

Minimize risk. Maximize potential.

Modeling Unit Learning Curves from Lot Data

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I@EAA Technology Showcase Webinar Series



Webinar Topics

- Learning Curve Fundamentals
- Power Model Review
- Working with Lot Data
 - Lot midpoint approximation
- Demonstration
 - Data transformation
 - Excel LINEST function and Goal Seek
 - @RISK implementation











As the number of units produced doubles, the **unit** cost/time to produce the doubled unit decreases by a constant percentage.

Unit (X)	Unit Cost (Y) or Time
1	100
2	90
4	81
8	72.9



$$Y = aX^b$$

where:

Y = cost/time of *Xth* unit

- $a = cost/time of first unit (T_1)$
- *X* = number of unit produced
- *b* = learning curve exponent

$$b = \frac{\ln(slope)}{\ln(2)} = \frac{\ln(.90)}{\ln(2)} = -.1520$$

$$Y = 100X^{-.1520}$$



Intrinsically Linear Transformations





Power Model Basics



 ϵ in power model is a multiplicative error term that is lognormally distributed, but in the transformed model ϵ is additive and normally distributed with same characteristics assumed as an ordinary least squares linear regression error term



 $if Y \sim N(\mu, \sigma)$ then $X = e^{Y} \sim \ln(\mu, \sigma)$ and $E(X) = e^{(\mu + \sigma^{2}/2)}$ $V(X) = [e^{(2\mu + \sigma^{2})}](e^{\sigma^{2}} - 1)$

to model in @RISK use: RiskLognorm (E(X), {V(X)}^{.5}) or RiskLognorm2 (μ,σ)

For fuller coverage of modeling regression error terms go to https://go.palisade.com/Modeling-Regression-Errors-with-RISK.html



Determining T₁ and Slope from Lot Data: Problem Statement

A critical subassembly has experienced the following production hours for the first five lots. Assuming learning follows the unit theory, find the T1 and slope for the appropriate learning curve and estimate the production hours needed for the 45 units in Lot 6.

Lot	Size	Units	Cost (man-hrs)
1	15	1 - 15	35000
2	10	16 – 25	20500
3	60	26 – 85	85000
4	30	86 - 115	39500
5	50	116 - 165	58000



Determining T₁ and Slope from Lot Data: Lot Midpoint (LMP)

- The Lot Midpoint (LMP), also called the Algebraic Lot Midpoint or Lot Plot Point, is defined as the theoretical unit whose cost (or time) is equal to the Average Unit Cost (AUC) or Lot Average Cost for that lot on the learning curve
- The LMP value represents the X value and the AUC represents the Y value when working with lot data

"The difficulty in finding the LMP is that you must know the slope of the learning curve in order to determine the LMP, but you need all the LMPs to find the slope of the learning curve."





Determining T₁ and Slope from Lot Data: LMP Approximation

$$LMP = \left[\frac{(L+.5)^{b+1} - (F-.5)^{b+1}}{(b+1)(L-F+1)}\right]^{1/b}$$

where:

- b = slope coefficient
- F = first unit # in lot
- L = last unit # in lot

See backup slides for a six-term alternate LMP approximation formula that ACE-IT's CO\$TAT 7.5 uses



Demonstration

General Steps:

- 1. Enter lot data in B4:F8
- 2. Enter starting slope assumption in C13
- 3. Run Goal Seek as shown in dialog box image
- 4. Enter new lot data in B9:E9
- 5. Run @RISK simulation with F11 as output

1	Α	B	С	D	E	F	G	Н		J
1										
2							Х	Y	In X	In Y
3		Lot	Size	First Unit	Last Unit	Cost (man-hrs)	LMP	AUC	In(LMP)	In(AUC)
4		1	15	1	15	35000	5.9465	2333.3	3 1.7828	7.7551
5		2	10	16	25	20500	20.2467	2050.0	3.0080	7.6256
6		3	60	26	85	85000	51.9132	1416.	3.9496	7.2561
7		4	30	86	115	39500	100.0391	1316.	4.6056	7.1829
8		5	50	116	165	58000	139.5821	1160.0	4.9387	7.0562
9		6	45	166	210	50074.5	187.4466			
10						1.0028				
11						50214.7				
40			Accumed	Degraceed				_		
12			Assumeu	Regresseu						
12		slope	85.3%	85.3%		(Data> W	hat-If A	nalysis> Go	al Seek
12 13 14		slope b	85.3% -0.2288	85.3% -0.2287	-0.0001	(Data> W	hat-lf A	nalysis> Go	al Seek
12 13 14 15		slope b <i>In</i> a	85.3% -0.2288	85.3% -0.2287 8.2114	-0.0001	(Data> W Goal See	/hat-lf A	nalysis> Go ?	al Seek
12 13 14 15 16		slope b In a a	85.3% -0.2288	85.3% -0.2287 8.2114 3682.8	-0.0001		Data> W Goal See	/hat-lf Ai	nalysis> Go ?	al Seek
12 13 14 15 16 17		slope b <i>In</i> a a	85.3% -0.2288	85.3% -0.2287 8.2114 3682.8	-0.0001		Data> W Goal See	/hat-lf Ai	nalysis> Go ? E14	al Seek
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12 13 14 15 16 17 18 19 20 21		slope b In a a slope SE _{slope}	ASSUITED 85.3% -0.2288 LIN -0.2287 0.0292	85.3% -0.2287 8.2114 3682.8 EST 8.2114 0.1119	-0.0001 intercept SE _{intercept}		Data> W Goal Seel S <u>e</u> t cell: To <u>v</u> alue: By <u>c</u> hangi	hat-If Ank	E14 0 \$C\$13	Aal Seek
12 13 14 15 16 17 18 19 20 21 22		slope b In a a slope SE _{slope} R ²	LIN -0.2287 0.0292 0.9534	85.3% -0.2287 8.2114 3682.8 EST 8.2114 0.1119 0.0748	-0.0001 intercept SE _{intercept} SE _v		Data> W Goal See S <u>e</u> t cell: To <u>v</u> alue: By <u>c</u> hangi	/hat-If Ar k ng cell: OK	E14 0 SCS13 Cance	Aal Seek
12 13 14 15 16 17 18 19 20 21 22 22 23		slope b In a a slope SE _{slope} R ² F	ASSUMED 85.3% -0.2288 -0.2287 -0.2287 -0.2287 -0.0292 -0.9534 61.3536	85.3% -0.2287 8.2114 3682.8 EST 8.2114 0.1119 0.0748 3	-0.0001 intercept SE _{intercept} SE _y df _{error}		Data> W Goal See S <u>e</u> t cell: To <u>v</u> alue: By <u>c</u> hangi	/hat-If Ar k ng cell: OK	E14 0 \$C\$13 Cance	al Seek
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Thank/You!

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Back Up Slides

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As the number of units produced doubles, the **cumulative average** cost/time to produce the doubled number of units decreases by a constant percentage.

Unit (X)	Cum Avg Cost (Y) or Time
1	100
2	90
4	81
8	72.9



$$\bar{Y} = aX^b$$

where:

 \overline{Y} = cum avg cost/time for X units

- $a = cost/time of first unit (T_1)$
- *X* = cum number of units produced
- *b* = learning curve exponent

$$b = \frac{\ln(slope)}{\ln(2)} = \frac{\ln(.90)}{\ln(2)} = -.1520$$

$$\bar{Y} = 100 X^{-.1520}$$



The rationale for the Rate Model is that unit cost may not only decrease as more units are produced, as modeled by Unit learning, but may also decrease as the rate of production increases and economies of scale are realized.



 $Y = aX^bQ^c$

where:

- Y = cost/time of *Xth* unit
- $a = cost/time of first unit (T_1)$
- X = number of unit produced
- *b* = learning curve exponent

Q = rate of production (qty per time period or lot)

c = rate exponent (rate slope = 2^c)



CO\$TAT 7.5 Six-Term LMP Approximation Formula

$$If PQ > 0 \Rightarrow LTQ_i = \sum_{Q=PQ+1}^{PQ+LQ} Q^b \cong \frac{(PQ+LQ)^{b+1} - (PQ)^{b+1}}{b+1} + \frac{(PQ+LQ)^b - (PQ)^b}{2} + \frac{b}{12} \left((PQ+LQ)^{b-1} - (PQ)^{b-1} \right)$$

If
$$PQ = 0 \Rightarrow LTQ_i = \sum_{Q=1}^{LQ} Q^b \cong \frac{(LQ)^{b+1}}{b+1} + \frac{(LQ)^b}{2} + \frac{b}{12}(LQ)^{b-1} - \frac{1}{b+1} + \frac{1}{2} - \frac{b}{12}$$

The true lot midpoint is then given by:

$$LMP_i = \left(\frac{LTQ_i}{LQ}\right)^{1/\ell}$$

From Shu-Ping Hu & Alfred Smith (2013), Accuracy Matters: Selecting a Lot-Based Cost Improvement Curve, Journal of Cost Analysis and Parametrics, 6:1, 23-42

