

Topic Area Project Management: Content

- VL2
 - Software Metrics
 - ↳ Properties of Metrics
 - ↳ Scales
 - ↳ Examples
- VL3
 - Cost Estimation
 - ↳ "Software Economics in a Nutshell"
 - ↳ Experts Estimation
 - ↳ Algorithmic Estimation
- VL4
 - Project Management
 - ↳ Project
 - ↳ Process and Process Modelling
 - ↳ Procedure Models
 - ↳ Process Models
- VL5
 - Process Metrics
 - ↳ CMMI Space
 - ...

Content

- Software Metrics
- ↳ Subjective Metrics
- ↳ Goal-Question-Metric Approach
- Cost Estimation
 - ↳ "Software Economics in a Nutshell"
 - ↳ Cost Estimation
 - ↳ Experts Estimation
 - ↳ The Delphi Method
 - ↳ Algorithmic Estimation
 - ↳ COCOMO
 - ↳ Function Points

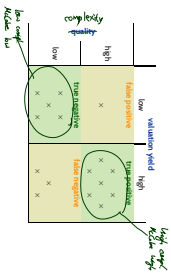
Kinds of Metrics: by Measurement Procedure

	objective metric	pseudo metric	subjective metric
Procedure	measurement counting possibly standardized	computation based on measurements or models	review by inspectors, verbal or by graphical
Advantages	exact, reproducible, can be obtained automatically	yield independent, directly usable statements on not characteristic characteristics	not subjective, plausible results, complex characteristics
Disadvantages	not always relevant, often subjective, body height as measure	hard to comprehend, pseudo-objective	assessment costly, quality of results depends on inspector
Example general	body height as measure	body mass index (BMI), weather forecast for the next day	health condition, weight for condition ("bad weather")
Example in Software Engineering	LOC or KCS, number of known bugs	LOC estimation by COCOMO	quality assessment, error weighting
Usually used for	collection of simple base measures	prediction/forecast estimations/ cost overall assessments	quality assessment, error weighting

(Ludewig and Lichten, 2019)

Recall: Can Pseudo-Metrics be Useful?

- Pseudo-metrics can be useful if there is a (good) correlation (with few false positives and few false negatives) between valuation yields and the property to be measured



- This may strongly depend on context information
- If LOC was (or could be made) non-subvertible (→ tutorial), then LOC/day could be a useful measure for e.g. project progress.

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example	problems	countermeasures
Statement The specification is available	Terms may be conditions are hardly possible	Allow only certain terms precisely
Assessment The modules implemented in a "clever way?"	Not necessarily comparable.	Only offer particular outcomes; put them on an (at least ordinal) scale.
Grading "Readability/s" graded 4.0"	Subjective; grading not reproducible.	Define criteria for grades; give examples how to grade; practice on existing artefacts

(Ludewig and Lohrer, 2013)

7/4

The Goal-Question-Metric Approach

8/4

Information Overload!

Now we have mentioned nearly 60 attributes one could measure...
Which ones should we measure?
It depends...



One approach: Goal-Question-Metric (GQM).

10/4

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

- The three steps of GQM:
- (i) Define the goals relevant for a project or an organisation.
 - (ii) From each goal, derive questions which need to be answered to check whether the goal is reached
 - (iii) For each question, choose (or develop) metrics which contribute to finding answers.
- Bring good wirt to a certain metric s (in general not an asset on its own. We usually want to optimise wirt **goals**, not wirt **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.

11/4

Example: A Metric for Maintainability

- Goal: assess maintainability
 - One approach: grade the following aspects, e.g. with scale $S = \{1, \dots, 10\}$. (Some aspects may be objective, some subjective (conductor review))
 - Norm Conformance
 - n1: stated user/provide(e)r etc.)
 - n2: labelling
 - n3: naming of identifiers
 - n4: design layout
 - n5: separation of blocks
 - n6: style of comments
 - Locality
 - l1: used parameters
 - l2: information hiding
 - l3: local flow of control
 - l4: design of interfaces
 - Readability
 - r1: data types
 - r2: comments
 - Testability
 - t1: test data
 - t2: test data preparation for test evaluation
 - t3: test data design
 - t4: dynamic consistency checks
 - Typing
 - ty1: type differentiation
 - ty2: type restriction
- Define: $m = \frac{m_1 + \dots + m_n}{n}$ (with weights: $m_j = \frac{m_{j1} + \dots + m_{jn}}{n_j}$, $G = \sum_{i=1}^{20} g_i$)
- Procedure:
 • Train reviewers on existing examples.
 • Do not over-interpret results of first applications.
 • Evaluate and adjust before putting to use, adjust regularly.
- (Ludewig and Lohrer, 2013)

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- Development of a pseudo-metric:
 (i) Identify aspect to be represented
 (ii) Devise a model for the metric
 (iii) Fix a scale for the metric
 (iv) Develop a definition of the pseudo-metric
 (v) Test how to compute the metric
 (vi) Test how to measure for all parameters of the definition.
 (vii) Apply and improve the metric.
- Define: $m = \frac{m_1 + \dots + m_n}{n}$ (with weights: $m_j = \frac{m_{j1} + \dots + m_{jn}}{n_j}$, $G = \sum_{i=1}^{20} g_i$)
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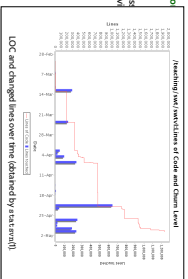
And Which Metrics Should One Use?

- 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, fixing, maintenance, etc.
- errors...
 - ... at least errors found during quality assurance, and errors reported by customer (can be recorded via standardized revision control messages)

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Measures derived from such basic measures may **highlight problems ahead early enough** and buy time to take appropriate countermeasures. 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).

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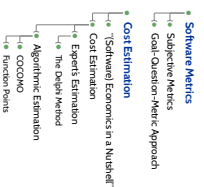
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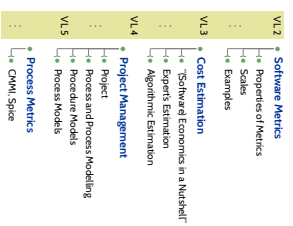
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Content

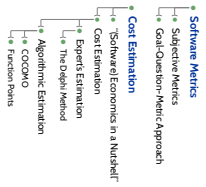


14.0

Topic Area Project Management: Content



15.0



16.0

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, comfort, flexibility, etc.

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

Costs	Possible Benefits
Labour during development (e.g. developer new test (e.g. later testing))	Use of fault (e.g. later testing)
New equipment (purchase, maintenance, depreciation)	Better equipment (maintenance: maybe expense from selling old)
New software purchases	Optimal use of new software
Conversion from old platform	Improvement of system, enhanced user interfaces
Received data gathering	Extended control
Training for employees	Increased productivity

19.0

"(Software) Economics in a Nutshell"



17.0

Costs: Economics in a Nutshell

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Costs

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

A first approximation:

cost (Kosten)	all disadvantages of a solution
Benefit (Nutzen) (or: positive costs)	all benefits of a solution

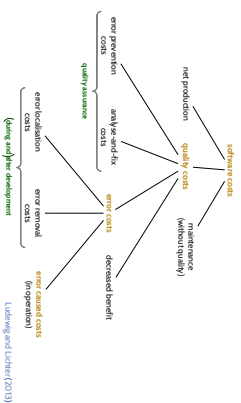
Note: costs / benefits can be subjective – and not necessarily quantifiable in terms of money...

Super-ordinate goal of many projects:

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

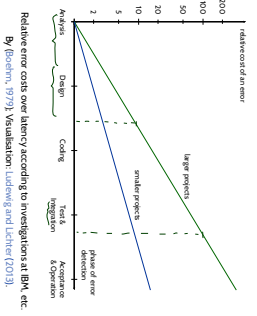
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Software Costs in a Narrower Sense



Software Engineering – the establishment and use of sound engineering practices (software engineering) is reliable and works effectively in real situations. (Lässig and Lehner (2013))

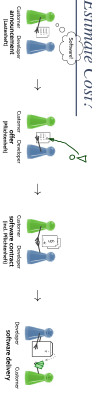
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Relative error costs over time according to investigations at IBM, etc. By Boehm, 1979; Vasilevskan, Ludewig and Lichte (2013)

22.10

Why Estimate Cost?



Lesson! Requirements Specification from Auftraggeber (customer) Gesamtheit der Funktionen an die Lieferanten und Leistungen eines Auftragnehmers (contractor) determine the success or failure of a project. The success or failure of a project is determined by the customer.

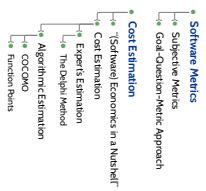
Developer can help with writing the requirements specification. In particular if customer is lacking technical background.

Prüfung! Feature Specifications from Auftraggeber (customer) Realisation of the requirements specification by the developer. The success or failure of a project is determined by the customer.

One way of getting the feature specification a **pre-project** may be subject of a designated contract. **Tip!** one and the same content can serve both purposes, then only the title defines the purpose.

25.10

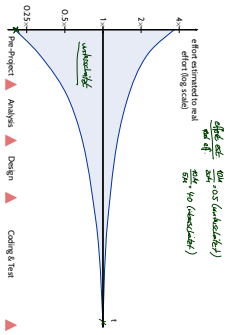
Cost Estimation



23.10

24.10

The "Estimation Funnel"



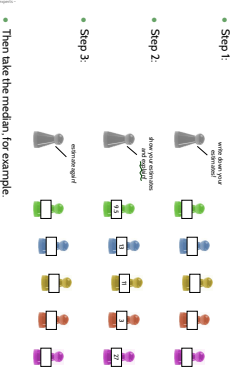
Uncertainty with estimations following (Boehm et al., 2000, p. 10) Vasilevskan, Ludewig and Lichte (2013)

26.10

Expert's Estimation

27.10

One approach: the Delphi method

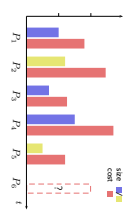


Then take the median, for example.

28.0

Algorithmic Estimation

29.0



Assume
 • Projects P_1, \dots, P_6 took place in the past.
 • Size S_i , cost C_i , and kind k_i ($0 = \text{blue}$ -ish, $1 = \text{yellow}$ -ish) have been measured and recorded.

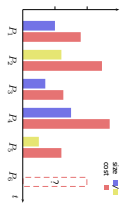
Question: What is the cost of the new project P_7 ?

Approach:
 (i) Try to find a function f such that $f(S_i, k_i) = C_i^*$ for $1 \leq i \leq 6$.
 (ii) Estimate size S_7 and kind k_7 .
 (iii) Estimate cost C_7 as $C_7 = f(S_7, k_7)$.

(In the artificial example above, $f(S, k) = S - 1.8 \pm k - 0.3$ would work, i.e. if P_6 is of kind yellow (thus $k_6 = 1$) and size estimate is $S_6 = 2.7$ then estimate C_6 as $f(S_6, k_6) = 5.16$.)

30.0

Algorithmic Estimation: Principle



Approach, more general:
 (i) Identify (measurable) factors F_1, \dots, F_n , which influence overall cost, like size in LOC
 (ii) Take a big sample of data from previous projects
 (iii) Try to come up with a formula f such that $f(F_1, \dots, F_n)$ matches previous costs
 (iv) Estimate values for F_1, \dots, F_n for a new project
 (v) Take $f(F_1, \dots, F_n)$ as cost estimate C for the new project.
 (vi) Conduct new project: measure C_1, \dots, C_n and cost C .
 (vii) Adapt f if C also identifies C .

Note:
 • The need for (expert's) estimation does not go away: one needs to estimate F_1, \dots, F_n .
 • Rationale: it is often easier to estimate technical aspects than to estimate cost directly.

30.0

Algorithmic Estimation: COCOMO

31.0

Algorithmic Estimation: COCOMO

- **Constructive Cost Model:**
 Formulate which fit a huge set of archived project data (from the late 70s)
- Flavours:
 • COCOMO 81 (Boehm, 1981) variants: basic, intermediate, detailed
 • COCOMO II (Fishman et al., 2000)
- All flavours are based on estimated program size S (measured in DSI (Delivered Source Instructions) or KDSI (1000 DSI))
- Factors like security requirements or experience of the project team are mapped to values for parameters of the formulae
- COCOMO examples:
 • textbooks like *Software and VLSI* (2013) (most probably made up)
 • an exceptionally vague example
 • COCOMO 81 for the Linux kernel (Wunderer, 2006) (and follow-ups)

32.0

Characteristic of the type		Dev.	Q	B	Software Project Type	
Size	Innovation	Complex	3.2	105	Open	
Small	Lite	Not tight	Scale			
100 LOC	Medium	Medium	3.0	112	Semi-detached	
1000 LOC	Large	Tight	Complex/HW	2.8	130	Embedded

- Basic COCOMO:**
- effort required: $E = a \cdot (S/HS)^b$ [PM (person-month)]
 - time to develop: $T = c \cdot E^d$ (months)
 - manpower: $H = E/T$ [FTE (full time employee)]
 - productivity: $P = S/E$ [DSI per PM] (← use to check for plausibility)

Intermediate COCOMO:

$$E = M_1 \cdot a_1 \cdot (S/HS)^{b_1} \cdot [person-months]$$

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

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

factor	very low	low	normal	high	very high	extra high
RELY	required software reliability	0.75	0.88	1	1.15	1.40
CPLX	product complexity	0.70	0.85	1	1.10	1.30
TIME	evolution time constant					1.66
ACAP	analytical capability	1.66	1.19	1	0.86	0.71
PCAP	programming capability	1.45	1.17	1	0.86	0.72
LEXP	programming language experience	1.14	1.07	1	0.95	
TOOL	use of software tools	1.24	1.10	1	0.91	0.83
SCEP	required development schedule	1.23	1.08	1	1.04	1.10

- Note what 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 – addition of Function Point approach for a minute; does not need completed architecture design
 - Post-Architecture Model – improvement of COCOMO 81 needs completed architecture design, and size of components estimatable

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

- Program size: $S = (1 + REVD) \cdot (S_{code} + S_{test})$
- requirements volatility: REVD, e.g. if new requirements make 10% of code unusable then REVD = 0.1
- S_{code} = estimated size minus size of re-used code.
- S_{test} = w/r , if writing new code takes r -times the effort of re-use.

Scaling factors:

$$X = \beta + \omega, \quad \omega = 0.01, \quad \beta = \frac{1}{100} \cdot (PRECC + ELLEX + RESL + TEAM + PMAT)$$

factor	very low	low	normal	high	very high	extra high
PRECC	predefined code with development process fixed	4.20	4.96	3.72	2.48	1.24
ELLEX	extensive experience	1.07	4.03	3.04	2.03	1.01
RESL	residual code and developer experience	1.07	1.55	4.24	2.83	1.41
TEAM	team cohesion/communication effort (in team)	1.48	4.38	3.29	2.19	1.10
PMAT	programmer/developer	2.89	3.24	4.91	3.17	1.54

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

group	factor	description
Product factors	RELY	required software reliability
	DATA	size of database
	CPLX	complexity of system
	ACAP	analytical capability of modules/components
System factors	PCAP	programming language experience
	LEXP	programming language experience
	TOOL	memory consumption constant
	PRECC	stability of development environment
Team factors	ELLEX	analyst capability
	RESL	continuity of involved personnel
	TEAM	experience with development environment
	PMAT	experience with programming language(s) and tools
Project factors	WTD	use of software tools
	SCEP	required development schedule

(also in COCOMO 81, now in COCOMO II)

Algorithmic Estimation: Function Points

Type	Complexity			Sum
	low	medium	high	
input	3	4	6	1
output	4	5	7	
query	3	4	6	
user data	7	10	15	
reference data	6	7	10	1
Unadjusted function points	UPF			
Value adjustment factor	VAF			1
Adjusted function points	AFP = UPF · VAF			

$$VAF = 0.65 + \frac{1}{100} \sum_{i=1}^{14} GSCF_i$$

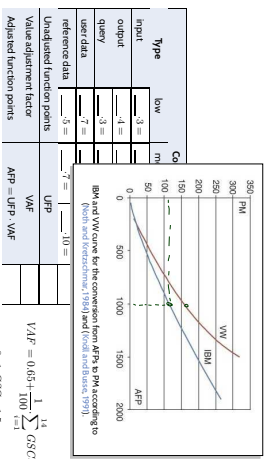
$$0 \leq GSCF_i \leq 5$$

39/40

Tell Them What You've Told Them...

- **Goal-Question-Metric approach:**
 - Define goals, derive questions, choose metrics
 - Evaluate and adjust
- Recall: It's about the goal, not the metrics.
- 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 (Lastenheft)
 - Feature specification (Pflichtenheft)
 - The latter (plus budget) is usually part of software contracts.
- Approaches
 - Expert's Estimation
 - Algorithmic Estimation (COCOMO, Function Points)
- → estimate cost indirectly, by estimating more technical aspects
- In the end, it's **experience** (and **experience** (and **experience**))

40/40



$$VAF = 0.65 + \frac{1}{100} \sum_{i=1}^{14} GSCF_i$$

$$0 \leq GSCF_i \leq 5$$

39/40

- Ludewig and Lütcher (2013) says:
- **Function Point approach** used in practice, in particular for **commercial software** (business software)
 - **COCOMO** tends to be **overestimating** in this domain, needs to be adjusted by **correcting factors**.

In the end, it's **experience, experience, experience**:
 "Estimate, document, estimate better" (Ludewig and Lütcher, 2013)

Suggestion: start to **explicitate your experience now**.

- **Take notes on your projects:**
 - e.g., Softwarepakulum, Bachelor Projekt, Master Bachelor's Thesis, Master Project, Master's Thesis, ...
 - Inrestimps, size of program created, number of errors found, number of pages written, ...
- **Try to identify factors** what hindered productivity, what boosted productivity, ...
- **Which detours and mistakes** were avoidable in hindsight? How?

40/40

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42/40

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43/40