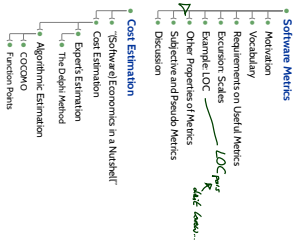


- Pseudo-metrics can be useful if there is a (good) correlation with low false positives and few false negatives) between values and the property to be measured.
-
- This may strongly depend on context information
 - If LOC was (or could be made non-subvertible (+> tutorials)), then **productivity** could be a useful measure for e.g. team performance.

Content



Recall: Pseudo-Metrics

- Some of the most interesting aspects of software development projects are **hard or impossible** to measure directly, e.g.:
- how **maintainable** is the software?
 - how much **effort** is needed until completion?
 - how is the **productivity** of my software people?
 - do all modules do **appropriate error handling**?
 - is the **documentation** sufficient and well usable?

Due to high relevance, people want to measure despite the difficulty in measuring. Two main approaches:

	expert review	code walkthrough	code walk	code walk	code walk	code walk	code walk
differentiated	✓	✓	✓	✓	✓	✓	✓
reproducible	✓	✓	✓	✓	✓	✓	✓
available	✓	✓	✓	✓	✓	✓	✓
reliable	✓	✓	✓	✓	✓	✓	✓
economical	✓	✓	✓	✓	✓	✓	✓
practicable	✓	✓	✓	✓	✓	✓	✓
robust	✓	✓	✓	✓	✓	✓	✓

- Note: not every derived measure is a pseudo-metric:
- average LOC per module: derived, **no pseudo** → we really measure average LOC per module.
 - measure maintainability in average LOC per module: derived, **pseudo** → we don't really measure maintainability, average LOC is only interpreted as maintainability. Not robust if easily subvertible (see exercises).

Can Pseudo-Metrics be Useful?

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Code Metrics for OO Programs (Childenber and Kanerey, 1994)

metric	computation
weighted method per class (WMC)	$\sum_{i=1}^n w_i$, w_i = number of methods, w_i = complexity of method i
depth of inheritance tree (multiple inheritance ?)	graph distance in inheritance tree (multiple inheritance ?)
number of children (NOC)	number of direct subclasses of the class
coupling between object classes (CBO)	$CBO(C) = K_C \cup K_I $, K_C, K_I = set of classes using, C = response for a class
lack of cohesion in methods (LCOM)	$L_{COM} = \max(P - Q , 0)$, P = methods using no common attribute, Q = methods using at least one common attribute

- direct metrics: DIT, NOC, CBO, pseudo-metrics: WMC, RFC, LCOM
- these seem to be agreement that it is more important to focus on empirical evidence for reliability of the proposed metrics than to propose new ones... (Frey, 2003)

Subjective Metrics

	example	problems	countermeasures
Statement	The specification is ambiguous	Terms may be ambiguous	Allow only certain statements, characterise them precisely
Assessment	The module is good in a certain way	Conditions are hard to measure. Not necessarily comparable	Only one particular criterion on an at least ordinal scale
Grading	Theability is graded + C	Subjective, grading not reproducible	Define criteria for grades, give examples how to grade, practice on existing artefacts

(Ludwig and Lehner, 2013)

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Example: A (Subjective) Metric for Maintainability

- Goal: assess maintainability.
 - One approach: grade the following aspects, e.g. with scale $S = \{0, \dots, 10\}$.
 - Item Conformance**
 - m_1 : size of units (modules etc)
 - m_2 : labelling
 - m_3 : naming of identifiers
 - m_4 : design layout
 - m_5 : style of comments
 - Locality**
 - l_1 : use of parameters
 - l_2 : information hiding
 - l_3 : design layout
 - l_4 : design of interfaces
 - l_5 : structure of control flow
 - l_6 : comments
 - Testability**
 - t_1 : test drive
 - t_2 : test data
 - t_3 : test cases for test evaluation
 - t_4 : diagnostic components
 - t_5 : dynamic consistency checks
 - Typing**
 - p_1 : data types
 - p_2 : structure of control flow
 - p_3 : comments
 - p_4 : type reduction
 - Define $m = \frac{m_1 + \dots + m_5}{5}$ (with weights: $m_j = \frac{m_1 + \dots + m_5}{5}$, $G = \sum_{i=1}^{20} g_i$)
 - Procedure:
 - Train reviewers on existing examples.
 - Do not over-interpret results of first applications.
 - Evaluate and adjust before putting to use, adjust regularly
- (Ludwig and Lehner, 2013)

8/10

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8/10

The Goal-Question-Metric Approach

- Information Overload?
- How we have mentioned nearly all attributes one could measure...
Which ones should we measure?
It depends...
- feasible
 - measurable
 - available
 - differentiated
 - economical
 - comparable
 - reproducible
 - robust
- One approach: Goal-Question-Metric (GQM)

9/10

Goal-Question-Metric (Basili and Weiss, 1984)

- The three steps of GQM
- Define the goals relevant for a project or an organisation
 - From each goal, derive questions which need to be answered to check whether the goal is reached
 - For each question, choose (or develop) metrics which contribute to finding answers.

10/10

- Being good wrt. to a certain metric is (in general) not an asset on its own.
We usually want to optimise wrt. goals, not wrt. metrics.
In particular critical pseudo-metrics for quality.
- Software and process measurements may yield **personal data** (Personenbezogene Daten).
Their collection may be regulated by laws.

11/10

And Which Metrics Should One Use?

Often useful: collect some basic measures in advance (in particular if collection is cheap / automatic), e.g.:

- size ...
- of newly created and changed code, etc. (automatically provided by revision control software)
- effort ...
 - ... for coding, review, testing, verification, etc.
- errors ...
 - ... at least errors found during quality assurance, and errors reported by customer

... at least errors found during quality assurance, and errors reported by customer (can be recorded via standardised revision control messages)

Measures derived from such basic measures may indicate problems ahead early enough and buy time to take appropriate counter-measures. E.g. track

- error rate per release, error density (errors per LOC)
- average effort for error detection and correction,
- etc.

over time. In case of unusual values: investigate further (maybe using additional metrics).



12.0

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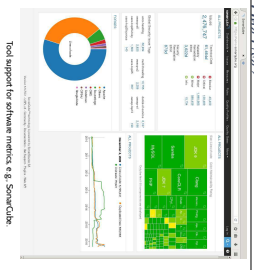
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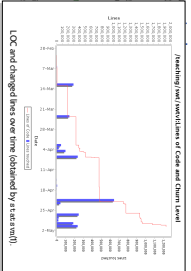
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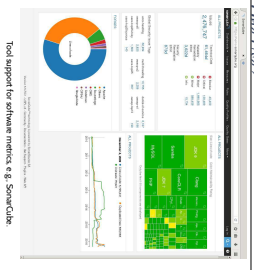
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12.0

Tell Them What You've Told Them...

- Software metrics are defined in terms of scales
- Use software metrics to
 - assess quality (metric, support decisions)
 - prescribe / describe (diagnose / propose)
- Whether a software metric is useful depends...
 - Not every software attribute is directly measurable
 - derived measures,
 - subjective metrics, and
 - pseudo metrics...
- ... have to be used with care – do we measure what we want to measure?
 - Metric examples:
 - LOC (code / systematic complexity)
 - more than 50 more metrics named
 - Goal-Question-Metric approach:
 - it's about the goal, not the metrics
 - Communicating figures: consider percentiles.

13.0

Topic Area Project Management: Content

- VL2 Software Metrics
 - Properties of Metrics
 - Scales
 - Examples
- VL3 Cost Estimation
 - Software Economics in a Nutshell
 - Expert Estimation
 - Algorithmic Estimation
- VL4 Project Management
 - Project
 - Process and Process Modelling
 - Process Models
 - Process Models
 - Process Metrics
 - CPWI, SPI
- VL5
- ...

14.0

“(Software) Economics in a Nutshell”

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- Distinguish current cost (laufende Kosten), e.g.
- wages
 - (business) management, marketing
 - rooms
 - computers, networks, software as part of infrastructure
 - ...
- and project-related cost (projektbezogene Kosten), e.g.
- additional temporary personnel
 - contract costs
 - expenses
 - hardware and software as part of product or system
 - ...
- Handwritten notes: *business infrastructure* (bracketed around the first list), *project related hardware* (bracketed around the second list).

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Costs

Next to 'Software', 'Costs' is one of the terms occurring most often in this book." Ludwig and Lohrer (2013)

A first approximation:

cost (Kosten) benefit (Nutzen) (see negative costs)	all disadvantages of a solution all benefits of a solution
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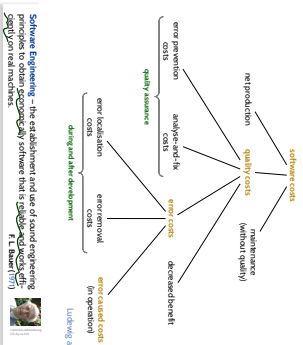
Note: costs / benefits can be subjective – and not necessarily quantifiable in terms of money.–

Super-ordinate goal of many projects:

- Minimize overall costs, i.e. maximise difference between benefits and costs. (Equivalent: minimize sum of positive and negative costs.)

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Software Costs in a Narrower Sense



19/10

Costs vs. Benefits: A Closer Look

The benefit of a software is determined by the advantages achievable using the software. It is influenced by:

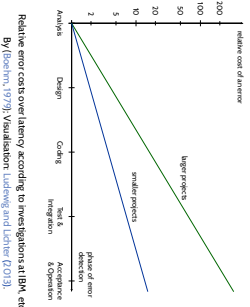
- the degree of concordance between product and requirements
- additional services, support, flexibility, etc.

Some other examples of cost/benefit pairs. (inspired by Jones (1990))

Costs	Possible Benefits
Labor during development (e.g. later savings)	Use of seat (e.g. later savings)
New equipment (purchase, maintenance, depreciation)	Better equipment (purchase, maintenance, depreciation)
New software purchase (Other uses of new software)	Improved system (Other uses of new software)
Increased data gathering	Increased control
Training for employees	Increased productivity

17/10

Discovering Fundamental Errors Late Can Be Expensive



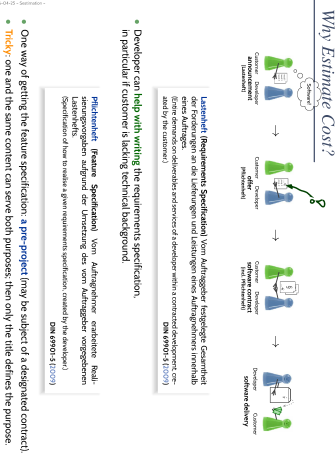
20/10

Cost Estimation



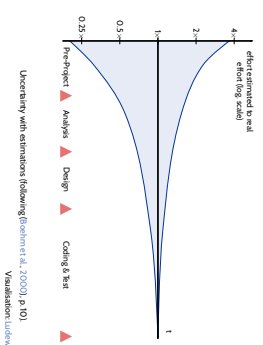
24.0

Why Estimate Cost?



22.0

The "Estimation Funnel"



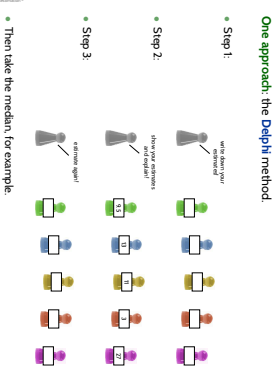
23.0

Plan

Expert's Estimation

25.0

Expert's Estimation

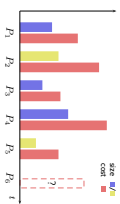


26.0

Algorithmic Estimation

27/10

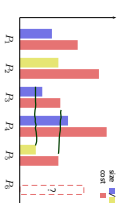
- Approach, more general:**
- Identify (measurable) factors F_1, \dots, F_n , which influence overall cost. Use size in LOC
 - Try to come up with a sample of data from previous projects
 - Try to come up with a formula $f(S, F_1, \dots, F_n)$ matches previous costs
 - Estimate values for F_1, \dots, F_n for a new project
 - Take $f(S, F_1, \dots, F_n)$ as cost estimate C for new project
 - Conduct new project, measure F_1, \dots, F_n , and cost C
 - Adjust f if $C \neq C'$
- Note:**
- The need for (expert) estimation does not go away, one needs to estimate F_1, \dots, F_n
 - Rationale: It is often easier to estimate technical aspect than to directly estimate cost



Algorithmic Estimation: Principle

27/10

- Assume:**
- Projects P_1, \dots, P_n took place in the past
 - Size S_i , costs C_i , and index k_i ($0 = \text{blue} = 1 = \text{yellow}$) have been measured and recorded
- Question:** What is the cost of the new project P_7 ?
- Approach:**
- Try to find a function f such that $f(S_i, k_i) = C_i$, for $1 \leq i \leq 6$
 - Estimate size S_7 and find k_7 **estimate**
 - Estimate C_7 as $C_7 = f(S_7, k_7)$
- (In the artificial example above, $f(S, k) = 8 \cdot 1.8 \cdot k + 0.3 \cdot \text{index}$, let if P_7 is index yellow $k_7 = 1$ and size estimate $S_7 = 27$ then $f(S_7, k_7) = 5.10$)



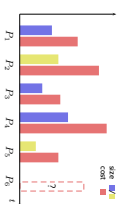
28/10

- Constructive Cost Model:**
- Formula which fits a large set of archived project data (from the 70s)
 - Factors:
 - COCOMO 81 (Boehm, 1988) basic, intermediate, detailed
 - COCOMO II (Boehm et al., 2000)
 - All based on estimated program size S measured in DSI or KDSI (thousands of Delivered Source Instructions)
 - Factors like security requirements or experience of the project team are mapped to values for parameters of the formulae
- COCOMO examples:**
- textbooks like Lathrop and Lathrop (2013) [probably outdated]
 - an exceptionally large example: COCOMO 81 for the Linux kernel (Wheeler, 2009) (and follow-up)

Algorithmic Estimation: Principle

28/10

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Algorithmic Estimation: Principle

28/10

- Algorithmic Estimation: COCOMO**
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COCOMO 81

30/10

Size	Characterized by Type of Development	Dvs. Environment	a	b	Schedule Project Type	
Small (500 LOC)	Little	Stable	3.2	1.05	Organic	
Medium (1000 LOC)	Medium	Medium	3.0	1.1	Semi-detached	
Large	Greater	Highly	Complex HW/Intelligence	2.8	1.20	Embedded

- Basic COCOMO:**
- effort required: $E = a \cdot (S/DSI)^b$ [PM (person-months)]
 - time to develop: $T = c \cdot E^d$ [months]
 - headcount: $H = E/T$ [FTE (full-time employees)]
 - productivity: $P = S/E$ [DSI per PM] (← use to check for feasibility)
- Intermediate COCOMO**
- $$E = M \cdot a \cdot (S/DSI)^b$$
- [person-months]
- $$M = RELY \cdot CPLX \cdot TIME \cdot ACAP \cdot FCAP \cdot LEXP \cdot TOOL \cdot SCED$$

$$M = RELY \cdot CPLX \cdot TIME \cdot ACAP \cdot PCAP \cdot LEXP \cdot TOOL \cdot SCEP$$

factor	very low	low	normal	high	very high	extra high
RELY - required software reliability	0.75	0.98	1	1.15	1.40	1.65
CPLX - product complexity	0.70	0.95	1	1.11	1.30	1.64
TIME - execution time constraint	1.46	1.19	1	0.86	0.71	
ACAP - analyst capability	1.42	1.17	1	0.86	0.7	
PCAP - programmer capability	1.4	1.07	1	0.95		
LEXP - programming language experience						
TOOL - use of software tools	1.24	1.10	1	0.95	0.83	
SCEP - required developer experience	1.23	1.08	1	1.04	1.10	

* Note: while e.g. "extra high" TIME means, may depend on project context (Consider data from previous projects)

Consists of

- Application Composition Model - project work is configuring components, rather than programming
- Early Design Model - adaptation of Function Point approach (in a minute); does not need completed architecture design
- Post-Architecture Model - improvement of COCOMO II; needs completed architecture design, and size of components estimable

$$E = 2.94 \cdot S^k \cdot M$$

- Program size: $S = (1 + RELVD) \cdot (S_{new} + S_{prev})$
- requirements volatility: RELVD
- Equivalent requirements: 10% of code made then RELVD = 0.1
- S_{new} - estimated size minus size of re-use code
- S_{prev} = W/P weighting code takes; P=times the effort of re-use

Scaling factors:

$$X = \delta + \alpha \cdot \omega = 0.911, \delta = \frac{PMAT}{PHEC + FLEX + RESL + TEAM + PMAT}$$

factor	very low	low	normal	high	very high	extra high
PHEC - project effort (conformance with small project)	6.20	4.96	3.72	2.48	1.24	0.00
FLEX - development flexibility (development process stability)	5.07	4.05	3.04	2.03	1.01	0.00
RESL - Architecture/requirements management, architecture size (controlled)	7.07	5.65	4.24	2.83	1.41	0.00
TEAM - team cohesion (communication)	5.48	4.38	3.29	2.19	1.10	0.00
PMAT - project maturity (low CMM)	7.80	6.24	4.69	3.12	1.56	0.00

$$M = RELY \cdot DATA \cdot \dots \cdot SCEP$$

group	factor	description
Product factors	RELY	required software reliability
	DATA	size of database
	CPLX	complexity of system
	DOCU	amount of required documentation
Algorithm factors	TIME	execution time constraint
	STOR	memory consumption cost limit
	TOOL	stability of development environment
	ACAP	analyst capability
Team factors	PCAP	programmer capability
	LEXP	programming language experience
	RESL	experience with application domain
	TEAM	experience with programming language and tools
Project factors	TOOL	use of software tools
	STOR	memory consumption cost limit
	TIME	execution time constraint

(Adapted from COCOMO II, version COCOMO II.3)

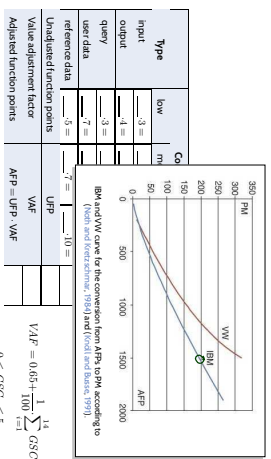
Function Points

classify data points

Type	low	medium	high	Sum
input	3 =	4 =	6 =	
output	4 =	5 =	7 =	
query	3 =	4 =	6 =	
user data (reference data)	7 =	10 =	15 =	
Undispatched function points (reference data)	5 =	7 =	10 =	
Value adjustment factor	AFP = UFP · VAF			

Adjusted function points

$$VAF = 0.65 + \frac{1}{100} \sum_{i=1}^{14} GS C_i, \quad 0 \leq GS C_i \leq 5$$



- Ludewig and Lehter (2013) says:
- Function Point approach used in practice in particular for commercial software business software)
 - COCOMO (Cost Model) overestimate in this domain: needs to be adjusted by corresponding factors.

In the end, it's **experience, experience, experience**.

"Estimate, document, estimate better." (Ludewig and Lehter, 2013)

Suggestions start to replicate your experience **more**.

- Take notes on your projects (e.g. Softwarepraktikum, Bachelor Projekt, Master Bachelor's Thesis, Master Projekt, Masters Thesis, ...)
- Examples: size of program created, number of errors found, number of pages written, ...
- Try to identify factors which hinder productivity, which boost productivity, ...
- Which factors and mistakes were **undidable** in hindsight? How?

- For software costs, we can distinguish:
 - net production,
 - quality costs,
 - maintenance.
- Software engineering is about being **economic** in all three aspects.
- Why estimate?
 - Requirements specification (User-Info)
 - Feature specification (Pflichtenheft)
- The latter (but is budget) is usually part of **software contracts**.

- Approaches:
- Expert's Estimation
 - Algorithmic Estimation
 - COCOMO
 - function Point
- estimate cost **indirectly** by estimating more **technical aspects**
- In the end, it's **experience**.

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