QSM Research: Understanding the Physics of Software Projects

Topics to include:

How to Develop a Data-driven Estimation Model Small Teams Deliver Lower Cost & Higher quality History is the key to Estimation Success

Presentation Agenda

- About QSM
- How it all began
 - Review Larry Putnam, Sr.'s original research & data
- Does the initial research stand the test of time?
 - Follow up with new data
 - Regularly published research almanacs
- Does current agile data exhibit similar behavior?
- Summary



Quick Facts

- Industry Leadership
 - Established in 1978 by Larry Putnam, Sr., a pioneer, and renowned thought leader in the software estimation field
 - Headquartered in McLean, VA, with affiliate offices in Europe and across the U.S.
- Industry Leading Tools and World's Largest Database
 - 13,000+ completed projects providing deep insights into fundamental relationships at play in software and systems development
 - Industry- leading SLIM[®] modeling tools enabled by proven parametric algorithms
- 40 Years of Proven Field Success
 - Experience working with thousands of clients in every industry and sector
 - Measurable impact on hundreds of consulting engagements
 - Experts at helping customers build their own internal estimation capacity
 - \$30M+ in product R&D investment guided by client needs and input over the years



Success is Not an Option - Or is It?

- A Simple, But Clear Mission To Help Clients Succeed:
 - By empowering them to negotiate well-informed decisions and execute successful software development and deployment
 - Through measuring and increasing efficiencies in their operations
 - Optimizing productivity and quality across their portfolio of projects

• We Do That By:

- Offering state-of-the-art tools and training coupled with comprehensive consulting services and outstanding customer support
- Turning data into defensible and actionable management information
- Identifying project constraints; understanding organizational objectives; applying metrics, assumptions and proven algorithms to accurately identify realistic goals and alternatives
- Providing transparency to all stakeholders to enable fact-based decisions
- Staying on the leading edge through relevant and timely research in the estimation industry





Our Founder – Lawrence H. Putnam, Sr.

- Started out as a career army officer
 - Trained as a nuclear affects engineer
 - Took Fortran programming to do nuclear calculations
 - Selected to run the Army's IT budgeting operation at the Pentagon (because of Fortran programming expertise)
 - Lost \$10 million at first congressional budget hearing
 - Decided better solutions were needed to manage software projects
- Retired from Army and started QSM in 1978
 - Created the Software Lifecycle Model (SLIM)
 - Expanded into a suite of tools over time
- Retired from QSM in 2007
 - Still comes into the office every day

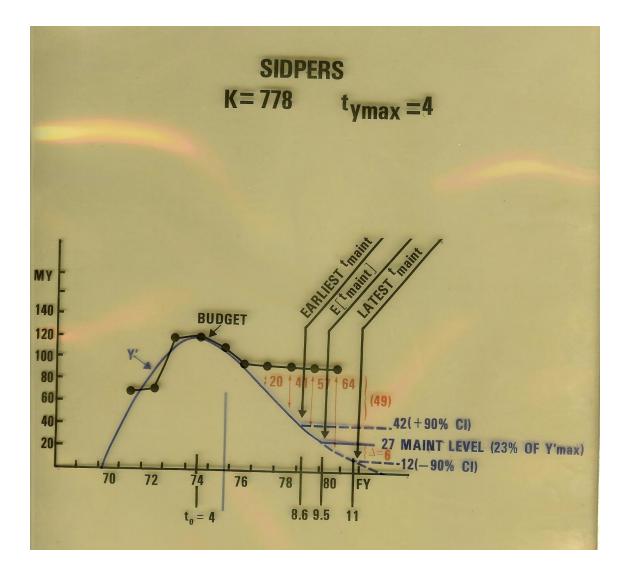


Software Equation Background

- Time Frame 1975
- SIDPERS \$10 Million Budget Loss
- Discovery of Norden-Rayleigh Equation
- Research to Discover the Rayleigh Parameters (Time and Effort)

Who had some data and what did it look like?

SIDPERS Budget Data



This is the project that lost Larry Sr. \$10 Million out of the US Army's data processing budget.





Original Data Set Larry Had to Work With

	USACSC	SUSTEM CHADACTEDISTICS					
	USACSC SYSTEM CHARACTERISTICS		-	Number of			
	Life Cycle	Size Development	Files	Rpts.	App1.	Progs	
System	K (MY)	Time t _d (YRS)	×ı	×2	×3		
MPMIS	73.6	2.28	94	45	52		
MRM	84	1.48	36	44	31		
ACS	33	1.67	11	74	39		
SPBS	70	2.00	8	34	23		
COMIS	27.5	1.44	14	41	35		
AUDIT	10	2.00	11	5	5		
CABS	7.74	1.95	22	14	12		
MARDIS	91	2.50	6	20	27		
MPAS	101	2.10	25	95	109		
CARMOCS	153	2.64	13	109	229		
SIDPERS	700	3.65	172	179	256		
VTAADS	404	3.50	155	101	144		
BASOPS-SUP.	591	2.73	81	192	223		
SAILS AB/C	1028	4.27	540	215	365		
SAILS AB/X	1193	3.48	670	200	398		
STARCIPS	344	3.48	151	59	75		
STANFINS	741	3.30	270	228	241		
SAAS	118	2.12	131	152	120		
COSCOM	214	4.25	33	101	130		

The Original Army Data Set (19 Good projects)

1. Homogenous Data Set

Same organization

Same application type

Same tooling

Complete data (size-time-effort)

Man was I LUCKY !





Relationships Between Size, Time & Effort

(X1,X2,X3 are Size Metrics)

There are relationships other than the difficulty that have significantly high correlation with measures of the product. These relationships are:

$$K/t_d^4 = f_1 (x_1, x_2, x_3)$$

$$K/t_d^3 = f_2 (x_1, x_2, x_3)$$

$$K/t_d^2 = f_3 (x_1, x_2, x_3)$$

$$K/t_d = f_4 (x_1, x_2, x_3)$$

$$K = f_5 (x_1, x_2, x_3)$$

$$Kt_d = f_6 (x_1, x_2, x_3)$$

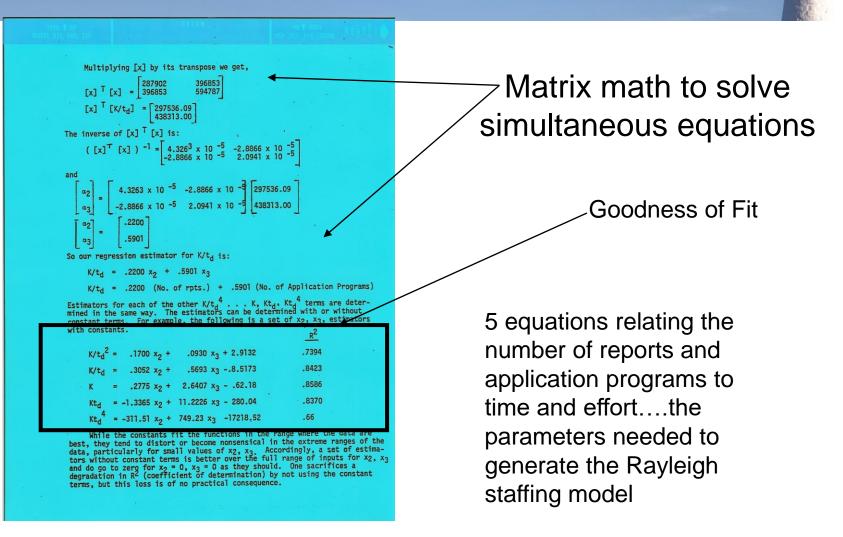
$$Kt_d^4 = f_7 (x_1, x_2, x_3)$$

where K and t_d are the parameters of the Rayleigh equation.

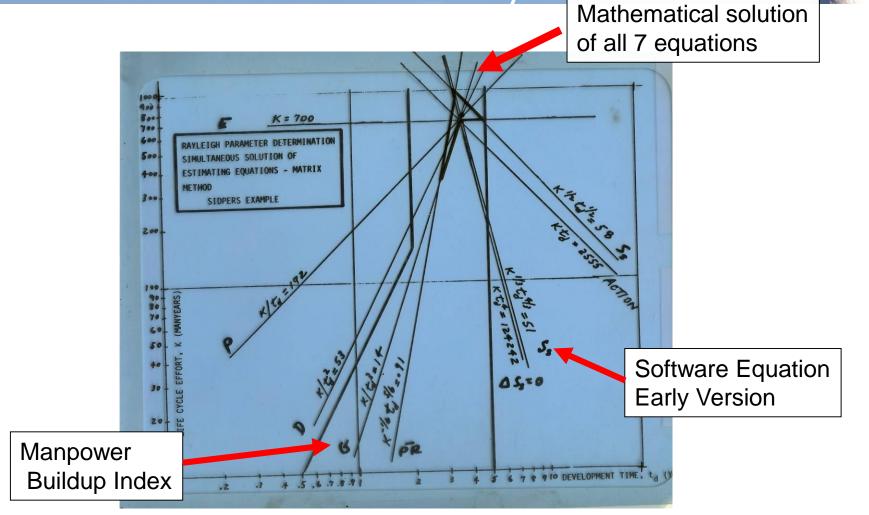
These are possible equations that would have to intersect (any 2) to provide the parameters for time and effort (K and td)

QSM[°]





Graphical Curve Fitting and Some of the Equations We Still Use Today







Schedule Milestone Determination

		NKE	LATION T	U tymax					
			FRACTION OF tymax						
SYSTEM	^t ymax	K	DESIGN CERTIF.	S.I.T.	P.E.T.	EXTEN			
IFS-1	4	495	.25	.65	.79	.85			
VTAADS	4	422	.58	.75	1.0	1.28			
SIDPERS	4	778	.48	.60	.73	.80			
SAILS A/B Conus	3	1064	.38	.71	.74	.90			
SAILS AB/X	3	1020	.53	.60	.64	.72			
			DESIGN	<u>s.i.t</u> .	P.E.T.	EXTEN.			
AVERAGE			.43	.66	.80	.93			
ô			.12	.07	.12	.20			
1.6 $\hat{\sigma}$ RANGE ABOUT MEAN			.2462	.5577	.6199	.61-1.2			

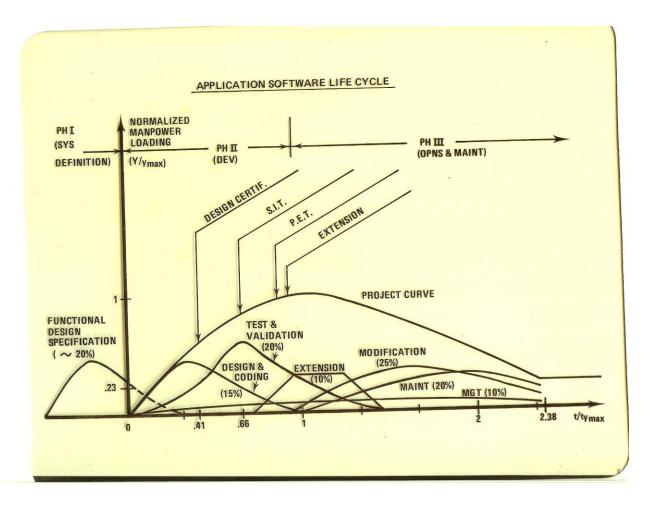
Milestone determination
 based on the average of these 5 Army Systems





The Putnam-Norden Rayleigh Model

End Product Estimate (Schedule, Effort and Milestones)





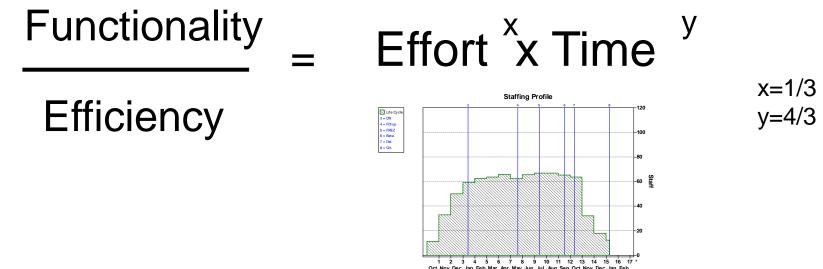
Key Thoughts

- Rayleigh Model is an excellent model of human design processes
- Software equation captures the time effort trade-off relationship
- The algorithms have withstood the test of time
- This suggests we must be close to the fundamental behavior of software development
 - How humans solve complicated design problems
 - Can be applied to other domains that exhibit similar behavior
 - Hardware, firmware, infrastructure, etc.

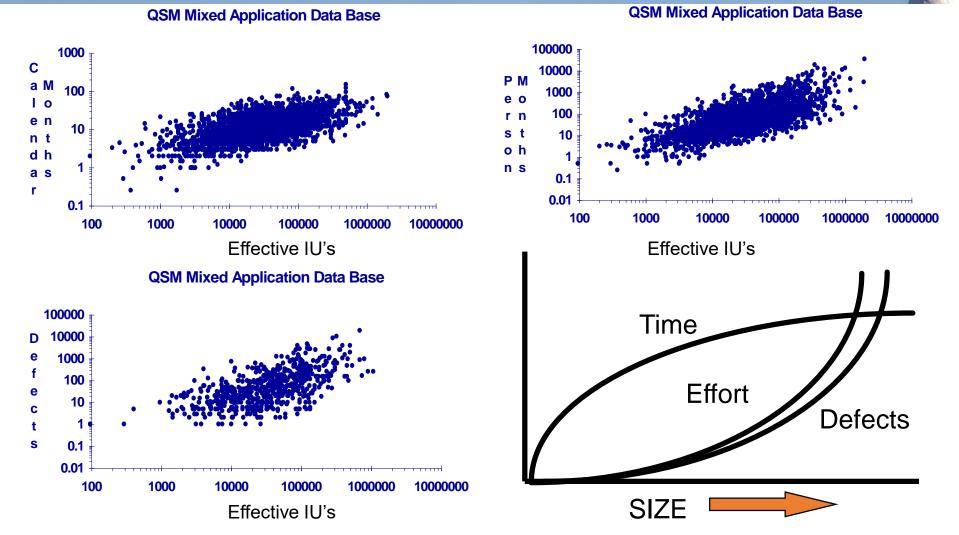


In Software (Design Processes) Everything is Non-linear

- Schedule & effort as a function of size
 - As the size and scope changes there is a non-linear change in the schedule and effort required to develop it
- Time effort trade-off
 - Models human communication complexity
 - Exponential increase in unique communication paths as we add people



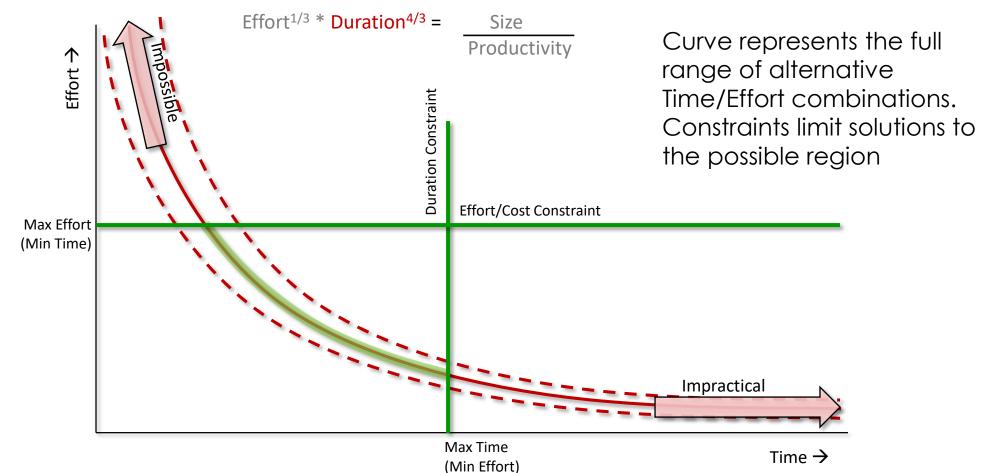
Software's Non-linear Behavior with Respect to Size





Time/Effort Trade-Off & Constraints

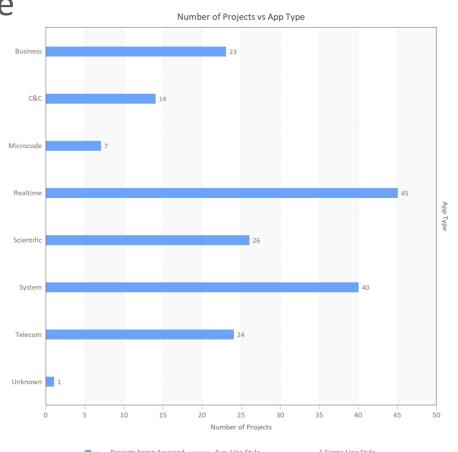






Confirmation with RADC Data

- Additional data became available from the Rome Air Development Center
 - Larger sample of data (180 projects) spanning multiple organizations
 - More diverse application complexity
 - Included Business/IT, Engineering & Realtime systems
 - Data covered a larger size range with multiple languages
- Similar Behavior Exhibited
 - Non-linear change in schedule & effort as a function of size/scope
 - Non-linear schedule/effort trade-off of different

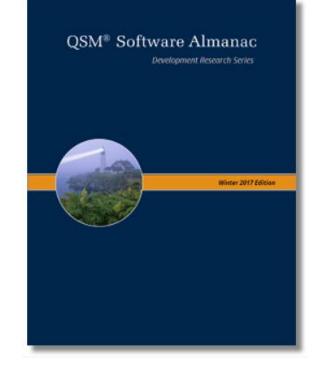






Do These Trends Stand the Test of Time

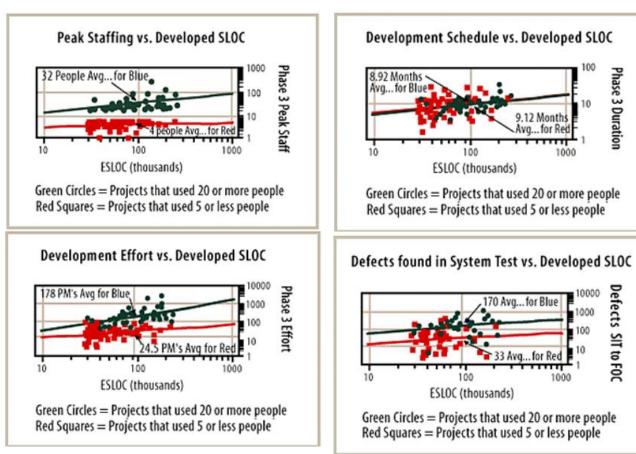
- QSM continues to collect industry data and revalidate the trade-off relationships
- We regularly publish new research in the form of QSM Software Almanac
 - http://www.qsm.com/resources/research/qsm-almanacs
- Work with clients to incorporate their own data
 - Custom sizing models
 - Productivity calibration
 - Methodologies used
 - Phase tuning
 - Staff loading
 - Skill allocations





Revalidated the Trade-Off Relationship in the Mid-1990's

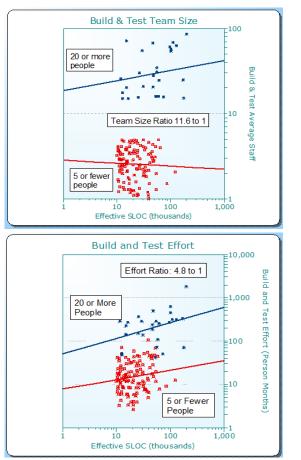
- "Haste Makes Waste" article
 - http://www.qsm.com/risk 02.html

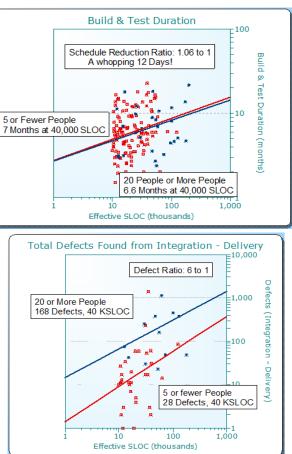




Revalidated the Trade-Off Relationship in the Mid-2000's

- 2006 QSM Almanac
 - <u>http://www.qsm.com/resources/research/qsm-almanacs</u>



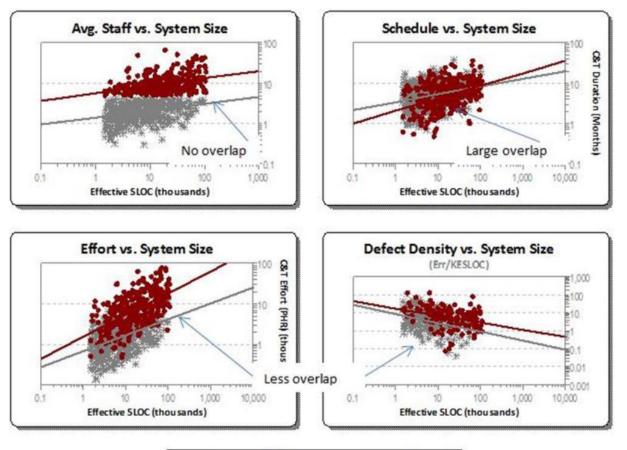




Revalidated the Trade-Off Relationship in 2017

• 2017 QSM Almanac

http://www.qsm.com/resources/research/qsm-almanacs



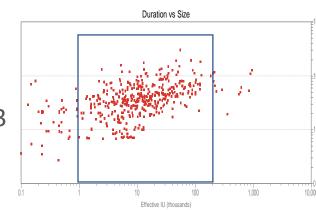
- Small Teams • - Large Teams

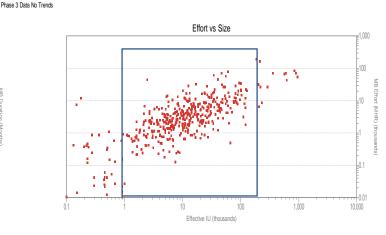
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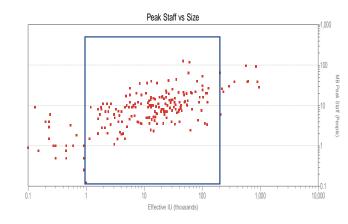


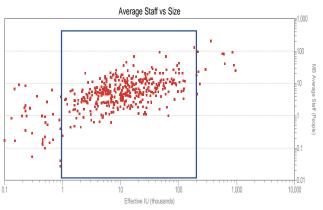
Do We See a Similar Pattern in Current Agile Data?

- Created two datasets with overlapping size data
 - Overall Sample
 - Avg staff median 5.36, mean 9.33
 - Size 100 IU and 953k IU
 - Number of projects 436
 - Large Team
 - Avg staff >= 9 people
 - Size >=1k IU and <= 200k IU
 - Number of projects 119
 - Small Team
 - Avg staff <= 4 people
 - Size >=1k IU and <= 200k IU
 - Number of projects 129







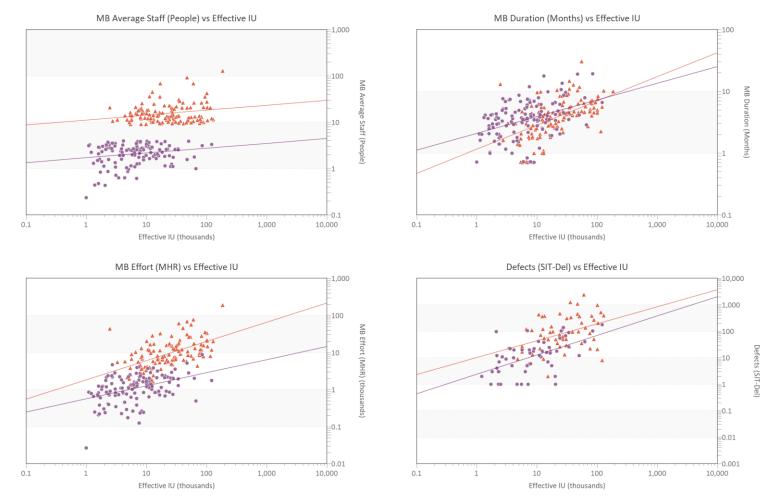






Trade-Off Relationship with Agile Data

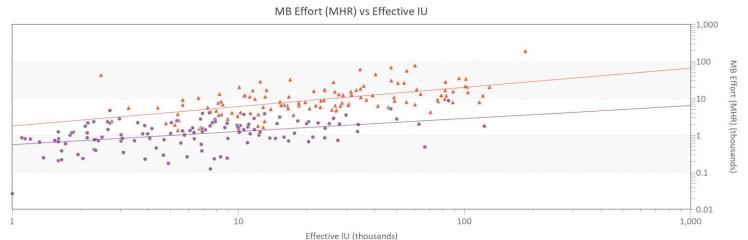
Phase 3 Small-Large Team Trends





Effort Trade-Off Relationship with Agile Data

Dataset Comp - Phase 3 Effort



Comparison of Large Teams to Small Teams MB Effort (MHR) vs. Effective IU

	MB Effort (MHR) Values							
	at Min Effective IU: 2464	at 25% Quartile Effective IU: 10600	at Median Effective IU: 23045	at 75% Quartile Effective IU: 47461	at Max Effective IU: 185500			
Benchmark Data Set: Small Teams	778.70	1303.05	1713.83	2211.45	3577.41			
Comparison Data Set: Large Teams	2924.72	6238.94	9337.63	13587.96	27577.17			
Difference From Benchmark	2146.02	4935.90	7623.80	11376.51	23999.77			

On average 4-8 times more effort expended by large teams for the same amount of functionality than the small team approach

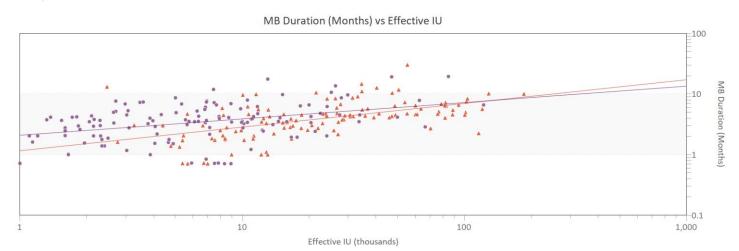
Comparison breakpoints based on min, max, median and quartile values for the data set: Large Teams





Schedule Trade-Off Relationship with Agile Data

Dataset Comp - Phase 3 Duration



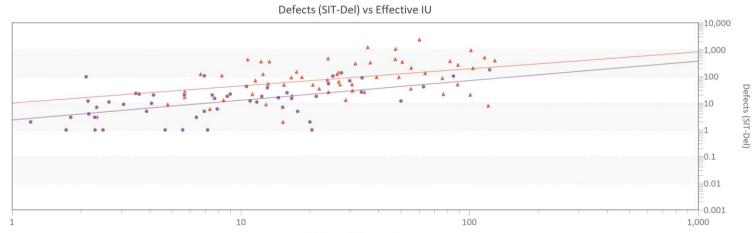
Comparison of Small Teams to Large Teams MB Duration (Months) vs. Effective IU

	MB Duration (Months) Values					
	at Min Effective IU: 1000	at 25% Quartile Effective IU: 2670	at Median Effective IU: 6840	at 75% Quartile Effective IU: 11743	at Max Effective IU: 122000	
Benchmark Data Set:						
Large Teams	1.16	1.70	2.46	3.04	7.59	
Comparison Data Set:						
Small Teams	2.09	2.72	3.51	4.07	7.66	
Difference From Benchmark	0.93	1.02	1.05	1.03	0.07	

A month or less difference in schedule between large and small team data sets







Effective IU (thousands)

Comparison of Small Teams to Large Teams Defects (SIT-Del) vs. Effective IU

Defects (SIT-Del) Values

	at Min Effective IU: 1100	at 25% Quartile Effective IU: 3263	at Median Effective IU: 7672	at 75% Quartile Effective IU: 15557	at Max Effective IU: 122000	
Benchmark Data Set: Large Teams	10.80	21.66	37.42	58.81	219.57	
Comparison Data Set: Small Teams	2.55	5.67	10.62	17.85	80.99	
Difference From Benchmark	-8.25	-15.99	-26.80	-40.97	-138.58	

3-4 times as many defects created by large teams than small team projects



Trends in Modern Development Methods

Modern agile & lean development methods seem to embrace these principles exploiting these positive economic trends - **Finally!**

- There is an emphasis on using small cross function teams
- Constant review and re-prioritization of functional content
 - Recognizes that change is inevitable
 - Emphasis on concept of "content must add value"
- Deliberately plan for short cycle times and strive for continuous delivery
 - Recognize that there is some minimum value that the customer will accept
 - Minimum Marketable Features (MMF) or Minimum Viable Product (MVP)



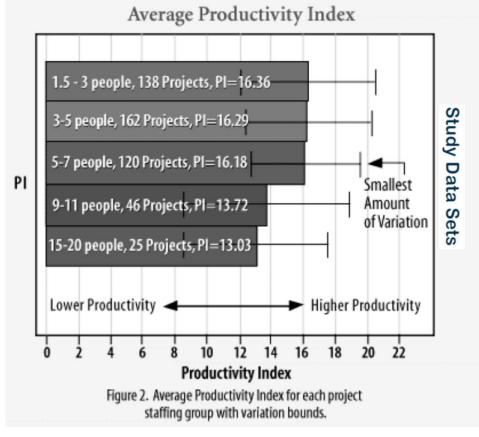
Trends in Team Size

• From the Scrum Guide Today

Development Team Size

Optimal Development Team size is small enough to remain nimble and large enough to complete significant work within a Sprint. Fewer than three Development Team members decrease interaction and results in smaller productivity gains. Smaller Development Teams may encounter skill constraints during the Sprint, causing the Development Team to be unable to deliver a potentially releasable Increment. Having more than nine members requires too much coordination. Large Development Teams generate too much complexity for an empirical process to be useful. The Product Owner and Scrum Master roles are not included in this count unless they are also executing the work of the Sprint Backlog.

• From QSM Research in 1997



Two Pizza Rule

- Developed by Jeff Bezos to keep meetings more productive
- The science behind it matches what we see in the software data
- As we add people to the team we exponentially increase potential communication paths

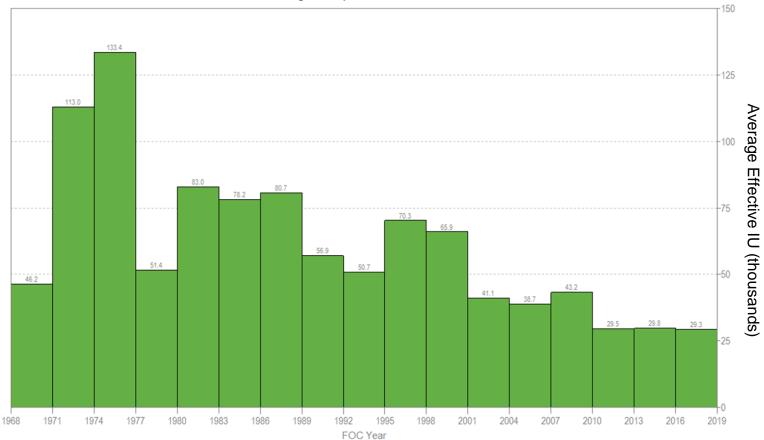






Trends in Project Size

Average Project Size Over Time



General trend is that size has been declining over time. Early on this may have been more due to more powerful languages and technology.

More recently we believe this size reduction is more driven by reuse and a conscious effort to keep project size as small as possible.



Summary

- Larry's original research and estimating models are as relevant today as they were close to 40 years ago
- They have been adapted and improved to make them applicable to modern development practices and methodologies
 - More configuration and fundamental changes to the model
- Current agile data shows very similar non-linear behavior
- Current agile development methodologies seem to be exploiting these behaviors
 - Promotes small cross functional teams
 - Promotes keeping scope as small as possible to fit short iterations

