Architecting System Performance all slides

by Gerrit Muller
TNO-ESI

Abstract

Architecting System Performance applies and elaborates the course Architectural Reasoning Using Conceptual Modeling to architect performance of systems. We teach an architecting method based on many views and fast iteration of the views. Visual models, functional models, and mathematical models in all views are the means to communicate about the system, to discuss specification and design choices, to reason about consequences, and to make decisions.

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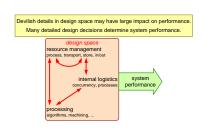
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Architecting System Performance; Course Overview

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Abstract

Course overview of the course Architecting System Performance.

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Nuggets Architecting System Performance

| 1. Course introduction |
|---------------------------------|
| 2. Managing system performance |
| 3. Course didactics |
| 4. Connecting breadth and depth |
| 5. Performance Modeling |
| 6. Level of Abstraction |
| 7. Visualizing Dynamic Behavior |

| 8. Emerging Behaviour |
|---------------------------------|
| 9. Budgeting |
| 10. Modeling Paradigms |
| 11. Applications and Variations |
| 12. Model Analysis |
| 13. Reasoning Approach |
| 14. Defining Performance |

time-oriented performance

| 15. Measuring Performance |
|--------------------------------|
| 16. Resource Management |
| 17. Greedy and Lazy Pattern |
| 18.Scheduling |
| 19. Robust Performance |
| 20. Bloating, Waste, and Value |



Assignments in Face-to-Face Module

| 0. elevator case | | | | | | |
|--|----------------------------|--|-----------------|---|--|--|
| supersystem | system | | subsystem | | | |
| 1. sketch the problem goal use case | key performance parameters | | main oncepts | critical technologies | | |
| 3. explore customer and business | situa • m • m si • qu | 2. make conceptual model of the current situation model dynamic behavior model 0-order kpp using functions (as simple as possible) quantify contribution to kpp using observed data | | | | |
| develop story model workflow and performar model customer value as functions | nce | | | | | |
| | solut • m | ions odel the | foreseen sol | el of potential lution ernative solutions | | |
| 5. list questions and uncertainties, reformulate problem and goal, and formulate gaps and options | | | | | | |
| 6. develop an elevator pitch to report you findings and recommendations to management | | | | | | |



Architecting System Performance; Course Material

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Abstract

Listing the course material for Architecting System Performance

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July 3, 2023 status: planned version: 0.1 logo TBD The ASPTM course is partially derived from the EXARCH course developed at *Philips CTT* by *Ton Kostelijk* and *Gerrit Muller*.

Extensions and additional slides have been developed at *ESI* by *Teun Hendriks*, *Roland Mathijssen* and *Gerrit Muller*.



Elevator: Hands-on Intro to Performance Modeling

core

Physical Models of an Elevator

http://www.gaudisite.nl/info/ElevatorPhysicalModel.info.html

optional

Teaching conceptual modeling at multiple system levels using multiple views http://www.gaudisite.nl/CIRP2014_Muller_TeachingConceptualModeling.pdf Understanding the human factor by making understandable visualizations http://www.gaudisite.nl/info/UnderstandingHumanFactorVisualizations.info.html



Course Didactics

core

Architecting System Performance; Course Didactics

http://www.gaudisite.nl/info/ASPcourseDidactics.info.html

optional

DSRP: https://en.wikipedia.org/wiki/DSRP

Assumptions: "Systems Engineering and Critical Reflection: The Application of Brookfield and Goffman to the Common Experiences of Systems Engineers" by Chucks Madhav; proceedings of INCOSE 2016, in Edinburgh, GB

70/20/10:

http://charles-jennings.blogspot.nl/

https://www.trainingindustry.com/wiki/entries/the-702010-model-for-learning-anddevelopment.aspx

http://jarche.com/2015/11/the-bridge-from-education-to-experience/

Reflection: "The Reflective Practitioner: How Professionals Think In Action" by Donald Schon, ISBN-10: 0465068782, Basic Books USA

Assumptions and beliefs:

https://pivotalthinking.wordpress.com/tag/ladder-of-inference/

http://stwj.systemswiki.org/?p=1120



Greedy and Lazy Patterns

core

Architecting System Performance; Greedy and Lazy Patterns http://gaudisite.nl/info/ASPgreedyAndLazy.info.html

optional

Fundamentals of Technology

http://gaudisite.nl/MAfundamentalsOfTechnologyPaper.pdf



Measuring

core

Architecting System Performance; Measuring

http://www.gaudisite.nl/info/ASPmeasuring.info.html

optional

Performance Method Fundamentals

http://www.gaudisite.nl/PerformanceMethodFundamentalsPaper.pdf

Measurement issues; From gathering numbers to gathering knowledge by Ton

Kostelijk http://www.gaudisite.nl/MeasurementExecArchSlides.pdf

Modeling and Analysis: Measuring

http://www.gaudisite.nl/MAmeasuringPaper.pdf

Exploring an existing code base: measurements and instrumentation

http://www.gaudisite.nl/info/ExploringByMeasuringInstrumenting.info.html



Architecting System Performance; Managing System Performance

by Gerrit Muller TNO-ESI, University of South-Eastern Norway]

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Abstract

This presentation presents the ideas behind the course Architecting System Performance. A number of frameworks and mental models show the context of this course and the approach to performance advocated in this course.

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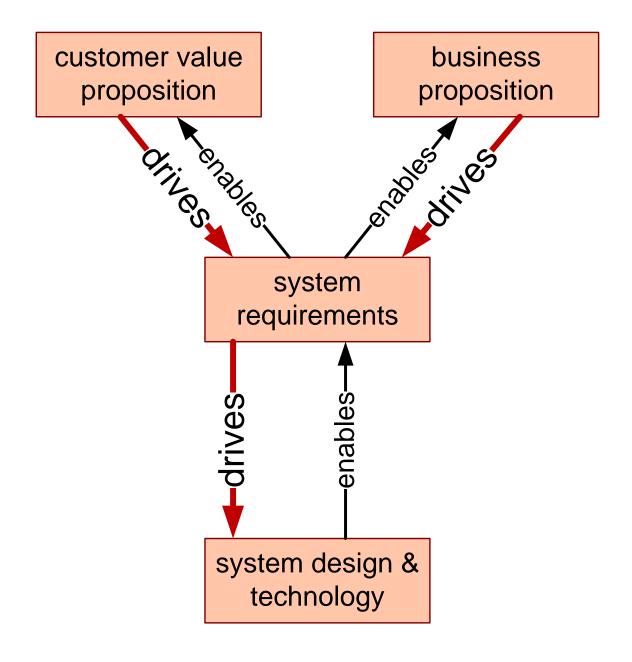
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Architecture Top View





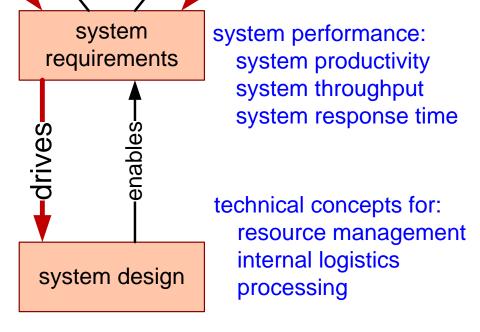
Performance Playing Field

consumer experience
enterprise performance:
enterprise productivity
enterprise throughput
enterprise response time

customer value proposition business proposition competitiveness service response time service cost

Performance attributes
require means for analysis
evaluation, and creation of
structure (parts and interfaces)
and dynamic behavior (functions)
at all levels.

Hence, we need conceptual modeling at all levels.

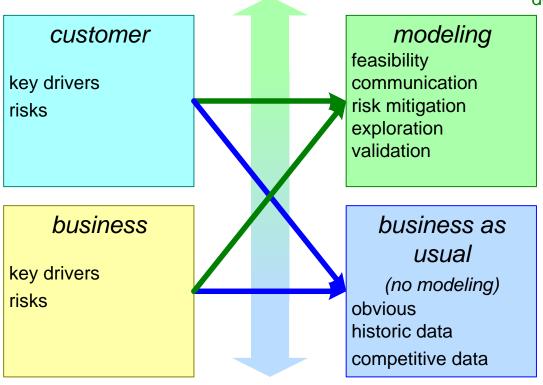


What and Why to Model

how well is the customer served?
how credible becomes the solution?
how much are time and effort reduced?
how much is the risk reduced?
how much is the solution improved?

purpose and type of model depend on project life cycle

type of model and views depend on purpose



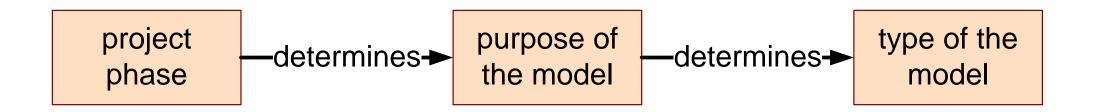
how much effort is needed to create model(s)?
how much effort is needed to use and maintain model(s)?
how much time is needed to obtain useful result?

decision factors:
accuracy of model
credibility of results
level of abstraction
working range
calibration of model
robustness of model
time to first results and feedback
effort
evolvability
(adaptation to new questions)



Modeling Evolves over Time

understanding exploration optimization verification



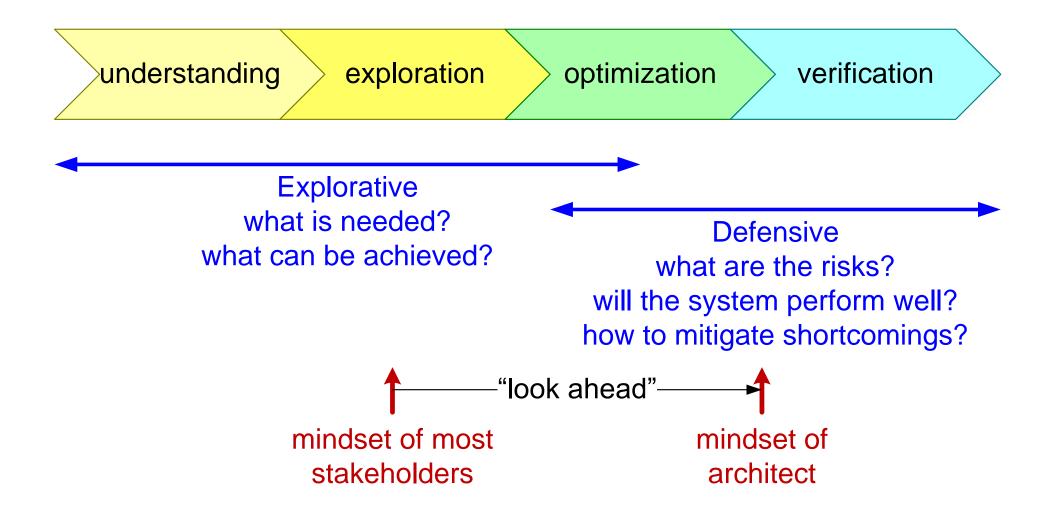
The Modeler's Mindset Evolves too

Understanding exploration optimization verification

Explorative what is needed? Defensive what can be achieved? what are the risks? will the system perform well? how to mitigate shortcomings?

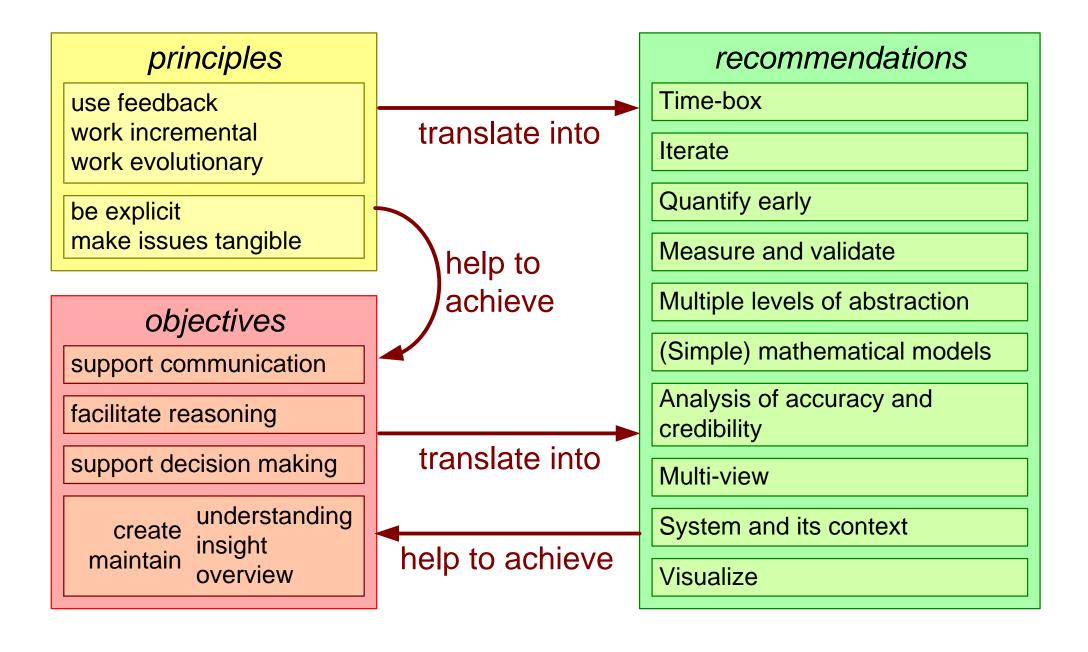


The Architect Can Be "Out of Phase"



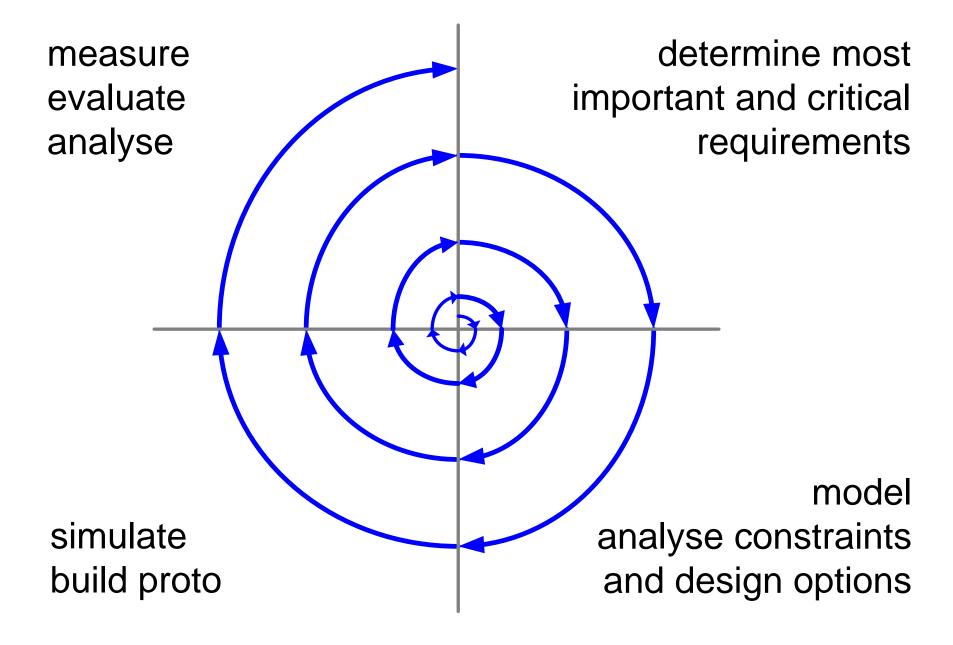


10 Fundamental Recommendations



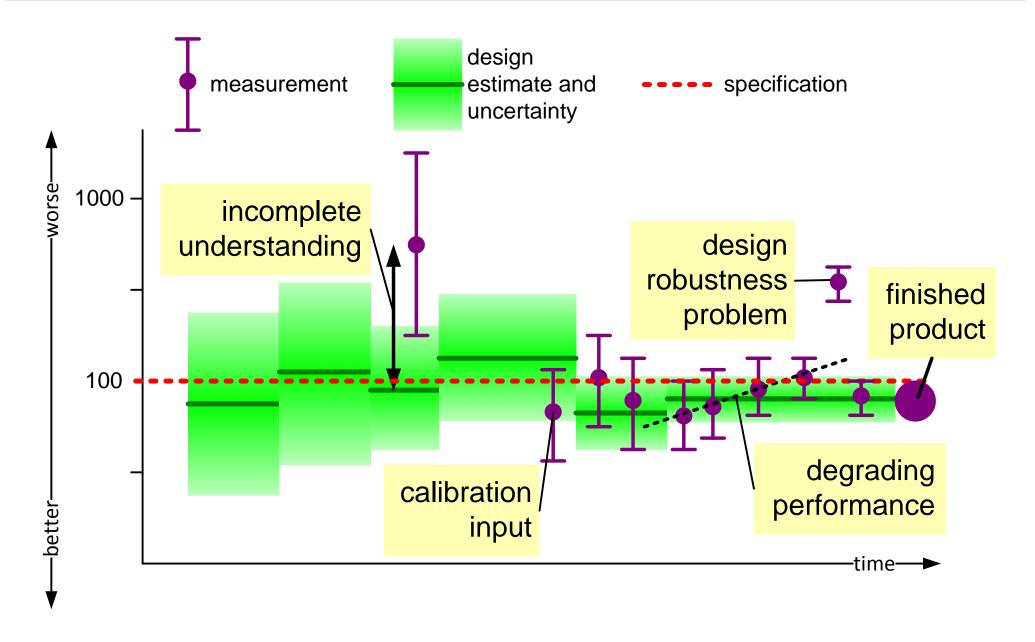


Iterative Performance Management during Development



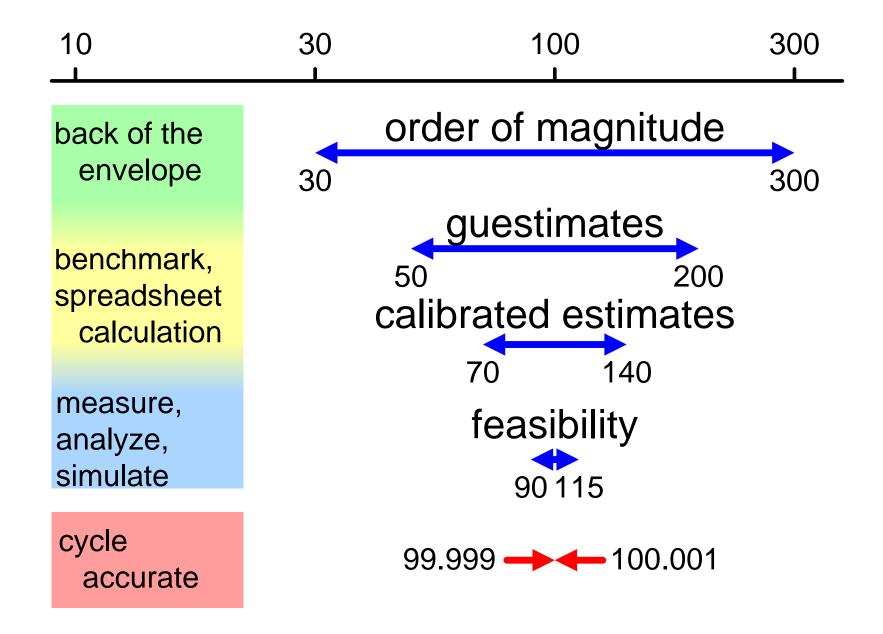


Managing Performance during Product Development





Quantification Steps





Architecting System Performance; Course Didactics

by Gerrit Muller TNO-ESI, University of South-Eastern Norway]

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Abstract

The didactics behind a course like Architecting System Performance is a challenge, because the learning goals relate mostly to attitude and ways of thinking. At the same time, the material covers methods, techniques, tools, and concepts, which may lure participants in mechanistic approaches. Core in the didactic approach is reflection. This presentation offers some "thinking models" to assist reflection.

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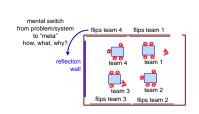
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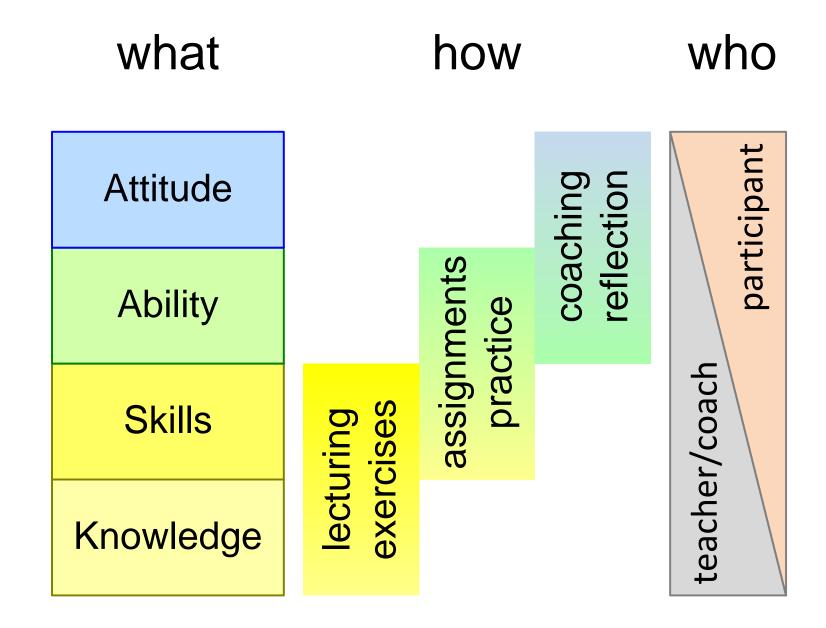
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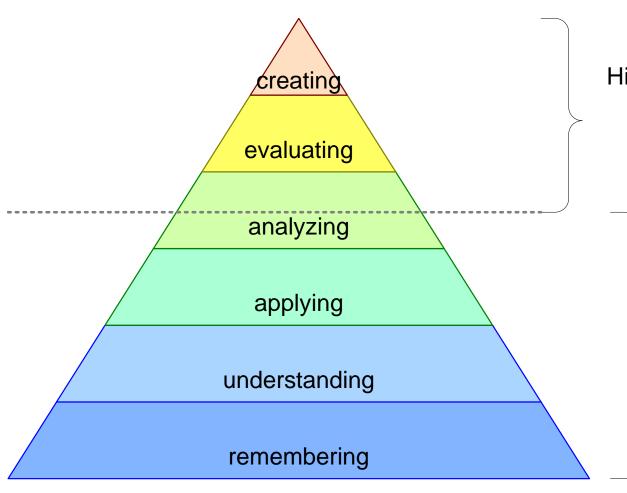


Competence Requires Various Learning Styles





Bloom's Taxonomy and Higher Order Thinking Skills



Higher Order Thinking Skills more difficult to teach more valuable takes time to develop

> must be mastered before, however when missing can be acquired fast

Lower Order Thinking Skills people can acquire them fast



Course Assumption:

This course focuses on Higher Order Thinking Skills.

We assume

that you have appropriate knowledge

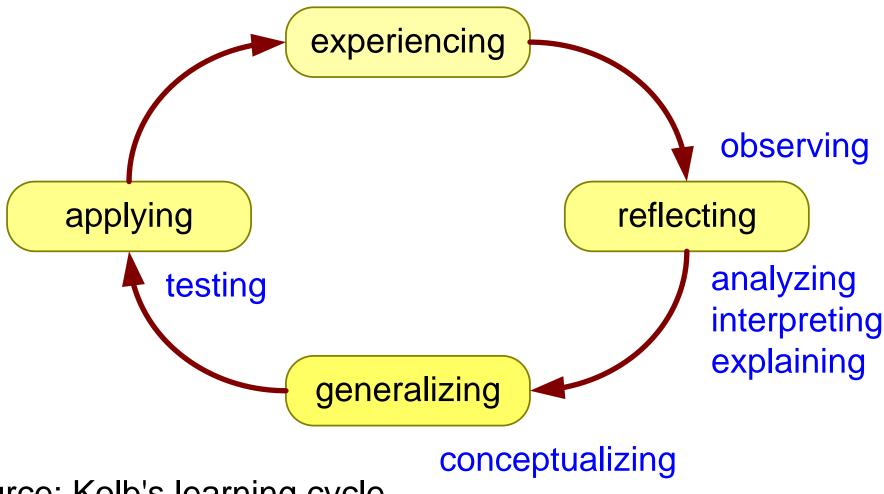
and

that you are able to find and absorb

required specific knowledge fast.



Problem-Based Learning Using Reflection



source: Kolb's learning cycle

http://www.infed.org/biblio/b-explrn.htm



Role of Experience in Learning

70:20:10 learning model

70: Experience

20: Exposure

10: Education

Modeling

Coaching

Scaffolding

Articulation

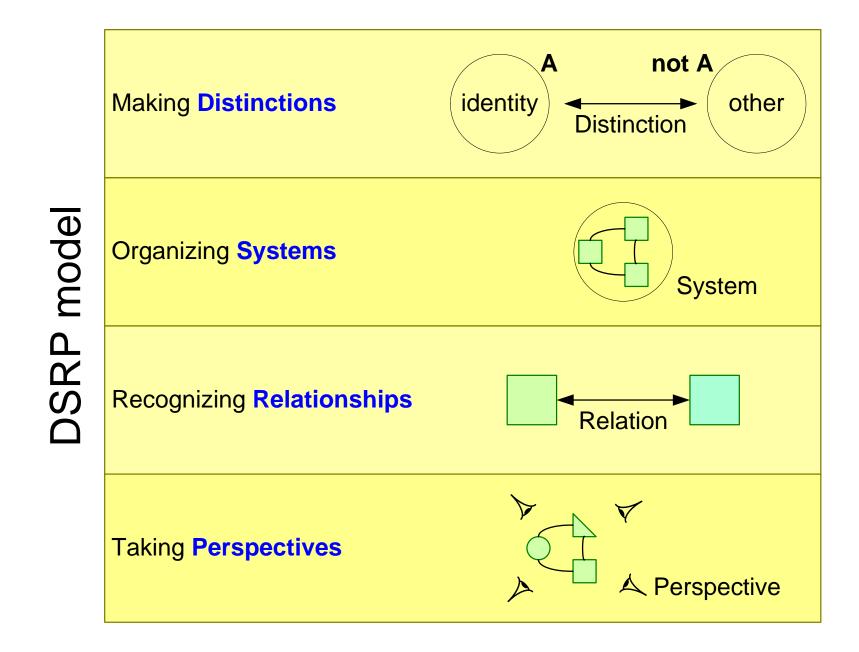
Reflection

Exploration

https://en.wikipedia.org/wiki/ Cognitive_apprenticeship

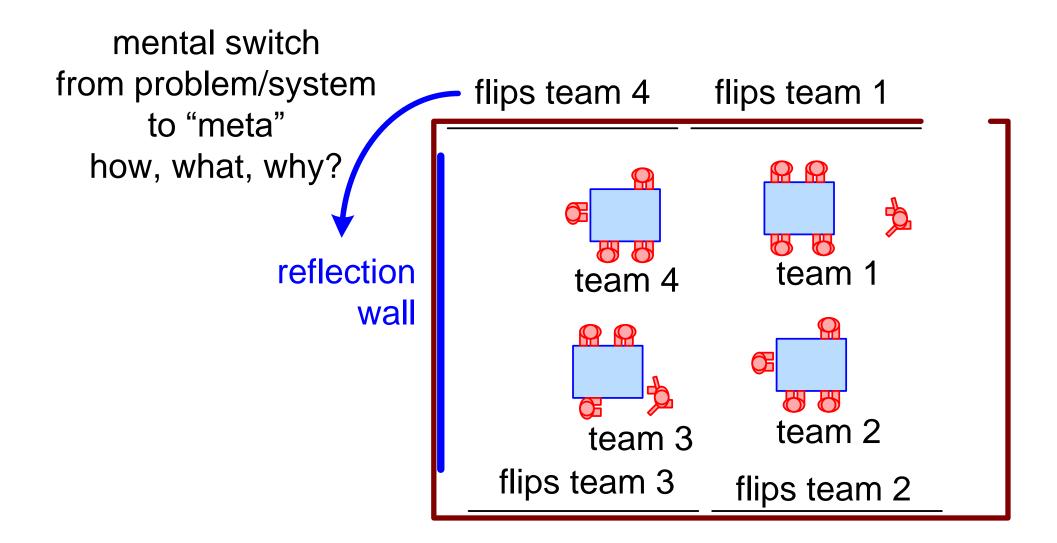


DSRP Model





Separate Reflection Wall





Scope and Topic of Reflection

operational or life cycle context

system of interest

component or function of interest

organization project team individual

principle

process or method

procedure or technique

tool or notation

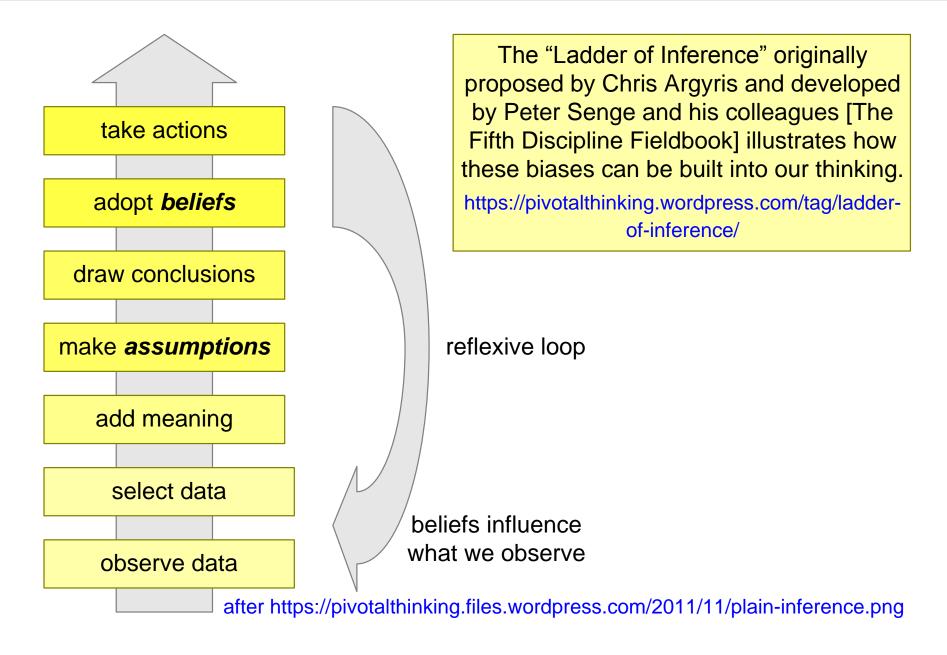
technical

psychosocial

means



The Role of Assumptions and Beliefs in Thinking





Architecting System Performance; Connecting Breadth and Depth

by Gerrit Muller TNO-ESI, University of South-Eastern Norway]

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Abstract

System Performance plays a crucial role in the customer value proposition and the business proposition. Minor details deep down into the system may have a large impact on system performance, and hence on both value propositions. Challenge in architecting system performance is to connect both worlds, which are mentally far apart.

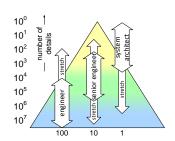
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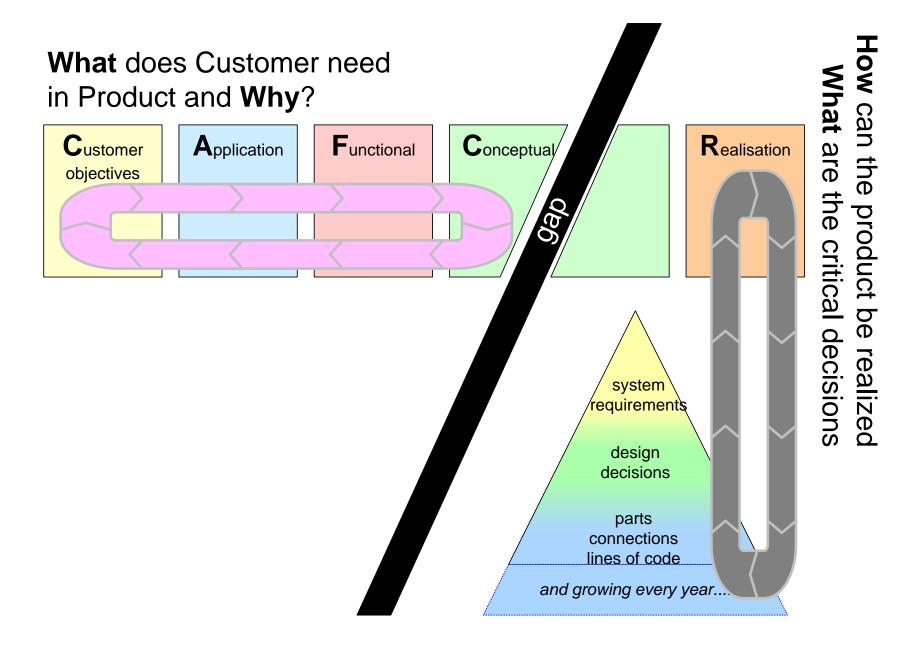
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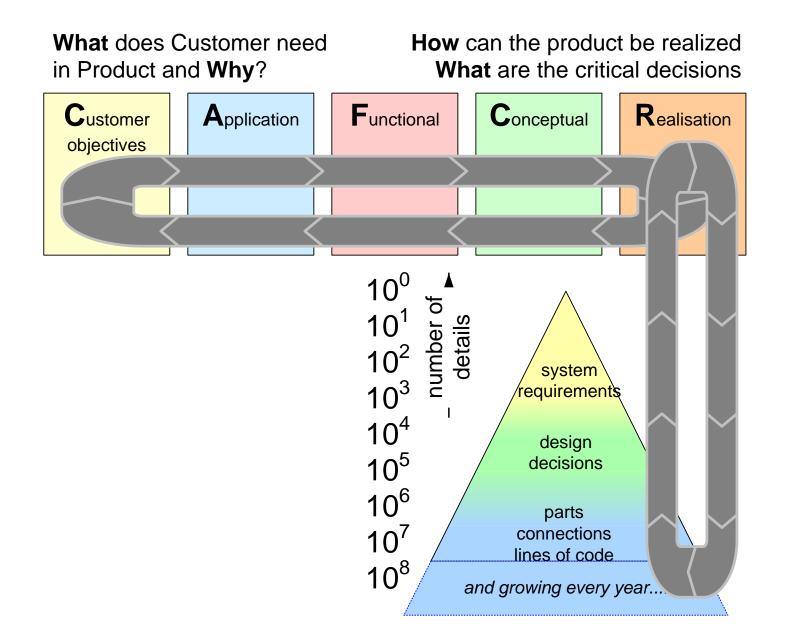


Organizational Problem: Disconnect



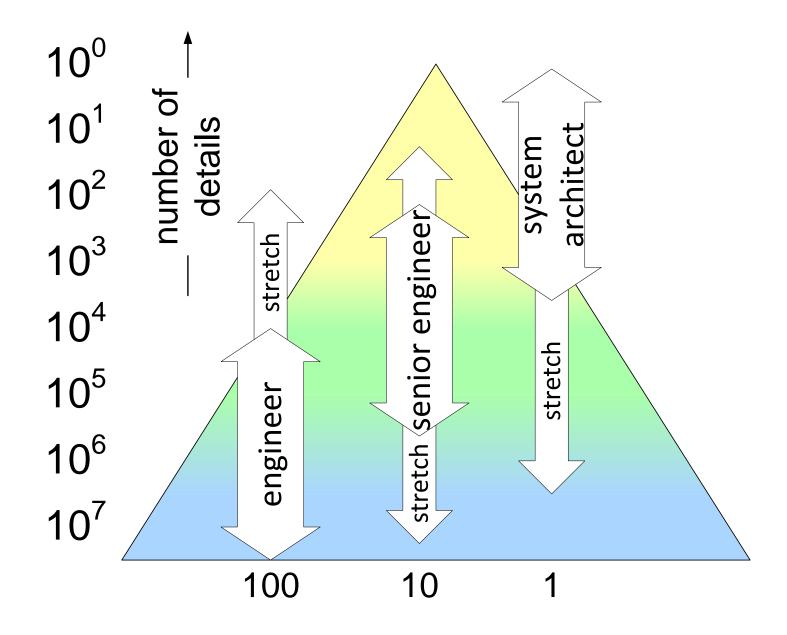


Architect: Connecting Problem and Technical Solution



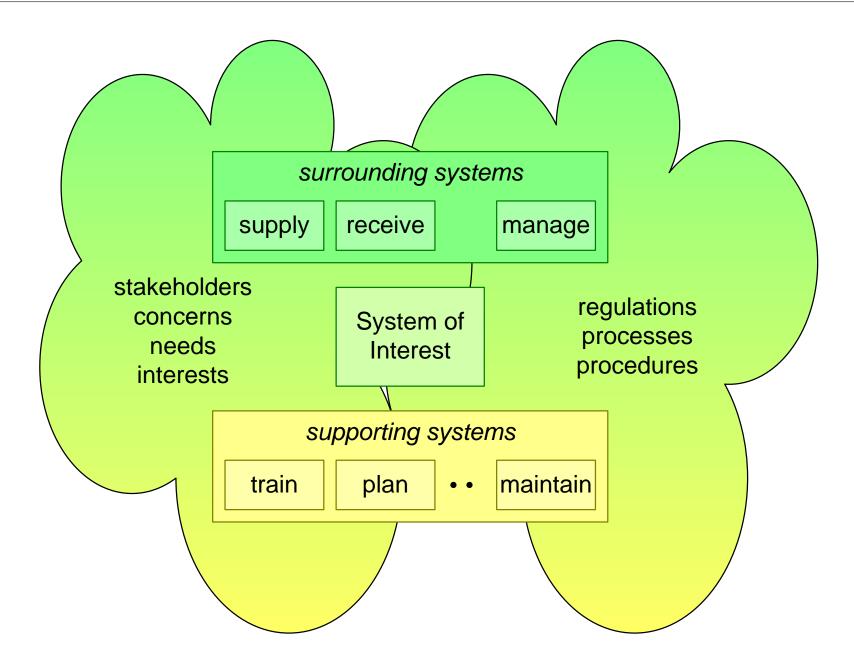


Major Bottleneck: Mental Dynamic Range





Breadth

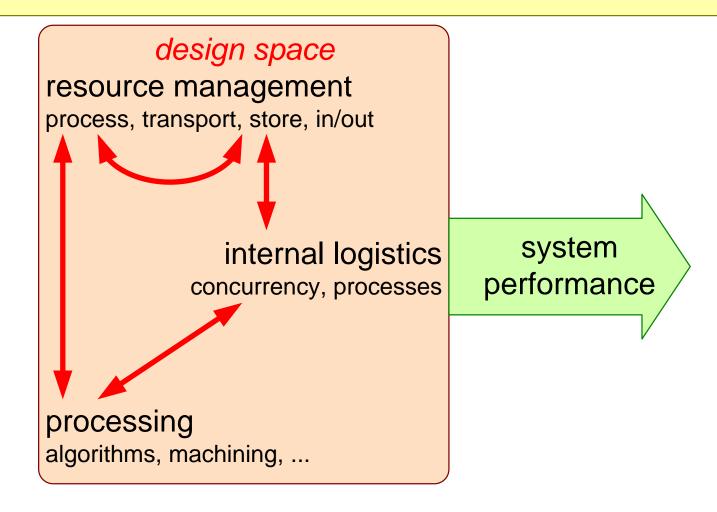




Depth

Devilish details in design space may have large impact on performance.

Many detailed design decisions determine system performance.





Modeling and Analysis; Performance Modeling

by Gerrit Muller TNO-ESI, University of South-Eastern Norway]

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Abstract

Principles and concepts of modeling performance.

Distribution

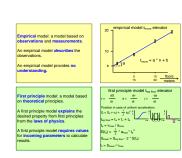
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Empirical versus First Principle Models

Empirical model: a model based on observations and measurements.

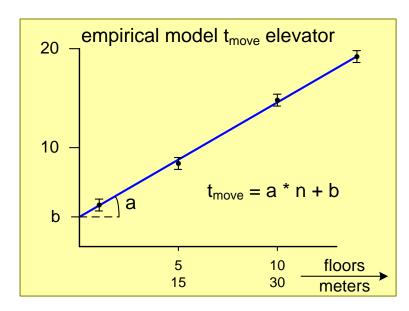
An empirical model **describes** the observations.

An empirical model provides no understanding.

First principle model: a model based on theoretical principles.

A first principle model explains the desired property from first principles from the laws of physics.

A first principle model requires values for incoming parameters to calculate results.

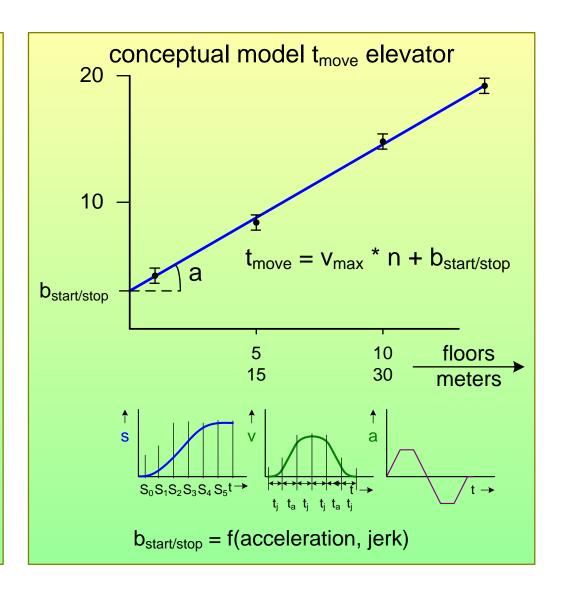


$$\begin{aligned} & \text{first principle model } t_{\text{top floor}} \text{ elevator} \\ v &= \frac{dS}{dt} \qquad a = \frac{dv}{dt} \qquad j = \frac{da}{dt} \\ & \text{Position in case of uniform acceleration:} \\ S_t &= S_0 + v_0 t + \frac{1}{2} a_0 t^2 \int_{S}^{t} \int_{t_0}^{t} \int$$

Conceptual = Hybrid of Empirical and First Principle

Conceptual model: a model explaining observations and measurements using a selection of first principles.

A conceptual model is a hybrid of empirical and first principle models; simple enough to understand and to reason, realistic enough to make sense.





From Zero to Higher Order Formulas

Oth order

main function main parameters

most simple order of magnitude

constant velocity

 $t_{top floor} = S_{top floor} / v_{max}$

1st order

add most significant secondary contributions improved estimation

constant acceleration

$$t_{top floor} = S_{top floor} / v_{max}$$
$$- a_{max} * t_a^2 / v_{max} + 2 * v_{max} / a_{max}$$

2nd order add next level of contributions

more accurate, understanding

constant jerk

$$t_{top floor} \sim = S_{top floor} / v_{max} - a_{max} * t_a^2 / v_{max}$$

+ 2 * $v_{max} / a_{max} + 2 * a_{max} / j_{max}$



Architecting System Performance; Level of Abstraction

by Gerrit Muller TNO-ESI, University of South-Eastern Norway]

e-mail: gaudisite@gmail.com

www.gaudisite.nl

Abstract

A recurring question in modeling and perfromance analysis is when to stop digging. What level of detail is needed to achieve acceptable performance? What level of abstraction result in credible and sufficiently accurate results? How to cope with many levels of abstraction?

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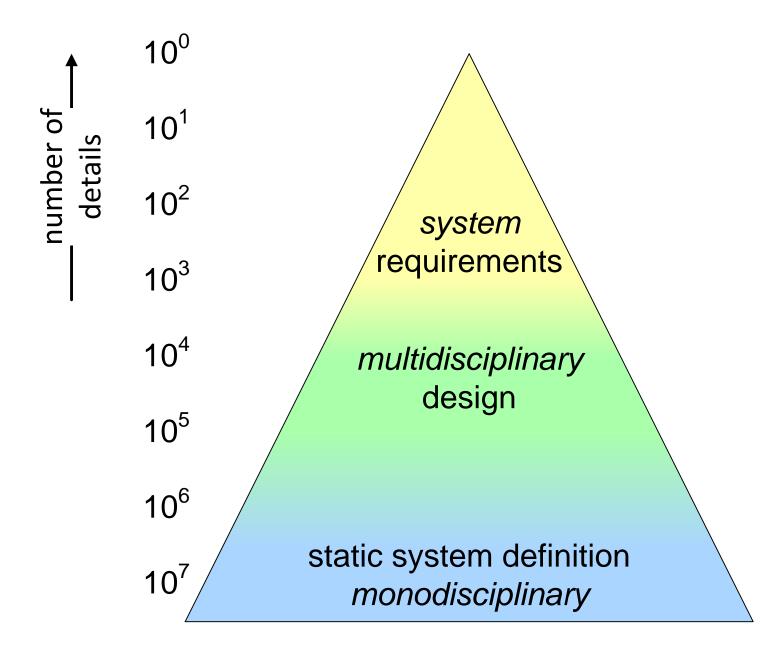
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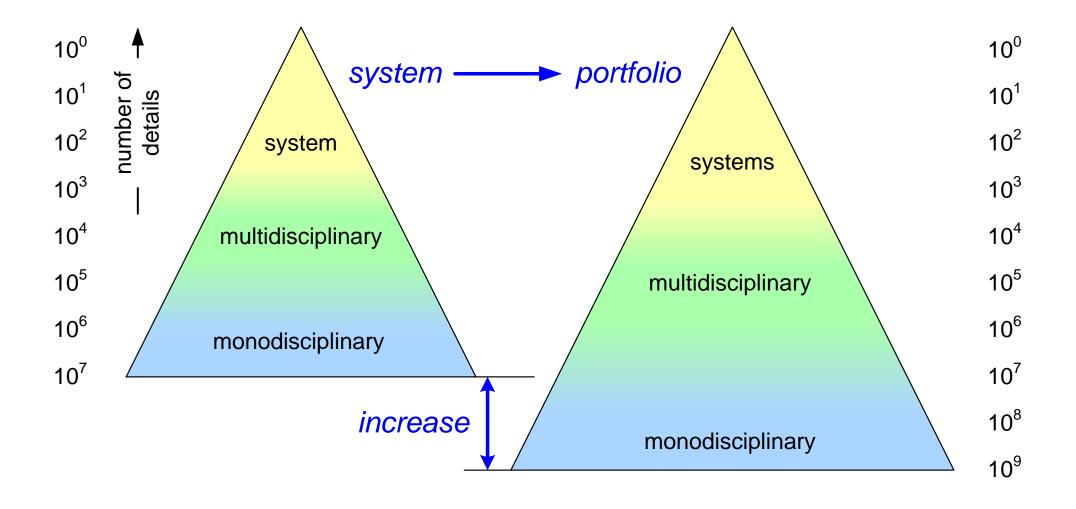


Level of Abstraction Single System



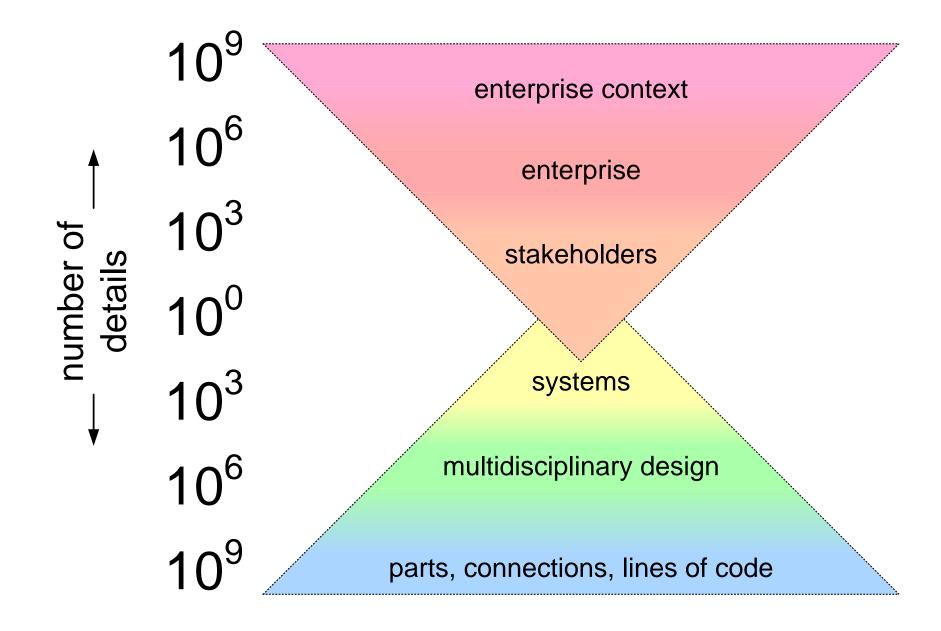


From system to Product Family or Portfolio



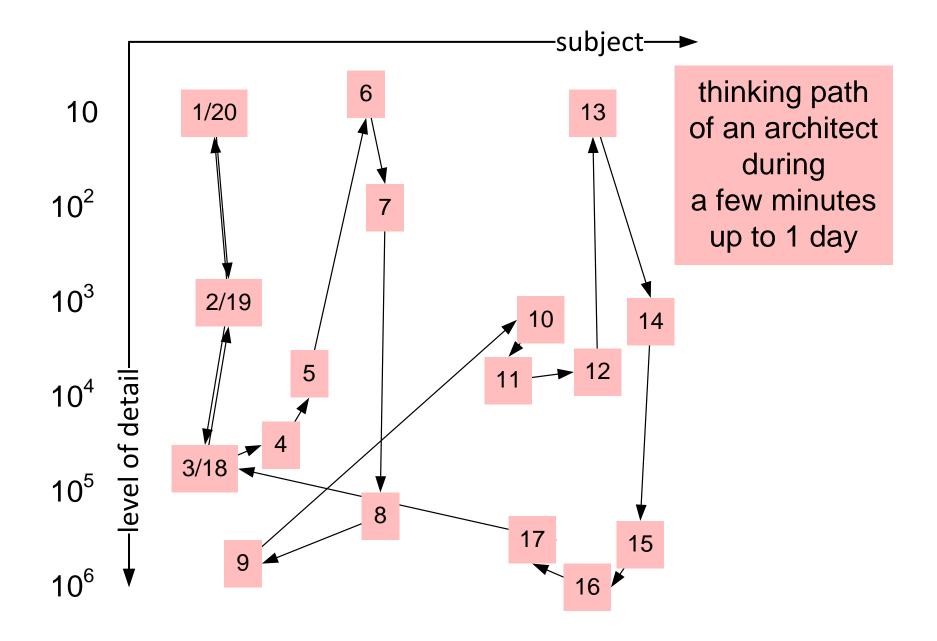


Product Family in Context



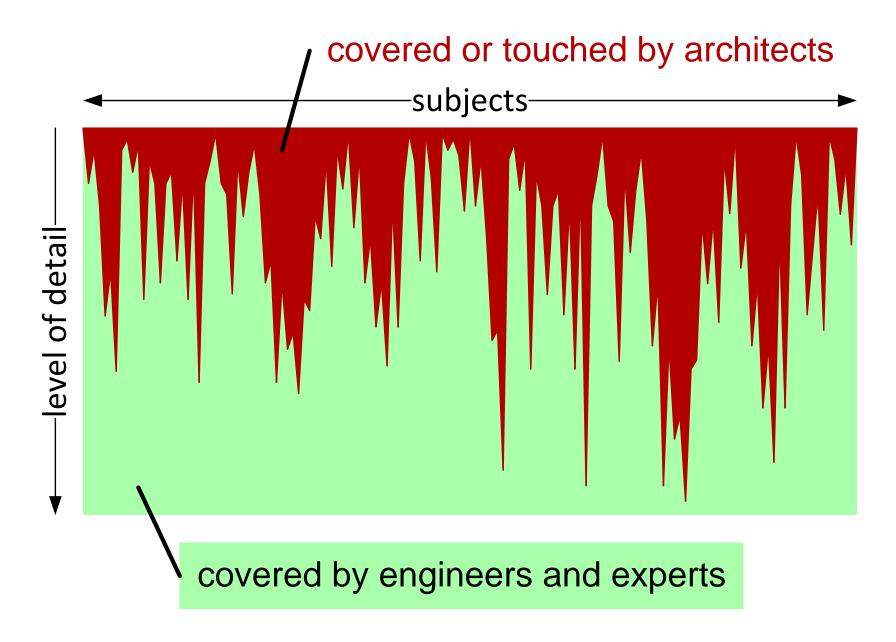


The seemingly random exploration path



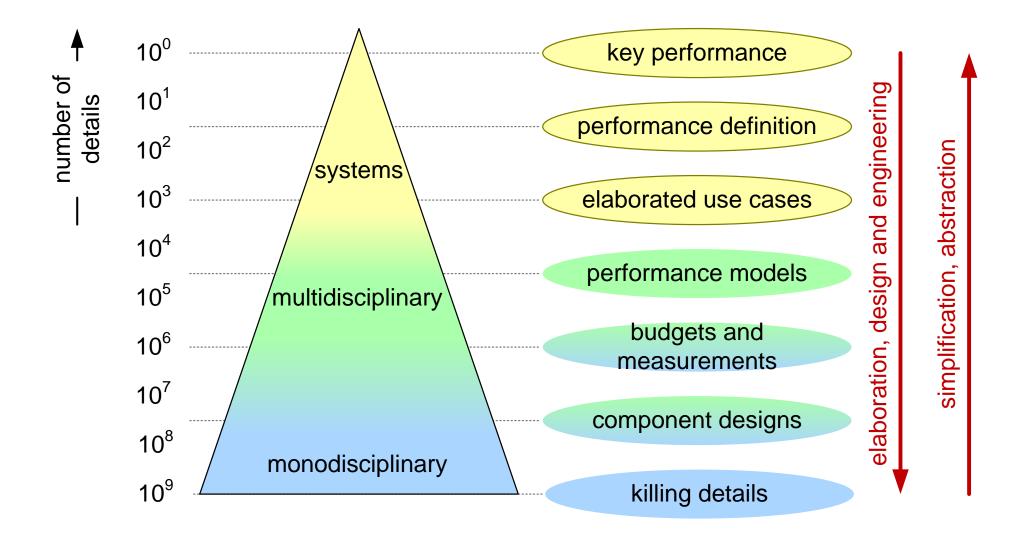


Coverage of problem and solution space



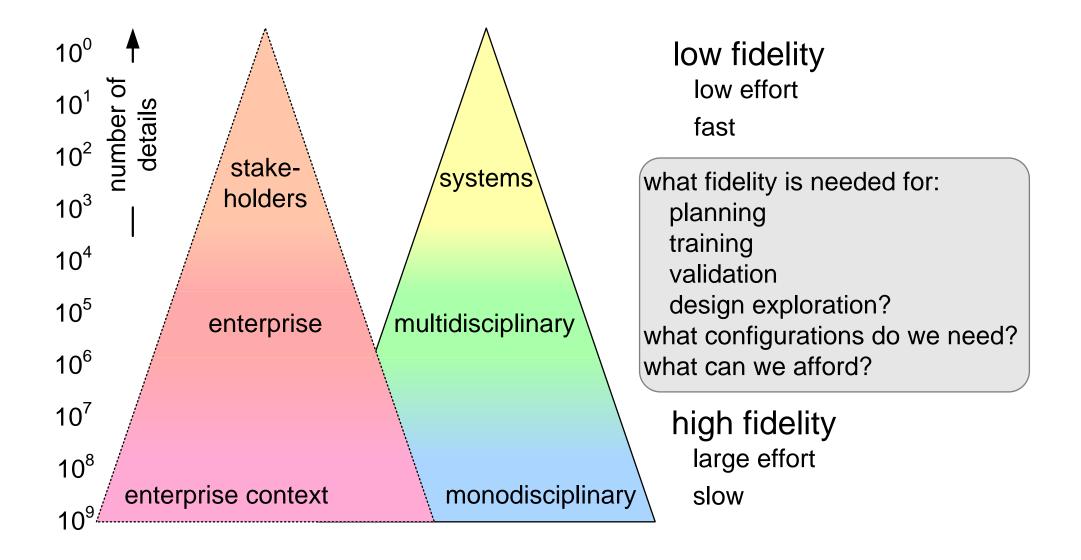


Many Levels of Abstraction





Fidelity Properties





Visualizing Dynamic Behavior

by Gerrit Muller TNO-ESI, University of South-Eastern Norway]

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Abstract

Dynamic behavior manifests itself in many ways. Architects need multiple complementary visualizations to capture dynamic behavior effectively. Examples are capturing information, material, or energy flow, state, time, interaction, or communication.

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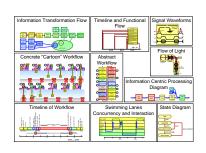
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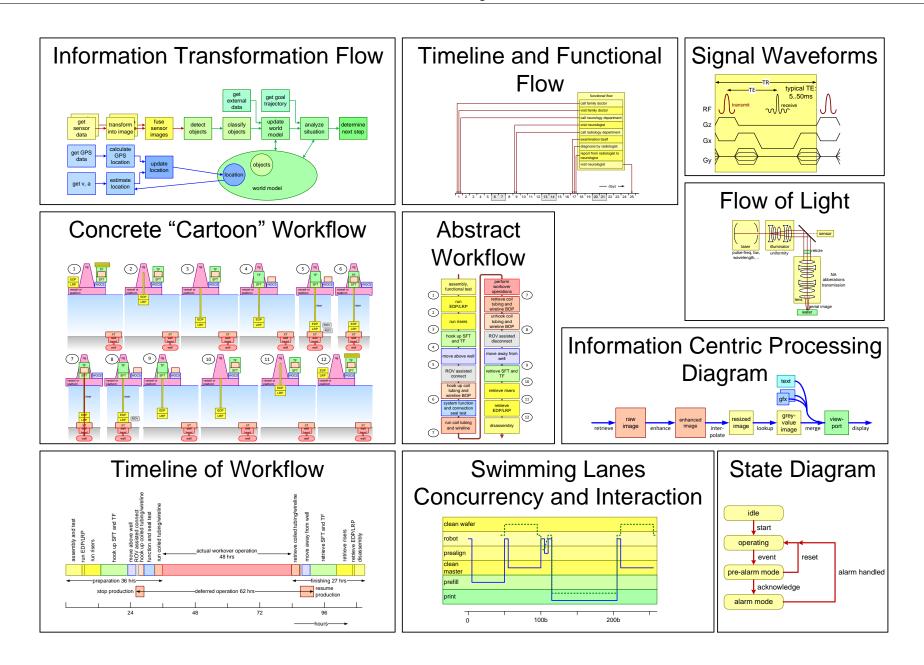
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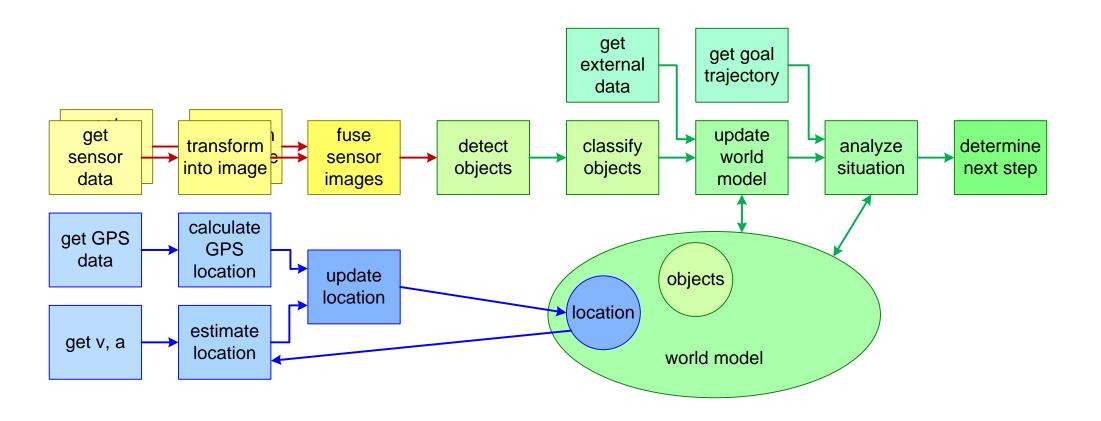
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Overview of Visualizations of Dynamic Behavior

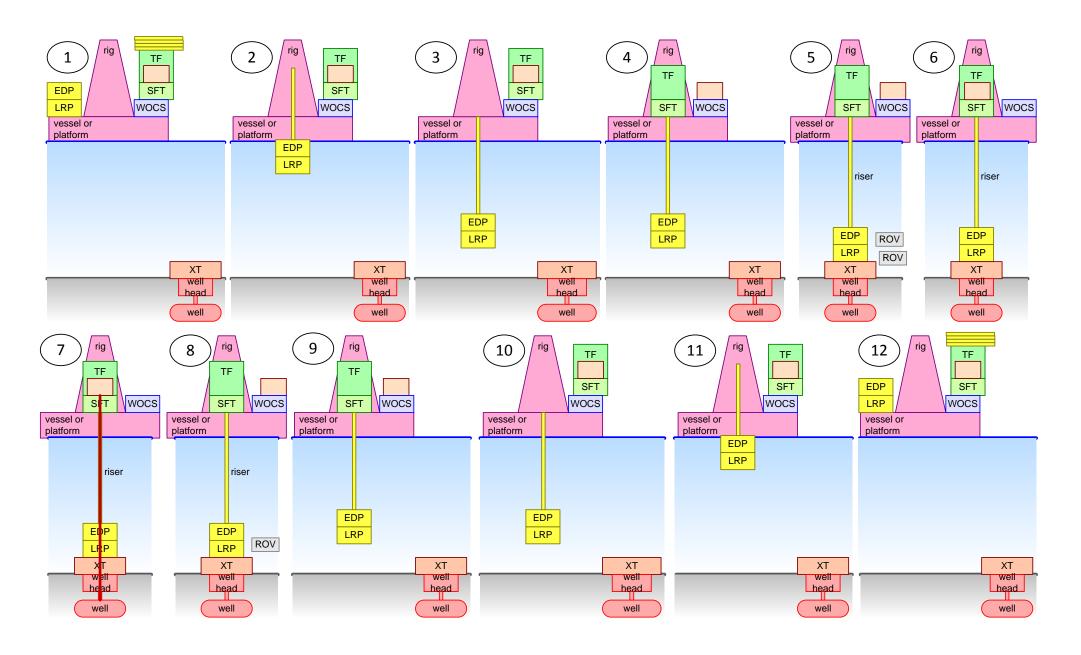


Example Functional Model of Information Flow



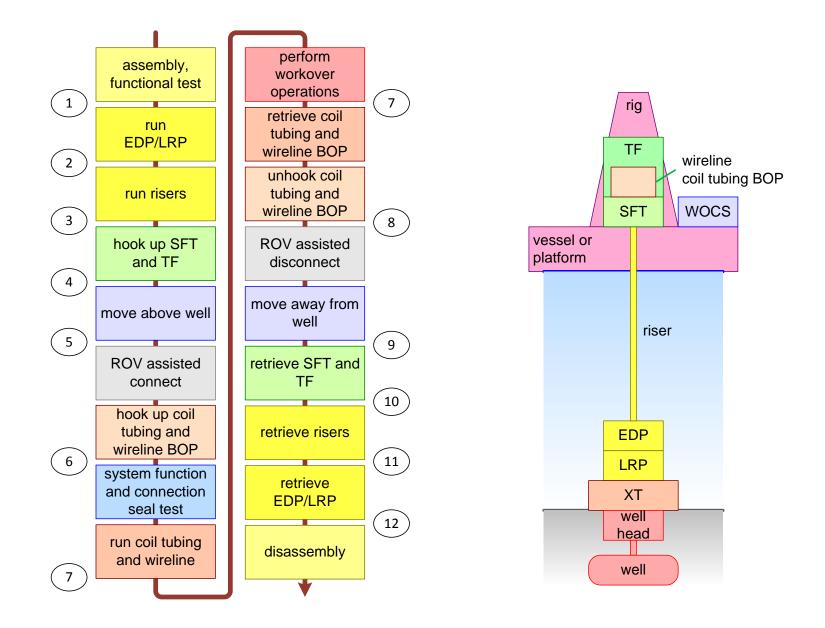


"Cartoon" Workflow



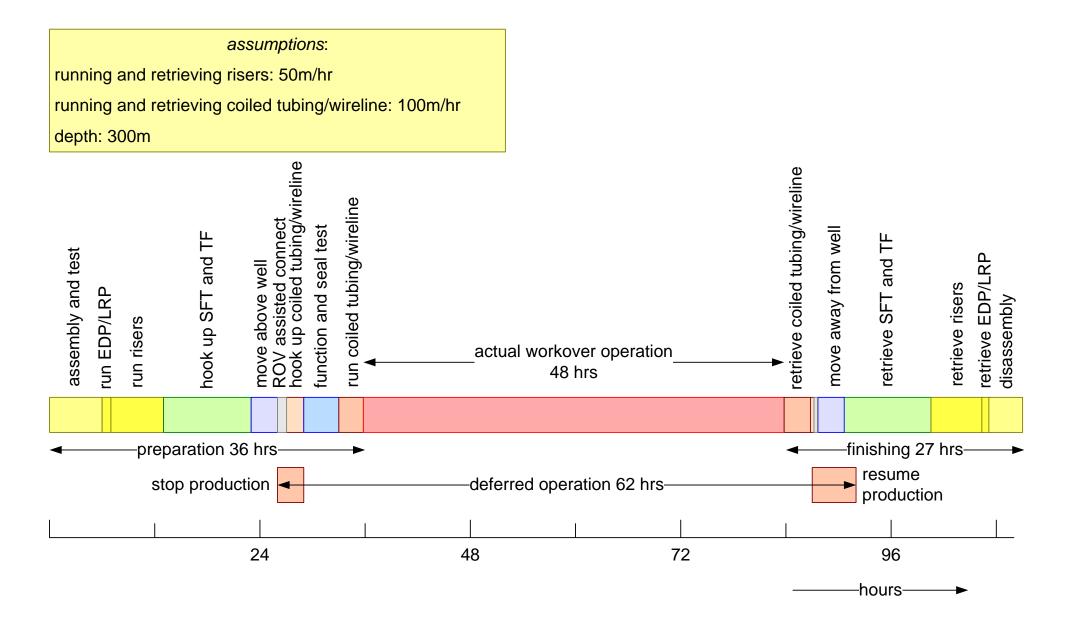


Workflow as Functional Model



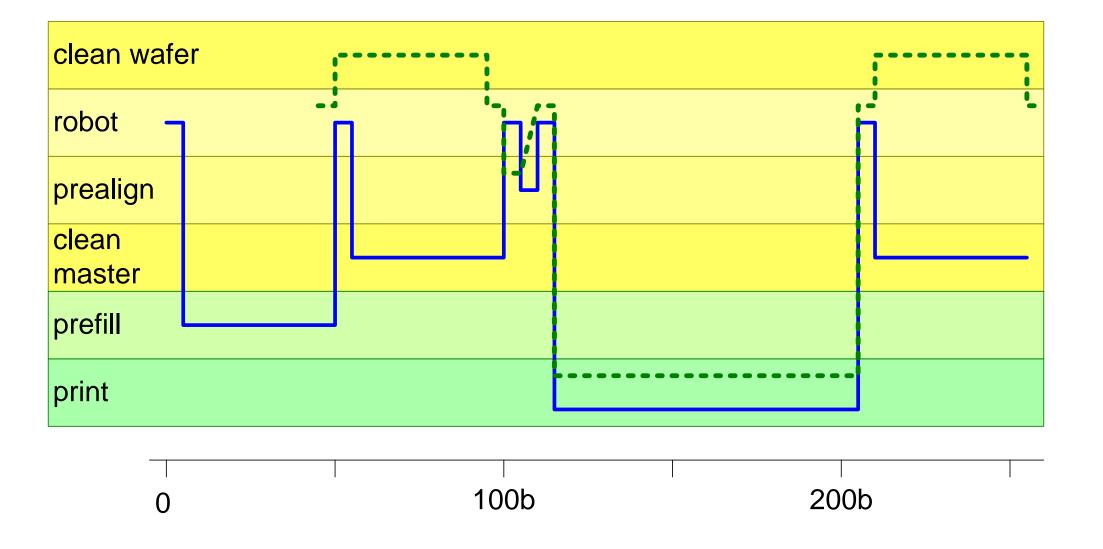


Workflow as Timeline



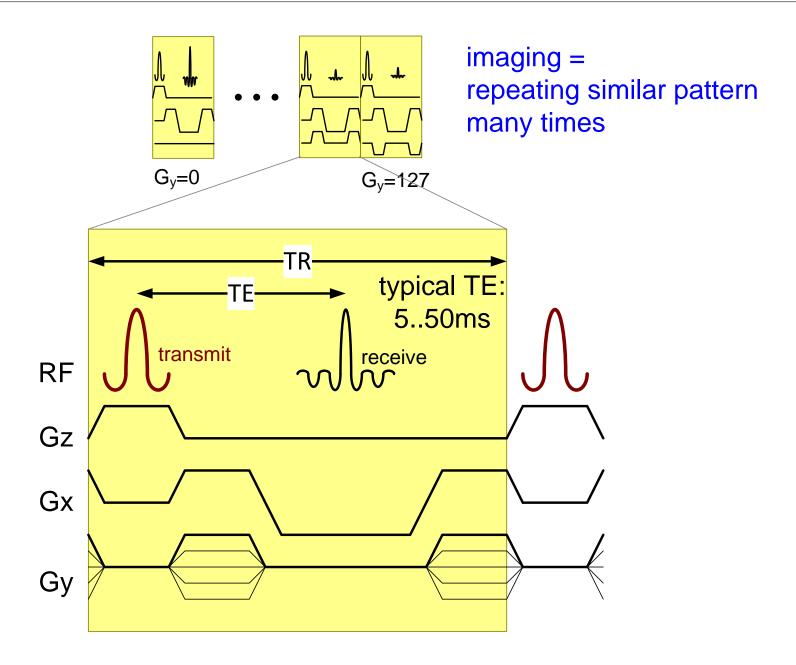


Swimming Lane Example



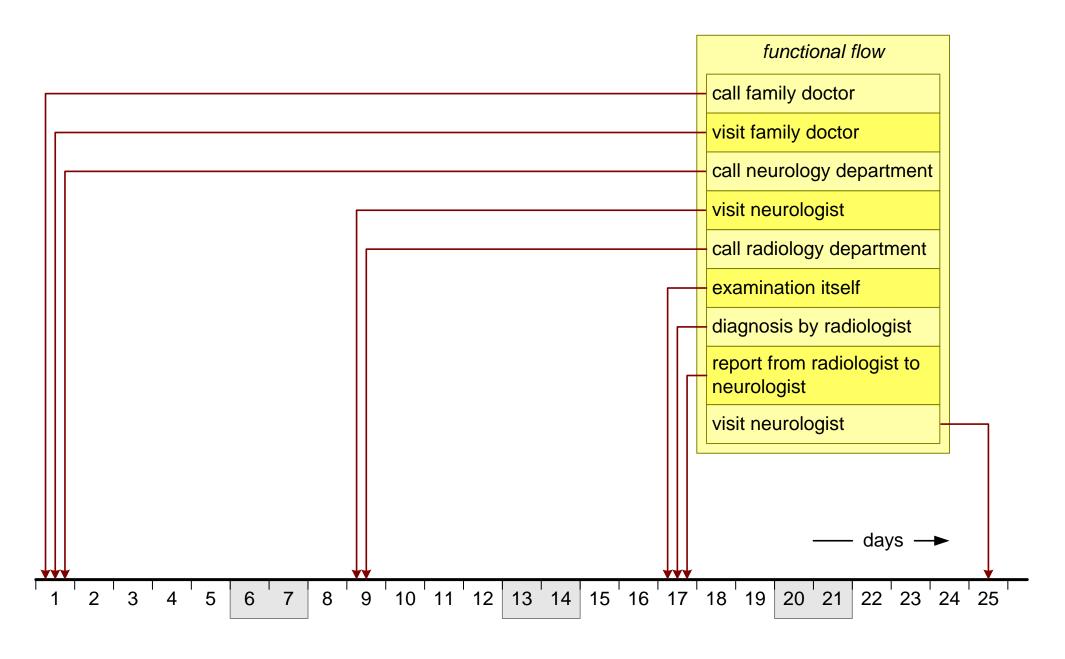


Example Signal Waveforms



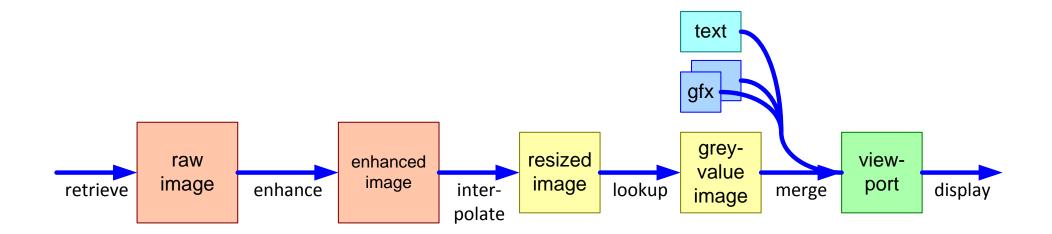


Example Time Line with Functional Model



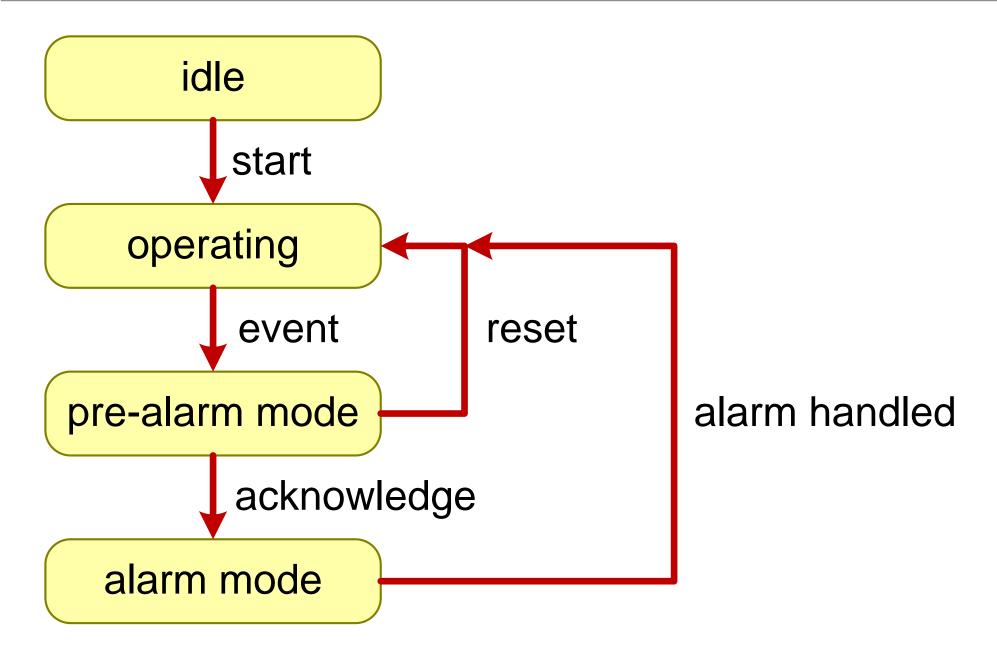


Information Centric Processing Diagram



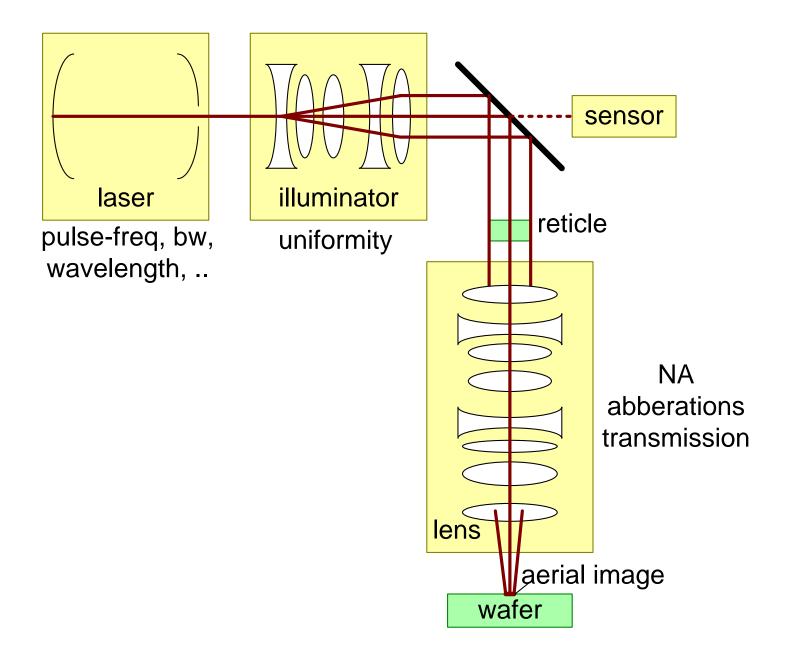


Example State Diagram





Flow of Light (Physics)





Dynamic Behavior is Multi-Dimensional

How does the system work and operate?

Functions describe what rather than how.

Functions are verbs.

Input-Process-Output paradigm.

Multiple kinds of flows:

physical (e.g. hydrocarbons, goods, energy)

information (e.g. measurements, signals)

control

Time, events, cause and effect

Concurrency, synchronization, communication

multi-dimensional information and dynamic behavior



Modeling and Analysis: Emerging Behavior

by Gerrit Muller TNO-ESI, University of South-Eastern Norway]

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Abstract

The essence of a system is that the parts together can do more than the separate parts. The interaction of the parts results in behavior and properties that cannot be seen as beloning to individual parts. We call this type of behavior "emerging behavior".

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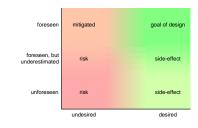
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Emergence is Normal and Everywhere

emergent behavior and properties =

function of

dynamic interaction between

parts in the system and

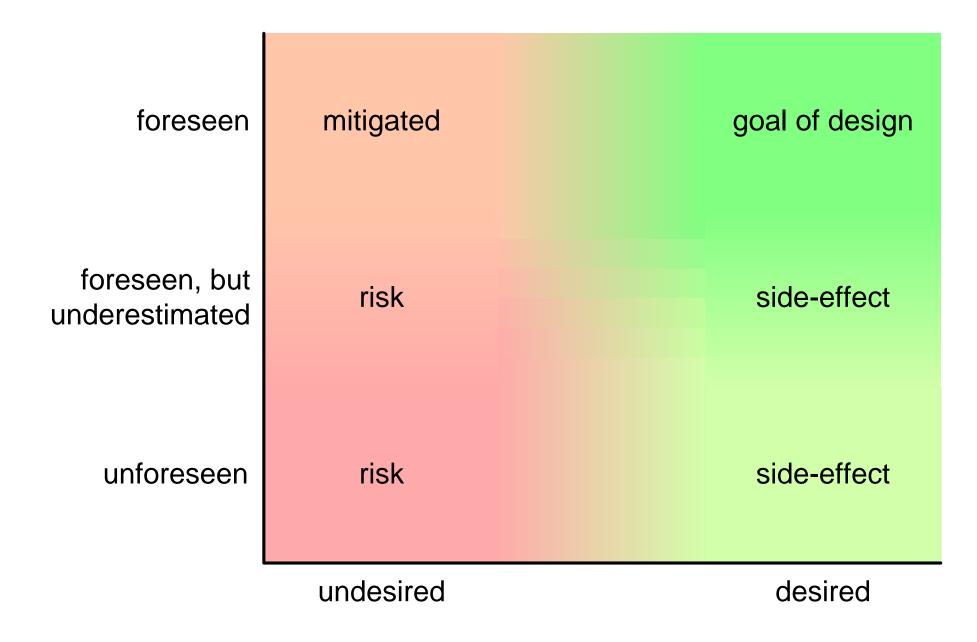
context of the system

examples

- flying and stalling of an airplane
- Tacoma bridge resonance



Emergence, Desire, and Foreseeing





Modeling and Analysis: Budgeting

by Gerrit Muller TNO-ESI, HSN-NISE

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www.gaudisite.nl

Abstract

This presentation addresses the fundamentals of budgeting: What is a budget, how to create and use a budget, what types of budgets are there. What is the relation with modeling and measuring.

Distribution

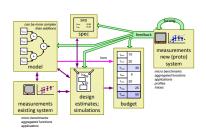
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Budgeting

content of this presentation

What and why of a budget

How to create a budget (decomposition, granularity, inputs)

How to use a budget



A **budget** is

a quantified instantation of a conceptual model

A **budget** can

prescribe or describe the contributions

by parts of the solution

to the system quality under consideration

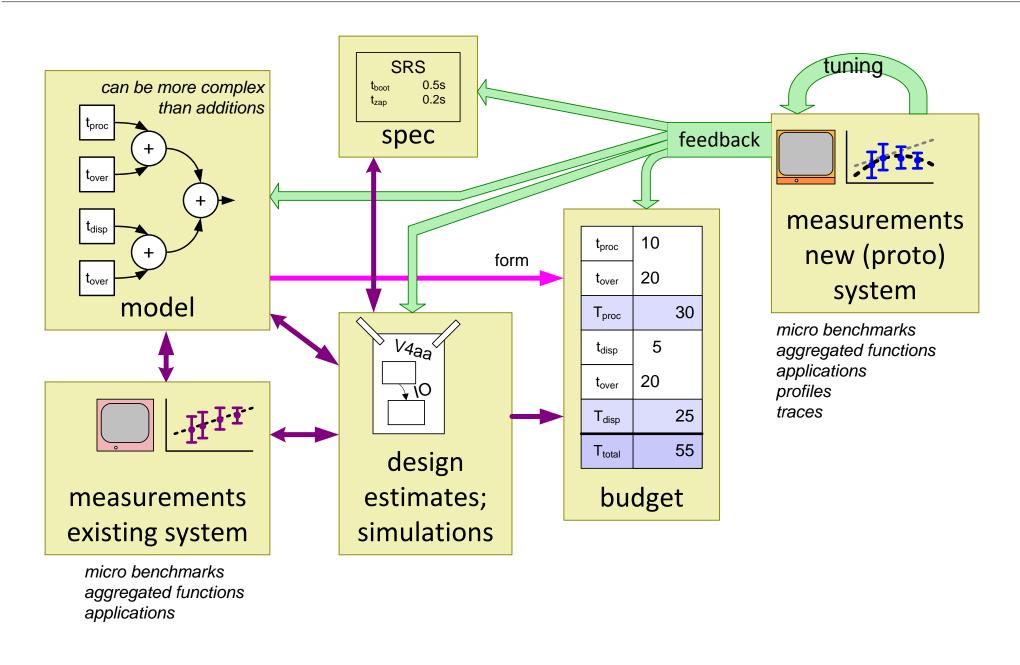


Why Budgets?

- to make the design explicit
- to provide a baseline to take decisions
- to specify the requirements for the detailed designs
- to have guidance during integration
- to provide a baseline for verification
- to manage the design margins explicitly



Visualization of Budget Based Design Flow





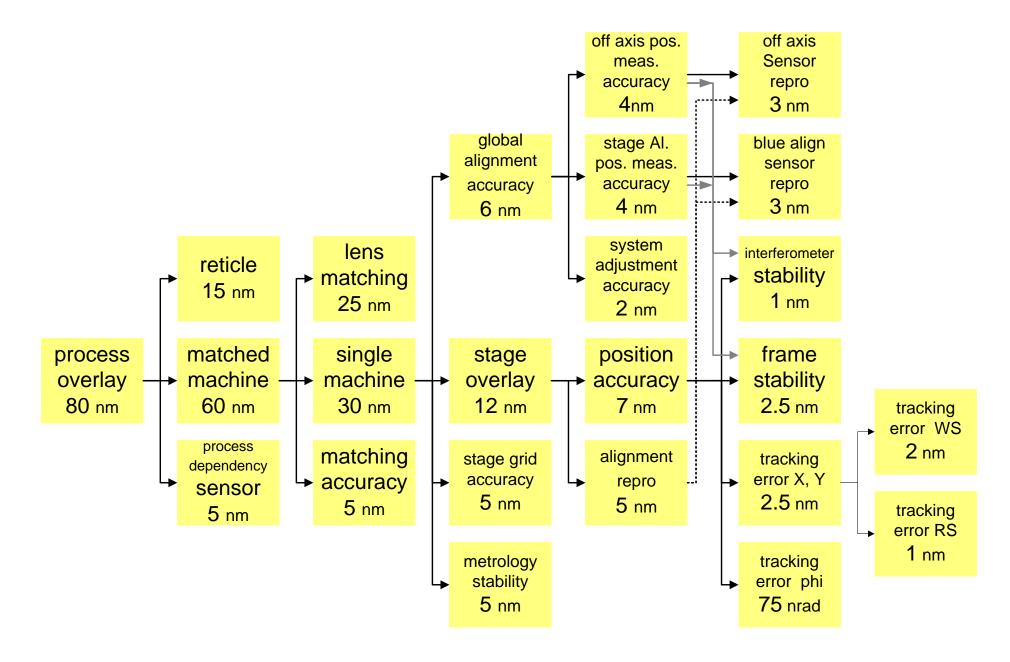
Stepwise Budget Based Design Flow

step example

| 1A measure old systems | micro-be | nchmarks, aggregated functions, applications |
|--|----------|--|
| 1B model the performance starting with old s | systems | flow model and analytical model |
| 1C determine requirements for new system | 1 | response time or throughput |
| 2 make a design for the new system | | explore design space, estimate and simulate |
| 3 make a budget for the new system: | measur | models provide the structure ements and estimates provide initial numbers specification provides bottom line |
| 4 measure prototypes and new system | micro-be | nchmarks, aggregated functions, applications profiles, traces |
| 5 Iterate steps 1B to 4 | | |



Budgets Applied on Waferstepper Overlay



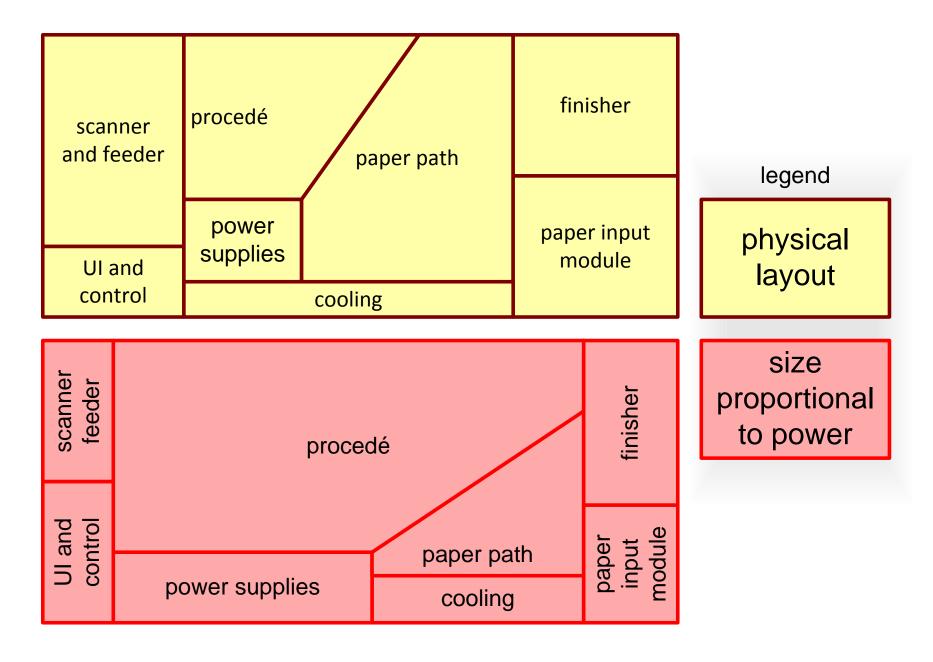


Budgets Applied on Medical Workstation Memory Use

| memory budget in Mbytes | code | obj data b | ulk data | total |
|--|--|---|--|---|
| shared code User Interface process database server print server optical storage server communication server UNIX commands compute server | 11.0 0.3 0.3 0.3 0.3 0.3 0.3 | 3.0 3.2 1.2 2.0 2.0 0.2 0.5 | 12.0 3.0 9.0 1.0 4.0 0 6.0 | 11.0 15.3 6.5 10.5 3.3 6.3 0.5 6.8 |
| application SW total | 0.3 13.4 | 0.5 12.6 | 35.0 | 0.8 61.0 |
| UNIX Solaris 2.x file cache | | | | 10.0 |
| total | | | | 74.0 |

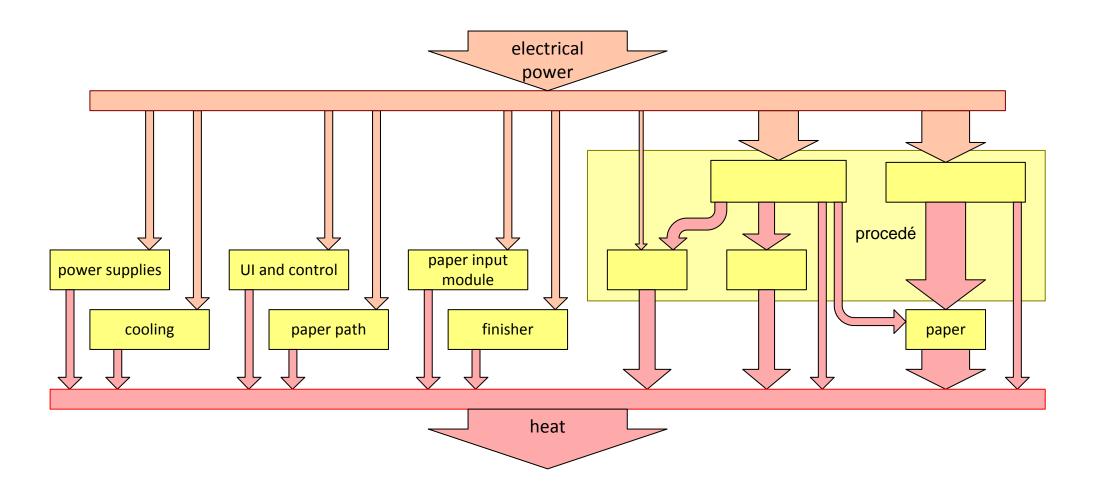


Power Budget Visualization for Document Handler





Alternative Power Visualization





Evolution of Budget over Time

fact finding through details
aggregate to end-to-end performance
search for appropriate abstraction level(s)

from coarse guesstimate

to reliable prediction

from typical case

to boundaries of requirement space

from static understanding

to dynamic understanding

from steady state

to initialization, state change and shut down

from old system

to prototype

to actual implementation

time ----

start

later

only if needed



Potential Applications of Budget based design

- resource use (CPU, memory, disk, bus, network)
- timing (response, latency, start up, shutdown)
- productivity (throughput, reliability)
- Image Quality parameters (contrast, SNR, deformation, overlay, DOF)
- cost, space, time



What kind of budget is required?

| static | dynamic |
|--------------|------------|
| typical case | worst case |
| global | detailed |
| approximate | accurate |

is the budget based on wish, empirical data, extrapolation, educated guess, or expectation?



Summary of Budgeting

A budget is a quantified instantiation of a model

A budget can prescribe or describe the contributions by parts of the solution to the system quality under consideration

A budget uses a decomposition in tens of elements

The numbers are based on historic data, user needs, first principles and measurements

Budgets are based on models and estimations

Budget visualization is critical for communication

Budgeting requires an incremental process

Many types of budgets can be made; start simple!



Colophon

The Boderc project contributed to Budget Based Design. Especially the work of

Hennie Freriks, Peter van den Bosch (Océ),

Heico Sandee and Maurice Heemels (TU/e, ESI)

has been valuable.



Modeling and Analysis; Modeling Paradigms

by Gerrit Muller TNO-ESI, University of South-Eastern Norway]

e-mail: gaudisite@gmail.com

www.gaudisite.nl

Abstract

The word modeling is used for a wide variety of modeling approaches. These approaches differ in purpose, level of detail, effort, stakeholders, degree of formaility, and tool support.

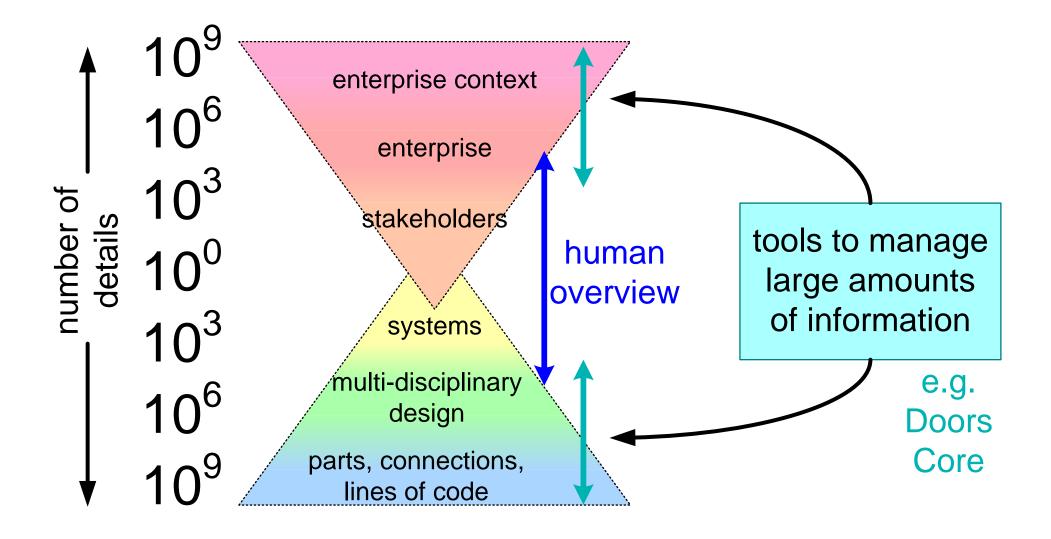
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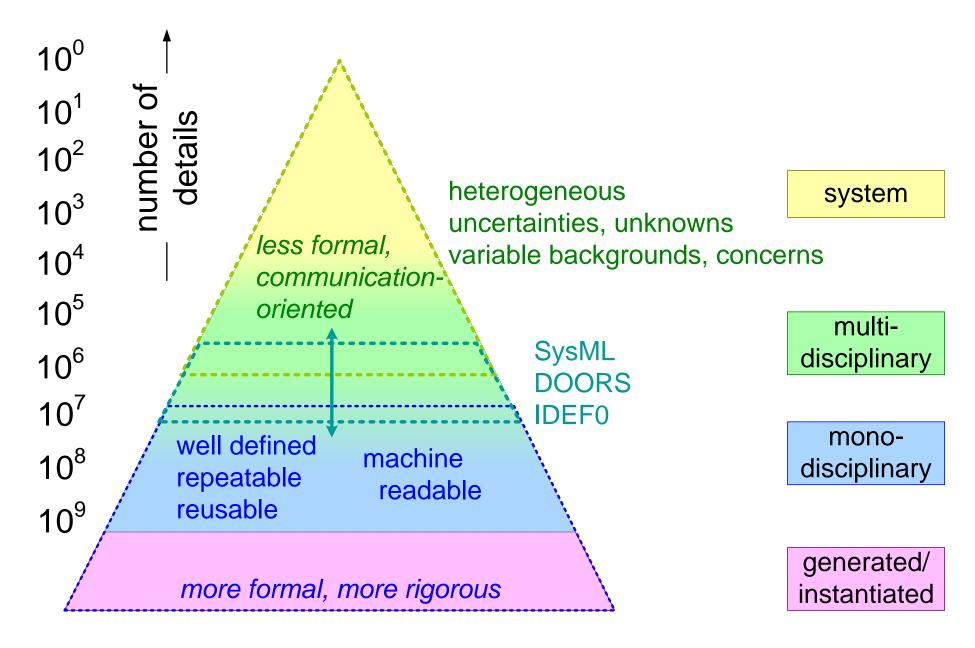


Human Thinking and Tools





Formality Levels in Pyramids





Modeling Paradigms

| paradigm | purpose |
|--|--|
| Conceptual system modeling | architecting understanding, evaluating, creating reasoning, communicating, decision making |
| SysML | formal capture of structure and behavior integrating other tools simulating |
| Design for 6 sigma | quality improvement in repeatable environments black box oriented |
| Conceptual information modeling | understanding and formalizing information |
| Design Framework | capturing and tracing architecture decisions |
| Matlab | modeling and analyzing designs and algorithms simulation and code generation |
| CAD | mechanical and electrical design interoperates with dedicated analysis, e.g. thermal, structural |
| Formal specification and design (model checkers) | verification |



Modeling and Analysis: Applications and Variations

by Gerrit Muller TNO-ESI, University of South-Eastern Norway]

e-mail: gaudisite@gmail.com

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Abstract

Models are used for a wide variation of purposes. Stakeholders can get confused between "reality" and the virtual counterparts. In practice, many hybrids between "real" and virtual systems exist. For example, planning and training systems using real algorithms and data, and physical systems using a world model for situation awareness.

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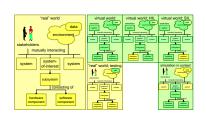
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Model Applications and Variations

sales acquisition

capability analysis

Systems of Systems apply all asynchronously

development

verification validation

operation

understanding exploration optimization

test data comparison trouble shooting

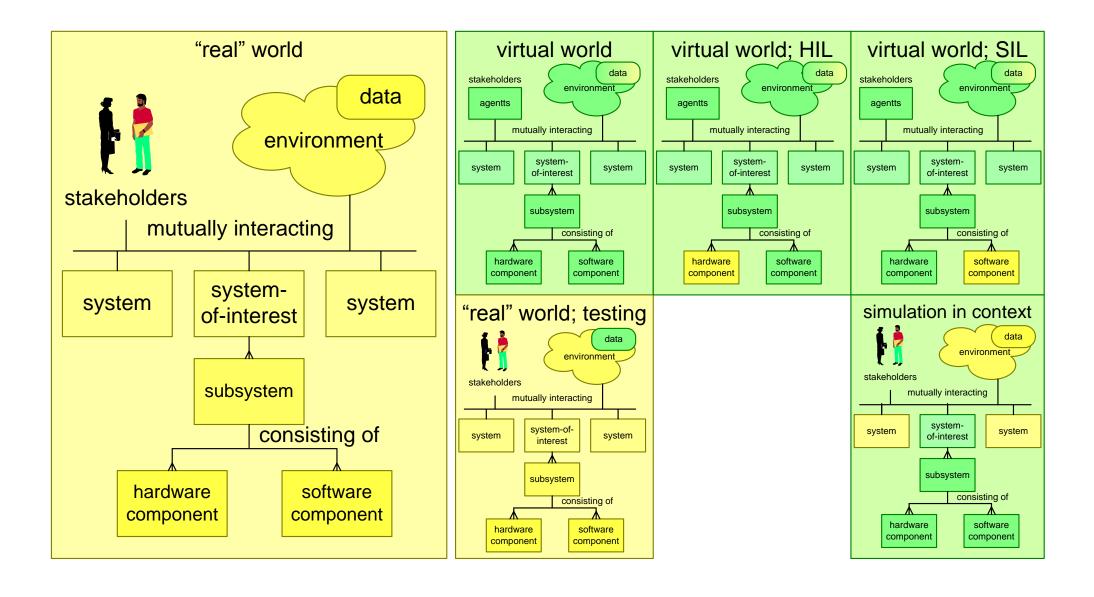
mission planning training health monitoring

in system
situation awareness
planning
training
health monitoring

evolvabilityall phases repeatwith same needs

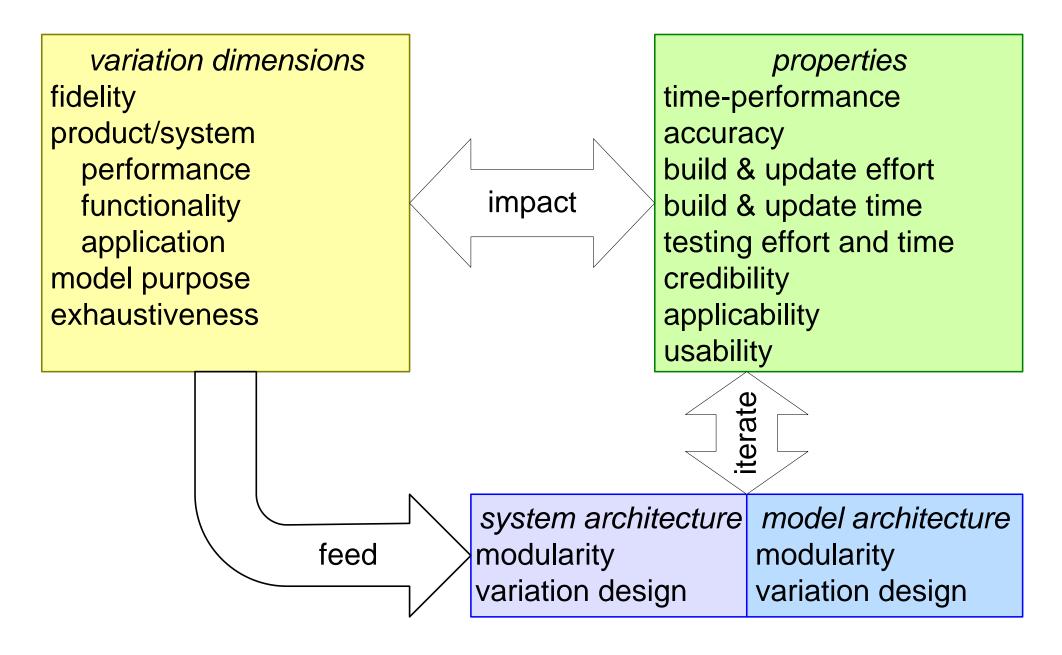


Spectrum from Real to Virtual Systems





Architecting for Variations





Modeling and Analysis: Model Analysis

by Gerrit Muller TNO-ESI, USN-NISE

e-mail: gaudisite@gmail.com

www.gaudisite.nl

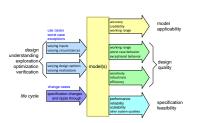
Abstract

Models only get value when they are actively used. We will focus in this presentation on analysis aspects: accuracy, credibility, sensitivity, efficiency, robustness, reliability and scalability.

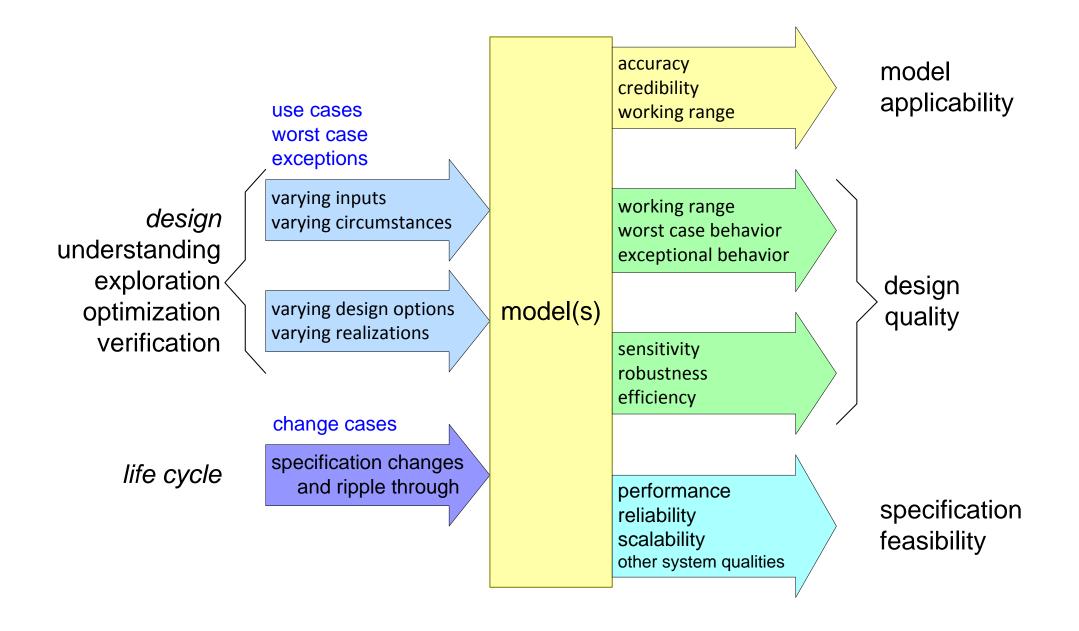
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What Comes out of a Model





Applicability of the Model

+**ε**1 - **ε**2

input accuracy credibility measurements abstraction assumptions facts

model(s)

accuracy credibility working range

abstraction credibility working range

abstraction usage context specifications designs realizations

model realization credibility propagation



How to Determine Applicability

try out models

be aware of accuracy, credibility and working range

simple and small models

1. Estimate accuracy of results

based on most significant inaccuracies of inputs and assumed model propagation behavior

2. Identify top 3 credibility risks

identify biggest uncertainties in inputs, abstractions and realization

3. Identify relevant working range risks

identify required (critical) working ranges and compare with model working range

substantial models

systematic analysis and documentation of accuracy, credibility and working range



Common Pitfalls

discrete events in continuous world

discretization artefacts e.g. stepwise simulations

(too) systematic input data

random data show different behavior e.g. memory fragmentation

fragile model

small model change results in large shift in results

self fulfilling prophecy

price erosions + cost increase (inflation) -> bankruptcy



Worst Case Questions

Which design assumptions have a big impact on system performance?

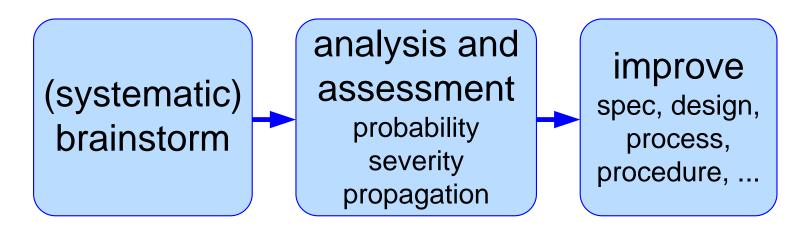
What are the worst cases for these assumptions?

How does the system behave in the worst case?

- a. poor performance within spec
- b. poor performance not within spec
- c. failure -> reliability issue



FMEA-like Analysis Techniques



| safety hazard analysis | potential hazards | damage | measures |
|---------------------------|---------------------------------|----------------------|-----------|
| reliability FMEA | failure modes exceptional cases | effects | measures |
| security | vulnerability risks | consequences | measures |
| maintainability | change cases | impact, effort, time | decisions |
| performance | worst cases | system behavior | decisions |



Brainstorming Phases

wave 1: the obvious

wave 2: more of the same

wave 3: the exotic, but potentially important

don't stop too early with brainstorming!



Different Viewpoints for Analysis

usage context

new product
e.g. WoW extension
merger
automated access

new functions new interfaces new media new standards

system

cache/memory trashing garbage collection critical sections local peak loads intermittent HW failure

power failure network failure new SW release roll back to old SW release

life cycle context



Modeling and Analysis: Reasoning Approach

by Gerrit Muller TNO-ESI, HSN-NISE

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Abstract

We make models to facilitate decision making. These decisions range from business decisions, such as Service Level Agreements, to requirements, and to detailed design decisions. The space of decisions is huge and heterogeneous. The proposed modeling approach is to use multiple small and simple models. In this paper we discuss how to reason by means of multiple models.

Distribution

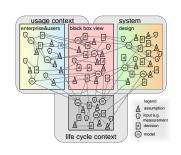
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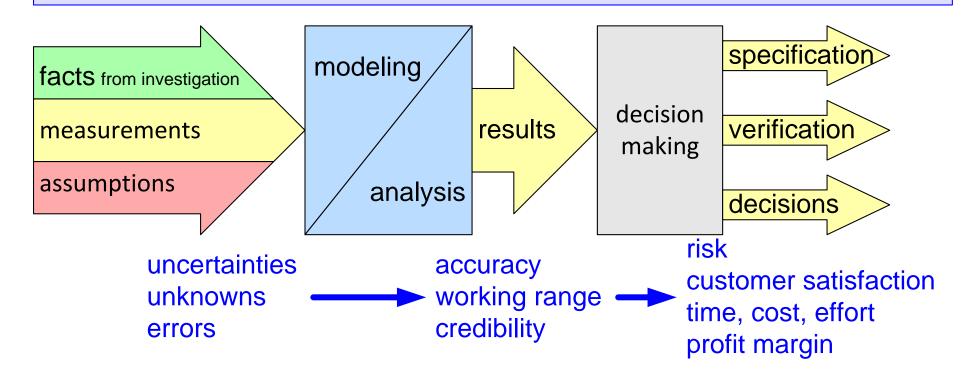
Purpose of Modeling

How to use multiple models to facilitate decisions?

How to get from many fragments to integral insight?

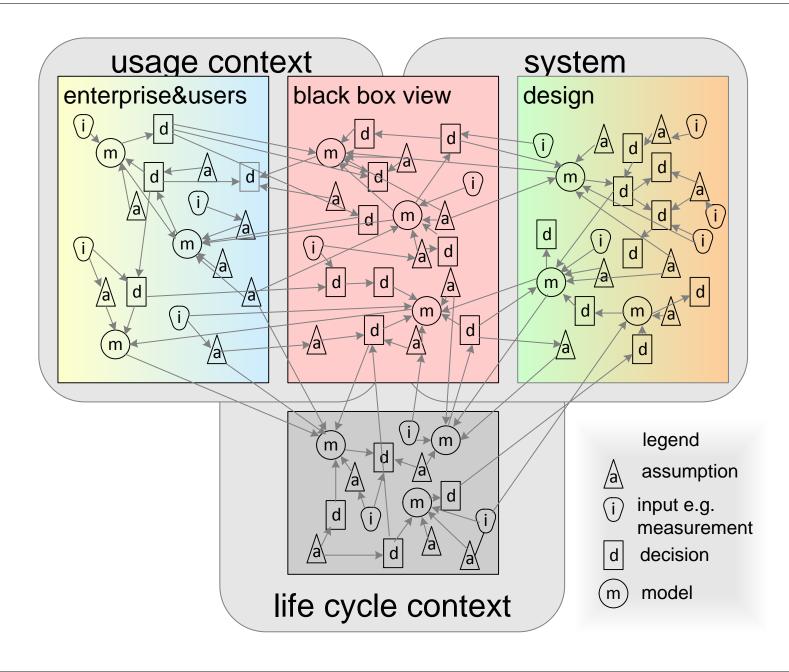
How many models do we need?

At what quality and complexity levels?



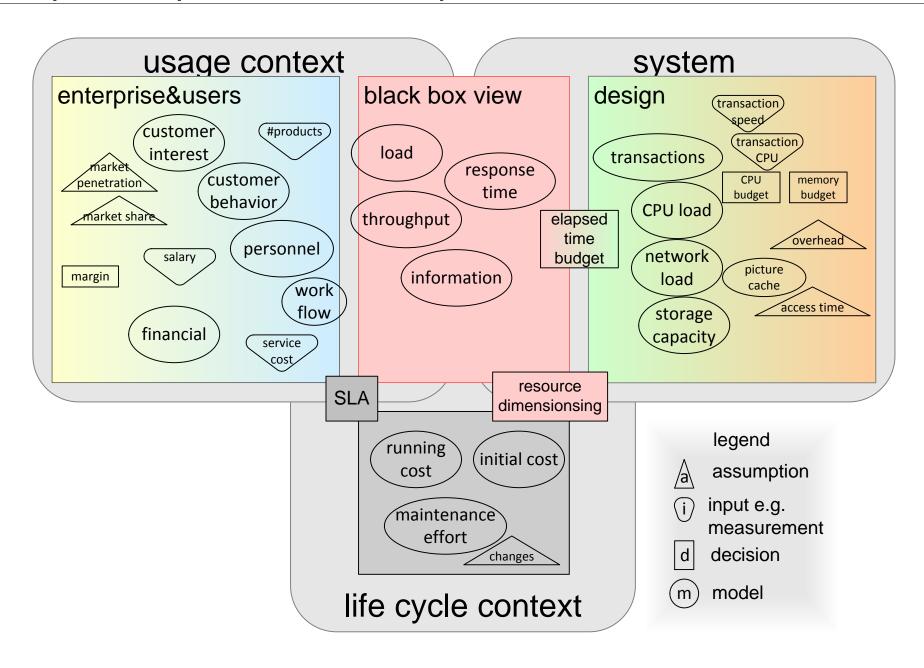


Graph of Decisions and Models



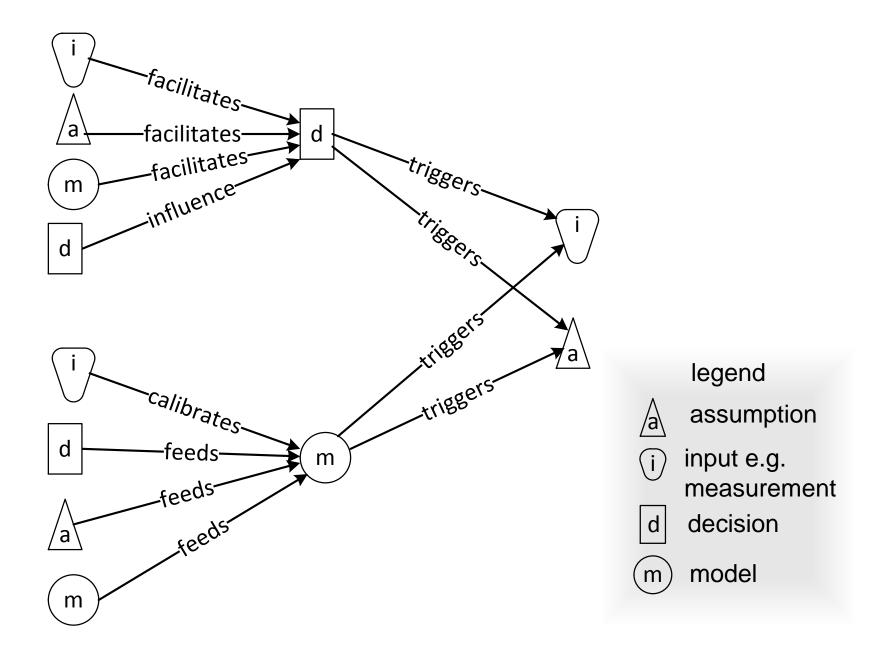


Example Graph for Web Shop



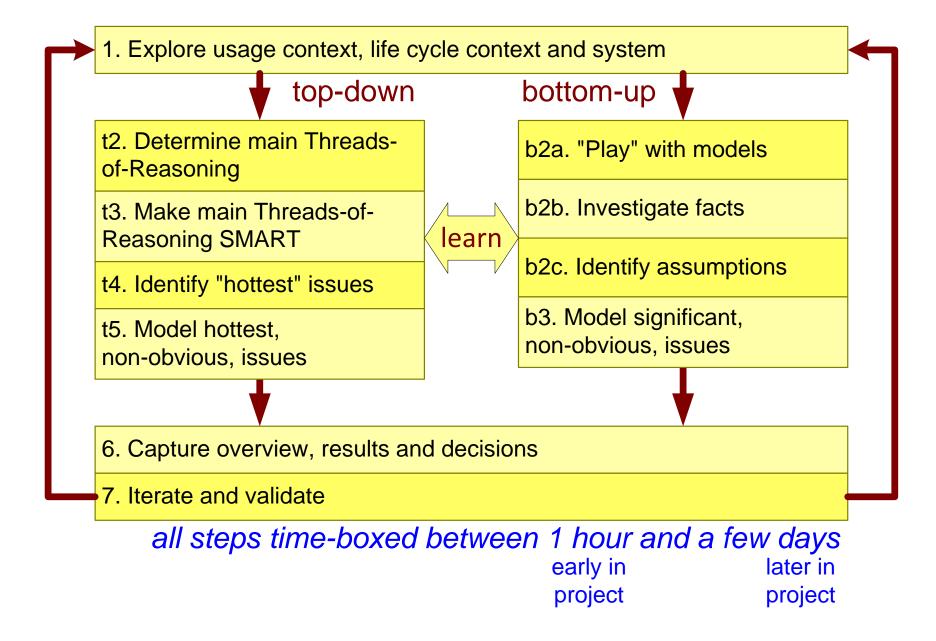


Relations: Decisions, Models, Inputs and Assumptions



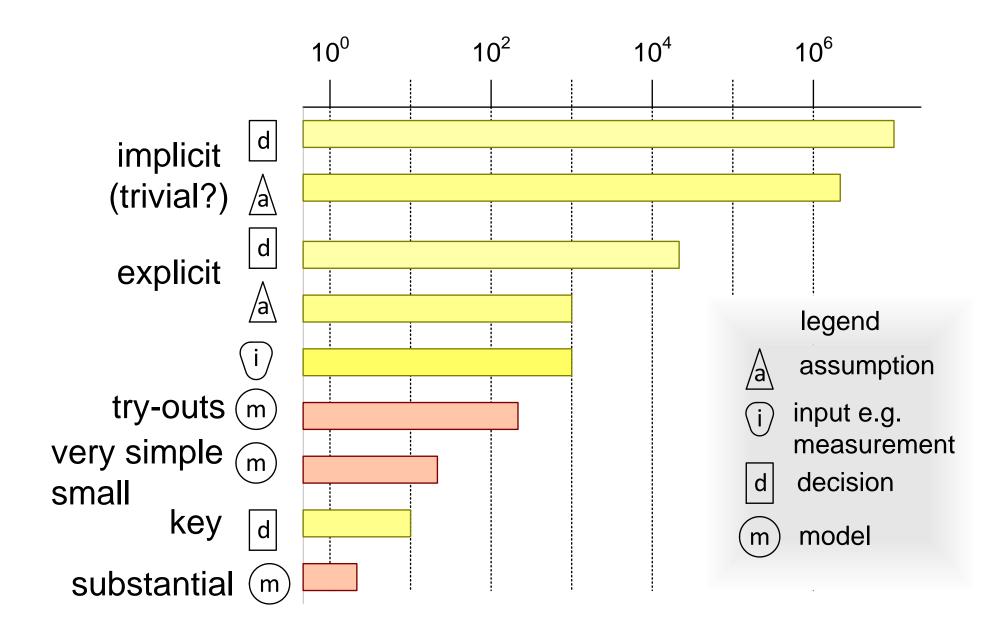


Reasoning Approach



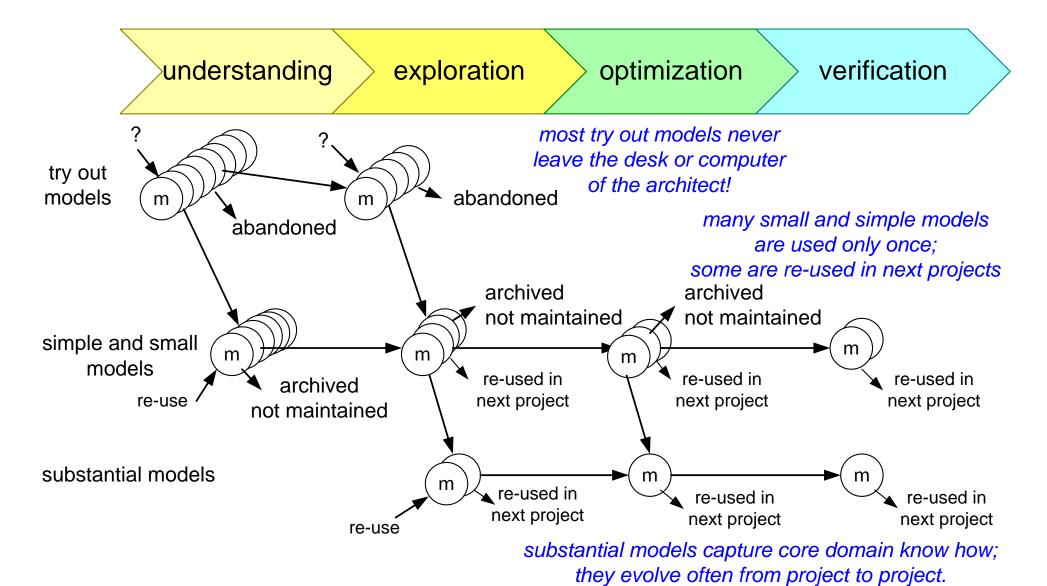


Frequency of Assumptions, Decisions and Modeling





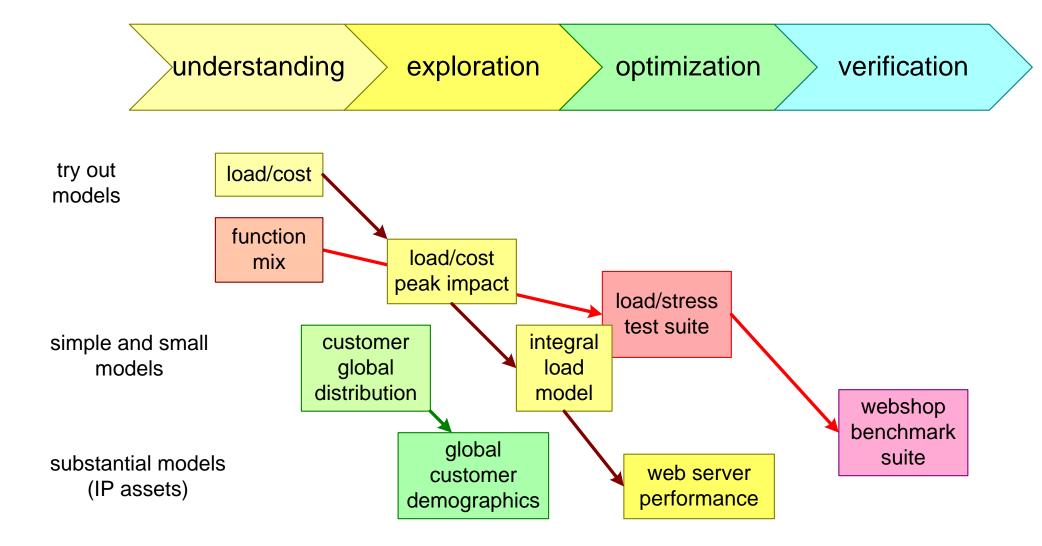
Life Cycle of Models





creation and evolution of intellectual property assets

Examples of Life Cycle of Models





Architecting System Performance; Defining Performance

by Gerrit Muller TNO-ESI, University of South-Eastern Norway]

e-mail: gaudisite@gmail.com

www.gaudisite.nl

Abstract

Performance is a broad term. Each domain has its own key performance parameters. Performance can be used to indicate time-oriented performance, such as response time, throughput, or productivity. However, more broadly, it may be used for aspects like image quality, spatial performance (f.i. positioning accuracy), energy or power properties, sensitivity and specificity of algorithms, or reliability and availability.

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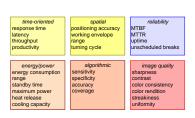
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Performance Attributes

time-oriented response time latency throughput productivity

spatial
positioning accuracy
working envelope
range
turning cycle

reliability
MTBF
MTTR
uptime
unscheduled breaks

energy/power
energy consumption
range
standby time
maximum power
heat release
cooling capacity

algorithmic sensitivity specificity accuracy coverage

image quality
sharpness
contrast
color consistency
color rendition
streakiness
uniformity



Defining Performance

```
performance is a function of:
  context
                depends on individual human characteristics
    perception
    circumstances
    operation of interest
  system of interest
    specification
                     generic, valid for the class of systems
    design
                     normal and special cases
                        (worst case, degraded, exceptions, ...)
    configuration
    version
                     instance specific
    history
```



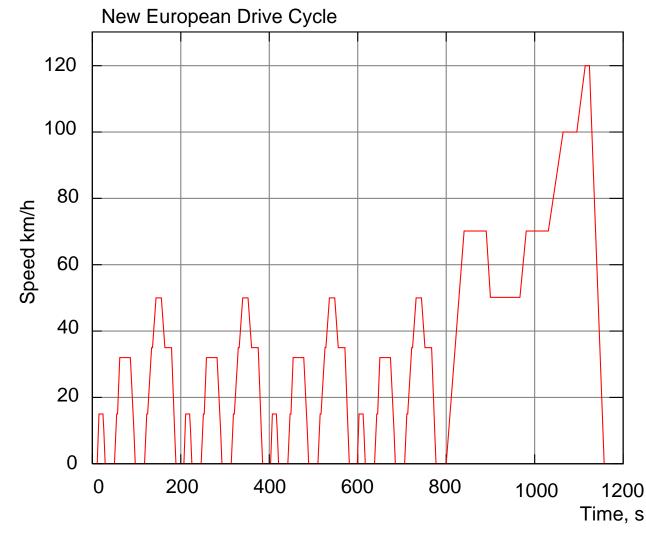
¹a use case in this context is rich (includes quantifications) and broad (covers the operation of interest, not a single function)

Example EV Range Definition

Electric Vehicle Driving Range

Range = f(
v(t),
Circumstances,
Driving style,
Car load,
Charging state,
Battery age)

A quantified Use Case defines under what circumstances the EV will achieve the specified range.

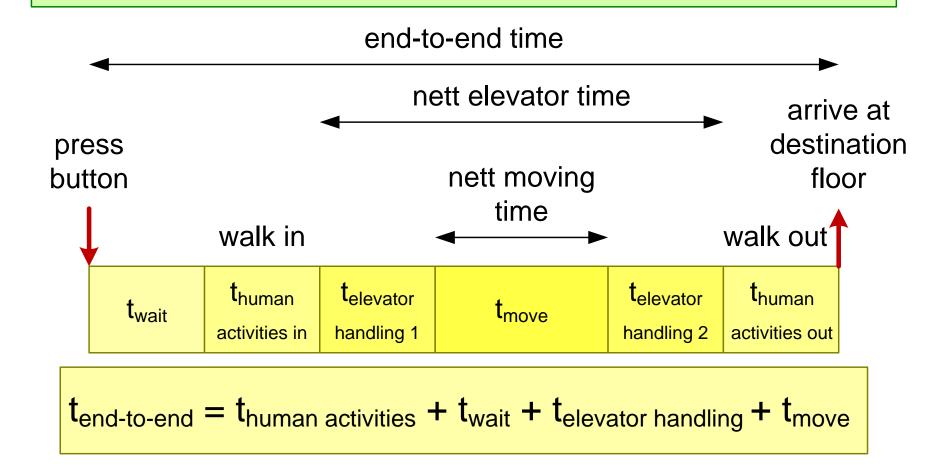


http://en.wikipedia.org/wiki/New_European_Driving_Cycle#/media/File:New_European_Driving_Cycle.svg
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End-to-End Performance

The **end-to-end** performance is the relevant performance as the **stakeholder** experiences it: from **initial trigger** to **final result**.





Architecting System Performance; Measuring

by Gerrit Muller TNO-ESI, University of South-Eastern Norway]

e-mail: gaudisite@gmail.com

www.gaudisite.nl

Abstract

Measuring is an essential part of architecting performance. Measurements provide quantified insight in actual behavior and performance. In this presentation, we discuss measuring, benchmarking, and instrumentation.

Distribution

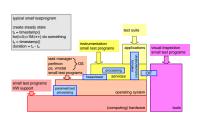
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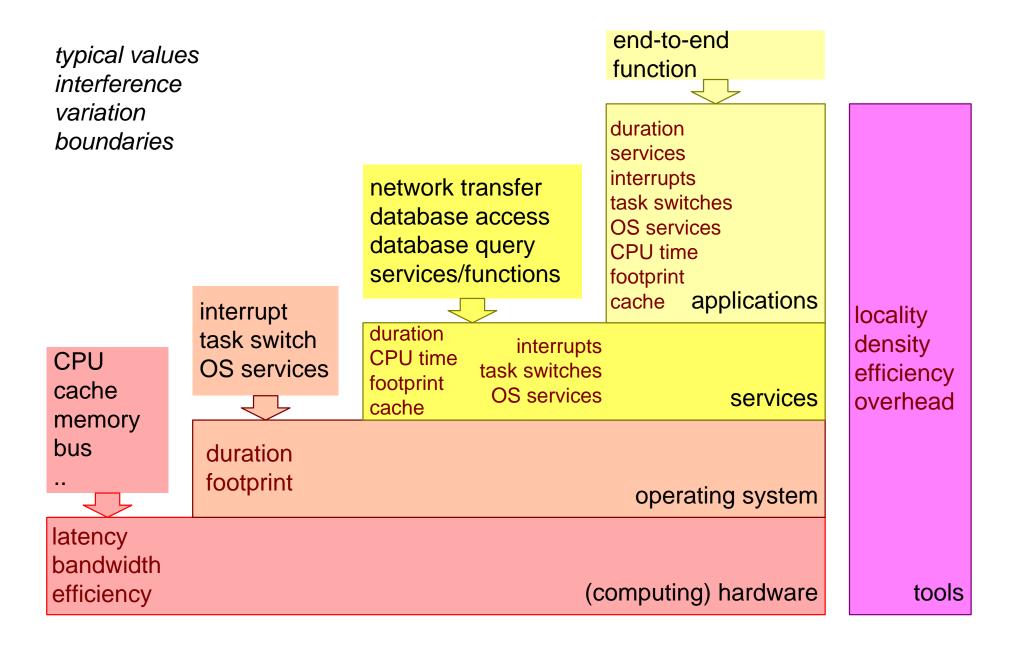
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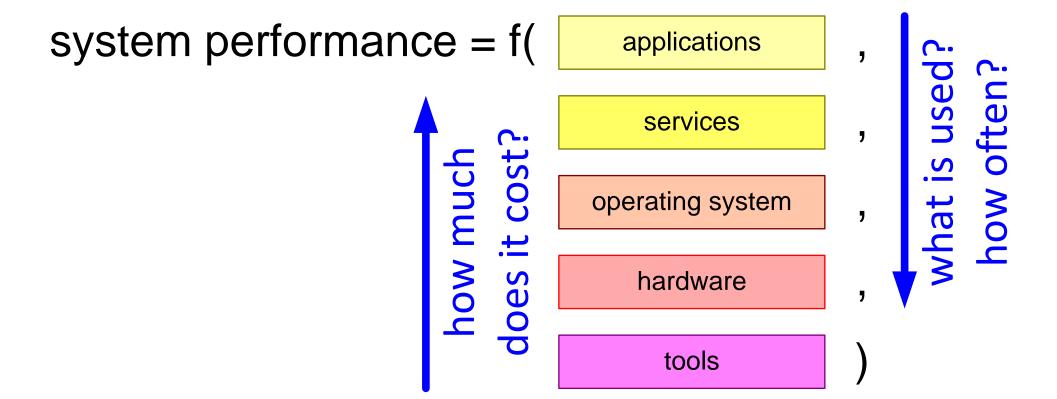


Performance Attributes in the Benchmark Stack





Performance as Function of the Layers



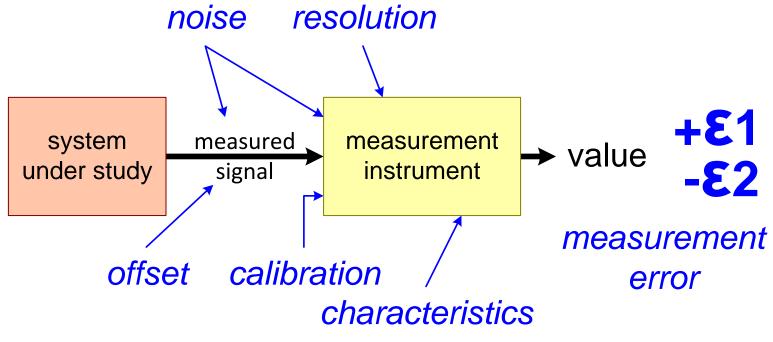


Example μ Benchmarks for Software

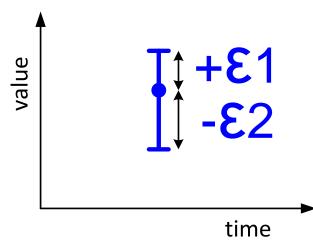
| | infrequent operations, often time-intensive | often repeated operations |
|-------------------------|---|--|
| database | start session finish session | perform transaction query |
| network, I/O | open connection close connection | transfer data |
| high level construction | component creation component destruction | method invocation same scope other context |
| low level construction | object creation object destruction | method invocation |
| basic programming | memory allocation memory free | function call loop overhead basic operations (add, mul, load, store) |
| os | task, thread creation | task switch interrupt response |
| HW | power up, power down boot | cache flush low level data transfer |



Measurement Errors and Accuracy



measurements have stochastic variations and systematic deviations resulting in a range rather than a single value





Be Aware of Error Propagation

$$t_{duration} = t_{end} - t_{start}$$

$$t_{start} = 10 + / - 2 \mu s$$

$$t_{end} = 14 + /- 2 \mu s$$

$$t_{duration} = 4 +/- ? \mu s$$

systematic errors: add linear

stochastic errors: add quadratic



Intermezzo Modeling Accuracy

Measurements have

stochastic variations and systematic deviations

resulting in a range rather than a single value.

The inputs of modeling,

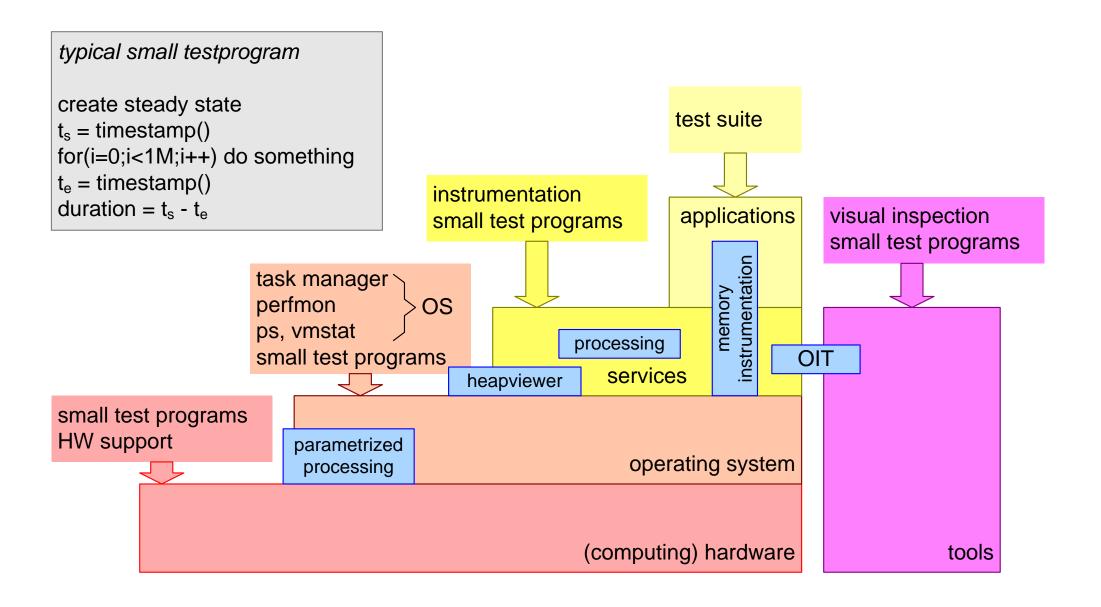
"facts", assumptions, and measurement results,

also have stochastic variations and systematic deviations.

Stochastic variations and systematic deviations propagate (add, amplify or cancel) through the model resulting in an output range.



Tools and Instruments in the Benchmark Stack





Architecting System Performance; Resource Management

by Gerrit Muller TNO-ESI, University of South-Eastern Norway]

e-mail: gaudisite@gmail.com

www.gaudisite.nl

Abstract

The management of the resources largely determines system performance. This document discusses concepts related to resource management, such as caching, concurrency, and scheduling.

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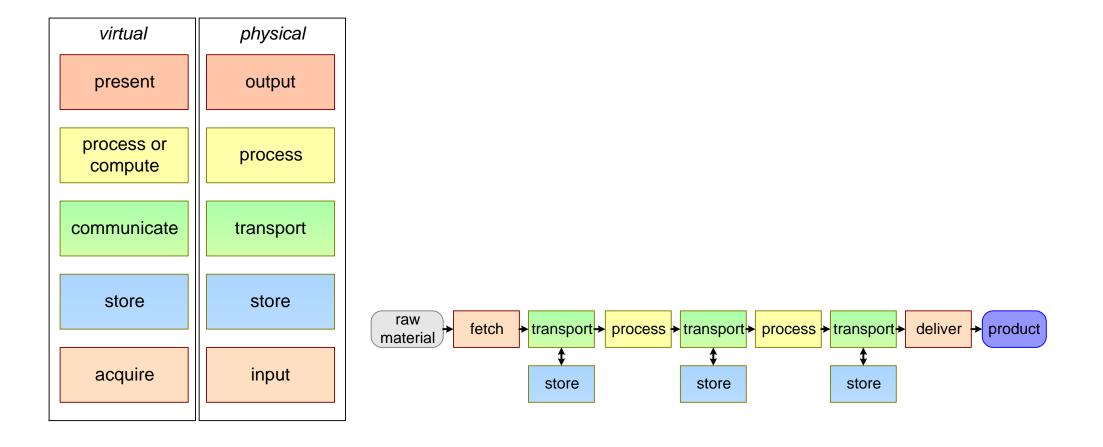
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Generic Resource Model





Design Considerations for Resource Management

Performance depends on resource utilization and management.

The design of the logistics, how does EMI¹ flow through the resources, is critical.

Critical design aspects are:

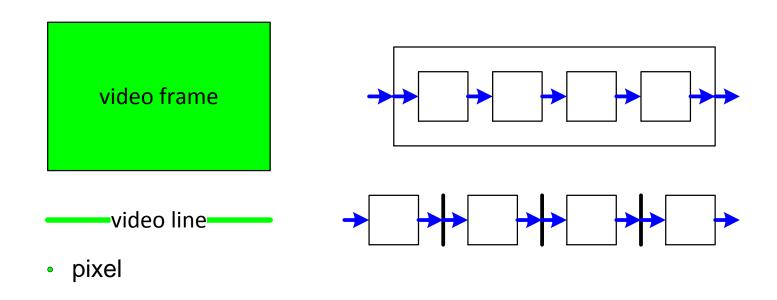
- concurrency (parallelism, pipelining)
- granularity of EMI
- scheduling (allocation of resources)

¹Energy Material Information



Granularity as Key Design Choice

unit of or unit of or synchronization <> unit of or processing <> unit of or l/O



fine grain: coarse grain: flexible rigid high overhead low overhead



Size versus Performance Trade off

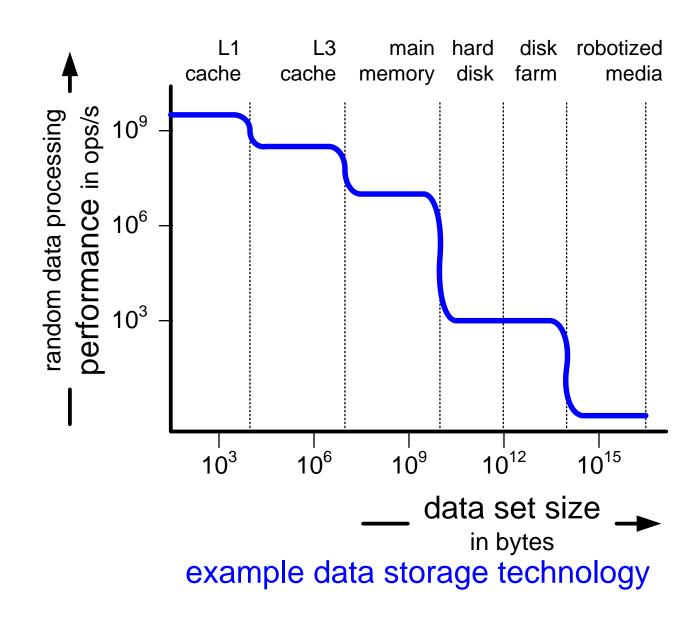
small capacity

fast technology small expensive

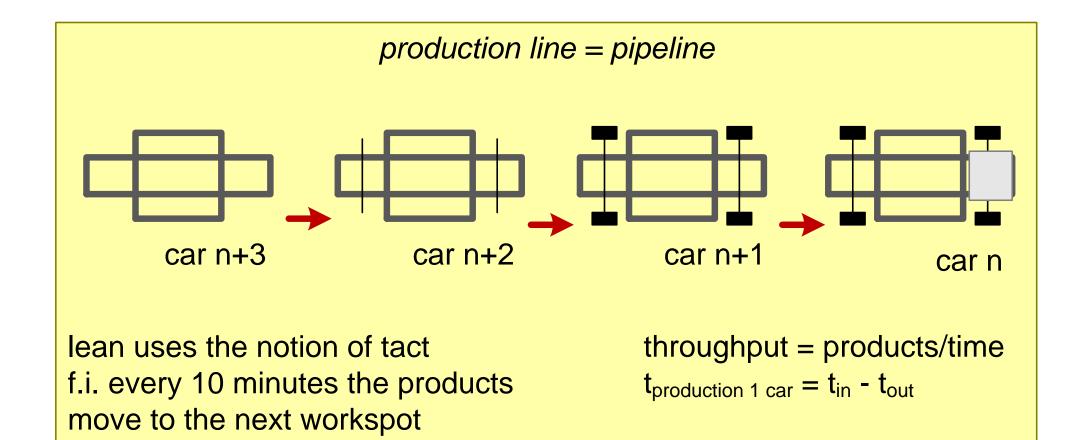
large capacity

slow technology large low cost

staircase effect: performance and size are non-linear with thresholds

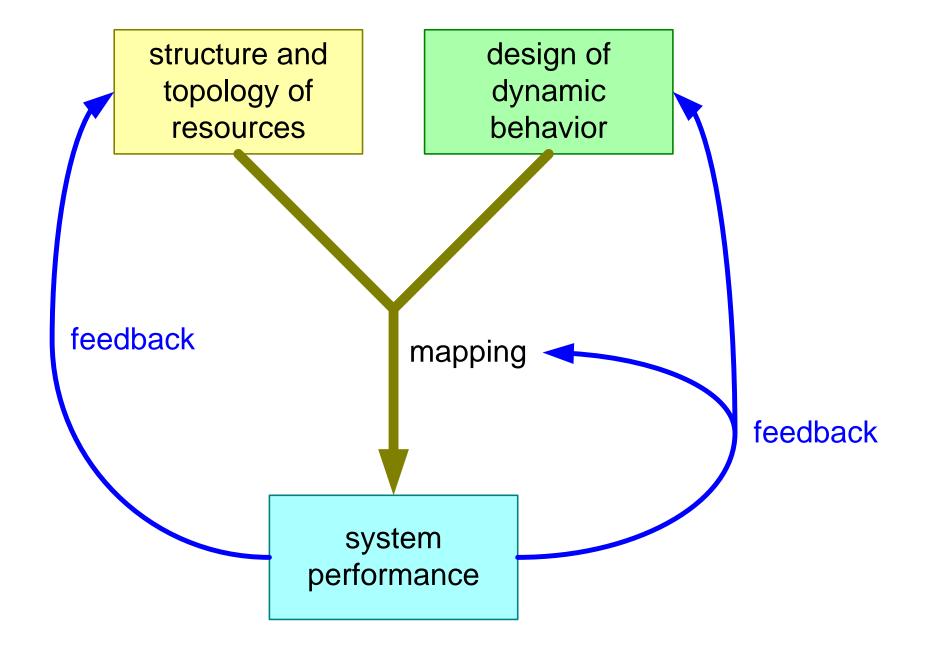


Pipeline pattern





Y-chart Pattern





Performance Pitfalls and Resource Management

Overhead (control, handling)

Starvation (underrun)

Saturation/stagnation (overrun)

Variation (duration, quality)

Serialization

Interference with other work

Unnecessary conversions or adaptations



Architecting System Performance; Greedy and Lazy Patterns

by Gerrit Muller TNO-ESI, University of South-Eastern Norway]

e-mail: gaudisite@gmail.com

www.gaudisite.nl

Abstract

Greedy and lazy are two opposite patterns in performance design. An extreme application of both patterns is start-up, where greedy starts as much as possible, and lazy as little as possible.

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logo

TBD

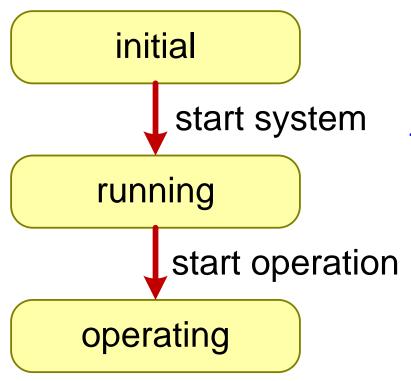
Greedy and Lazy Patterns

| | lazy | greedy |
|---------------|--|---|
| | (on demand, pull) | (push, forecast) |
| what | do nothing until someone needs it | prepare time consuming operations, when resources are idle |
| benefits | no resource usage unless needed | results are available immediately |
| disadvantages | time to result depends on execution time | some resource use is wasted |
| when | default | to achieve required performance (explore other concepts too!) |

this pattern applies to all domains (IT, goods flow, energy)



Start up of Systems as Example

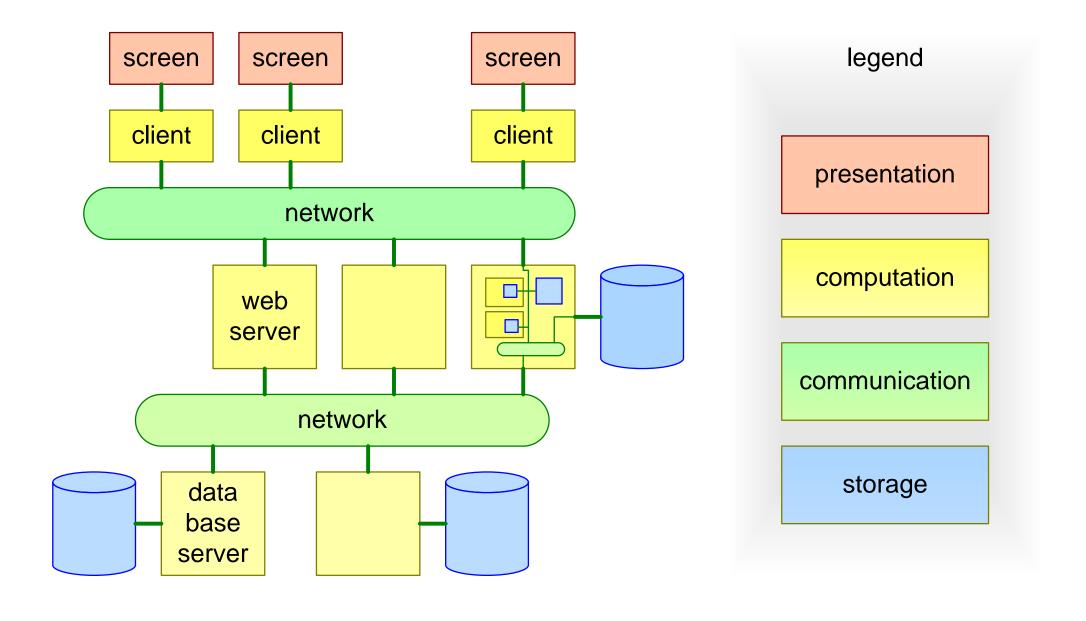


How much time does it take to start a laptop with Windows?

How much time does it take to start an application (e.g. Word)?



Example from Cloud Applications



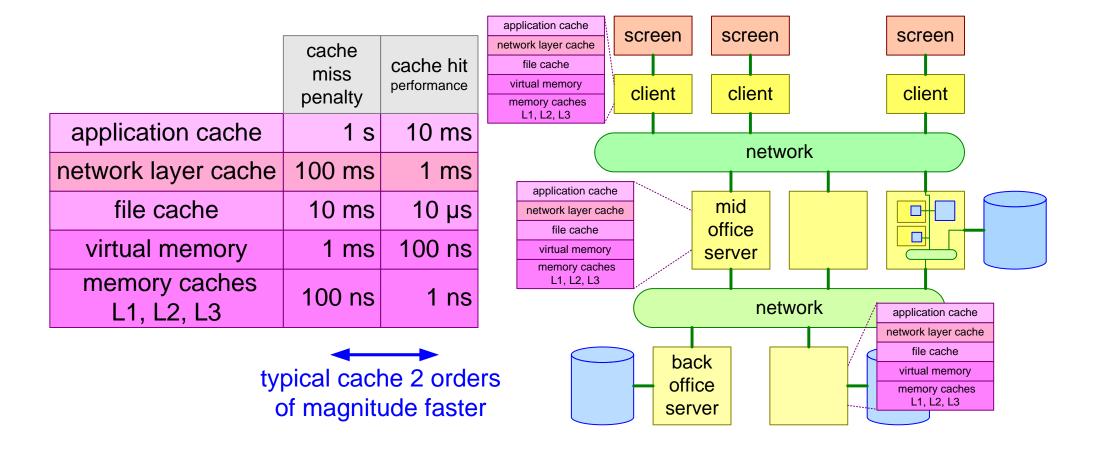


Caching Pattern (Physical Grab Stock)

performance issues solution patterns design parameters frequently used subset in fast local storage caching algorithm long latency (mass) storage low latency storage location long latency communication less communication cache size large chunks (less overhead) overhead communication chunk size processing once (keep results) resource intensive processing format



Many Layers of Caching





Disadvantages of Caching Pattern

robustness for application changes

ability to benefit from technology improvements

robustness for changing context (e.g. scalability)

robustness for concurrent applications

failure modes in exceptional user space

These patterns increase complexity and coupling.

Use only when necessary for performance.



Architecting System Performance; Scheduling

by Gerrit Muller TNO-ESI, University of South-Eastern Norway]

e-mail: gaudisite@gmail.com

www.gaudisite.nl

Abstract

Scheduling plays a crucial role in resource allocation to get desired system performance. This document discusses local and global scheduling.

Distribution

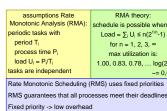
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Single Resource Scheduling

Scheduling of time critical operations on a single resource:

Earliest Deadline First

optimal

complex to realize

Rate Monotonic Scheduling

no full utilization

simple to realize



Earliest Deadline First

| Determine deadlines | in Absolute time (CPU cycles or msec, etc.) |
|---|---|
| Assign priorities | Process that has the earliest deadline gets the highest priority (no need to look at other processes) |
| Constraints | Smart mechanism needed for Real-Time determination of deadlines Pre-emptive scheduling needed |

EDF = Earliest Deadline First

Earliest Deadline based scheduling for (a-)periodic Processing

The theoretical limit for any number of processes is 100% and so the system is schedulable.

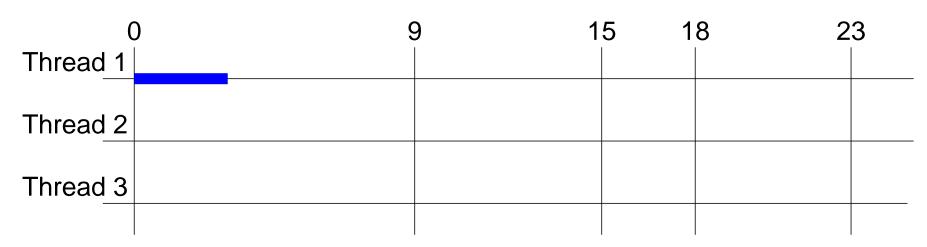


Exercise Earliest Deadline First (EDF)

Calculate loads and determine thread activity (EDF)

| Thread | Period = deadline | Processing | Load |
|----------|-------------------|------------|-------|
| Thread 1 | 9 | 3 | 33.3% |
| Thread 2 | 15 | 5 | |
| Thread 3 | 23 | 5 | |
| | | | |

Suppose at t=0, all threads are ready to process the arrived trigger.





Rate Monotonic Scheduling

| Determine deadlines (period) | in terms of Frequency or Period (1/F) |
|--|---|
| Assign priorities | Highest frequency (shortest period) ==> Highest priority |
| • Constraints | Independent activities Periodic Constant CPU cycle consumption Assumes Pre-emptive scheduling |

RMS = Rate Monotonic Scheduling

Priority based scheduling for Periodic Processing of tasks with a guaranteed CPU - load



Exercise Rate Monotonic Scheduling (RMS)

Calculate loads and determine thread activity (RMS)

| Thread | Period = deadline | Processing | Load |
|----------|-------------------|------------|-------|
| Thread 1 | 9 | 3 | 33.3% |
| Thread 2 | 15 | 5 | |
| Thread 3 | 23 | 5 | |
| | | | |

Suppose at t=0, all threads are ready to process the arrived trigger.

| (|) | 9 1 | 5 1 | 8 2 | 3 |
|----------|---|-----|-----|-----|---|
| Thread 1 | | | | | |
| - | | | | | |
| Thread 2 | | | | | |
| Thread 3 | | | | | |
| | | | | | |



RMS-RMA Theory

```
assumptions Rate
Monotonic Analysis (RMA):
periodic tasks with
period T_i
process time P_i
load U_i = P_i/T_i
tasks are independent
```

```
RMA theory:

schedule is possible when:

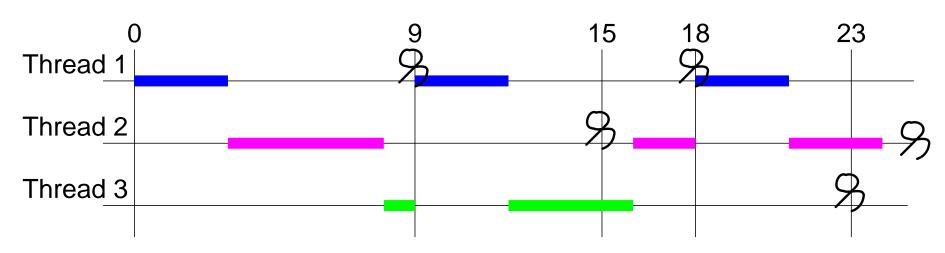
Load = \sum_{i} U_{i} \le n(2^{1/n}-1)
for n = 1, 2, 3, \infty
max utilization is:
1.00, 0.83, 0.78, ... log(2)
\sim= 0,69
```

Rate Monotonic Scheduling (RMS) uses fixed priorities
RMS guarantees that all processes meet their deadlines
Fixed priority -> low overhead



Answers: loads and thread activity (EDF)

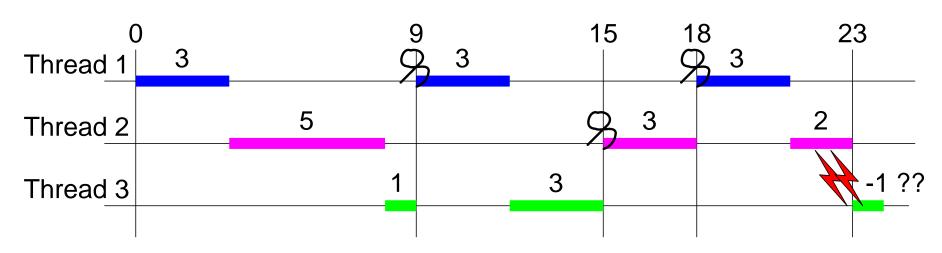
| Thread | Period = deadline | Processing | Load |
|----------|-------------------|------------|-------|
| Thread 1 | 9 | 3 | 33.3% |
| Thread 2 | 15 | 5 | 33.3% |
| Thread 3 | 23 | 5 | 21.7% |
| | | | 88.3% |





Answers: loads and thread activity (RMS)

| Thread | Period = deadline | Processing | Load |
|----------|-------------------|------------|-------|
| Thread 1 | 9 | 3 | 33.3% |
| Thread 2 | 15 | 5 | 33.3% |
| Thread 3 | 23 | 5 | 21.7% |
| | | | 88.3% |





From Simple-Local to Complex-Global

A perspective on dynamic behavior is to view the system as set of periodic behaviors.

Periodic behavior is easier to model and analyze, e.g. using RMS and RMA.

Modern systems and Systems of Systems consists of complex networks of concurrent resources.

Typically, a combination of more advanced **global** scheduling is combined with simple **local** scheduling.



Architecting System Performance; Robust Performance

by Gerrit Muller TNO-ESI, University of South-Eastern Norway]

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Abstract

Performance should be robust. The performance should be reproducable and it should be well-behaving in extreme conditions.

Distribution

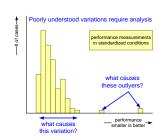
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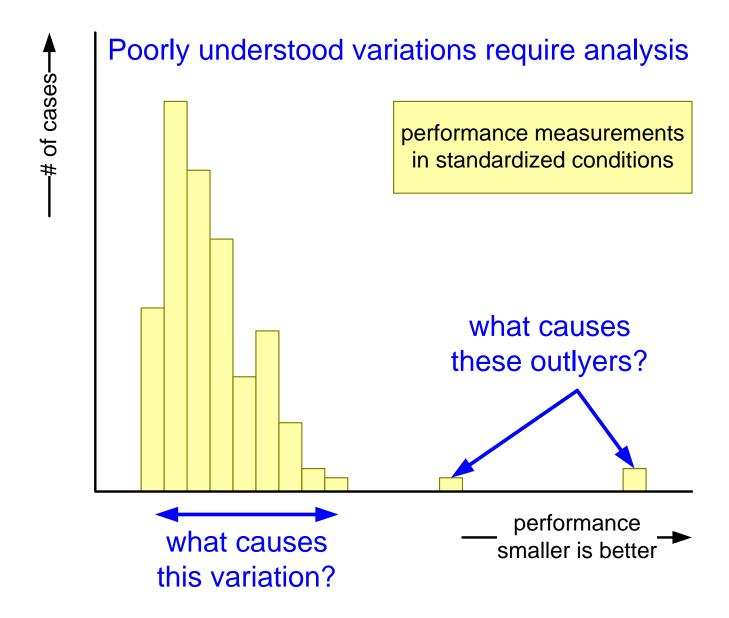
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Variations are Suspect





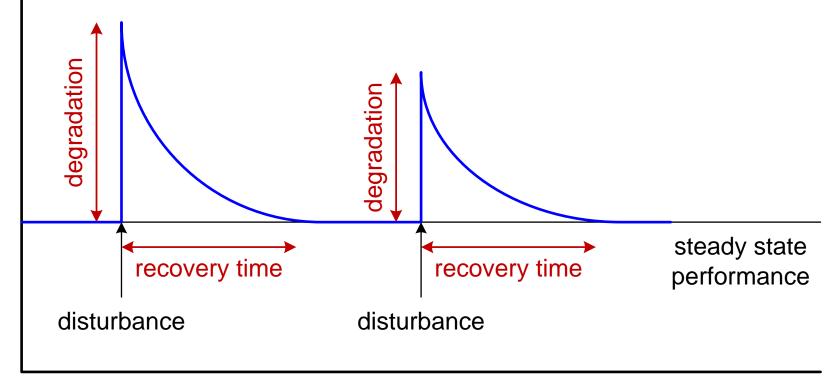
Coping with Disturbances



How does the system respond to disturbances?

How quickly does it recover?

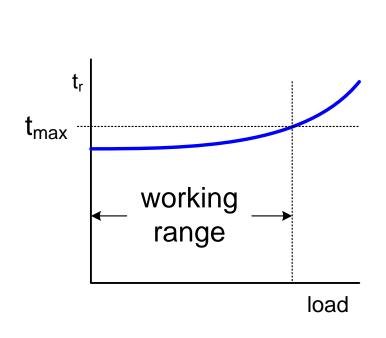
How far does performance degrade?

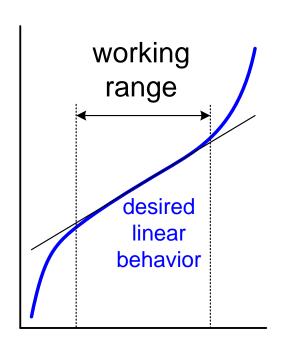






Measure to Validate Working Range





A system design assumption is often:
the performance of this function
{ is constant | is linear | doesn't exceed x | ...}

The working range is the interval where this assumption holds



Validate Understanding of System Performance

Characterize the system

use the system in varying conditions measure performance as function of the conditions

Stress testing

where does the design fail? (go beyond specified limits)

Load testing

keep the system in heavy load condition observe how it keeps performing measure variations

(Accelerated) Lifetime testing

age the system observe how it keeps performing



Bloating, Waste, and Value

by Gerrit Muller TNO-ESI, University of South-Eastern Norway]

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Abstract

A threat to performance is the combination of feature creep and technical debt. This combination causes bloating of the design. In Lean terms, the combination causes waste. A crucial question is where is the value, and is the value in balance with the potential degradation of performance.

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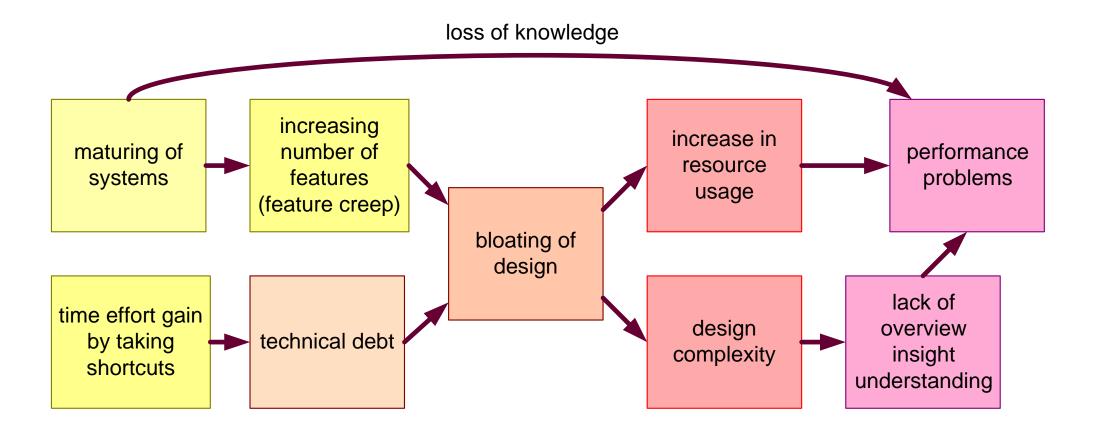
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From Feature Creep to Performance Problems





Technical Debt

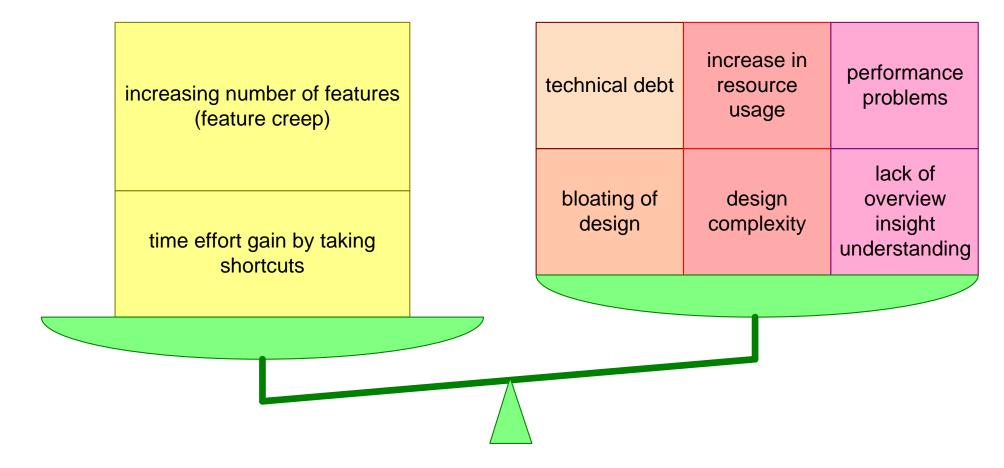
Technical Debt is a **metaphor** used within the software industry to communicate the consequences of **pragmatic design decisions** deviating from the **intended design** of a system

from: http://gaudisite.nl/INCOSE2016_Callister_Andersson_SMARTtechnicalDebt.pdf based on Cunningham http://c2.com/doc/oopsla92.html



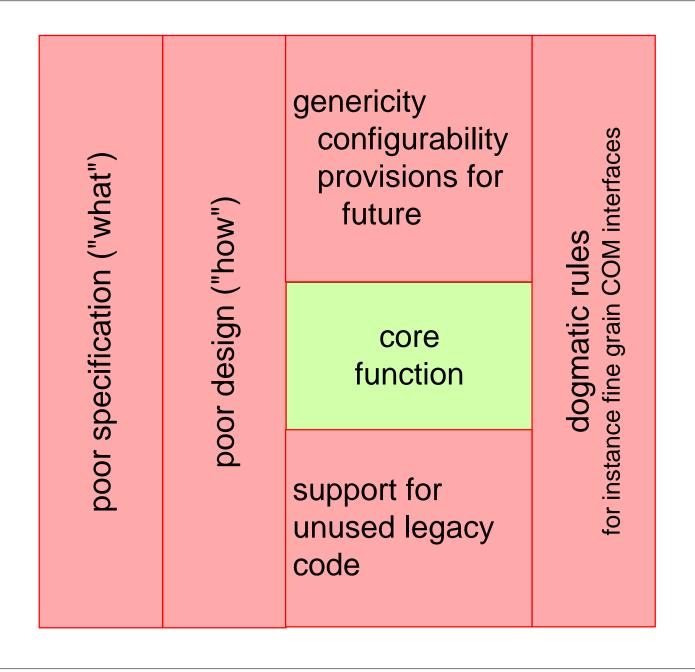
Value versus Performance Degradation

Are benefits (value) in balance with the costs (such as performance degradation)?





Exploring bloating: main causes



legenda

overhead

value



Necessary functionality ≫ the intended regular function

testing

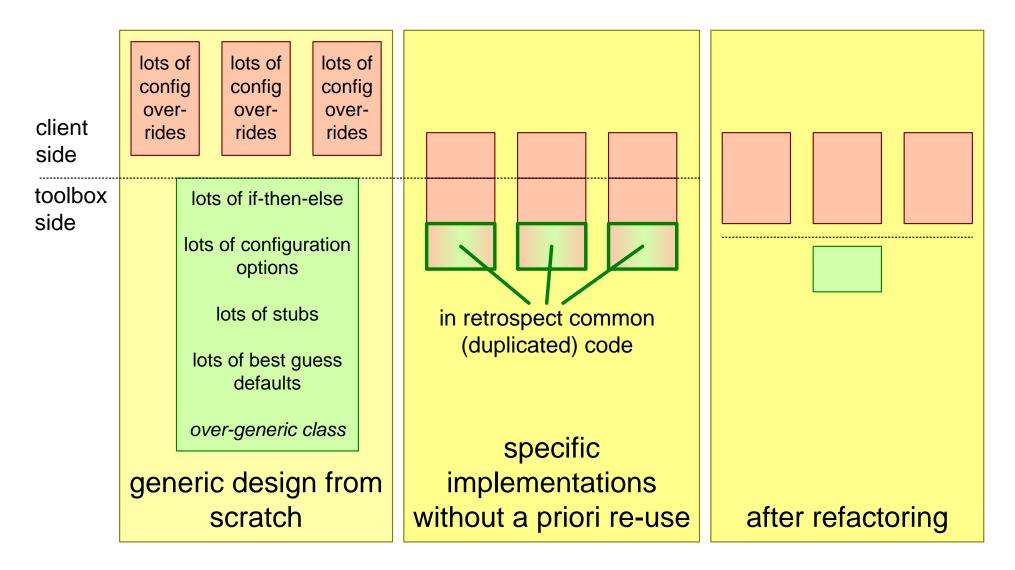
regular functionality

instrumentation diagnostics tracing asserts

boundary behavior: exceptional cases error handling



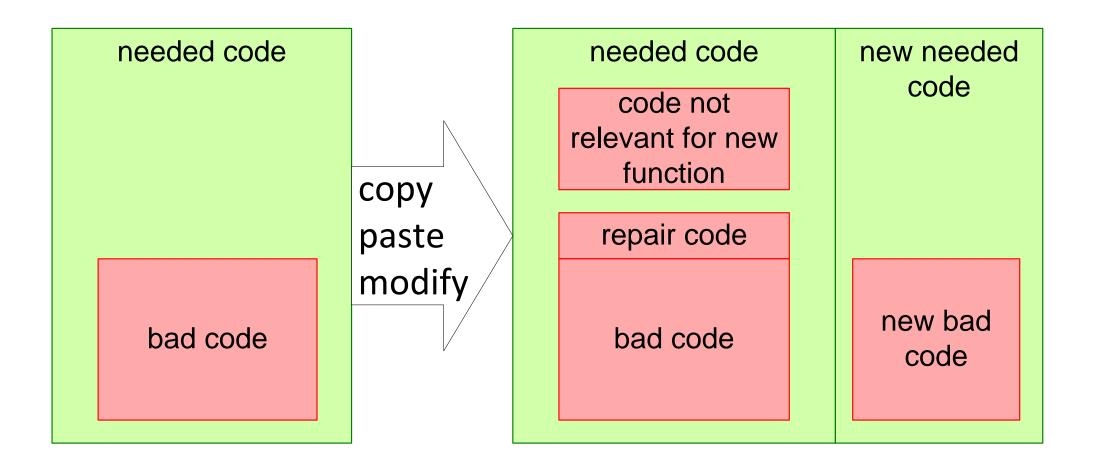
The danger of being generic: bloating



"Real-life" example: redesigned *Tool* super-class and descendants, ca 1994



Shit propagation via copy paste





Example of shit propagation

copy

paste

```
Class Old:
    capacity = startCapacity
    values = int(capacity)
    size = 0

def insert(val):
    values[size]=val
    size+=1
    if size>capacity:
        capacity*=2
    relocate(values,
        capacity)
```

```
Class New:

capacity = 1
values = int(capacity)
size = 0

def insert(val):
values[size]=val
size+=1
capacity+=1
relocate(values,
capacity)

capacity)
```

```
Class DoubleNew:

capacity = 1

values = int(capacity)

size = 0

def insert(val):

values[size]=val

size+=1

capacity+=1

relocate(values,

capacity)

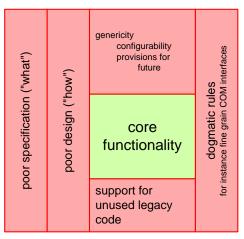
def insertBlock(v,len):

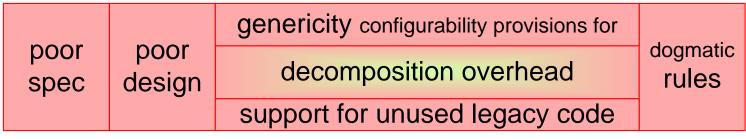
for i=1 to len:

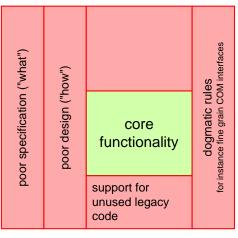
insert(v[i])
```

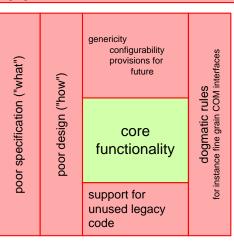


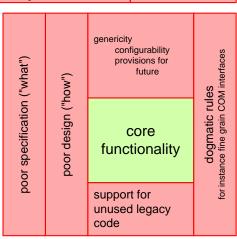
Bloating causes more bloating











legenda

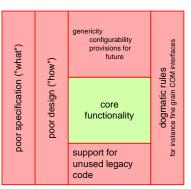
overhead

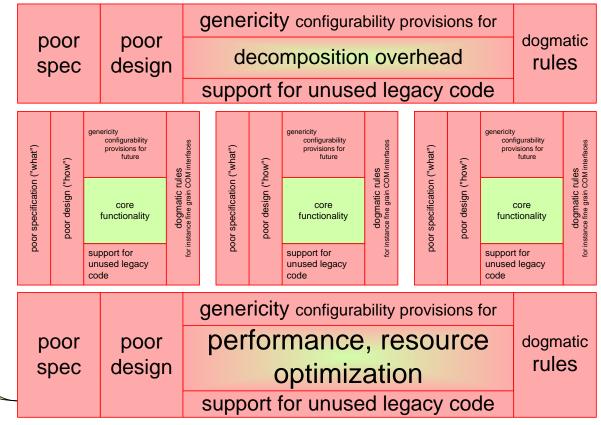
value



Causes even more bloating...

Bloating causes performance and resource problems.
Solution: special measures: memory pools, shortcuts, ...







legenda

overhead

value