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# Horizontal Segmentation of Process Performance Models

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and User Group

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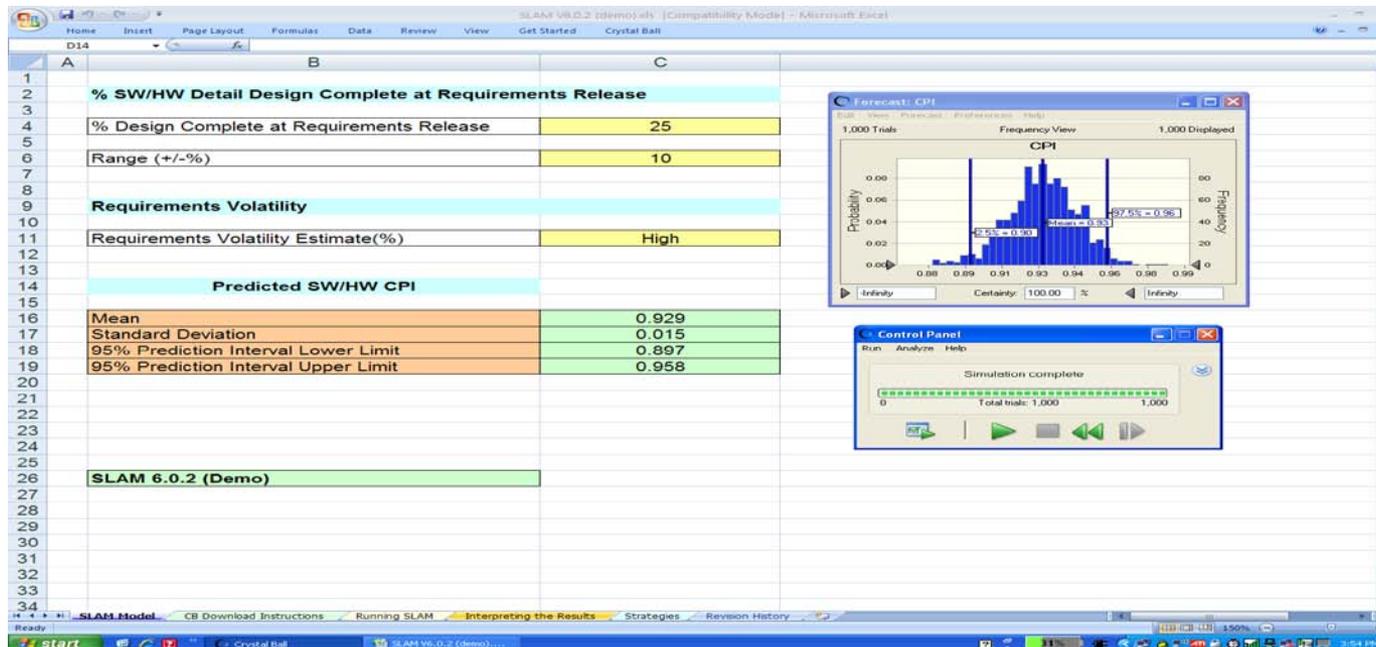
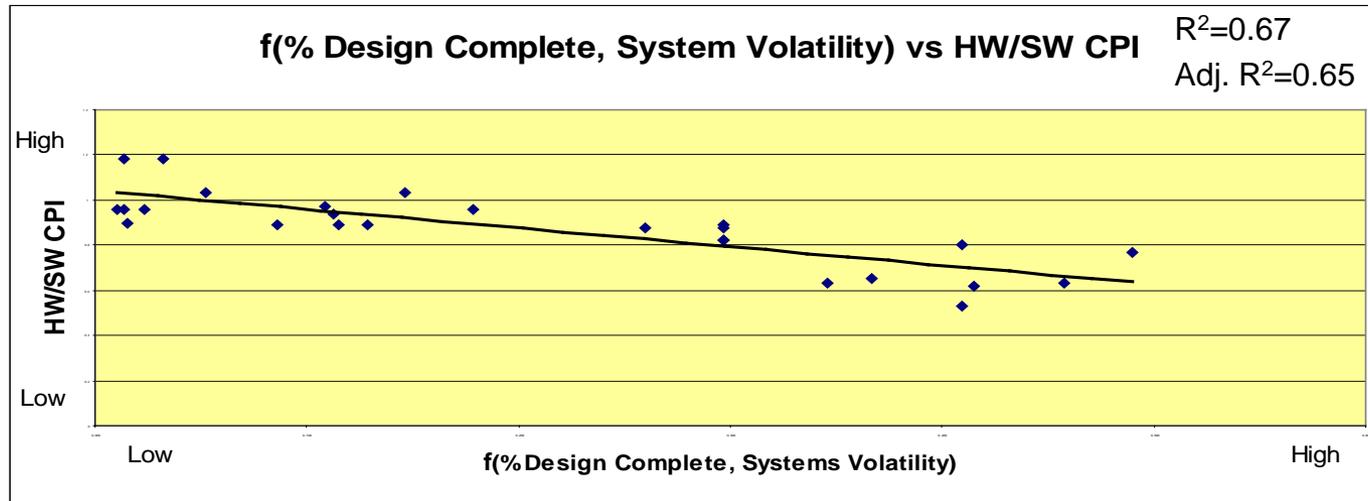
## Overview

- **Raytheon Integrated Defense Systems (IDS) introduced its 2<sup>nd</sup> Generation of Process Performance Models (2G PPMs) as part of its CMMI high maturity journey**
  - First generation models were developed to address individual lifecycle phase-specific issues
    - Designing for Quality, CAIV, cost models
    - Relationship to business and project objectives often implicit
  - 2G PPMs are used to manage risks related to business and project objectives (cost, schedule, quality) throughout the project lifecycle
    - Relationship of models to goals is explicit and enabled by model-embedded mitigation strategies
    - Supports ongoing, project grass-roots effort to achieve business and project objectives
  
- **As 2G PPMs began to proliferate, sociotechnical issues emerged**
  - Variation of individual needs became apparent
  - PPM questions related to project context manifested themselves as a series of requests for additional models and features
  - Caused model development and deployment issues related to model “packaging”
  
- **This presentation will review the use of “horizontal segmentation” as a strategy to support deployment of a “model family” that supports Raytheon business goals and model-user needs**

# System Lifecycle Analysis Model (SLAM)

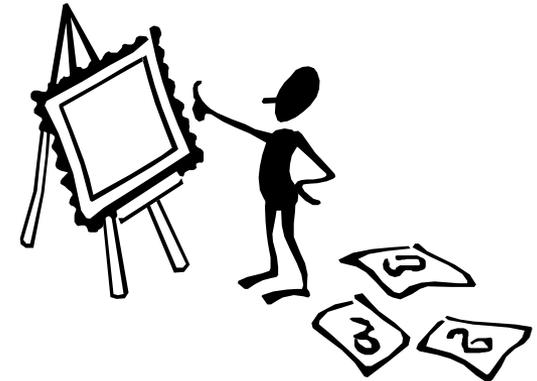
- *Initial 2G Process Performance Model*
- **Models influence of requirements volatility and requirements / design overlap on software / hardware development cost performance**
  - Enables project risk assessment & sensitivity analysis around the likelihood of achieving performance cost objectives, and the development / deployment of mitigation strategies.
- **SLAM Model Inputs**
  - Estimated % Design Complete at Systems Requirements Release
    - Confidence Range (+/- 5, 10, or 15%)
  - Requirements Volatility Estimate
    - Best estimate based on historical baseline for product line, process tailoring, etc.
    - Variance estimates built into model based on historical actuals
- **SLAM Model Outputs**
  - Projected Software / Hardware Cost Performance (CPI)
    - Mean, Standard Deviation
    - 95% Upper & Lower Prediction Interval Limits

# System Lifecycle Analysis Model (SLAM)



## Post-SLAM

- **Feedback from the SLAM deployment included requests for additional features**
  - Can we predict requirements volatility?
  - Can we predict rework?
  - Can we predict IVV performance?
  - Can we predict likelihood of meeting schedule?
- **The Process Performance Models Team began considering how to meet the needs of different users**
  - Build additional features into SLAM?
  - Create separate models?
  - Bundle separate models into one supermodel?
- **Decision made to create a schedule risk model**
  - Used a SLAM-derived, similar look and feel approach
  - Provided an opportunity for further feedback



# Scheduling Analysis of Variability Engine (SAVE)

- *Can we predict likelihood of meeting schedule?*
- **Models influence of individual task cycle time variation on our ability to deliver on-time against defined scheduling requirements**
  - Enables projects to statistically quantify risk associated in meeting schedule and perform sensitivity analysis
  - Enables project identification of process changes that reduce risk (model contains potential mitigation strategies)
  - May be used during planning, re-planning, Estimate at Complete (EAC) activity
- **SAVE Model Inputs**
  - Estimated individual task activity duration (typically defined in terms of their shortest, most likely and longest cycle time expectations)
  - Individual task activity predecessors (defines which tasks feed which tasks)
  - Target overall schedule duration (deadline)
  - Number of task activities
- **SAVE Model Outputs**
  - Probability of meeting target duration
  - Average Duration (Cycle Time)
  - 95% Upper & Lower Prediction Intervals
  - % of time each task is on the critical path

# Scheduling Analysis of Variability Engine (SAVE)

Microsoft Excel - SAVE 2.2 1\_100 SEI

File Edit View Insert Format Tools Data Window Help EasyFitXL

J36

(Note that for the calculations to work properly, the Iteration setting under Tools > Options > Calculation must be enabled.)

## Scheduling Analysis of Variability Engine (SAVE)

Target Duration	90	Update
# of Tasks	17	

Task ID#	Task Name	Predecessor ID#	Shortest Time	Most Likely Time	Longest Time	Percent on CP
100			1	1	1	100.0%
101		100	1	2	5	37.4%
102		100	1	3	5	62.6%
103		101,102	2	5	10	100.0%
104		103	1	5	10	100.0%
105		104	3	10	15	100.0%
106		105	5	10	20	18.9%
107		105	5	10	15	81.1%
108		107	1	2	5	0.2%
109		107	2	3	5	80.9%
110		109	1	2	3	80.9%
111		108,110	1	2	3	81.1%
112		106	1	2	4	18.9%
113		111,112	2	4	8	100.0%
114		113	1	3	5	100.0%
115		104,114	1	3	5	100.0%
116		115	10	20	60	100.0%
117						-
118						-
119						-
120						-
121						-
122						-
123						-
124						-
125						-
126						-

Prediction	
Probability of meeting target duration	72.2%
Mean	83.09
95% Prediction Interval Lower Limit	63.80
95% Prediction Interval Upper Limit	107.29

You are 95% confident that the total task duration will be between 63.8 and 107.3 days

**Forecast: Total Duration Time**

1,000 Trials      Frequency View      998 D

**Duration Distribution**

Mean = 83.09

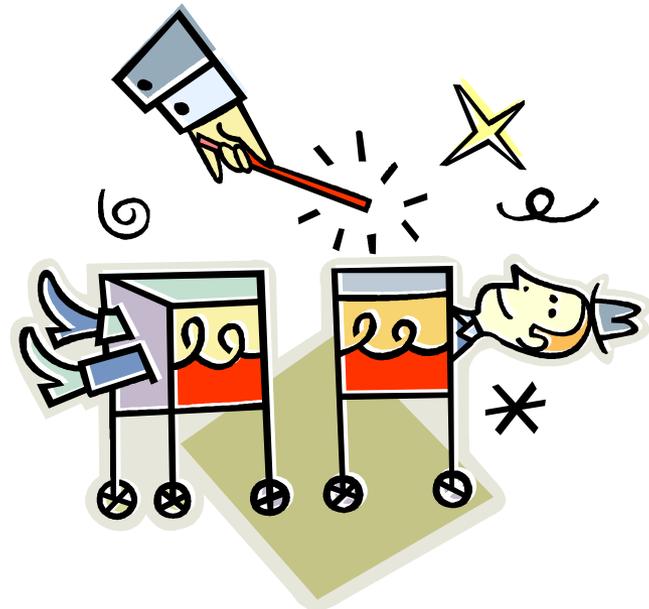
Certainty: 72.16 %      90.00

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start      > SAVE 2.2...      D:\Document...      Microsoft Po...      Microsoft Exc...      96%      9:56 AM

# Post-SAVE

- **Feedback from the SAVE deployment included requests for further model refinements**
  - Can we model probability of achieving cost targets?
  - Can we interface SAVE with Microsoft Project?
  - Can we integrate cost and schedule probability models?
  - Can we go below cost and schedule, and model size and productivity?
- **The answer to all the questions is “yes”, but returns us to the model/supermodel dilemma**
- **Solution: Horizontal Segmentation**



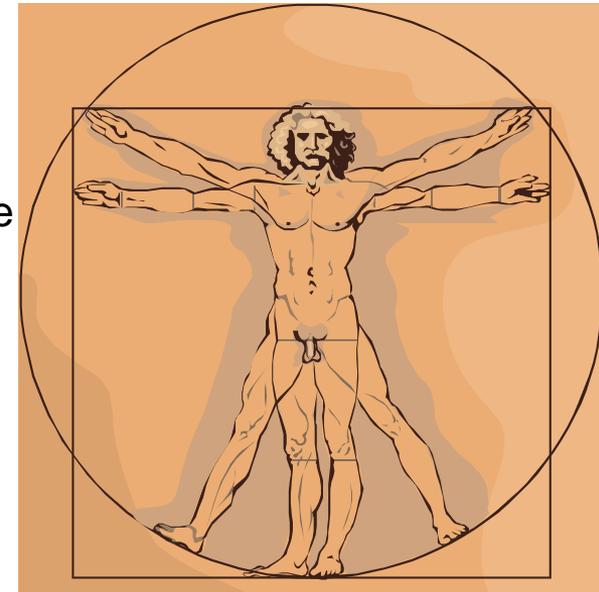
# Horizontal Segmentation

- **In 1986, Psychophysicist Dr. Howard Moskowitz was hired by Campbell's Soup Company to increase market share of Prego spaghetti sauce over arch-rival Ragu.**
  - Prego sales lagged and Campbell's wanted to change the sauce and improve its consumer appeal
  - Through earlier work in the 1970s with Pepsi and Vlasic Pickles, Moskowitz had suspected that a universal "one sauce fits all" solution was not the answer
  - The answer instead was in the understanding and leveraging of variability
  - Moskowitz had 45 different sauces created and taste-tested
    - Results indicated that consumers had different concepts of the perfect spaghetti sauce that could be segmented into groups
    - Main preferences: plain, spicy, extra chunky
    - No extra chunky sauce was marketed at the time
    - Campbell's made \$600M over the next 10 years on extra chunky
- **Moskowitz' breakthrough approach, known as Horizontal Segmentation, revolutionized the spaghetti sauce, soft drink and multiple other consumer markets**



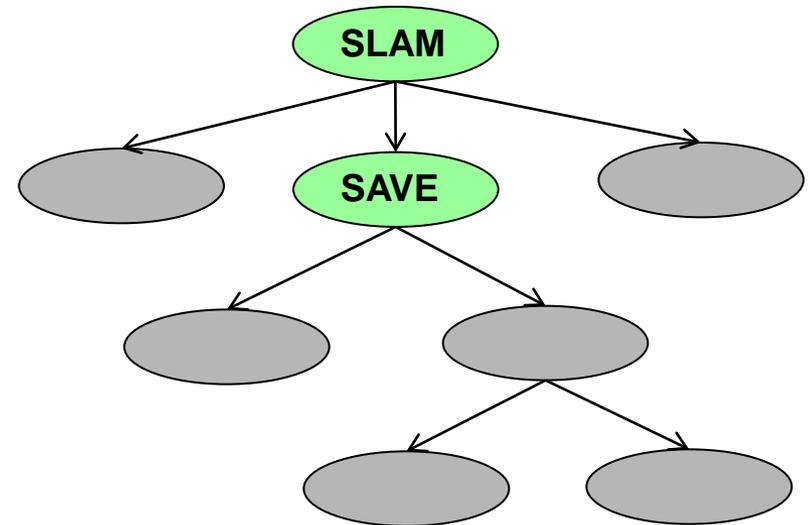
# Horizontal Segmentation

- **Horizontal Segmentation – groups people by preference patterns**
  - Contrasts with more traditional social stratifications (age, income, gender, race, etc.)
- **Lessons of Horizontal Segmentation**
  - There is no perfect sauce. There are perfect sauces.
  - There is no perfect pickle. There are perfect pickles.
  - There is no perfect process performance model. There are perfect process performance models.
- **Creating a suite of process performance models tailored to meet user preferences can facilitate deployment**
  - Offering users multiple model options empowers users
  - Enables model selection based on innate primal sensibilities (freedom of choice, personal preferences, contextual project differences)



# PPM Horizontal Segmentation

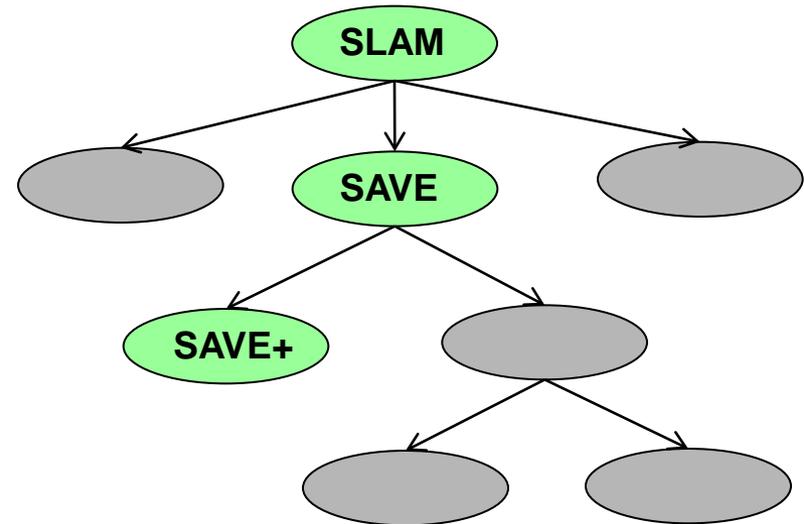
- **Users of process performance models want options that support personal or project preferences**
  - Combinations of cost, schedule, quality, integrated with other project tools
- **Based on feedback and the concept of Horizontal Segmentation, Raytheon IDS has created a cost & schedule model suite derived from SLAM and SAVE**
  - Supports business goals
  - Address individual project objectives, risks, and preferences



# SAVE+

## Can we integrate SAVE with Microsoft Project?

- Uses a Crystal Ball interface to MS Project
- Similar to Risk+



# SAVE+

The screenshot displays the Microsoft Project interface for a project named 'SAVE + Neal'. The main view is a Gantt chart showing task dependencies and durations from February to April 2010. Two windows are open over the chart:

**Sensitivity: End: Finish** (Contribution to Variance View, 2,000 Trials):

Task	Contribution to Variance
Complete Sale: Duration	15.6%
Coordinate / Conduct DD250...	15.3%
Packaging Complete: Duration	14.5%
Crate: Duration	13.2%
Cut / Label / Prepacjage: D...	10.4%
Pre-package Complete: Duration	8.0%

**Forecast: End: Finish** (Frequency View, 2,000 Trials, 1,991 Displayed):

This window shows a histogram of the project's end date. The x-axis represents the date (4/30/2010 12:00 AM) and the y-axis represents Probability (0.00 to 0.18) and Frequency (0 to 350). The forecast indicates a certainty of 31.05% by 2/24/2010 3:59 PM.

# Predictive Cost Modeling (PCM)

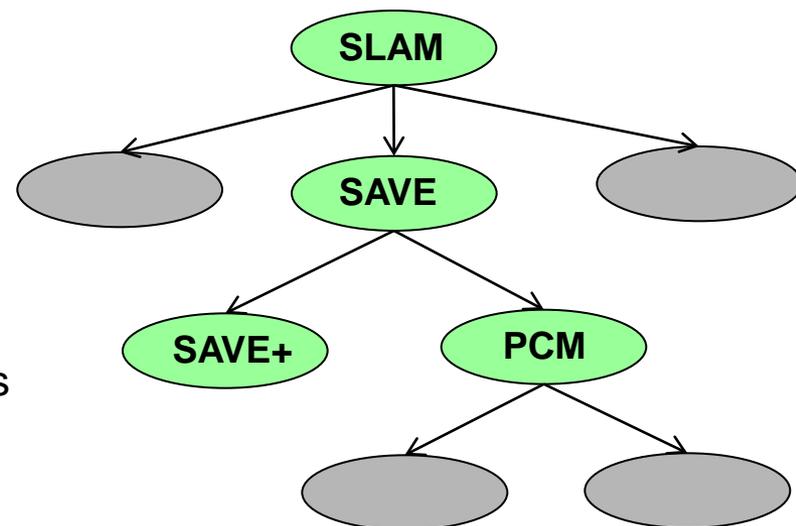
- *Can we model probability of achieving cost targets?*
- **PCM models the influence of individual cost element variation on our ability to meet cost targets / budget requirements**
  - Enables projects to statistically quantify risk associated in meeting cost targets / budgets and perform sensitivity analysis
  - Enables project identification of process changes that reduce risk
  - May be used during planning, re-planning, EAC activity

## ■ Predictive Cost Model Inputs

- Estimated mean cost for each individual cost element
- Associated expected range of performance for each individual cost element (min / max for the default triangular distribution)
- Overall cost target

## ■ Predictive Cost Model Outputs

- Predictive average cost
- 95% Upper & Lower Prediction Intervals
- Probability of meeting cost target / budgets
- Total cost estimate per task



# Predictive Cost Modeling (PCM)

PCM 1.3 application example.xlsm - Microsoft Excel

(Note that for the calculations to work properly, the iteration setting under Tools > Options > Calculation must be enabled.)

### Cost Analysis of Variability

Clear Output Fields

Target (\$ or Hours)	240000	
# of Tasks	25	Update

Task Name	Range(-%)	Cost Estimate	Range(+%)	Comments
	10%	1058	10%	
	10%	1079	10%	
	10%	1079	10%	
	10%	935	10%	
	10%	2014	10%	
	10%	1007	10%	
	10%	949	10%	
	10%	1029	10%	
	10%	956	10%	
	10%	3118	10%	
	10%	832	10%	
	10%	2014	10%	
	10%	3461	10%	
	10%	51583	10%	
	10%	624	10%	
	10%	3043	10%	
	10%	2240	10%	
	10%	624	10%	
	10%	6726	10%	
	10%	1938	10%	
	10%	15459	10%	
	10%	14011	10%	
	10%	9007	10%	
	10%	105946	10%	
	10%	8642	10%	

Cost Prediction	
Probability of meeting cost target	54%
Mean	239458.72
95% Prediction Interval Lower Limit	229914.06
95% Prediction Interval Upper Limit	248870.02

#### Forecast: Total Cost Distribution

1,000 Trials      Frequency View      998 Displayed

USL = 240,000.00

Certainty: 54.09 %

#### Sensitivity: Total Cost Distribution

1,000 Trials      Contribution to Variance View

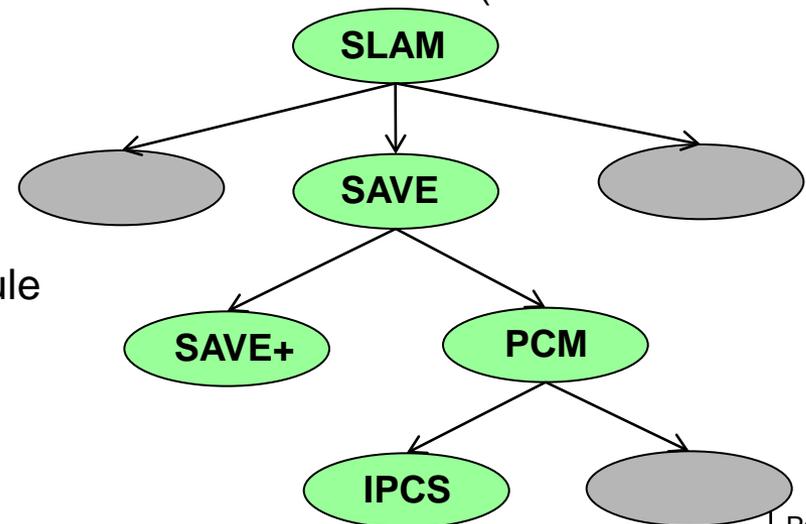
Task	Contribution to Variance
=BF32	74.3%
=BF22	18.1%
=BF29	4.1%
=BF21	-0.7%
=BF27	0.8%

PCM 1.2    CB Download Instructions    Rev History

Ready    start    Neal A Mackertic...    Welcome to one...    Enabling Cost & ...    Raytheon IDS Sc...    Crystal Ball    PCM 1.3 applicati...    90%    4:50 PM

# Integrated Predictive Cost & Scheduling (IPCS)

- *Can we integrate cost and schedule probability models?*
- **IPCS integrates the Schedule Analysis of Variability Engine (SAVE) and the Predictive Cost Model (PCM)**
  - Enables projects to statistically quantify risk associated in meeting cost & schedule targets and perform integrated sensitivity analysis
  - Enables project identification of process changes that reduce risk
  - May be used during planning, re-planning, EAC activity
- **IPCS Cost Model Inputs**
  - Estimated mean cost and duration expectations for each individual element
  - Individual task activity predecessors (defines which tasks feed which tasks)
  - Associated expected range of performance for each individual element (min / max for the default triangular distribution)
  - Overall cost & schedule targets
- **IPCS Cost Model Outputs**
  - Probability of meeting cost & schedule targets
  - Predictive mean expectation for cost & schedule
  - 95% Upper & Lower Prediction Intervals
  - % of time each task is on the critical path
  - Total cost estimate per task



# Integrated Predictive Cost & Scheduling (IPCS)

S & C AVE III\_fortuna12\_23.xls [Compatibility Mode] - Microsoft Excel

Security Warning: Some active content has been disabled. Options...

Q46

(Note that for the calculations to work properly, the iteration setting under Tools > Options > Calculation must be enabled.)

### Scheduling and Cost Analysis of Variability

Target Duration	38	Update	Target (\$ or Hours)	1250
# of Tasks	9			

Task ID#	Task Name	Predecessor ID#	Shortest Time	Most Likely Time	Longest Time	Percent on CP
100	A		3	5	9	100%
101	B	100	3	5	9	47%
102	C	100	3	5	9	54%
103	D	101,102	3	5	9	100%
104	E	103	3	5	9	49%
105	F	103	3	5	9	51%
106	G	104,105	3	5	9	100%
107	H	106	3	5	9	100%
108	I	107	3	5	9	100%
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Range(-%)	Cost Estimate	Range(+%)	Comments
5%	100	50%	
20%	120	20%	
5%	40	20%	
5%	55	40%	
15%	10	50%	
5%	500	20%	
5%	123	20%	
30%	100	75%	
5%	80	20%	

Clear Output Fields

Schedule Prediction	
Probability of meeting target duration	18.6%
Mean	41.01
95% Prediction Interval Lower Limit	34.80
95% Prediction Interval Upper Limit	47.45

Cost Prediction	
Probability of meeting cost target	87%
Mean	1203.55
95% Prediction Interval Lower Limit	1134.00
95% Prediction Interval Upper Limit	1289.81

**Forecast: Total Duration Time**

1,000 Trials | Frequency View | 996 Displayed

Duration Distribution

Mean = 41.01

Certainty: 18.63 %

**Forecast: Cost Prediction Distribution**

1,000 Trials | Frequency View | 998 Displayed

Cost Prediction Distribution

Mean = 1,203.55

USL = 1,250.00

Certainty: 86.59 %

**Sensitivity: Cost Prediction Distribution**

1,000 Trials | Contribution to Variance View

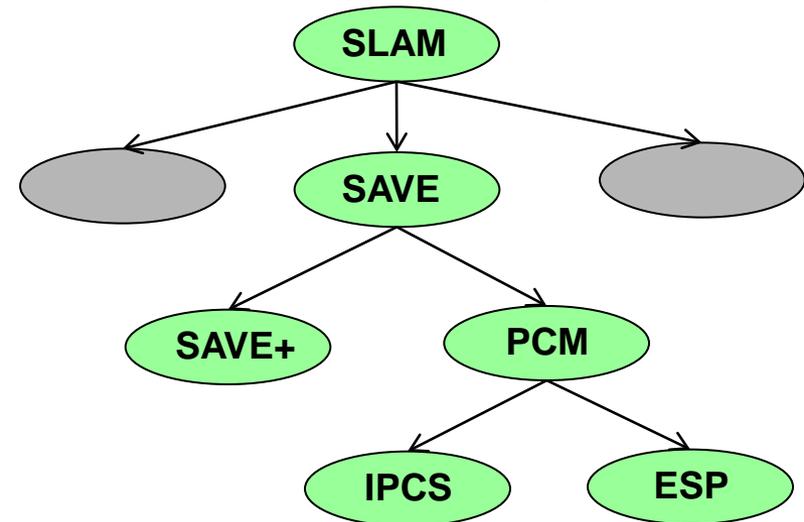
Sensitivity: Cost Prediction Distribution

Task	Contribution to Variance
F (BJ14)	47.2%
H (BJ16)	30.1%
A (BJ9)	9.4%
B (BJ10)	5.0%
C (BJ15)	3.0%

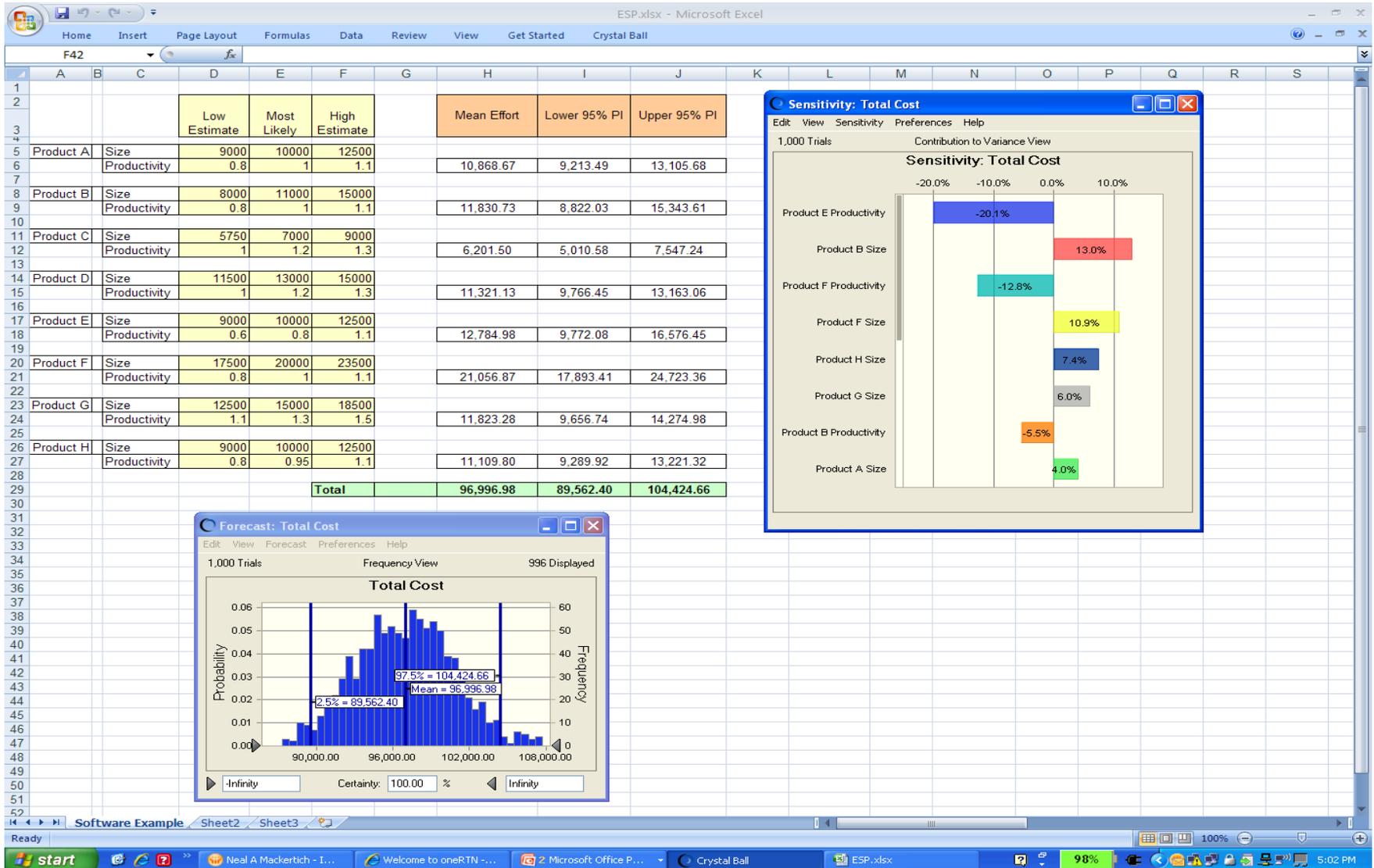
Ready Calculate | SAVE | CB Download Instructions | Rev History | 90% | 4:57 PM

# Effort based on Size & Productivity (ESP)

- *Can we go below cost and schedule, and model size and productivity?*
- **ESP models the influence of size and productivity variability on our ability to meet cost targets / budget requirements**
  - Enables projects to statistically quantify risk associated in meeting cost targets / budgets and perform sensitivity analysis
  - Enables project identification of process changes that reduce risk
  - May be used during planning, re-planning, EAC activity
- **Size and Productivity Cost Model Inputs**
  - Estimated mean size and productivity for each individual cost element
  - Associated expected range of performance for each size and productivity estimate (min / max for default triangular distribution)
  - Overall cost target
- **Size and Productivity Outputs**
  - Predictive average cost
  - 95% Upper & Lower Prediction Intervals
  - Probability of meeting cost target / budgets
  - Total cost estimate



# Effort based on Size & Productivity (ESP)



## Results & Benefits

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- Stakeholder groups and projects have found the family of developed 2G PPMs both easy to use and conceptually aligned with project issues.
- Projects have identified and implemented specific improvements as a direct result of the integrated deployment of 2G PPMs that have enabled their execution. Resulting improvement efforts include:
  - Increased up-front investment in *integrated* Engineering planning & analysis
  - Process Performance trade studies
  - Process redesign
  - Enhanced peer reviews
  - Resource reallocation and conflict resolution
- In addition to deployed projects, 2G PPMs have been used up-front during the bid & proposal phase and are utilized by Engineering Management during schedule negotiations with program management.
- Significant qualitative benefits from integrated deployment of these models cannot be underestimated as projects leads / teams are clearly thinking and behaving differently with respect to their analysis of risk & opportunity.

## Summary

- Oddly enough, while project practitioners may tend to *overestimate* contextual differences (your model does not apply to me because my project is different!), analysts may tend to *underestimate* contextual differences and preferences in search of universal solutions that may or may not exist.
- Leveraging the concept of Horizontal Segmentation enables our development and deployment of Cost & Schedule Risk Analysis Models by
  - Sharpening our focus on understanding and serving our projects
  - Enabling our understanding of contextual differences between differing product types, project challenges and personal preferences
  - Increasing project team understanding of the benefits of statistical modeling and igniting their passion for more...
  - Reinforcing the need for data stratification
  - Challenging and reinventing our existing proposal / project management measurement & analysis processes and tools

“There is no perfect Process Performance Model.  
There are perfect Process Performance Models\_.”

## References

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1. Mackertich, Campo, Beitz, “*System Lifecycle Analysis Model (SLAM)*”, 2<sup>nd</sup> SEI High Maturity Measurement and Analysis Workshop, November 2008.
2. Mackertich, Campo, “*Scheduling Analysis of Variability Engine(SAVE)*”, 3<sup>rd</sup> SEI High Maturity Measurement and Analysis Workshop, March 2009.
3. Campo, Mackertich, “*Goal Question Model*”, NDIA CMMI Technology Conference and User Group, November 2009.
4. Malcolm Gladwell on spaghetti sauce,  
[http://www.ted.com/talks/lang/eng/malcolm\\_gladwell\\_on\\_spaghetti\\_sauce.html](http://www.ted.com/talks/lang/eng/malcolm_gladwell_on_spaghetti_sauce.html)

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