

# Earned Schedule

## Deterministic Project Duration Forecasting



# Abstract

- Earned Schedule (ES), an extension to Earned Value Management (EVM), is a practice recognized in PMI<sup>®</sup> and ISO global standards. The time domain performance indicators for variance and efficiency, derived from ES, are deemed useful for project control. As well, ES facilitates project duration and completion date forecasting. Recent research indicates ES forecasting accuracy is generally improved when the performance factor equal to one ( $PF=1$ ) is used. However, the ES schedule performance index,  $SPI(t)$ , remains as accepted practice and, at times, provides the better deterministic forecast. It is postulated that there may be performance characteristics for identifying which of the calculation methods yields better accuracy. The presentation will discuss the investigation and research findings to discern those characteristics.



# Objective

- Understand ES and its facility to provide schedule performance information
- Learn about the research that has led to improvements in duration forecasting
- Able to apply ES deterministic forecasting, to include the calculation selection method for improving accuracy



# Overview

- Introduction
- Forecasting Research
- Why? - Investigation
- Analysis – Proposed Answer
- Simulated & Real Data Testing
- Summary/Conclusion
- Suggested Research



# Introduction

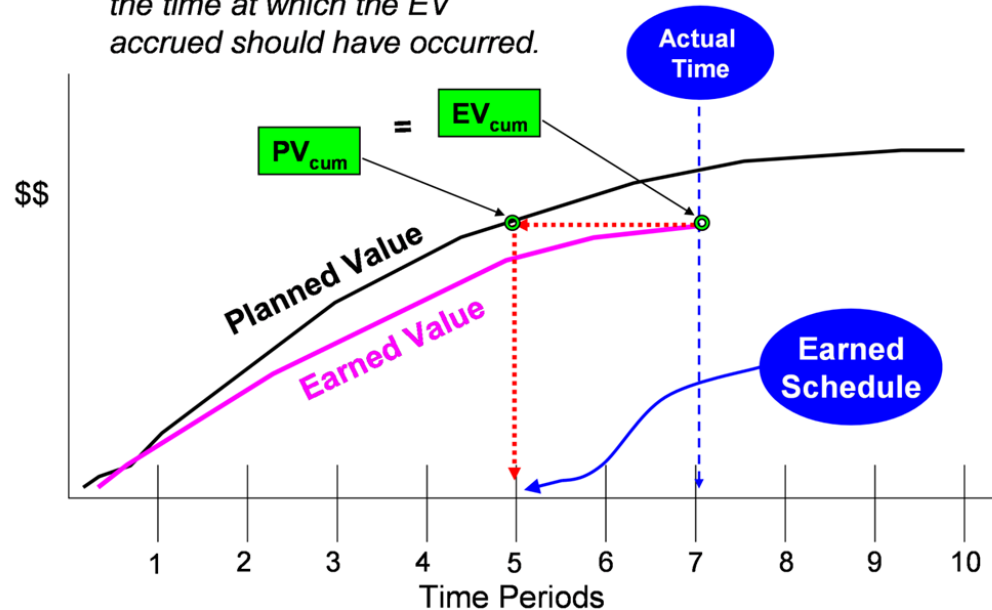
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## Earned Schedule

# Introduction / Earned Schedule



*The ES idea is to determine the time at which the EV accrued should have occurred.*



$$\text{Time based schedule performance efficiency: } SPI(t) = ES / AT$$



# Earned Schedule Indicators

- Formula

- $ES = C + I$

where:  $C$  = number of time increments for  $EV \geq PV_n$

$$I = (EV - PV_C) / (PV_{C+1} - PV_C)$$

- Indicators

- Schedule Variance:  $SV(t) = ES - AT$

- Schedule Performance Index:  $SPI(t) = ES / AT$



# Earned Schedule Forecasting

- Forecasting Formulas – Independent Estimate at Completion (time)

- $IEAC(t) = PD / SPI(t)$

where: PD = planned duration

- $IEAC(t) = AT + (PD - ES) / PF_s$

where:  $PF_s$  = performance factor - schedule



# Earned Schedule Terminology

<b>Metrics</b>	Earned Schedule	$ES_{cum}$	$ES = C + I$ number of periods (C), $EV \geq PV_C$ plus an incomplete portion (I)
	Actual Time	$AT_{cum}$	AT = number of periods executed
<b>Indicators</b>	Schedule Variance	$SV(t)$	$SV(t) = ES - AT$
		$SV(t)\%$	$SV(t)\% = (ES - AT) / ES$
	Schedule Performance Index	$SPI(t)$	$SPI(t) = ES / AT$
<b>Predictor</b>	To Complete Schedule Performance Index	$TSPI$	$TSPI = (PD - ES) / (PD - AT)$
			$TSPI = (PD - ES) / (ED - AT)$
<b>Forecasts</b>	Independent Estimate at Completion (time)	$IEAC(t)$	$IEAC(t) = PD / SPI(t)$
			$IEAC(t) = AT + (PD - ES) / PF(t)$
	Variance at Completion (time)	$VAC(t)$	$VAC(t) = PD - IEAC(t)$ or ED



# Forecasting Research

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2003 - 2015

# Forecasting Research/2003-15

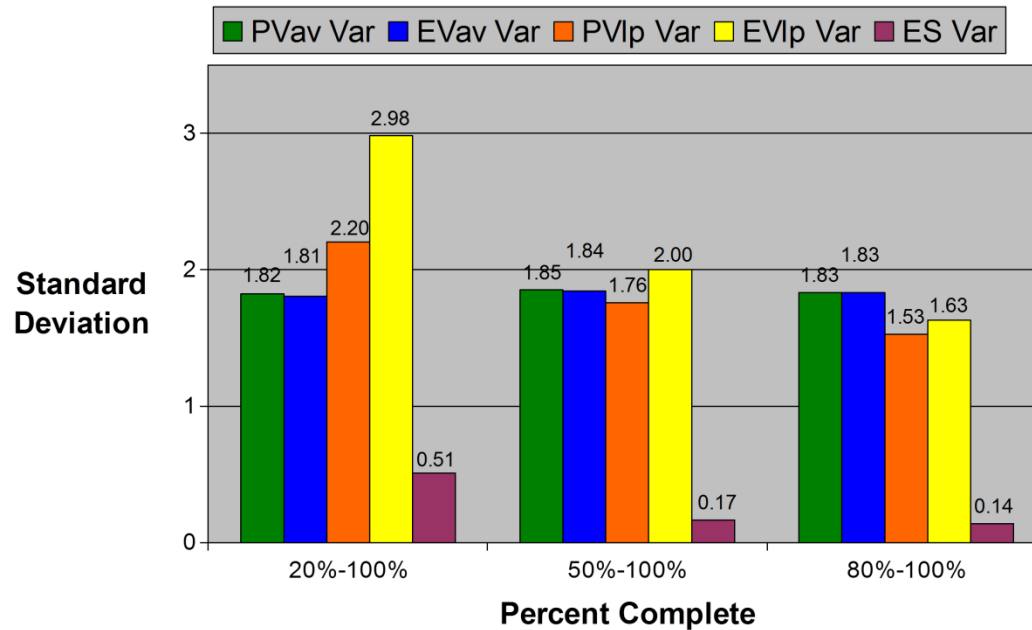


- Creation of ES, indicators, and implication to forecasting
- Kym Henderson application to project portfolio / forecasting
- Vanhoucke & Vandevoorde simulations
  - Comparison to EVM methods ...ES better
  - Schedule topology ...serial most accurate
- Comparison versus EVM time conversion methods
- 'Longest Path' resolution to highly parallel schedules
- Creation of statistical forecasting method

# Forecasting Research/2003-15



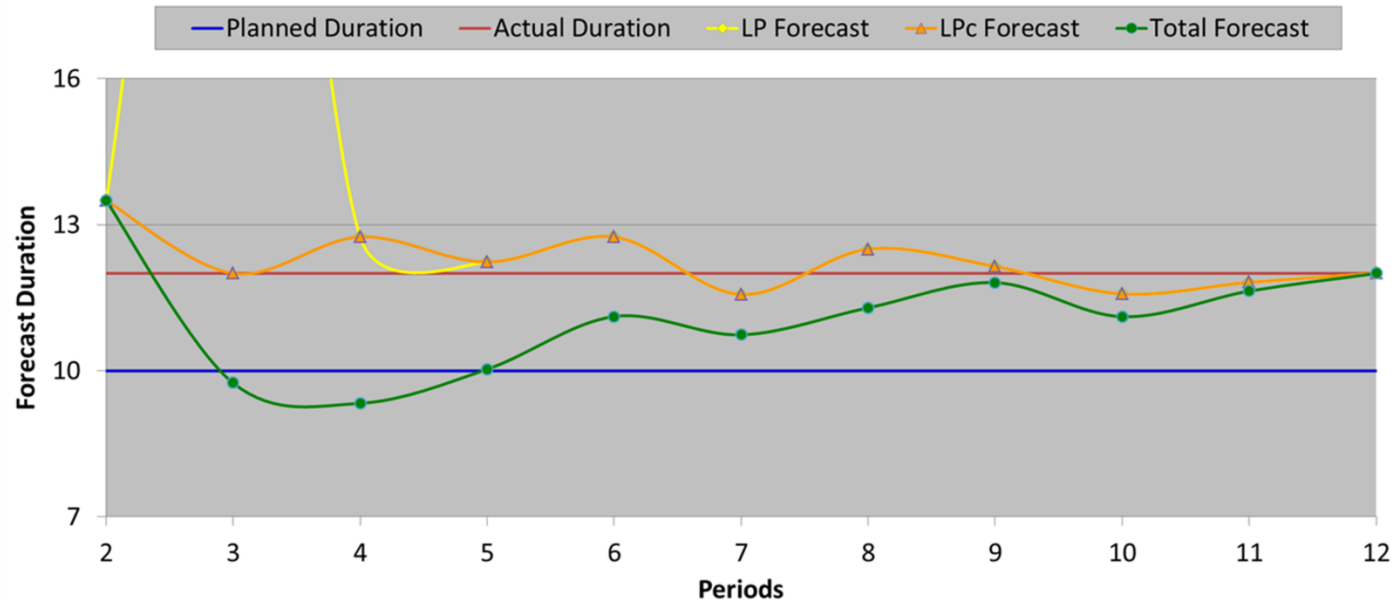
- Comparison of ES versus EVM time methods



# Forecasting Research/2003-15



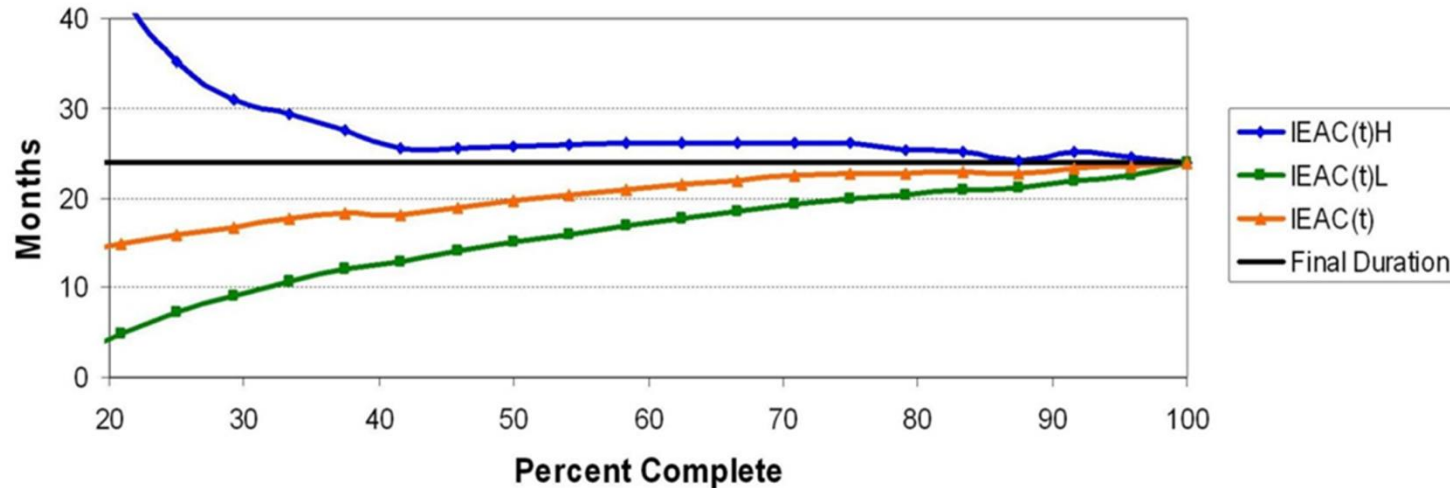
- ‘Longest Path’ resolution to highly parallel schedules



# Forecasting Research/2003-15



- Creation of statistical forecasting method

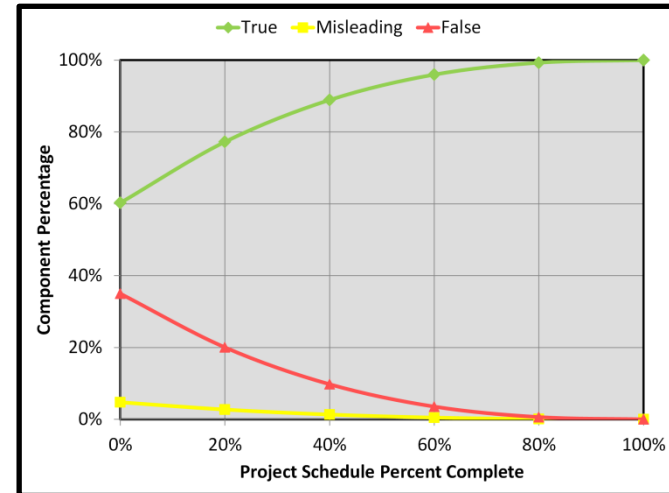


# Forecasting Research/2003-15



- Vanhoucke investigates misleading and incorrect forecasts
- Forecast convergence characteristic minimizes misleading and incorrect forecast concern

		Outcome		
		RD < PD	RD = PD	RD > PD
Indicator	SPI(t) > 1	1 True	4 Mislead	7 False
	SPI(t) = 1	2 Mislead	5 True	8 Mislead
	SPI(t) < 1	3 False	6 Mislead	9 True



# Forecasting Research/2003-15



- Resolution to forecasting inaccuracy caused by stop work & down time interruptions
- Crumrine verifies ES forecasting using major defense projects
- Milestone forecasting created for large projects





# Forecasting Research

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## Batselier & Vanhoucke (2015)

# Forecasting Research / 2015...



- B & V comprehensively evaluated forecasting using EVM data from 51 projects, predominantly construction
- Alarming, results of the research demonstrated the use of performance factor,  $PF=1$ , with the ES method often provides the more accurate deterministic forecast
- B&V finding was corroborated, using 16 projects (IT & High Tech) ... $PF=1$  was shown to provide better results for 12

# Forecasting Research / 2015...



- Real data concerns
  - Do the data sets represent a very localized set of conditions?  
The data may be biased toward PF=1 forecasting
  - Are re-plans, stop work, and down time in the data?  
Each increases variation, causing SPI(t) to appear worse

# Forecasting Research / 2015...



- Observations made in the 2<sup>nd</sup> study
  - When  $SPI(t)$  is constant, its forecast is superior
  - With highly variable performance,  $PF=1$  is more accurate
- Further research was suggested ...to gain an understanding of 'Why  $PF=1$  appears better?'



# Why $PF=1$ Better?

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## Investigation



# PF=1 Better? / Investigation

- Two step process
  - Step 1 - Compile data from simulating performance of 25 projects using 39 scenarios of variation and early/late bias
  - Step 2 - Inspect the tabulated results for patterns favoring one of the forecasting methods, PF=1 or SPI(t)

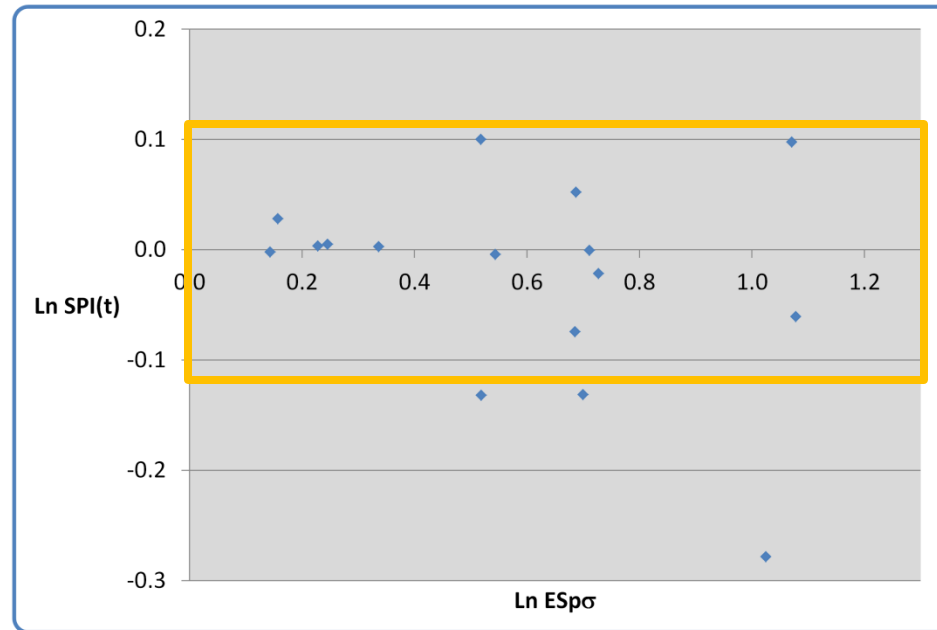
# PF=1 Better? / Investigation

- Step 1 – Compiled Results (example)

Scenario	Scenario Inputs							Outcome Averages			Error Results	
	SPI(t)	PD	Fixed Var	F or V	Var Mult	Bias "+"	Trigger	FD	lnESp $\sigma$	lnSPI(t)	MAPE(1)	MAPE(S)
B1	1.000	44	NA	V	0.10	0.10	0.10	46	0.157	-0.034	0.0297	0.0156
B2	1.000	35	NA	V	0.50	0.10	0.10	44	0.299	-0.199	0.1087	0.0463
B3	1.000	24	NA	V	0.90	0.10	0.10	36	0.678	-0.384	0.1694	0.0899
B4	1.000	45	NA	V	0.10	0.50	0.10	46	0.143	-0.002	0.0118	0.0126
B5	1.000	42	NA	V	0.50	0.50	0.10	43	0.336	0.003	0.0271	0.0413
B6	1.000	39	NA	V	0.90	0.50	0.10	40	0.711	-0.001	0.0427	0.0660
B7	1.000	48	NA	V	0.10	0.90	0.10	47	0.126	0.036	0.0126	0.0126
B8	1.000	52	NA	V	0.50	0.90	0.10	44	0.256	0.153	0.0836	0.0295
B9	1.000	56	NA	V	0.90	0.90	0.10	42	0.443	0.287	0.1510	0.0454

# PF=1 Better? / Investigation

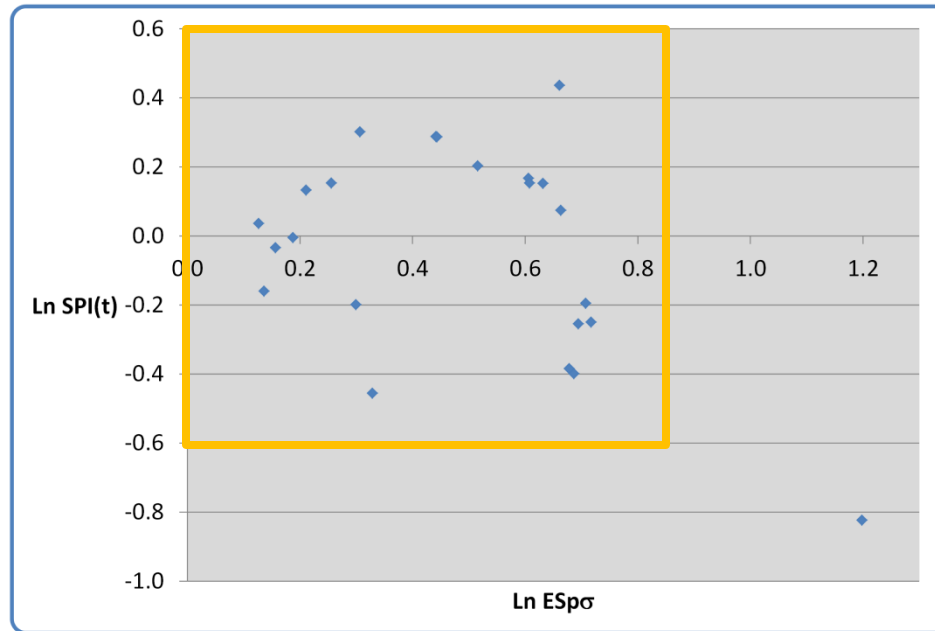
- Step 2 – PF=1 forecast better, 16 outcomes





# PF=1 Better? / Investigation

- Step 2 – SPI(t) forecast better, 23 outcomes





# Analysis

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# Proposed Answer

# Analysis / Answer

- Forecast formula selection rules

SPI(t) & PF=1 Selection Rules		
Use PF=1	when	$-0.1 \leq \ln \text{SPI}(t) \leq 0.1$ & $\ln \text{ESp}\sigma \leq 0.8$
Use SPI(t)	when	$0.6 \geq \ln \text{SPI}(t) > 0.1$ & $\ln \text{ESp}\sigma \leq 0.8$
Use SPI(t)	when	$-0.6 \leq \ln \text{SPI}(t) < -0.1$ & $\ln \text{ESp}\sigma \leq 0.8$
Use PF=1	when	$-0.6 > \ln \text{SPI}(t) > 0.6$ or $\ln \text{ESp}\sigma > 0.8$ = Out of Control



# Testing

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## Simulated & Real Data

# Testing / Simulated Data

Simulated Data										
Project	1	2	3	4	5	6	7	8	9	10
PD	24	56	66	39	25	29	35	39	35	52
FD	34	41	43	47	43	33	38	37	43	45
# Out of Cntrl	0	0	0	0	3	0	13	0	0	0
Err-SPI(t)	0.082	0.029	0.064	0.012	0.049	0.061	0.082	0.049	0.040	0.023
Err-PF1	0.162	0.167	0.213	0.082	0.209	0.050	0.030	0.033	0.099	0.074
Err-Selection	0.083	0.029	0.064	0.012	0.065	0.068	0.036	0.051	0.043	0.034
Err-Best	0.081	0.027	0.057	0.012	0.049	0.032	0.023	0.026	0.040	0.021

# Testing / Real Data

Real Data										
Project	1	2	3	4	5	6	7	8	9	10
PD	21	32	43	24	41	29	43	17	44	42
FD	24	38	47	24	50	30	50	23	50	50
# Out of Cntrl	3	0	23	0	22	0	31	0	20	2
Err-SPI(t)	0.187	0.090	0.134	0.041	0.077	0.067	0.058	0.138	0.156	0.100
Err-PF1	0.129	0.093	0.050	0.015	0.071	0.034	0.067	0.131	0.067	0.093
Err-Selection	0.156	0.093	0.111	0.031	0.086	0.063	0.064	0.149	0.124	0.096
Err-Best	0.125	0.075	0.039	0.015	0.044	0.034	0.039	0.103	0.056	0.070

# Testing / Composite

Simulated Data										
Project	1	2	3	4	5	6	7	8	9	10
S & 1 within 5%?	no	no	no	no	no	yes	no	no	no	no
S or 1 Better?	S	S	S	S	S	1	1	1	S	S
Select Improve?	yes	yes	yes	yes	yes	no	yes	no	yes	yes

Real Data										
Project	1	2	3	4	5	6	7	8	9	10
S & 1 within 5%?	no	yes	no	yes	yes	yes	yes	yes	no	yes
S or 1 Better?	1	S	1	1	1	1	S	1	1	1
Select Improve?	yes	yes	yes	yes	no	yes	yes	no	yes	yes



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# Summary / Conclusion



# Summary / Conclusion

- Two empirical studies reported that  $PF=1$  provides better project duration forecasts than  $SPI(t)$
- There are instances where  $SPI(t)$  was better ... thus, performance conditions may cause one method to forecast more accurately
- Simulation of 25 projects, applying 39 performance scenarios, revealed graphical patterns

# Summary / Conclusion

- From the graphical patterns selection rules were derived

SPI(t) & PF=1 Selection Rules			
Use PF=1	when	$-0.1 \leq \ln \text{SPI}(t) \leq 0.1 \ \& \ \ln \text{ESp}\sigma \leq 0.8$	
Use SPI(t)	when	$0.6 \geq \ln \text{SPI}(t) > 0.1 \ \& \ \ln \text{ESp}\sigma \leq 0.8$	
Use SPI(t)	when	$-0.6 \leq \ln \text{SPI}(t) < -0.1 \ \& \ \ln \text{ESp}\sigma \leq 0.8$	
Use PF=1	when	$-0.6 > \ln \text{SPI}(t) > 0.6 \ \text{or} \ \ln \text{ESp}\sigma > 0.8 = \text{Out of Control}$	

- Selection rules were tested using simulated and real data
- For both data sets, the selection rules for 8 of 10 projects yielded more accurate forecasts than at least one of the PF=1 and SPI(t) methods.



# Summary / Conclusion

- The two deterministic methods yield very comparable forecasts about 70 percent of the time
- The risk of using the  $PF=1$  method is there are instances when its forecast can be in error by greater than 10 percent
- The risk of exclusively using  $SPI(t)$  forecasting is it generally has larger error than  $PF=1$ , although the difference is small.



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# Suggested Research

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- The duration forecasting selection rules proposed should be further examined before general adoption
- It is suggested that those practitioners having real data apply the rules and report their findings
- Application of the selection rules to simulated EVM data is welcomed, and may yield refinements or a better approach
- For those interested, the forecast method selection spreadsheet is available from the ES website ([www.earnedschedule.com](http://www.earnedschedule.com))

# References

- Batselier, J., M. Vanhoucke (2015). “Empirical Evaluation of Earned Value Management Forecasting Accuracy for Time and Cost,” *Journal of Construction Management*, 141(11): 05015010, 1-13
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- Lipke, W. (2017). “Assessing Earned Value Management and Earned Schedule Forecasting,” *PMWorld Journal* (online), August, Vol VI, Issue 8
- Lipke, W. (2009). *Earned Schedule*, Raleigh, NC: Lulu Publishing
- Lipke, W. (2003). “Schedule is Different,” *The Measurable News*, Summer: 31-34

The logo consists of a yellow oval with a dark red border. Inside the oval, the letters 'LES' are written in a large, stylized, dark red serif font. The 'L' and 'S' are particularly large and overlap the 'E'.

Thank You!!

*TM*