484)		(90,696,944)	(58,422,023)	N/A	
						Quadratic	c	26,593,101		23,012,184	30,174,017	N/A	
						Total Applied	343	15.3% Data Reduction					
11	Nonlinear - Exponential 2P	TRL Improvement Level	Single	TRL Improvement	TRL Improvement 1	1		7,552,009	7,358,007	7,795,826	8,915,566	6,984,827	
					Summary of Fit	Ref Model Name: TRL NL - Exponential 2P		TRL Improvement 2	2		19,819,678	19,071,071	20,460,087
		AICc	12,885		TRL Improvement 3	3		52,015,246	49,293,157	53,562,645	64,957,507	44,268,887	
		BIC	12,901		TRL Improvement 4	4		136,510,080	127,045,553	139,852,153	174,791,139	110,039,589	
		SSE	3.643E+17		TRL Improvement 5	5		358,260,387	326,495,146	364,157,622	469,225,970	271,164,357	
		MSE	1.068E+15		Function	Form	Exponential 2P						
		RMSE	32,684,768			Equation	Cost = a x EXP(b x TRL Improv)			Lower 95%	Upper 95%		
		R-Square	0.6097699		Parameters	Scale	a	2877586.9		1,854,215	3,900,959	N/A	
		Coef. of Variation (CV) =	1.714			Growth Rate	b	0.9648616		1	1	N/A	
							Total Applied	343	15.3% Data Reduction				
12	Nonlinear - Exponential 3P	Hierarchy Level	Single	Hierarchy Rank	Hardware / Software / Mat'l.	1		2,372,496	2,103,266	2,381,757	3,179,024	1,652,995	
					Summary of Fit	Ref Model Name: Hier NL - Exponential 3P		Component / Part	2		2,434,288	2,121,663	2,423,068
		AICc	19,749		Assembly	3		3,443,119	2,172,173	2,770,150	4,872,234	864,528	
		BIC	19,766		Subsystem	4		19,913,385	5,784,901	8,615,831	21,727,389	488,200	
		SSE	3.316E+17		System	5		288,808,534	59,985,261	93,997,585	266,732,460	2,587,698	
		MSE	6.233E+14		Function	Form	Exponential 3P						
		RMSE	24,966,011			Equation	Cost = a+b x EXP(c x Hierarchy Rank)			Lower 95%	Upper 95%		
		R-Square	0.7436299		Parameters	Asymptote	a	2368463.9		-46553.5	4783481.2	N/A	
		Coef. of Variation (CV) =	2.070			Scale	b	246.95741		-254.6162	748.53104	N/A	
						Growth Rate	c	2.7927648		2.3862295	3.1993001	N/A	
						Total Applied	535	11.3% Data Reduction					
13	Nonlinear - Gompertz 4P	Hierarchy Level	Single	Hierarchy Rank	Hardware / Software / Mat'l.	1		1,354,510	945,930	1,172,449	1,930,128	461,332	
					Summary of Fit	Ref Model Name: Hier NL - Gompertz 4P		Component / Part	2		1,508,967	800,569	1,064,833
		AICc	19,754		Assembly	3		3,349,633	861,100	1,307,399	3,447,583	56,907	
		BIC	19,775		Subsystem	4		25,284,800	5,501,148	8,562,946	23,924,522	260,400	
		SSE	3.335E+17		System	5		286,685,634	71,940,045	109,630,329	291,585,161	4,530,021	
		MSE	6.281E+14		Function	Form	Gompertz 4P						
		RMSE	25,060,990			Equation	Cost = a + (b - a) x Exp(-Exp(-c x (Hierarchy Rank - d)))			Lower 95%	Upper 95%		
		R-Square	0.7421611		Parameters	Lower Asymptote	a	8.24E+14		-6.76E+14	2.32E+15	N/A	
		Coef. of Variation (CV) =	2.078			Upper Asymptote	b	1340361.5		-1157856	3838579	N/A	
						Growth Rate	c	-2.477964		-2.780436	-2.175492	N/A	
					Inflection Point	d	11.003063		11.003063	11.003063	N/A		
						Total Applied	535	11.3% Data Reduction					





JOHNS HOPKINS
APPLIED PHYSICS LABORATORY