



**4° International Conference on  
IT Data collection, Analysis and Benchmarking**  
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## **Data Driven Cost Estimating**

**And the Role of Industry  
and Private Data**



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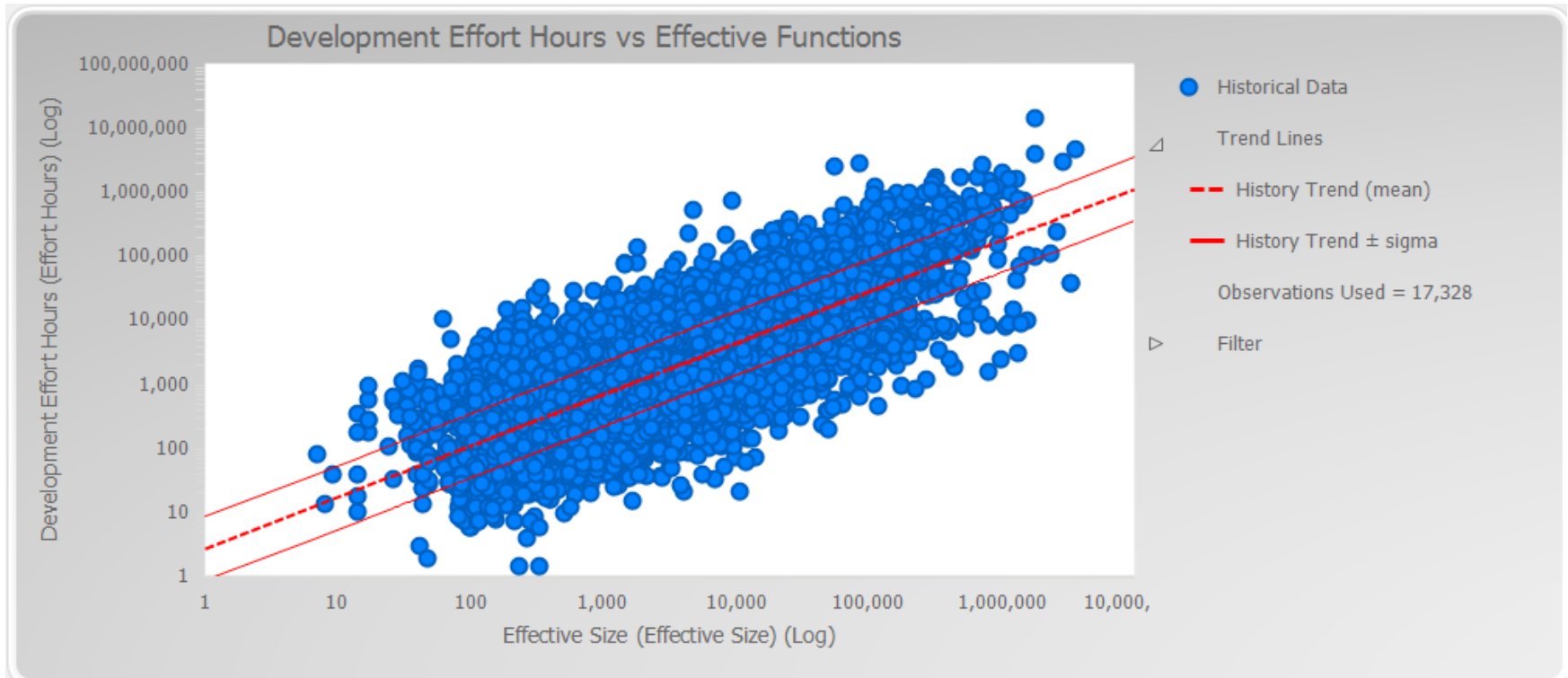
### In the Beginning



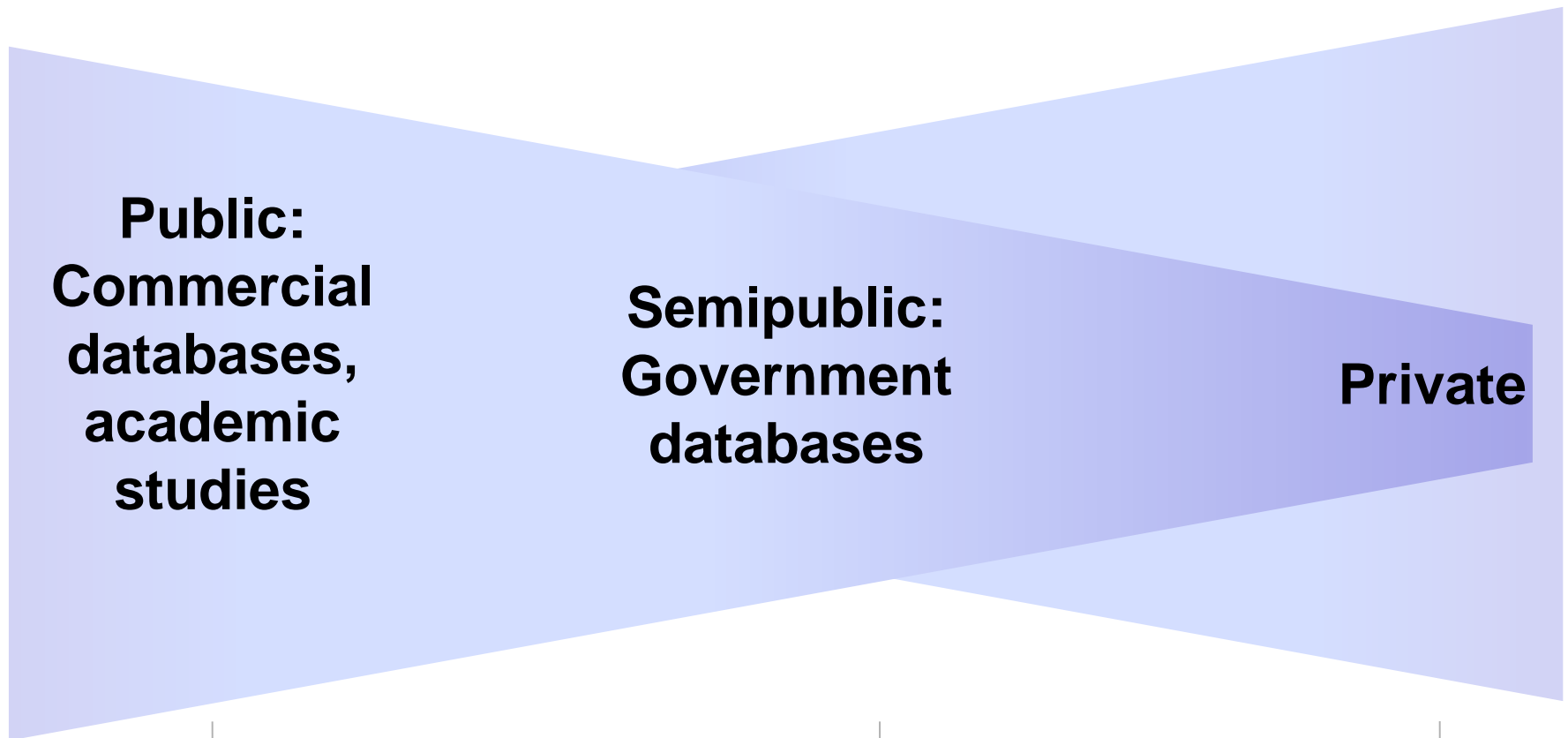
# Data Driven Cost Estimating

## And the Role of Industry and Private Data

Now



## Sources of Data



### Public versus Private – Comparisons Between...

#### Public Data

Can be **shared**

Supposedly **diverse**

Typically “**narrow**” data coverage

Project knowledge usually general

Several **Categorical** labels may fit

#### Private Data

Must be **sanitized**, if shareable at all

Decidedly **not diverse**

Typically “**wider**” data coverage

Project knowledge can be **intimate**

**Categorical** labels are more specific

### Data Gathering – Beggars Can't Be Choosy

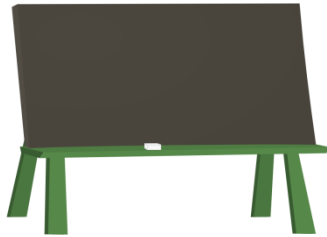
#### A List of “Nice to Haves”

Scope generally is considered most important; all else is negotiable. Though even scope can be traded away...

- Platform
- Application
- Scope
- Origin
- Effort
- Duration
- Cost
- Included labor
- Included activities
- Defects
- Staffing
- Complexity
- Extraordinary circumstances



### Private Data – Bonus Data Collection Over Time



**1. Up front:**

To gauge scope creep from start to finish. Very important for calibration, since project estimates also are made at the beginning.

**2. In process:**

During development for management, for internal use in identifying issues and gauging progress



**3. Post mortem:**

Upon completing development, this becomes the bedrock for a project repository

**4. In service:**

During maintenance to gauge life- cycle costs



### Private – The “Data Under Duress” Scenario

“The Data...”

- “...may air our dirty laundry.”
- “...reveals our labor rates.”
- “... is not relevant or is different from other projects”
- “...doesn’t exist.”
- “(...doesn’t exist, although we can’t tell you that.).”
- “...is not clean.”
- “...may threaten our bid strategy.”



### **Data Sharing – Sunlight Is the Best Policy**

- **Your laundry isn't so dirty. Everyone knows developing software is hard and takes more time than planned.**
- **You don't need to reveal labor rates.**
- **Your competitors won't ever see your data.**
- **You have more data than you know, even if you don't record effort and scope.**
- **Let us save you time and try and clean it for you.**
- **Resultant data products help you bid more successfully.**

### Data Aggregating – A Game of Least Common Denominator

Combining data, you rely on whatever all the data sets share.



Although with opportunity for side studies.

## Data Driven Cost Estimating

## And the Role of Industry and Private Data

### Task/Artifact Based Sizing - Creating Custom Metrics for Estimation

Create custom metrics by 'back firing' from accounting data. We support this offline and now directly in SEER-SEM.

DATA	
Task	Hours
PX1-1758	133
PX1-3658	28
PX0-7883	57
PX1-1913	115
PX1-6462	112
PX0-4531	53
PX0-3619	210
PX1-7922	16
PX0-3973	126
PX0-4620	8
PX0-3311	192
PX0-2488	117
PX1-4917	130
PX1-3708	140
PX0-4318	159
PX0-5449	204
SUMMARY METRICS	
Max	210
75%	154.25
50%	121.5
25%	54
Min	8

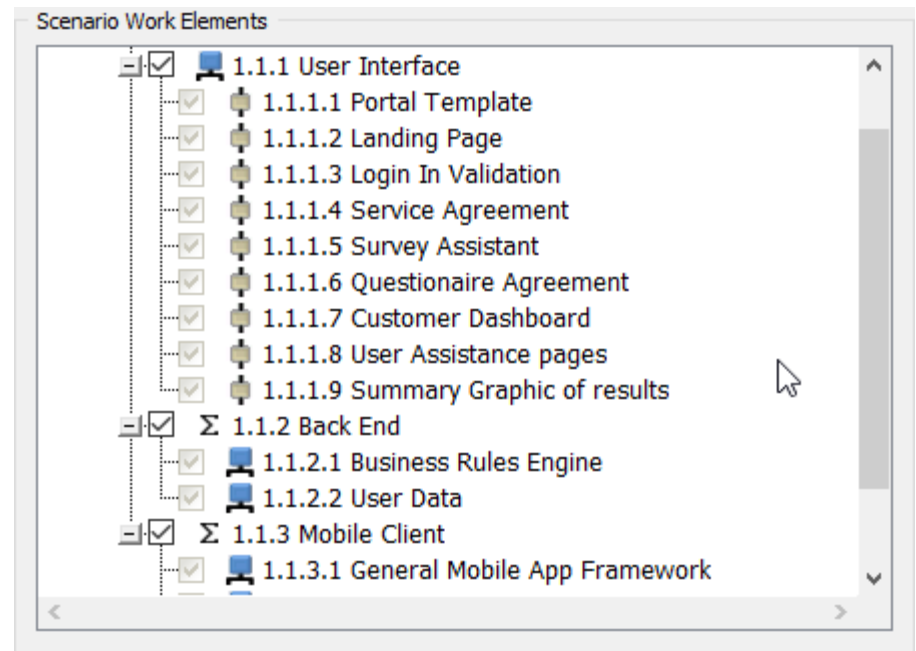
Size Metric Definition				
Size Metric	Mainframe Transaction Tasks			
File name	MFTT.PRX			
Entry Name	Value	Units	Metric	Assumptions
Simple Batch	54	Hours	3.45 FP	...
Nominal Batch	121.5	Hours	7.76 FP	...
Complex Batch	154.25	Hours	9.85 FP	...
		Hours		...
		Hours		...

PROGRAM: Blast Processing				
SIZE METRIC				
Size Metric Description				
NEW				
Simple Batch	1	1	2	
Nominal Batch	2	2	2	
Complex Batch	2	2	3	
Software phase at estimate				Code

### Meta-Data

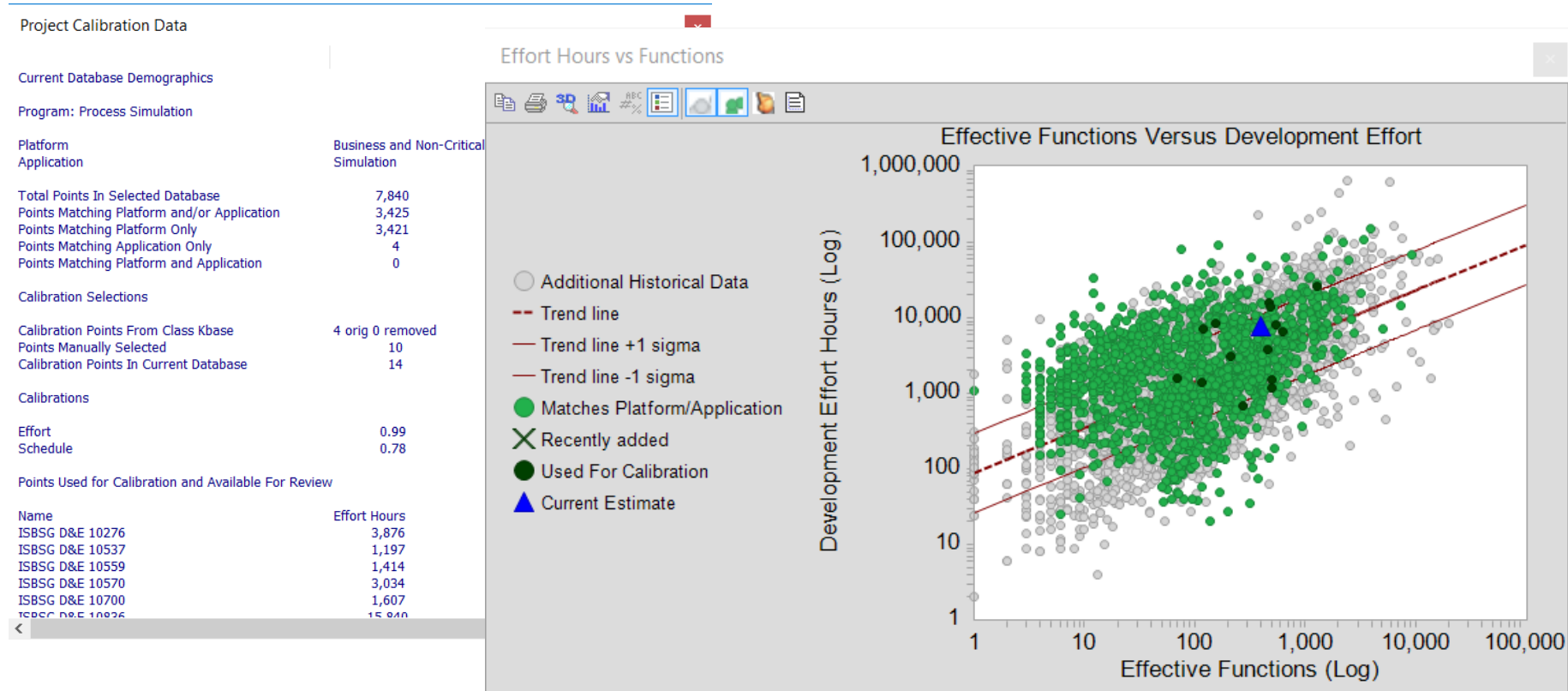
- Capture Project Patterns for estimation
  - Consistent structures make comparing projects easier
  - Avoids errors of omission in estimation
  - Facilitates data collection by identifying common categories



# Data Driven Cost Estimating

## And the Role of Industry and Private Data

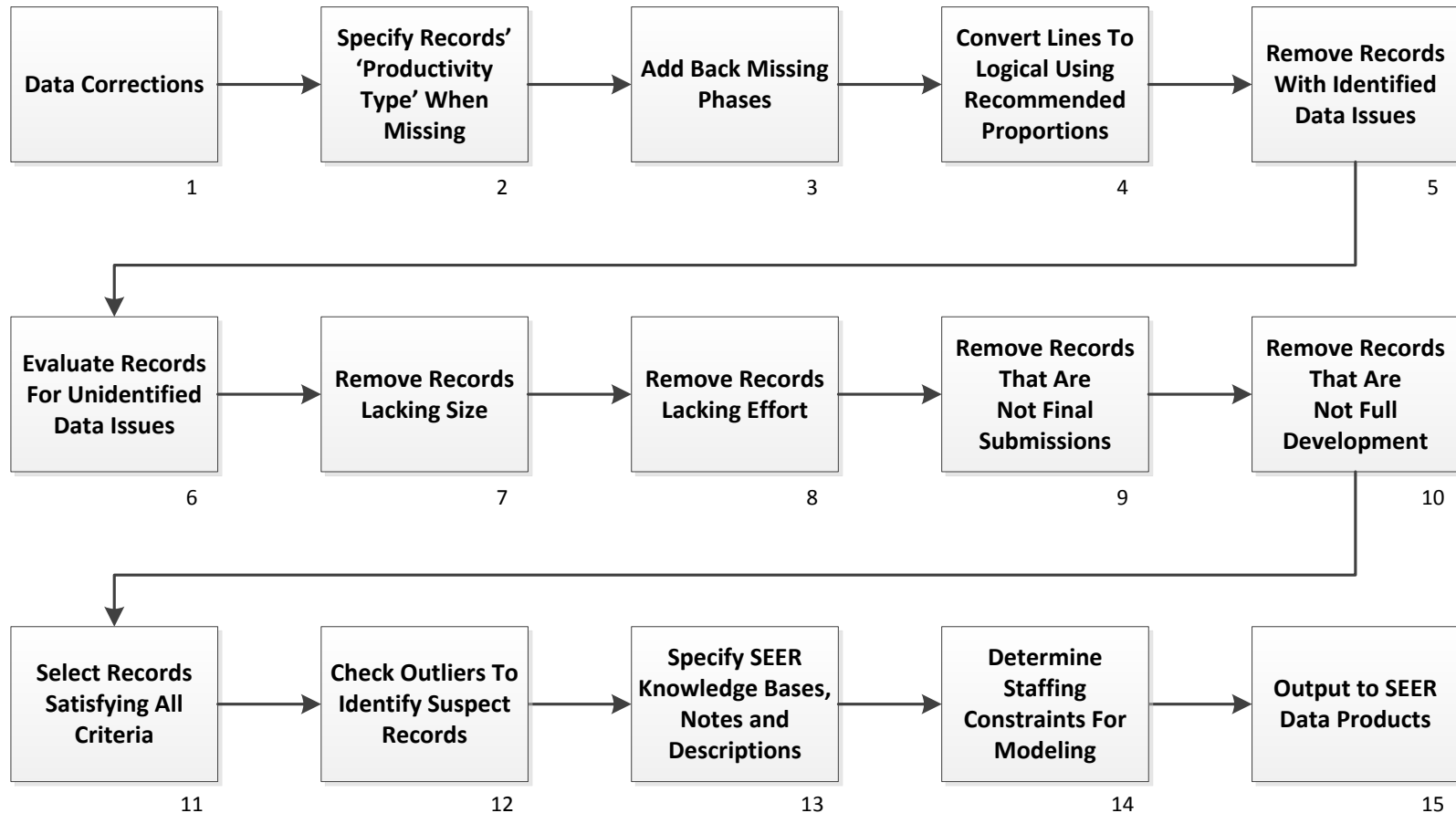
### Dynamic Calibration – making adjustments on the fly



## Static Calibration/CER Generation

- Requires data prep and data processing
- Running the regressions is the easy part
- Be prepared to rinse and repeat

### Process Shared Data For Analysis



### Forensics

**Schedule.** Was it “stop and start”? Were there schedule constraints?

**Resources.** Were there hard-hitting resource constraints?

**Volatility.** Did requirements undergo extraordinary evolution?

**Manager’s Objectives.** Was it to complete the project in *minimum time* or at *least cost*?

**Effort.** Are effort figures actually derived from cost figures?

When creating the estimate --- adjustments for *extraordinary* conditions may be possible *within the software estimating model*.



## Exploratory Data Analysis (ANOVA or regressions)

**Helps to determine highest explanatories and functional forms.**

VARIABLE NAME	ESTIMATED COEFFICIENT	T-RATIO	P-VALUE	DEFINITION OF VARIABLE
LR2	0.2842	5.52	0	1 when resource level is 2 or above * log UFPs
R2_DUM	-1.7423	-6.25	0	1 when resource level is 2 or above
LR4	0.0436	2.15	0.032	1 when resource level is 4 * log UFPs
FOURGL	-0.4255	-8.72	0	1 when 4GL language is used
LNUFP	0.6960	29.36	0	log of UFPs
PLAN	0.6842	2.38	0.017	1 when the Plan phase is included, according to revised list of phases
BUILD	-1.3634	-3.31	0.001	1 when the Build phase is included, according to revised list of phases
TST	1.1291	3.26	0.001	1 when the Test phase is included, according to revised list of phases
IMPL2	0.0907	1.68	0.094	1 when the Implementation phase is included, according to ORIGINAL list
LPLAN	-0.1511	-2.79	0.005	PLAN * log of UFPs
LSPEC	0.1249	7.33	0	SPEC * log of UFPs
LBUILD	0.1721	2.10	0.036	BUILD * log of UFPs
LTEST	-0.2344	-3.29	0.001	LTEST * log of UFPs
BUSINESS	-0.2778	-5.87	0	1 for non-critical Business projects
CLISERV	-0.1593	-1.76	0.078	1 for non-critical Client-Server projects
CONSTANT	4.0837	33.85	0	

**Principal Components Analysis** also is useful for reducing the number of variables.

Typical forms:

$$Effort = a * size^{entropy}$$


$$Duration = b * size^{duration\ entropy}$$

## And the Role of Industry and Private Data

## Automated classification:

- Enables consistency
- Reduces mistakes

SECTOR / ORG / APPLICATION LOOKUP TABLE				2	3	4	5	6	7	8	9	10
Industry Sector	Organisation Type	Application Group	Application Type									
				BusinessCrt	business	Client	clientser	cloud	embedded	erp	finproc	ground
	All industry organization typ	Business Application	Security/Authentication;	All	1							1
	All-purpose;	Business Application	Document management;	All-purpose;	1							
	All-purpose;	Business Application	Document management;Job, case,	All-purpose;	1							
	All-purpose;	Real-Time Application	Telecom & network management;	All-purpose;Real-Time Application	Telecom & network management;							1
Banking	Agriculture, Forestrg, Fishin	Business Application	Management Information System;	BankingAgri	1							
Banking	Banking;			BankingBanking;								
Banking	Banking;	Business Application		BankingBan	1							
Banking	Banking;	Business Application	Application Security Control;	Ba	1							
Banking	Banking;	Business Application	Auditing Management;	BankingBan	1							
Banking	Banking;	Business Application	Cards and Payments;	Ba	1						1	
Banking	Banking;	Business Application	Catalogue/register of things or ever	BankingBan	1							
Banking	Banking;	Business Application	Client Server;	BankingBanking;Business Appli		1						
Banking	Banking;	Business Application	Client/Server Customer Service app;	BankingBanking;Business Appli		1						
Banking	Banking;	Business Application	Customer billing/relationship manag	BankingBan	1							

ISBSG Corporate Release February 2015 mappings r14.xlsm																																		
	A	B	C	D	E	F	G	IT	IU	IW	IX	IV	IZ	JA	JB	JC	JD	JE	JF	JG	JH	JI	JJ	JK	JL	JM	JN	JO						
1								PLATFORM DERIVATION																										
2	ISBSG Corporate Release	6760 row													Source Retia	1		Source Retia:	1	2	3	4	5		Source Retia	1								
3	ISBSG Project ID	Rating	Rating	Software Age	Major Grouping	Major Grouping	Major Grouping	BUSINESS AREA TYPE		Source Retia	1	ARCHITECTURE				WEB DEVELOPMENT			DEVELOPMENT PLATF			Source Retia	1	TARGET PLATFORM										
4		Date Quality Ret	UPP Rating	Year of Project	Industry Sector	Organization Type	Application Group	App	Bus	Inf	Man	Pro	Reg	Sci	Tele	Web	App	App	Inf	Interna			App	Inf	Man	Pro	Reg	Sci	Tele	Web	App	App	Inf	Interna
5	10001 D	A			1998 Service Indus Recreation & Personnel Services;	Business Application		1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0					
6	10011 B	A			1996 Construction Construction;	Business Application		0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0	0	1					
7	10010 B	A			2002 Wholesale & Billin;	Business Application																												
8	10014 B	A			2004																													
9	10015 B	A			2000 Wholesale & Retail Trade;	Business Application											1	0	0				0	1	0	0	0	0	0	0	0	0	0	
10	10026 B	A			2000 Insurance Insurance;	Business Application		0	0	0	0	0	0	0	0	0	0	0	0				0	1	0	0	0	0	0	0	0	0	0	0
11	10029 B				2004 Banking Banking;	Business Application																	0	1	0	0	0	0	0	0	0	0	0	0

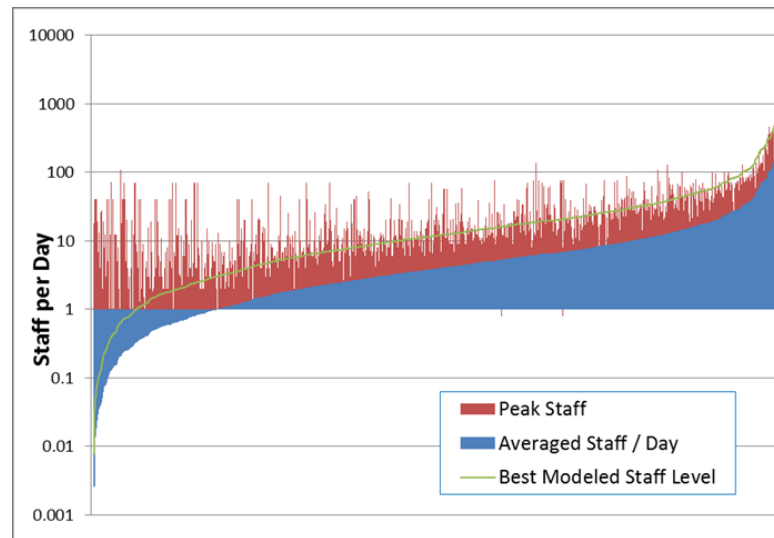
### Typical Normalizations

- Adjustment to Line of Code or Function Point measures
  - Methods vary
  - For private data, frequently no adjustment is made – if customer prefers a given metric “flavor”, we use it
- Adjustments to re-include missing activities or labor
  - Typically using simple proportions derived from other projects in the sample
- Inferring peak staffing
  - Informed by average staffing, when known or as calculated (example on next page)

### Peak Staff - Significant, Often Missing

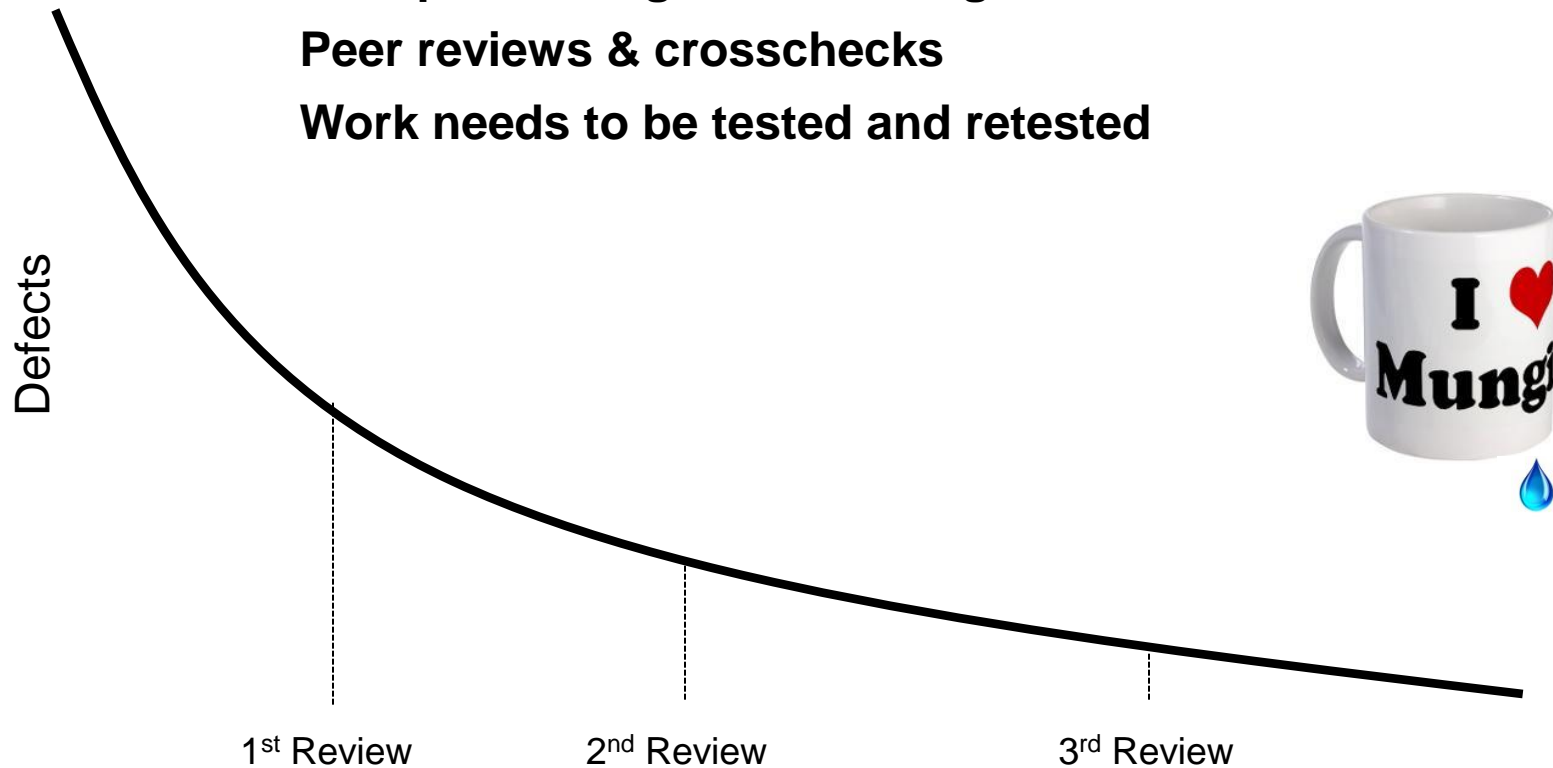
**Staffing constraints have a significant effect on estimating;** when not specified, they should be inferred when possible:

- Use reported peak staff when available
- When not reported, based on analysis, specify peak staff at  $X * \text{average staff}$  with no constraints below  $Y$  FTEs

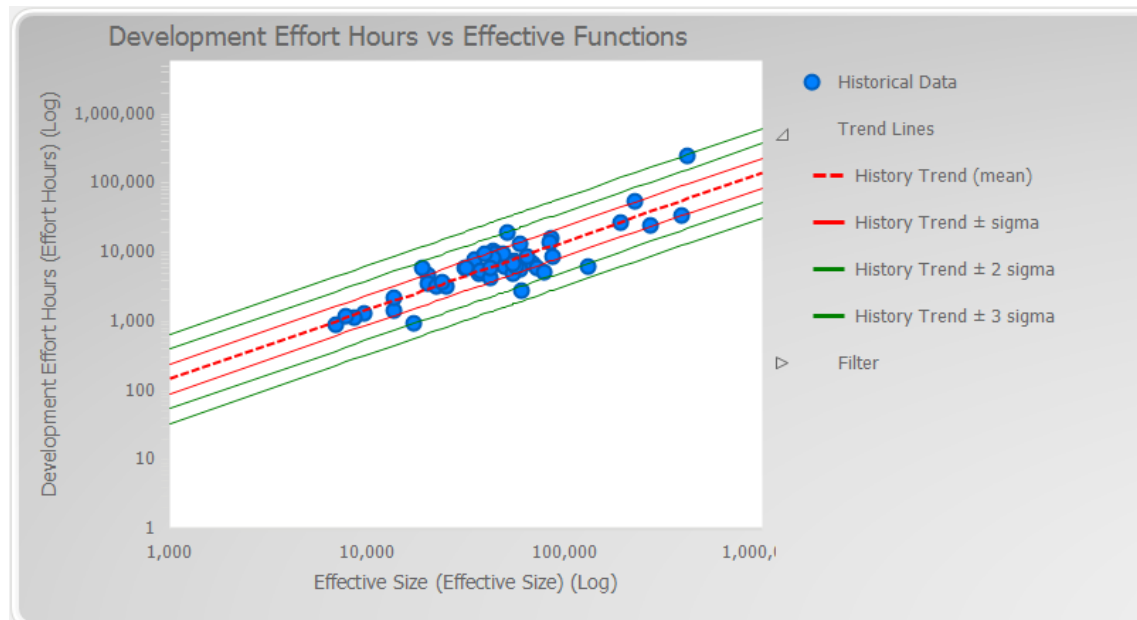


### Oversight

**Data processing is like coding:  
Peer reviews & crosschecks  
Work needs to be tested and retested**



## Sanity Checks



**SEER Metrics** is used to plot data within probability bands.  
Points outside 2 sigma are examined.

## More vs. Less Data – Double-Edged Sword

### Less Data

**Harder** to spot outliers

**Easier** to explain them

**Harder** to systemically correct

### More Data

**Easier** to spot outliers

**Harder** to explain them

**Easier** to systemically correct

# Data Driven Cost Estimating

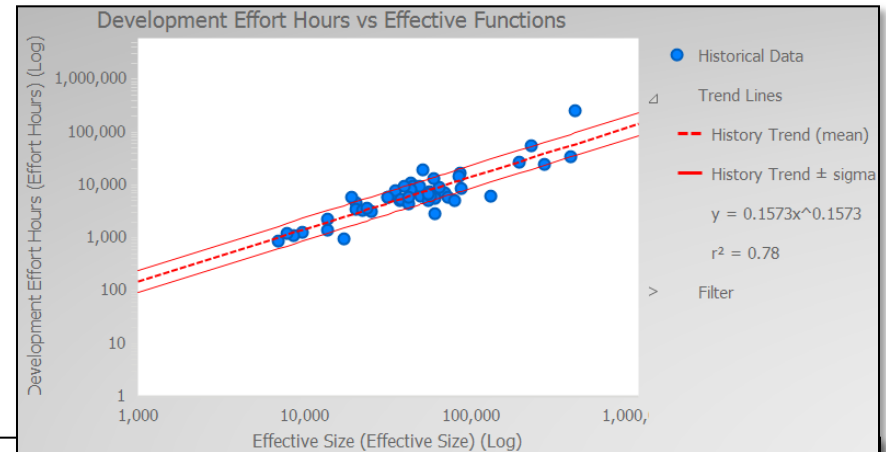
# And the Role of Industry and Private Data

## Calibration Automation

Automation for solving form  $y=a*s^b$

OLS using SEER Metrics

LAE using Excel



Customer Data Set					Effective Size (Effective Size) (Log)													
Sample size		168		2 - Obtain calibration values, interacting with SEM					LAE REGRESSION									
				Obtain Calibration Values														
OLS regression results for comparison				Effort Exponent					1.07									
Exponents				Schedule Exponent					0.53									
Effort		Duration		Effort Multiplier					0.76									
1.0792		0.5243		Schedule Multiplier					1.78									
Multipliers																		
Effort		Duration																
0.2895		0.0393																
0 - A SEM project needs to be created beforehand																		
1 - Reset sheet and receive new data																		
Reset Sheet and Receive Data																		
					COMBINED PERFORMANCE													
Effective Size	Development Actual Effort	Development Actual Schedu	Development Hours	Development Schedule M.	Development Actual Schedu	Development Hours	Development Schedule M.	SEM Calibration Performance		Effort	Duration	4.2523	0.51022	4.2522	0.0000	1.6340	1.1238	
25900	17875.2	3.24	54109.81	32.84	3.24	39020.3	45.54	SEM uncalibrate	Median absolute deviation	min	122%	92%	4.2523	1.49973	4.2283	0.0239	1.6221	0.1224
24600	17875.2	31.60	50867.27	32.17	31.60	36928.39	4.43E+01	SEM uncalibrate	Mean absolute deviation	min	373%	1700%	3.676	0.61067	3.6888	0.0128	1.3527	0.7420
7700	4742.4	4.08	12621.33	20.21	4.08	10656.26	2.39E+01											



## Measuring Calibration Efficacy

Calibration efficacy can be:

- **Specific** – “Does this stratification produce enough predictions within X percent?”
- **Relative** – “Does another stratification work better while capturing the requisite estimating scenarios?”

Stratification A				Stratification B			
		Effort	Duration			Effort	Duration*
Median absolute deviation	min	51%	78%	Median absolute deviation	min	37%	96%
Mean absolute deviation	min	80%	163%	Mean absolute deviation	min	48%	146%
Standard error (normalized)	min	88%	75%	Standard error (normalized)	min	90%	53%
Prediction within 10%	max	6%	13%	Prediction within 10%	max	18%	5%
Prediction within 25%	max	18%	25%	Prediction within 25%	max	34%	22%
R-squared	max	66%	14%	R-squared	max	94%	0%
Number of records: 109				Number of records: 103			
		Effort	Duration			Effort	Duration
Multiplier		0.95	2.04	Multiplier		0.90	5.04
Exponent		0.84	0.23	Exponent		0.93	0.10
Original multiplier		4.65	2.19	Original multiplier		2.06	23.20
OLS Multiplier		4.66	1.51	OLS Multiplier		3.65	23.20
OLS Exponent		0.83	0.26	OLS Exponent		0.89	0.05

### Typical Statistics

**Mean Magnitude of Relative Error (MMRE)**

Percentage variation between actual and estimate with actual as divisor. *Closer to zero the better.*

**Average Estimate Ratio**

The average of (actual / estimate) ratios. *Closer to one the better.*

**Median Estimate Ratio**

The median of (actual / estimate) ratios. *Closer to one the better.*

**Standard Deviation**

The standard deviation of estimate vs. actual variances. *Closer to zero the better.*

**Prediction (x)**

The percent of estimates falling within x% of actual outcomes. *Closer to 100% the better.*

**R-squared**

Goodness of fit. (The goal is to purely “fit the data” and so this is a good measure.) *Closer to one the better.*

**Other statistics**

Chow for structural breaks (such as may occur as scope grows), F test to detect difference between two samples (such as estimate versus actual),

### Side Studies

An example of the studies possible, in addition to classic size-effort and duration-effort relationships:

1. Size growth studies, to understand estimated versus actual system size
2. The impact of staffing on project productivity
3. Determinants of project duration
4. Ability of project requirements to determine effort and duration
5. Variations in productivity by program, division/contractor, maturity rating, programming language, toolset, etc.
6. Determinants of project phases' relative durations
7. Standardized, multiple tags for projects, yielding more descriptive information and more data for specific queries
8. Ability to predict maintenance and sustainment
9. What taxonomy has the best explanatory versus descriptive power?

### Private/Local Data

- Use to shape the estimation process
  - Sizing
  - Project Patterns
  - Productivity Tuning
  - Economic Factors
  - Estimate Catalogs
- Focus is consistency, repeatability and accuracy

### Public/Industry Data

- Use to target major trends
  - Calibration/CER Generation
  - Productivity Benchmarking
- Sanity Check Your Estimate
- Vendor/Competitor Evaluation
- Evaluate Industry Demographics

### Applying Data To Estimation

- It's not a lights out process
  - Requires analysts, SMEs, humans to process evaluate and identify where it can add value
- Getting started can be slow and bumpy
  - As the process is repeated, it gets easier
  - Automate as much as possible
- Revisit processes
- Measure the benefits
  - Accuracy improvements, estimate turnaround, confidence in estimates