



COMP 5700/6700/6706

Software Process

Spring 2016
David Umphress

Estimation

- Lesson: Estimation
- Strategic Outcomes:
 - To lay a foundation for estimating end-product size and effort
 - To learn the mechanics of estimating by component
- Tactical Outcomes:
 - To know common size estimating approaches in use today
 - To comprehend the use of components in estimation
 - To comprehend how to map size to effort
 - To apply an estimation technique to a sample problem

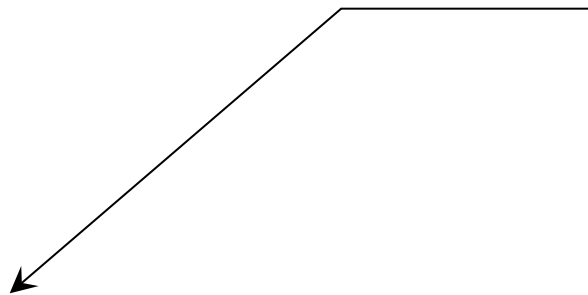
- Support material:
 - "Estimation" Stellman, A. and J. Greene. 2005. *Applied Software Project Management*. O'Reilly.
 - Fingers in the Air: A Gentle Introduction to Software Estimation. Asproni, G. 2005. *Methods and Tools, Winter 2005*. pp 2-13
 - Fundamentals of Function Point Analysis. Longstreet, D. Downloaded Jan 09. www.softwaremetrics.com.
- Instant take-aways:
 - High-level coverage of common size estimation approaches
 - Component-based estimation

- Bookshelf items

- McConnell, S. 2006. *Software Estimation: Demystifying the Black Art*. Microsoft Press
- Humphrey, W. 1995. *A Discipline for Software Engineering*. Addison Wesley.
- Boehm, B.W., 1981. *Software Engineering Economics*. Prentice-Hall

Syllabus

- Software engineering raison d'être
- Process foundations
- Common process elements
- Construction
- Reviews
- Refactoring
- Analysis
- Architecture
- Estimation
- Scheduling
- Integration
- Repatterning
- Measurements
- Process redux
- Process descriptions*
- Infrastructure*
- Retrospective



- **Rationale**
- **Principles of estimation**
- **Common techniques**
- **Component-based estimation**
 - **process**
 - **size**
 - **effort**
 - **confidence**
 - **setup**

COMP5700/6700/6706 Goal Process

Minimal Guiding Indicators

Goal	Indicator
Cost:	None
Schedule:	PV/EV > .75
Performance:	
Product:	none
NFR:	none
FR:	100% BVA
Process:	pain < value

Minimal Sufficient Activities

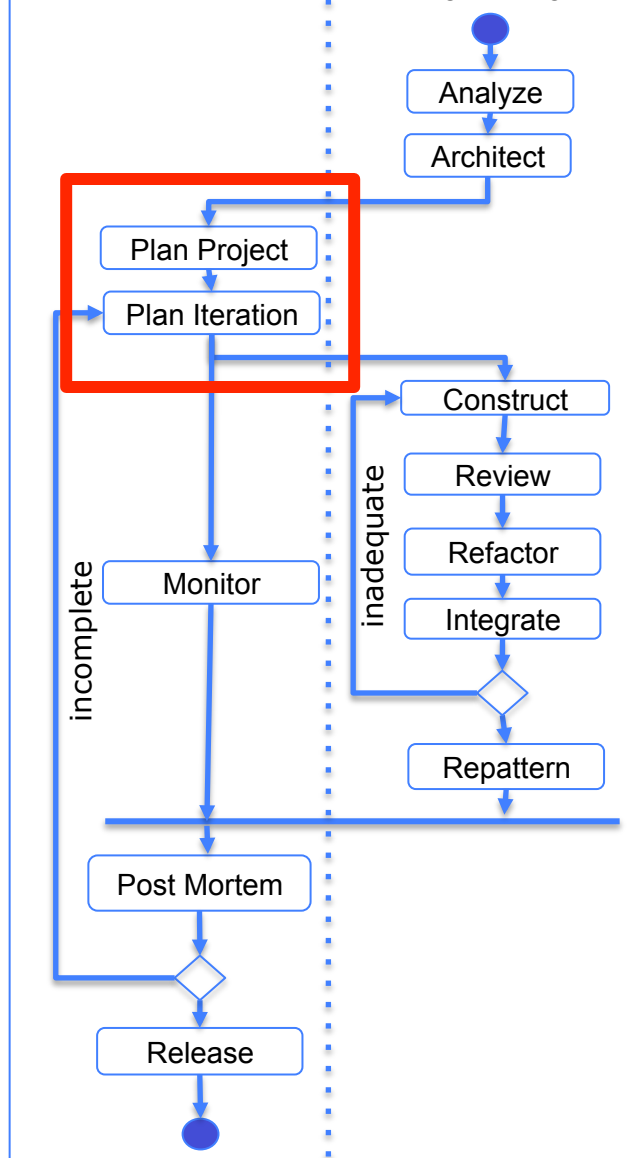
Engineering Activities

Envision
Analyze
Synthesize
Architect
Articulate
Construct
Refactor
Interpret
Review
Integrate
Repattern

Operational Activities

Plan
 Plan project
 Plan iteration
Monitor
Release

Minimal Viable Process



Minimal Effective Practice

MSA	MEP
Analyze	Scenarios
Architect	CRC
Plan Project	Component-based estimation
Plan Iteration	Component-iteration map
Construct	TDD
Review	Review checklist Test code coverage
Refactor	Ad hoc sniffing
Integrate	Ad hoc
Repattern	Ad hoc
Monitor	Time log Change log Burndown
Post Mortem	PV/EV
Release	Eclipse zip spreadsheets

How do you estimate effort?

Rationale

- Why do we need to estimate effort?
 - Effort estimate \sim defect rate
 - can predict quality of product, testing effort
 - Effort estimate \sim resources
 - can forecast hardware/software/peopleware needed
 - Effort estimate \sim cost
 - can predict bucks
 - Effort estimate \sim assessment mechanism
 - can measure progress using estimated vs actual
 - Effort estimate \sim buy-in
 - can use estimating efforts to gain worker/management buy-in

Rationale

- What makes us think we can?
 - Estimating models in other fields
 - large base of history
 - in wide use
 - generate detailed planning data
 - Software estimating experience
 - 100% + errors are normal
 - few developers make estimates
 - fewer still use orderly methods
 - But ...
 - empirical studies show relationship between size and resources
 - we must choose an appropriate method
 - based on empirical evidence within organization
 - may (will?) differ from place to place, culture to culture

A Good Estimate is

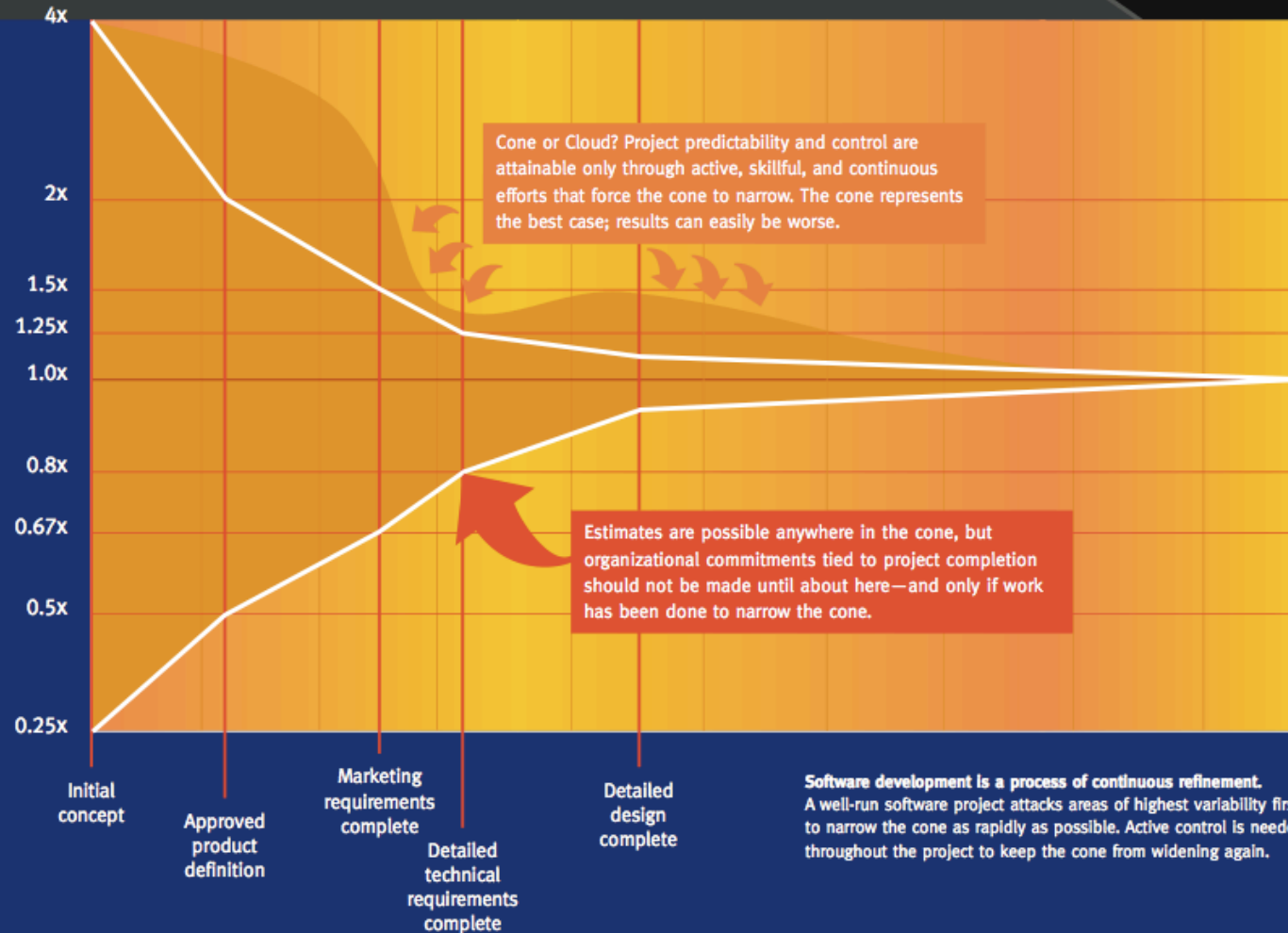
- "... an estimate that provides a clear enough view of the project reality to allow the project leadership to make good decisions about how to control the project to hit its targets."

Steve McConnell

Cone of Uncertainty:

Remaining variability
in project scope
(cost, size, or features)

All software projects are subject to inherent errors in early estimates. The Cone of Uncertainty represents the best-case reduction in estimation error and improvement in predictability over the course of a project. Skillful project leaders treat the cone as a fact of life and plan accordingly.



Construx®

Effort Estimation

- Foundation
 - Planning process requires us to gauge the magnitude of work.
 - Our approach will be to
 - identify components, then
 - estimate size, then
 - estimate effort
 - Our goal in do so
 - measure functionality delivered to customer
 - provide a measure of product quantity understood by users and developers
 - provide a measure that is available early in the product life-cycle
 - serve as a metric for project planning, project control, process improvement
 - But, must decide what “size” means

[Humphrey 1995]

Size

- Common measures of size
 - physical:
 - LOC
 - normalized LOC
 - screens
 - pages
 - scripts
 - reports
 - function:
 - function points
 - feature points

Common Estimation Techniques

- By guessing
 - By expert opinion
 - correlate estimates of size from multiple sources
 - e.g., Wideband-delphi
 - By function
 - predict by examining features
 - e.g., Function points
 - By analogy
 - compare to similar system of known size
 - e.g., Fuzzy-logic method
- By decomposition
 - divide and conquer
 - e.g., Standard component method
 - By parametric models
 - mathematically model historical data
 - e.g., Proxy estimation

By guessing

- RHE



By expert opinion

- Background
 - Uses several estimators
 - “size estimation by consensus”
- Method: Wideband-Delphi
 - multiple estimators make independent estimate, submit to coordinator
 - coordinator calculates average estimate, indicates (anonymously) other estimates
 - repeat until estimates stabilize
 - size estimate = average estimate; variation = range of original estimates

Wideband-Delphi

- E.g.
 - Round 1
 - Estimates:
 - A - 13,800 LOC
 - B - 15,700 LOC
 - C - 21,000 LOC
 - Coordinator calculates average of 16,833 LOC and resubmits
 - Round 2
 - Estimators then meet and discuss the estimates. New estimates:
 - A - 18,500 LOC
 - B - 19,500 LOC
 - C - 20,000 LOC
 - Coordinator calculates average of 19,333 LOC and resubmits
 - Round 3
 - Estimators agree on Round 2 figures

Wideband-Delphi

- Agile equivalent: Planning Poker

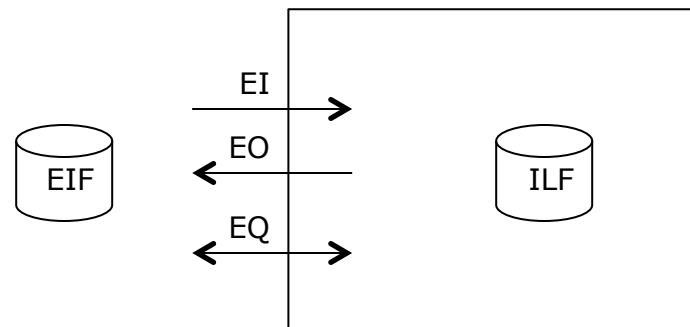


Wideband-Delphi

- Perspective
 - pros
 - can produce very accurate results
 - utilizes organization's skills
 - can work for any sized product
 - cons
 - relies on experts
 - is time consuming
 - is subject to common biases

By function

- Background
 - function point is an arbitrary unit of “functionality”
 - function point analysis seeks to assess a measure of functionality by examining interfaces to system
 - feature point analysis is fp + algorithmic analysis
- Method: function point estimation
 - determine project boundary
 - count files (data)
 - internal logical files
 - external interface files
 - count transactions
 - external inputs
 - external outputs
 - external inquiries
 - determine system complexity factor



Function Point Estimation

CHECKLIST FOR IDENTIFYING INTERNAL LOGICAL FILES (ILFs)				
Data Group:				
Definitions: An Internal Logical File (ILF) is a <i>user identifiable</i> group of logically related data or <i>control information</i> maintained within the boundary of the application. <i>User identifiable</i> refers to the specific user requirements that an experienced user would define. <i>Control information</i> is data used by the application to ensure compliance with business function requirements specified by the user. <i>Maintained</i> refers to the ability to modify data through an <i>elementary process</i> . An <i>elementary process</i> is the smallest unit of activity that is meaningful to the end-user in the business (e.g., add, change, delete).				
Rules:	Yes	No	N/A	Remarks
Is the group of data or control information a logical, or user identifiable, group of data that fulfils specific user requirements?				
Is the group of data maintained within the application boundary?				
Is the group of data modified, or maintained, through an elementary process of the application?				

Example ILFs:

- logical file of error messages
- logical file of audit information
- logical file of application data

Not ILFs:

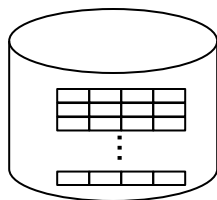
- temporary/working files
- data for normal backup and recovery

[it.toolbox.com]

Function Point Estimation

- Sample counting: Internal logical files
 - Quantifying functionality to service input file

Number of Record Element Types	Number of Data Element Types		
	1-19	20-50	51+
1	7fp	7fp	10fp
2 to 5	7fp	10fp	15fp
6+	10fp	15fp	15fp



1 record type
4 data element types
= 7fp

[it.toolbox.com]

Function Point Estimation

- E.g.,
 - New program has the following requirements
 - analyze marketing performance
 - project the likely sales
 - allocate sales to regions and time periods
 - produce monthly report
 - Raw function points estimate is 54

System Components		Complexity			Total
		Low	Average	High	
Inputs	EI	3	4	6	
Outputs	EO	4	5	7	
Inquiries	EQ	3	4	6	
Files	ILF	7	10	15	
Interfaces	EIF	5	7	10	

Function Point Estimation

- E.g., (con't)
 - function point estimate adjusted for environment is

$$\text{adjusted count} = \text{raw} \times (\text{system char} \times .01 + .65)$$

General System Characteristics			
Data communications	<u>2</u>	On-line update	0
Distributed functions	0	Complex processing	0
Performance	0	Reusability	<u>3</u>
Heavily-used configuration	0	Installation ease	0
Transaction rate	0	Operational ease	0
On-line data entry	<u>4</u>	Multiple sites	0
End user efficiency	0	Facilitate change	0
Total			<u>9</u>
Not present = 0		Average influence = 3	
Incident influence = 1		Significant influence = 4	
Moderate influence = 2		Strong influence = 5	

$$\begin{aligned}
 \text{adjusted count} &= \\
 &54 \times (9 \times .01 + .65) \\
 &= 54 \times .74 \\
 &= 39.96 \\
 &= 40 \text{ fp}
 \end{aligned}$$

[IFPUG 1990]

Function Point Estimation

- Perspective
 - pros
 - usable in the earliest requirements phases
 - independent (sorta) of programming language, product design, or development style
 - large body of historical data
 - well documented method
 - active users group
 - cons
 - automated counting not always possible
 - do not reflect language, design, or style differences
 - designed for estimating commercial data processing applications

By analogy

- Method: fuzzy-logic size estimation
 - Gather size data on previously developed programs
 - Subdivide data into size categories:
 - very large, large, medium, small, very small
 - establish size ranges
 - include all existing and expected products
 - Subdivide each range into subcategories
 - Allocate the available data to the categories
 - Establish subcategory size ranges
 - When estimating a new program, judge which category and subcategory it most closely resembles

Fuzzy-Logic Size Estimation

- E.g.
 - given 5 programs:
 - a file utility of 1,844 LOC
 - a file management program of 5,834 LOC
 - a personnel record keeping program of 6,845 LOC
 - a report generating package of 18,386 LOC
 - an inventory management program of 25,943 LOC
 - establish 5 size ranges VS, S, M, L, VL:

Range	Low	Midpoint	High	Representative System
Very Small	1,325	1,946	2,566	file utility
Small	2,566	3,768	4,970	no members
Medium	4,970	9,298	9,626	file mgmt; personnel records sys
Large	9,626	14,134	18,641	report generator
Very Large	18,641	27,373	36,104	inventory mgmt

Fuzzy-Logic Size Estimation

- E.g. (con't)
 - In comparing the new program to the historical data you observe
 - more complex application than file management or personnel programs
 - not as complex as the inventory management program
 - more function than the report package
 - You conclude that the new program is in the lower end of “very large,” or from 18 to 26 KLOC.

Fuzzy-Logic Size Estimation

- Perspective
 - pros
 - based on relevant historical data
 - easy to use
 - requires no special tools or training
 - provides reasonably good estimates where new work is similar to prior experience
 - cons
 - requires familiarity with historical data
 - not useful for new program types
 - not useful for programs much larger or smaller than historical data

By decomposition

- Background
 - estimates size based on principal units of decomposition
- Method: standard component sizing
 - identify common building blocks (subsystems, modules, screens, etc.)
 - associate sizes with building blocks
 - decompose new development into blocks and compare with history
- E.g.
 - assume a historical database of
 - data input component 1,108 LOC
 - output component 675 LOC
 - file component 1,585 LOC
 - control component 2,550 LOC
 - computation component 475 LOC

Standard Component Sizing

- E.g. (con't)

- estimate of maximum, minimum, and likely numbers of components of new product

• data input component	1	4	7
• output component	1	3	5
• file component	2	4	8
• control component	1	2	3
• computation component	1	3	7

- calculate likely code sizes based on history

• data input component	1108	4432	7756
• output component	675	2025	3375
• file component	3170	6340	12680
• control component	2550	5100	7650
• computation component	475	1425	3325

Standard Component Sizing

- E.g. (con't)
 - calculate the minimum, likely, and maximum LOC of the new product
 - minimum 7,978
 - likely 19,322
 - maximum 34,786
 - distribute the size*
 - $LOC = (7978 + 4 \cdot 19322 + 34786) / 6 = 20,009$ LOC
 - the standard deviation is $(34786 - 7978) / 6 = 4468$
 - the estimate range is $20,009 + 4468 = 15,500$ to $24,500$ LOC

*(low + 4*mid + high)/6 gives Gaussian distribution with a standard deviation of (high - low)/6

Standard Component Sizing

- Perspective
 - pros
 - based on relevant historical data
 - easy to use
 - requires no special tools or training
 - provides a rough estimate range
 - cons
 - decomposition must mirror end product to be accurate
 - must use large components early in a project

[Humphrey 1995]

By parametric model

- Background
 - use statistical model of past estimation efforts to predict size
- Method: Component-based estimation
 - Concept
 - estimators rarely think in detail at project beginning
 - use two-phase approach to estimating:
 - identify major components, then
 - determine size/duration based on components, then
 - adjust according to past performance
 - Component characteristics
 - should correlate closely to development costs
 - should be easy to visualize early in development
 - should be countable by automated means

[derived from Humphrey 1995]

Component-Based Estimation

- Method
 - begin
 - identify components from architecture
 - forecast effort based on historical components
 - adjust effort via stat model of past projects
 - end

Principles of Size Estimation

- General principles

- repeatability
 - separate estimation efforts must have same result \pm delta
- consistency
 - must be performed and used the same way across the organization
- empiricism
 - must have measure of estimation error*

What makes an estimation technique "good"?

*What is a reasonable bound on estimation error?

A measure of “goodness”:

- **V** is actual
- **V_i** is estimate for project at time i
- **n** is the number of estimating points
- the lower the value of ε , the better the estimation process
- high ε 's across projects indicate an estimation problem

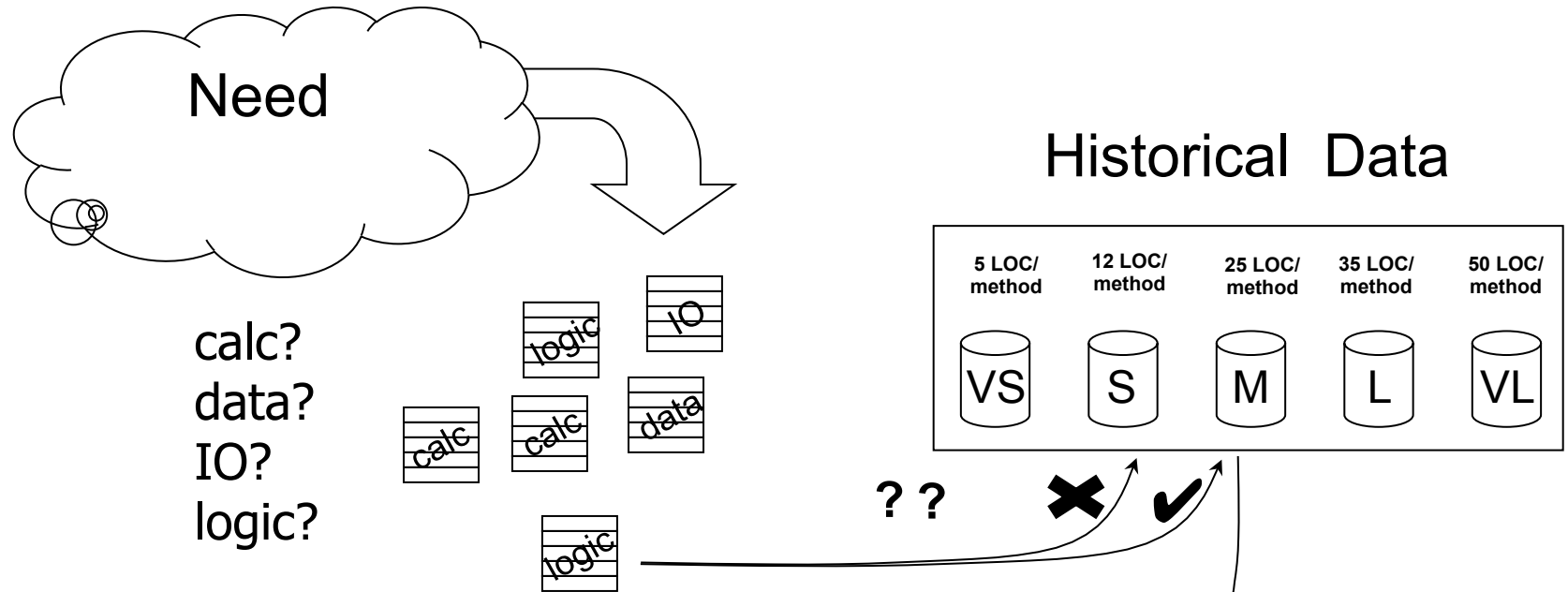
$$\varepsilon = \sqrt{\frac{1}{n} \sum_{i=1}^n \left(\frac{V_i - V}{V} \right)^2}$$

Let's Pick ...

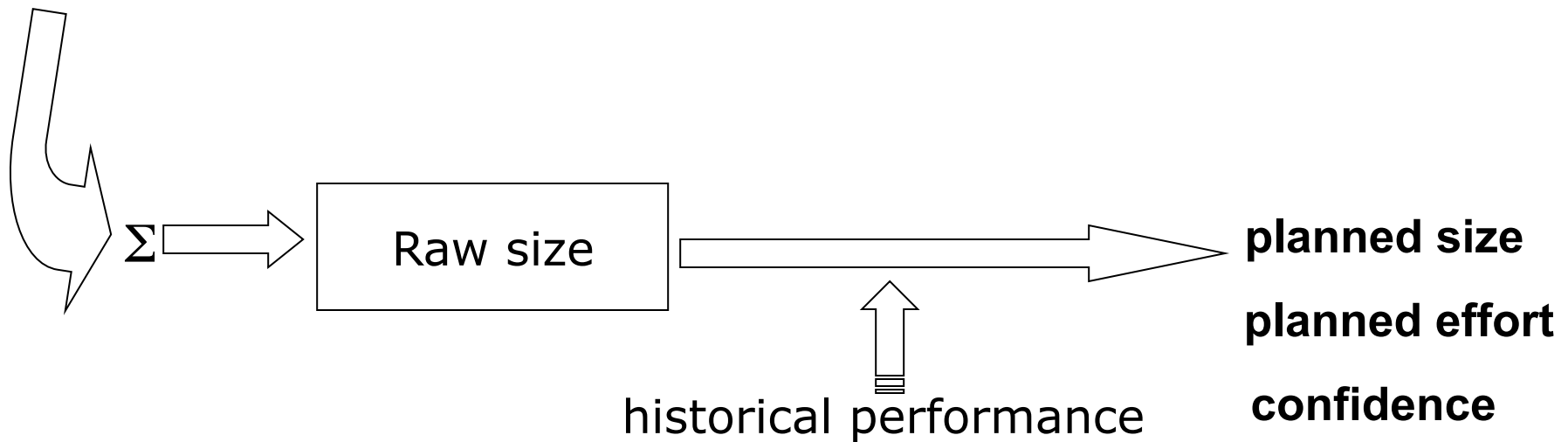
- By RHE
- By expert opinion
- By function
- By analogy
- By decomposition
- By component



... this one



Component size = relative size * methods

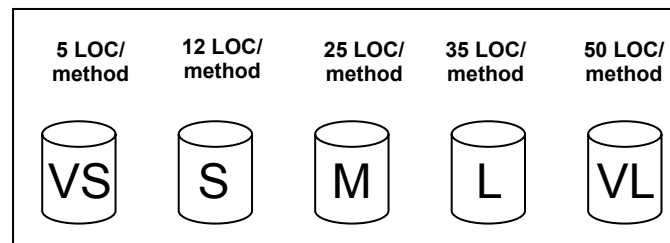


Component-Based Estimation Setup

- Need the following information to estimate size:
 - Data on historical objects
 - Actual lines of code (LOCa)
 - Data on historical projects:
 - Raw lines (LOCr)
 - Actual lines (LOCa)
 - Actual effort (Ea)

1. Putting stuff in buckets

Historical Data



Building relative size matrix

- Goal: to determine bucket sizes
- To calculate size ranges
 - determine ln of LOC per method for each component
 - calculate the mean and standard deviation
 - distribute sizes as*

Size	Lower	Midpoint	Upper
VS	1	$e^{\text{mean}-2*\text{std}}$	$e^{\text{mean}-1.5*\text{std}}$
S	$e^{\text{mean}-1.5*\text{std}}$	$e^{\text{mean}-\text{std}}$	$e^{\text{mean}-0.5*\text{std}}$
M	$e^{\text{mean}-0.5*\text{std}}$	e^{mean}	$e^{\text{mean}+0.5*\text{std}}$
L	$e^{\text{mean}+0.5*\text{std}}$	$e^{\text{mean}+\text{std}}$	$e^{\text{mean}+1.5*\text{std}}$
VL	$e^{\text{mean}+1.5*\text{std}}$	$e^{\text{mean}+2*\text{std}}$	big

* All values are rounded up to nearest integer

Size matrix (con't)

- Example

- given the following components

- 3 methods, 39 total LOC
 - 5 methods, 127 total LOC
 - 2 methods, 64 total LOC
 - 3 methods, 28 total LOC
 - 1 method, 23 LOC
 - 2 methods, 44 total LOC

- LOC per method =

13, 25.4, 32, 9.3, 23, 22

- ln of LOC per method =

2.56, 3.23, 3.46, 2.23,
3.13, 3.09;

mean = 2.954; std = 0.461

	Lower	Mid	Upper
VS	1	8	10
S	10	12	15
M	15	19	24
L	24	30	38
VL	38	48	big

$$e^{(\text{mean}-0.5*\text{std})}$$

$$=e^{(2.954-0.5*0.461)}$$

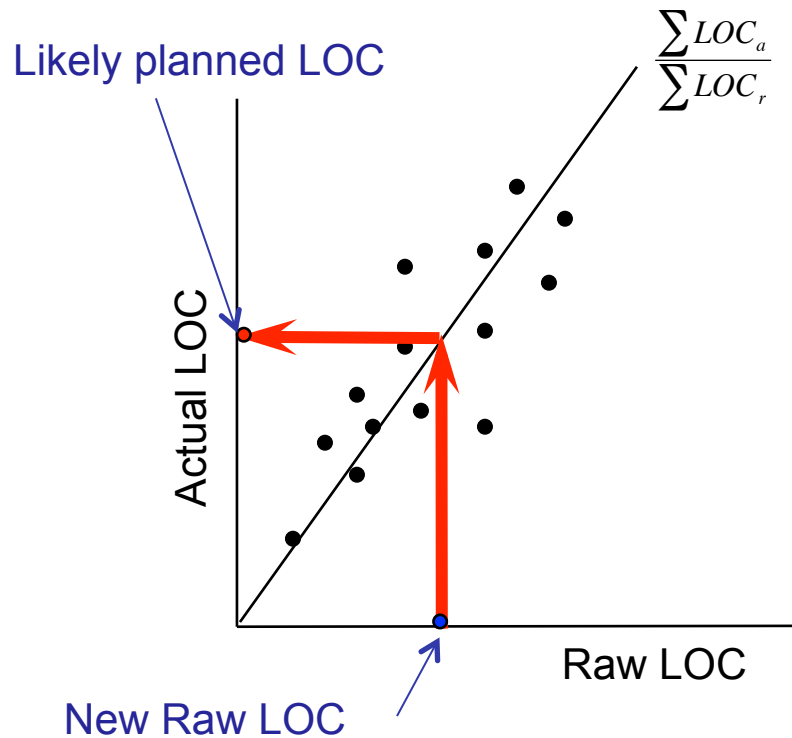
$$e^{(\text{mean})}=e^{2.954}$$

$$e^{(\text{mean}+0.5*\text{std})}$$

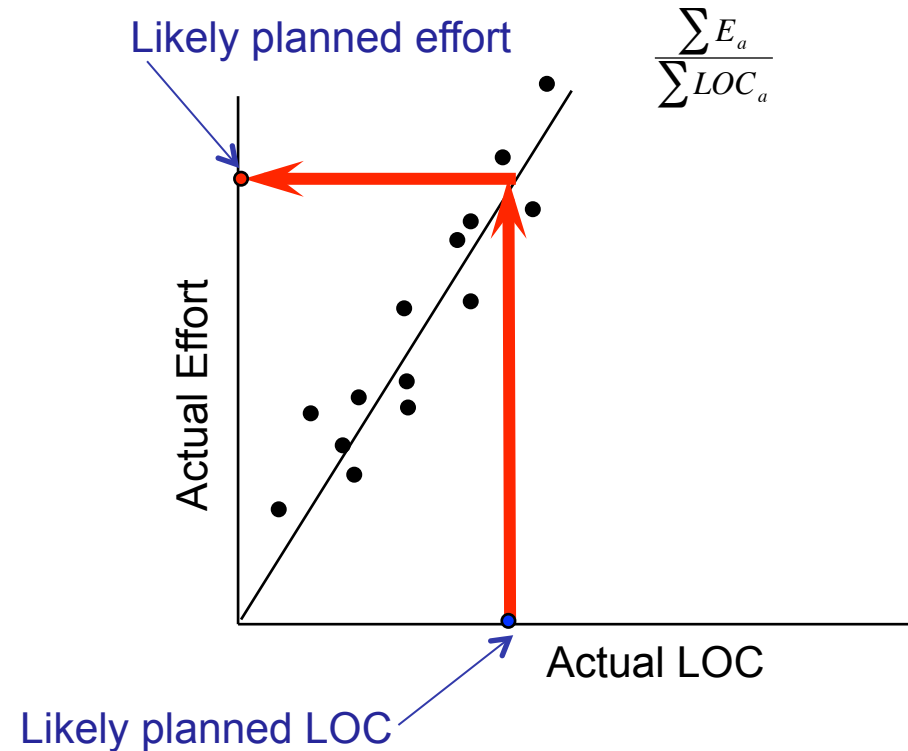
$$=e^{(2.954+0.5*0.461)}$$

2. Calculating historical trend

Estimation fundamentals



$$\left[NewLOC_r \times \frac{\sum LOC_a}{\sum LOC_r} \right] = NewLOC_p$$



$$\left[NewLOC_p \times \frac{\sum E_a}{\sum LOC_a} \right] = NewE_p$$

Historical trends

- Example
 - Given the historical database:

LOC _r	LOC _a	E _a
229	240	1100
382	255	1020
215	270	900
336	240	760
316	227	800
316	322	1200
133	120	260
433	412	1500
266	170	500
296	353	1700

$$\frac{\sum LOC_a}{\sum LOC_r} = .89$$

If the raw size is 342 LOC, then the planned size is $\text{ceil}(342 * .89) = 306$

$$\frac{\sum E_a}{\sum LOC_a} = 3.73$$

and the planned effort is $\text{ceil}(306 * 3.73) = 1142$

3. Calculating confidence

Confidence

- Prediction interval
 - ... provides a likely range around the estimate
 - Note: Humphrey suggests determining linear regression coefficients, then calculating statistical range:

$$Range = t(\alpha/2, n-2) \sqrt{\frac{1}{n-2} \sum_{i=1}^n (y_i - \beta_0 - \beta_1 x_i)^2} \sqrt{1 + \frac{1}{n} + \frac{(x_k - x_{avg})^2}{\sum_{i=1}^n (x_i - x_{avg})^2}}$$

- My opinion: this is statistical overkill
- We will take a different, although Humphrey-compatible approach

Confidence

- Prediction interval ... our approach
 - ... involves calculation of lower prediction interval (LPI) and upper prediction interval (UPI) ...
 - for time estimates
 - LPI = Estimate * fastest historical productivity
 - UPI = Estimate * slowest historical productivity
- Confidence on prediction interval
 - ... involves calculating correlation coefficient
 - $1.0 \geq r^2 \geq .75$ high
 - $.75 > r^2 \geq .50$ med
 - $.50 > r^2$ low

Confidence

- Size prediction interval

LOCr	LOCa	Ea
229	240	1100
382	255	1020
215	270	900
336	240	760
316	227	800
316	322	1200
133	120	260
433	412	1500
266	170	500
296	353	1700

240/229

LOCa/LOCr
1.05
0.67
1.26
0.71
0.72
1.02
0.90
0.95
0.64
1.19

LOCr=342
LOCp=306

UPI=431

worst
underestimate
 $342 \times 1.26 =$

LPI=219

worst
overestimate
 $342 \times 0.64 =$

Note:

set UPI to LOCp if UPI < LOCp
set LPI to 1 if LPI > LOCp
also set confidence to low

Confidence

- Correlation on predication interval
 - calculate correlation coefficient

LOCr	LOCa	Ea
229	240	1100
382	255	1020
215	270	900
336	240	760
316	227	800
316	322	1200
133	120	260
433	412	1500
266	170	500
296	353	1700

$$r(x,y)^2 = \left(\frac{n \sum_{i=1}^n x_i y_i - \sum_{i=1}^n x_i \sum_{i=1}^n y_i}{\sqrt{\left[n \sum_{i=1}^n x_i^2 - \left(\sum_{i=1}^n x_i \right)^2 \right] \left[n \sum_{i=1}^n y_i^2 - \left(\sum_{i=1}^n y_i \right)^2 \right]}} \right)^2$$

= 0.52

= medium confidence

$1.0 \geq r^2 \geq .75$ high
 $.75 > r^2 \geq .50$ med
 $.50 > r^2$ low

Confidence

- Time prediction interval - 1

LOCr=342
LOCp=306
LPI=219
UPI=431

LOCr	LOCa	Ea
229	240	1100
382	255	1020
215	270	900
336	240	760
316	227	800
316	322	1200
133	120	260
433	412	1500
266	170	500
296	353	1700

1100/240 →

Ea/LOC
4.58
4.00
3.33
3.17
3.52
3.73
2.17
3.64
2.94
4.82

Ep=1142
LPI=665

UPI=1475

Note:

set UPI to Ep if UPI < Ep
set LPI to 1 if LPI > Ep
also set confidence to low

Confidence

- Correlation on predication interval
 - calculate correlation coefficient

LOCr	LOCa	Ea
229	240	1100
382	255	1020
215	270	900
336	240	760
316	227	800
316	322	1200
133	120	260
433	412	1500
266	170	500
296	353	1700

$$r(x,y)^2 = \left(\frac{n \sum_{i=1}^n x_i y_i - \sum_{i=1}^n x_i \sum_{i=1}^n y_i}{\sqrt{\left[n \sum_{i=1}^n x_i^2 - \left(\sum_{i=1}^n x_i \right)^2 \right] \left[n \sum_{i=1}^n y_i^2 - \left(\sum_{i=1}^n y_i \right)^2 \right]}} \right)^2$$

= 0.85

= high confidence

$1.0 \geq r^2 \leq .75$ high
 $.75 > r^2 \leq .50$ med
 $.50 > r^2$ low

Confidence

- Time prediction interval -2
 - recall size estimate:
 - $LOC_p = 306$
 - $UPI = 431$
 - $LPI = 219$
 - calculate a "fallback" prediction
 - $Ep\ LLPI = LOC_p LPI * 2.17 = 475$
 - $Ep\ UUPI = LOC_p UPI * 4.82 = 2077$

Ea/LOC

4.58

4.00

3.33

3.17

3.52

3.73

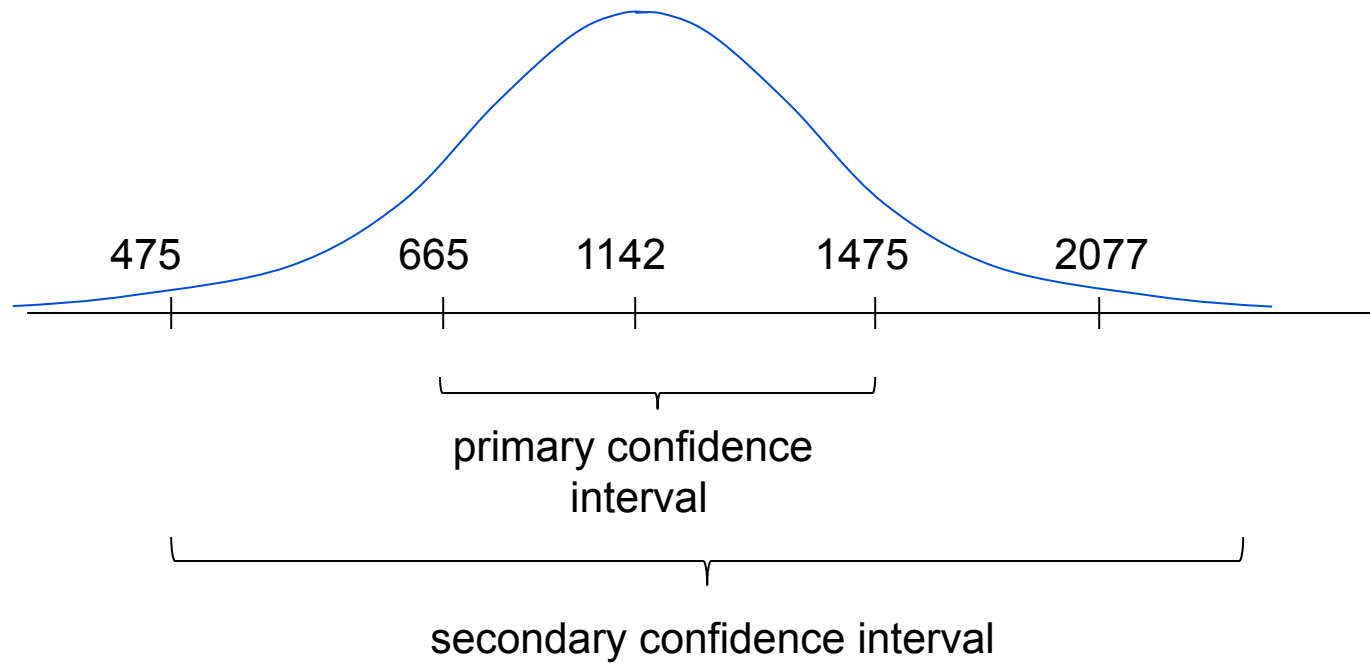
2.17

3.64

2.94

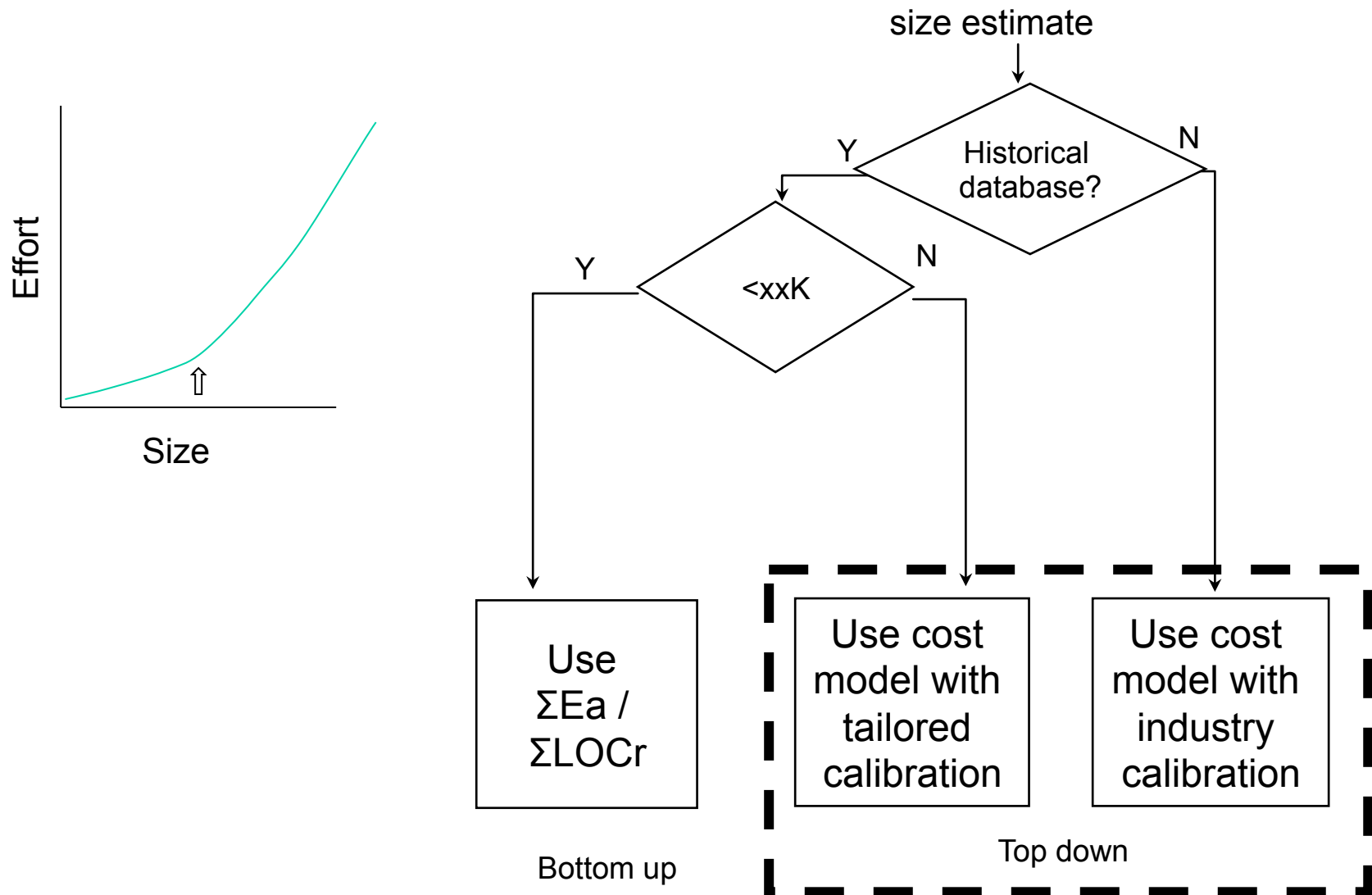
4.82

Result



4. Calculating E_p when $LOCr$ is very large

Project Duration



Cost Models

- Size/cost models

- history based

- static single-variable model

- $\text{Resource} = c_1 \times (\text{estimated characteristic})^{c_2}$

- » where resource is effort, project duration, staff size, etc.

- » c_1 and c_2 are constants derived from historical data

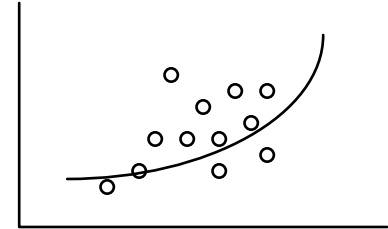
- » estimated characteristic describes production environment

- static multivariable model

- $\text{Resource} = c_1 e_1 + c_2 e_2 + \dots$

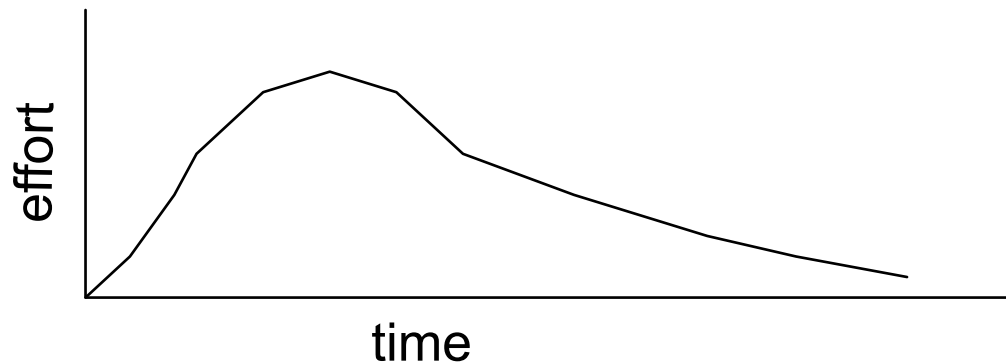
- » where e_i is the i th software characteristic

- » c_1, c_2 are empirically derived constants for the i th characteristic



Cost Models

- Size/cost models
 - theoretic based
 - projects resource requirements as a function of time (typically using Rayleigh-Norden curve)



Sample Model

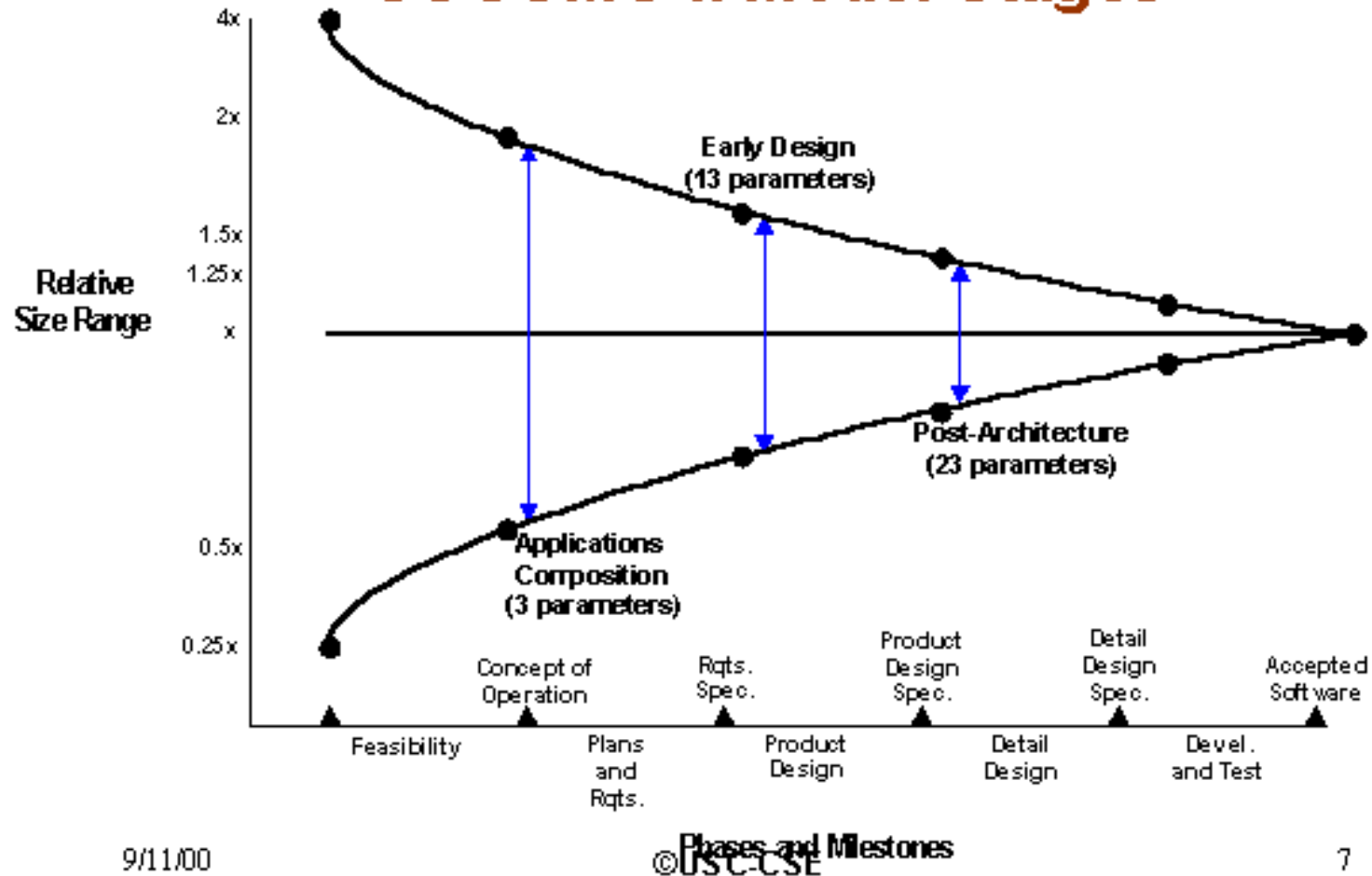
- COCOMO II
 - COConstructive COst Model
 - Widely known
 - Regression analysis of industry projects
 - History
 - Introduced in 1981
 - Analysis of TRW projects
 - Three models
 - Basic COCOMO: computes effort as a function of size
 - Intermediate COCOMO: computes effort as a function of size and cost drivers
 - Advanced COCOMO: intermediate COCOMO plus assessment of cost driver impact on lifecycle phase
 - Revised in 2000
 - referred to as COCOMO 2000 or COCOMO II (vs COCOMO 81)
 - Same basic models

Barry sez ...



University of Southern California
Center for Software Engineering

COCOMO II Model Stages



COCOMO II formula

- Schedule

- $PM = A \times \text{Size}^{(B + 0.01 \times \sum SF_i)} \times \prod EM_i$

- where

- PM = person months
 - Size = estimated size of system in KSLOC
 - SF = scale factors
 - EM = effort multipliers
 - A = calibration data = 2.94
 - B = calibration data = 0.91

COCOMO II formula

- Schedule ... basic model
 - $PM = A \times Size^{(B + 0.01 \times \sum SF_i)}$
 - where
 - PM = person months
 - Size = estimated size of system in KSLOC
 - SF = scale factors
 - A = calibration data = 2.94
 - B = calibration data = 0.91

Scale Factors

Scale Factor	Very Low	Low	Nom	High	V High	Extra High
Precedent	Surprise! 6.20	4.96	3.72	2.48	1.24	very familiar 0.00
Flexibility	rigorous 5.07	4.05	3.04	2.03	1.01	general goals 0.00
Risk resolution	little 7.07	5.65	4.24	2.83	1.41	full 0.00
Team Cohesion	difficult interaction 5.48	4.38	3.29	2.19	1.10	seamless interactions 0.00
Process Level	CMMI L1 7.80	6.24	4.68	3.12	1.56	CMMI L5 0.00

COCOMO II formula

- Schedule ... intermediate model

- $PM = A \times \text{Size}^{(B + 0.01 \times \sum SF_i)} \times \prod EM_i$

- where

- PM = person months
 - Size = estimated size of system in KSLOC
 - SF = scale factors
 - EM = effort multipliers
 - A = calibration data = 2.94
 - B = calibration data = 0.91

Effort Multipliers - Full Dev Model

- Product factors
 - Required software reliability
 - Database size
 - Product complexity
 - Reuse development
 - Documentation
- Platform factors
 - Execution time constraint
 - Storage constraint
 - Platform volatility
- Personnel factors
 - Analyst capability
 - Programming capability
 - Personnel continuity
 - Applications experience
 - Platform experience
 - Language/tool experience
- Project factors
 - Use of software tools
 - Multisite development
 - Required dev schedule

For example ...

Effort Multiplier	Very Low	Low	Nom	High	V High	Extra High
RELY	annoyance 0.82	0.92	1.00	1.10	1.26	human rated 1.26
ACAP	15th pctl 1.42	35th 1.19	55th 1.00	75th 0.85	90th 0.71	100th 0.71
PCAP	15th perc 1.34	35th 1.15	55th 1.00	75th 0.88	90th 0.76	100th 0.76
PCON	48%/yr 1.29	24% 1.12	12% 1.00	6% 0.90	3% 0.81	0% 0.81
APEX	≤2 mo 1.22	6 mo 1.10	1 yr 1.00	3 yrs 0.88	6 yrs 0.81	>6 yrs 0.81

Case Study -- Airborne Radar System

- Scale factors

- Precedentedness Nominal
- Flexibility: Very Low
- Risk Resolution: Very High
- Team Cohesion: Nominal
- Process Maturity: High

$$\begin{aligned} &\text{given KSLOC} = 94 \\ &\text{PM} = A \times \\ &\quad \text{Size}^{(B \times 0.01 \times \sum \text{SFi})} \times \\ &\quad \prod \text{EM}_i \\ &= 526.5 \text{ person months} \end{aligned}$$

- Effort Multipliers

- | | | | | | |
|--------|----|------|----|------|----|
| – RELY | VH | STOR | H | PLEX | VH |
| – DATA | N | PVOL | N | LTEX | N |
| – CPLX | VH | ACAP | H | TOOL | VH |
| – RUSE | VH | PCAP | H | SITE | N |
| – DOCU | N | PCON | H | SCED | VL |
| – TIME | VH | APEX | VH | | |

COCOMO II formula II

- Time
 - $TDEV = C \times PM^{(D + 0.2 \times 0.01 \times \sum SF_i)}$
 - where
 - TDEV = development calendar time
 - PM = person months
 - SF = Scale factor
 - C = calibration data = 3.67
 - D = calibration data = 0.28
- Airborne Radar System
 - TDEV = 17.5 months

Summary

Topics

- Rationale
- Principles of estimation
- Common techniques
 - Wideband-delphi
 - Function points
 - Fuzzy-logic
 - Standard component
 - Component
- Component-based estimation
 - process
 - size
 - effort
 - confidence
 - setup

Key Points

- “Size” is a key measurement of the overall software process
- There are a number of ways to estimate size, all of which are tenable EXCEPT guessing
- Component-based estimation is our approach this semester
 - entails sizing by component
 - adjusted by historical bias