

# Algorithms for context prediction in ubiquitous systems

Lecture in WS08/09

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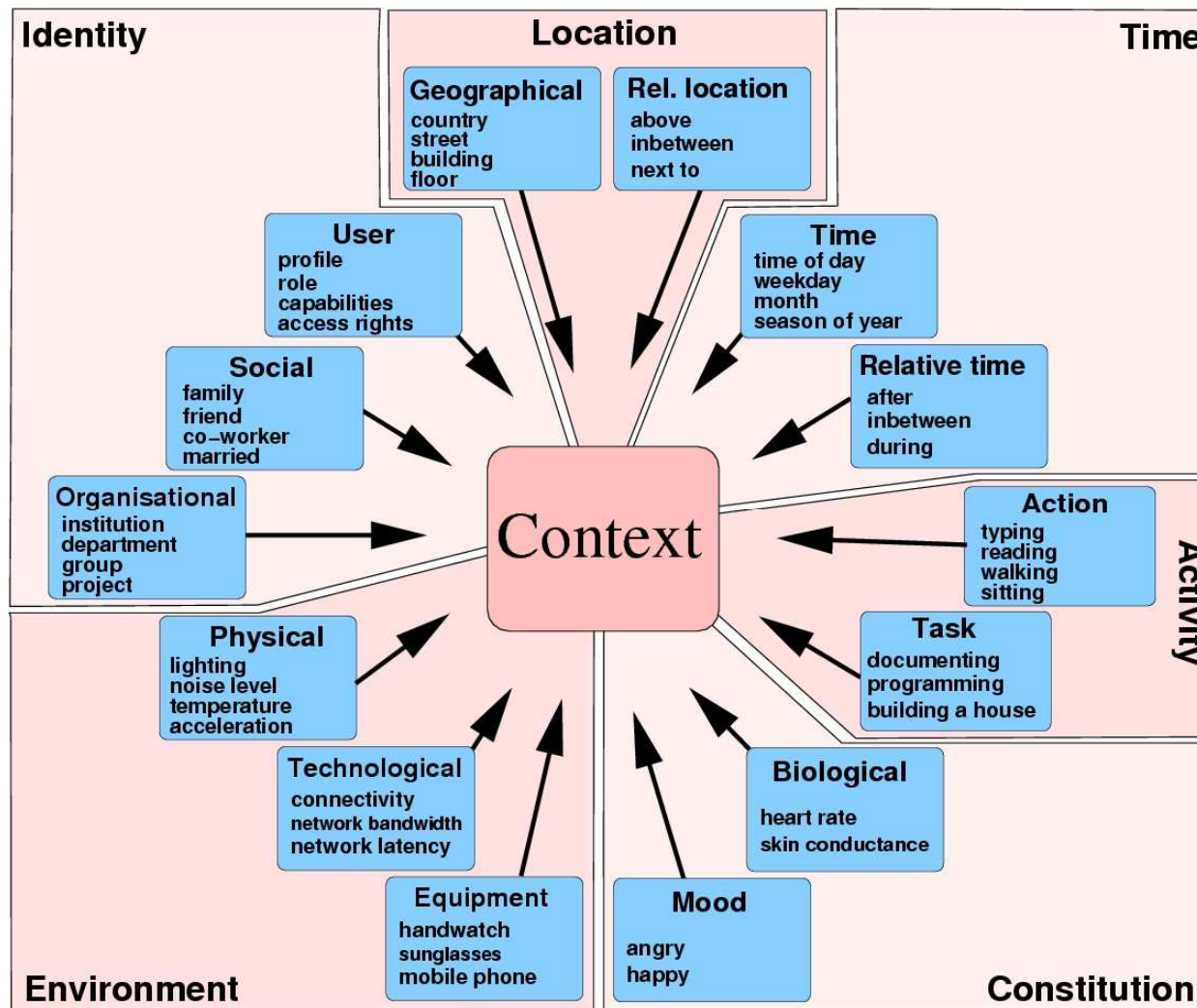
**Michael Beigl**

TU Braunschweig

Institute of Operating Systems  
and Computer Networks

[www.ibr.cs.tu-bs.de/dus](http://www.ibr.cs.tu-bs.de/dus)

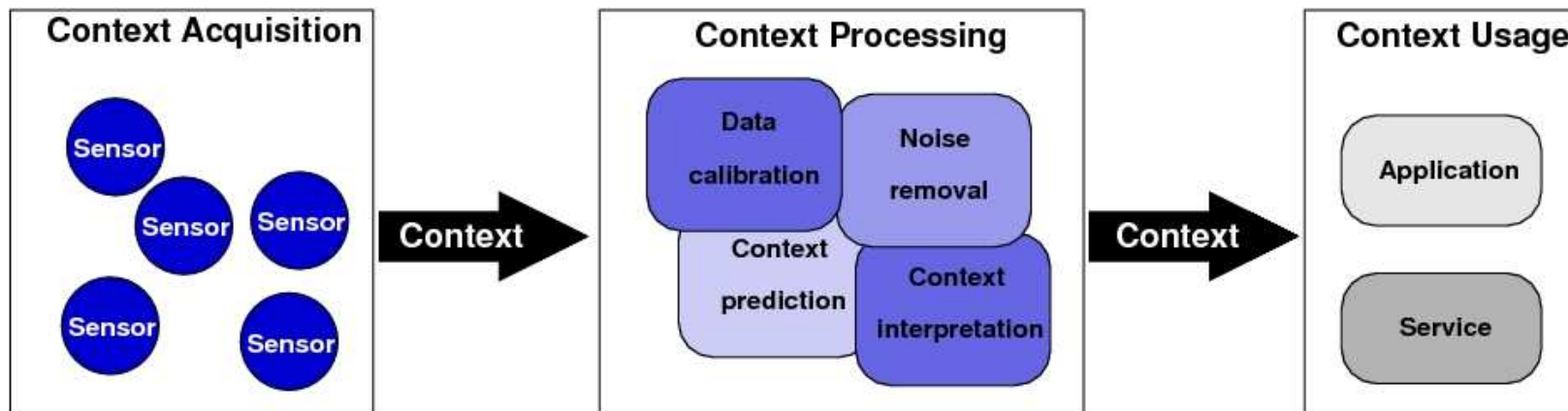
# What is context?



# Example sensors

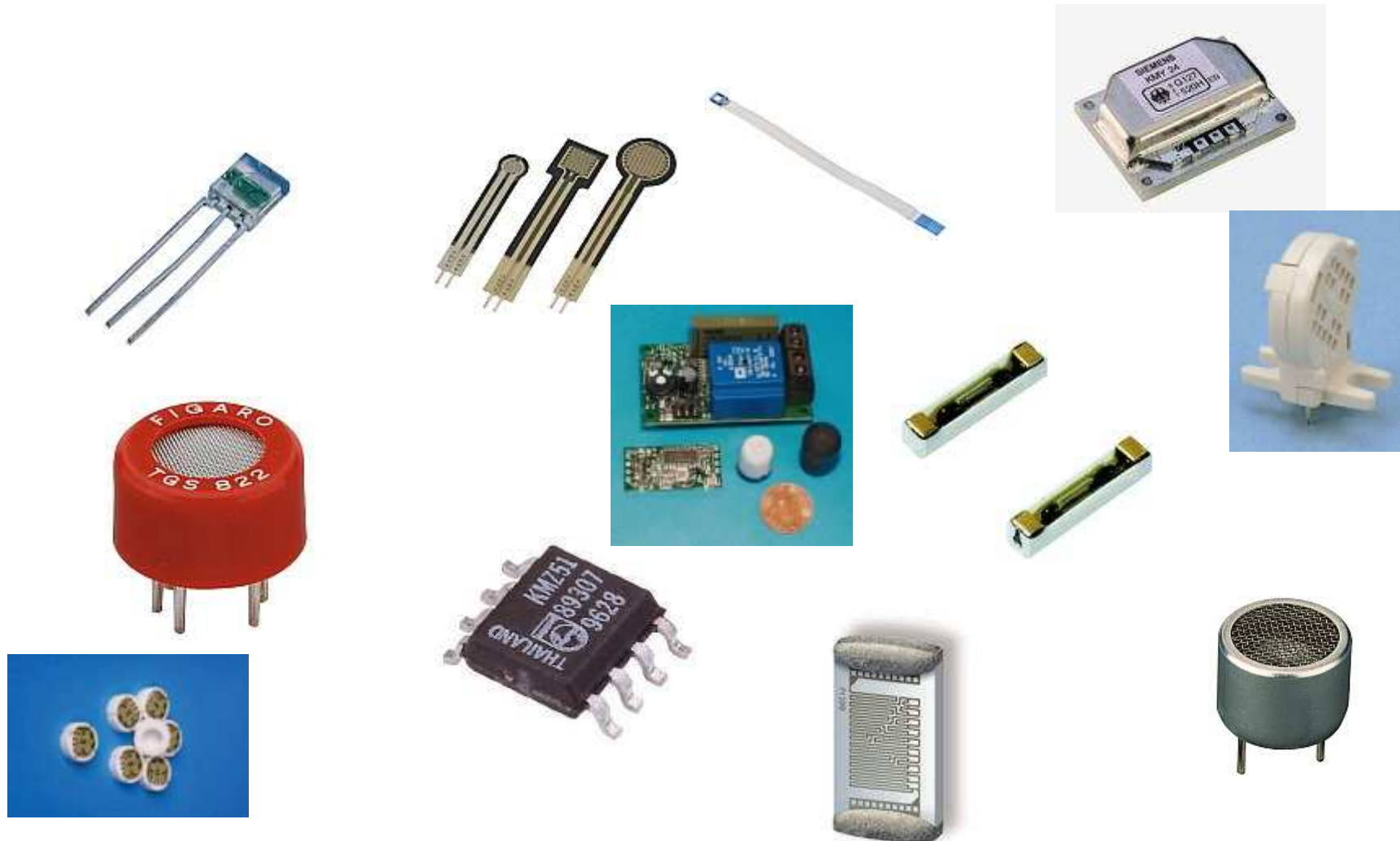
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## Context processing and context utilisation



# Example sensors

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Stephan Sigg

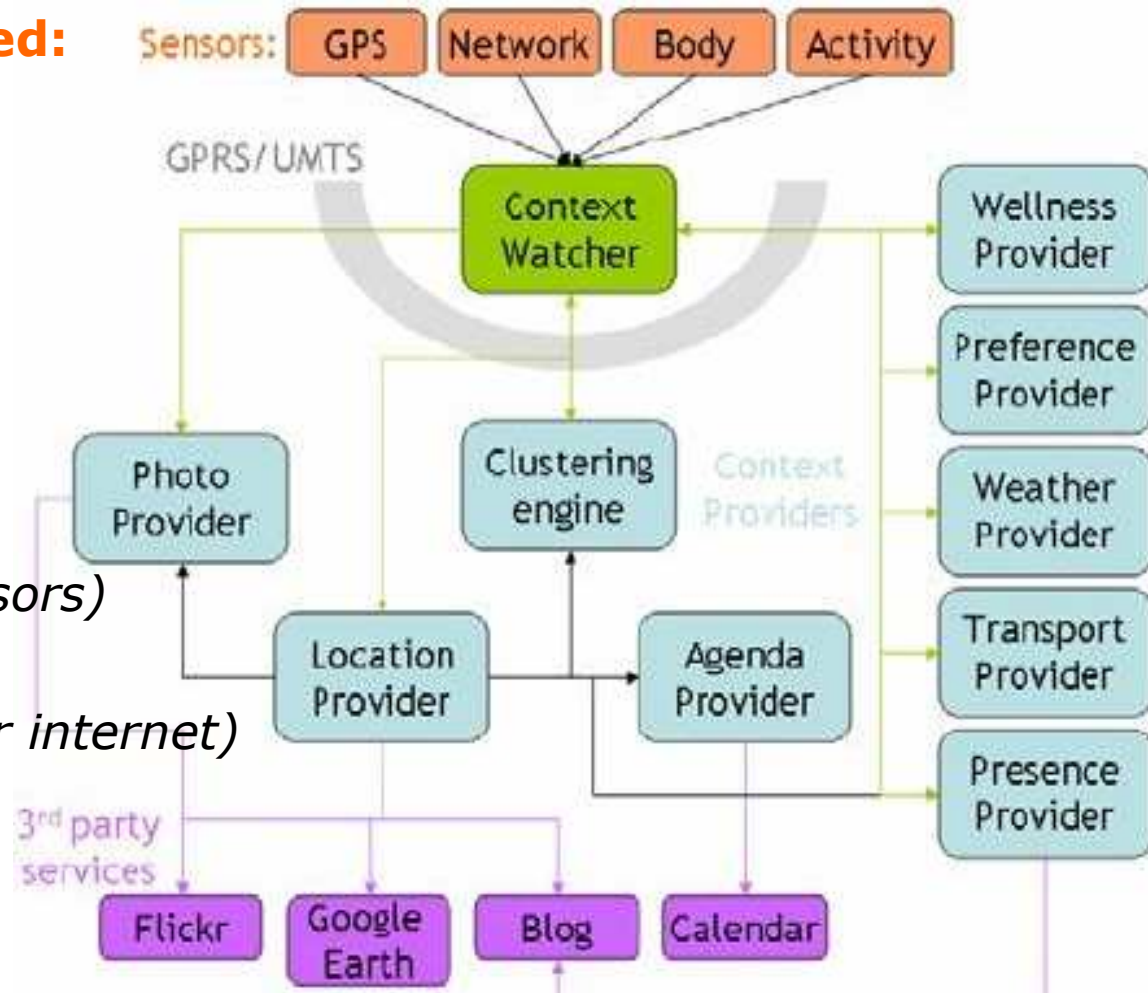
Context prediction algorithms, Winterseil

1-4

# Example: Context Watcher

## Context sources utilised:

- Location
  - (*GSM cell-ID; GPS*)
- Mood
  - (*user input*)
- activity
  - (*calender based*)
- Bio-data
  - (*heart and foot sensors*)
- Weather
  - (*location based over internet*)
- Photo/picture
  - (*camera*)



# Example: Context Watcher



## Context Data

cell id: 10571  
altitude: 59.4  
speed: 115.1 km/h  
course: 246.6  
pos: (52.279, 6.503)  
range: 1 m  
street: E30  
postal code: 7462  
city: Rijssen (NL)



## Johan's blog

📅 Saturday, March 24, 2007

### A day in Papendrecht

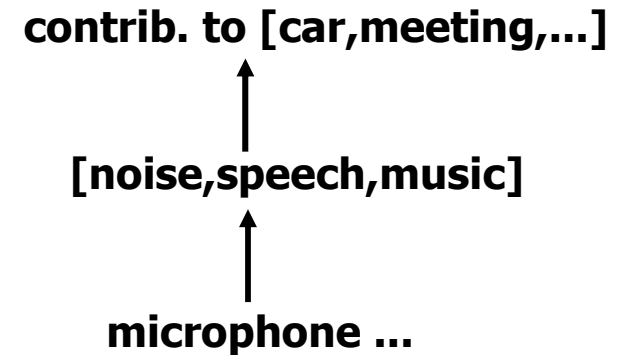
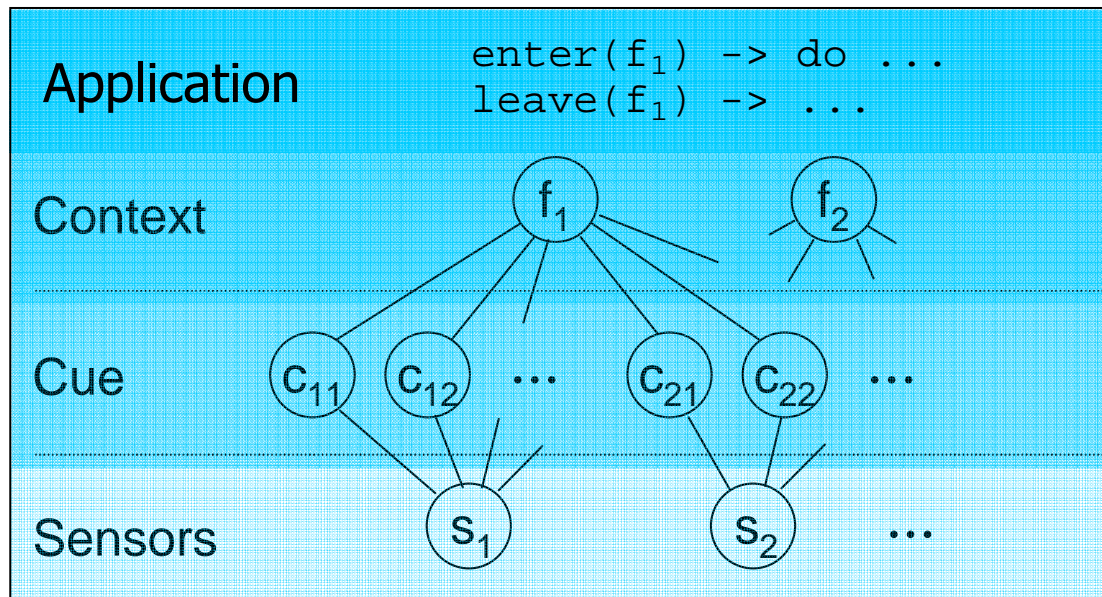
The weather that I enjoyed today: it has been rather cloudy in Alblasserdam, 1/9°C, with a relative humidity of 93%, a gentle breeze was blowing from north to northeast. The cities that I visited today: Papendrecht (7.4h), Dordrecht (1.6h), Alblasserdam (4.5h). The max of speed that I had today: 104.9. The photos that I took today:



# Context recognition

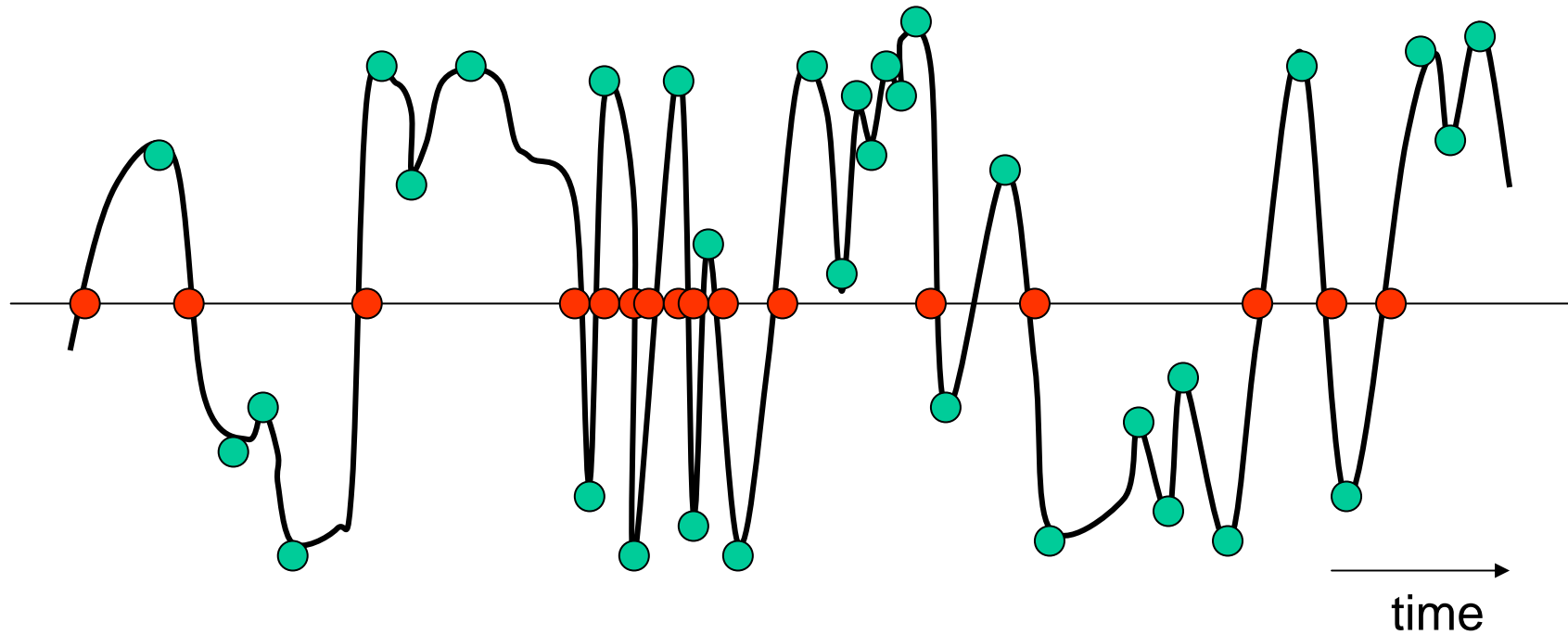
## Context processing stages

- Raw electrical signals
- Interpretation of signals as electric values
- Aggregation, first abstraction of signals
- Further abstraction based on semantics
- Interpretation of abstracted data to contexts



# Processing ratio

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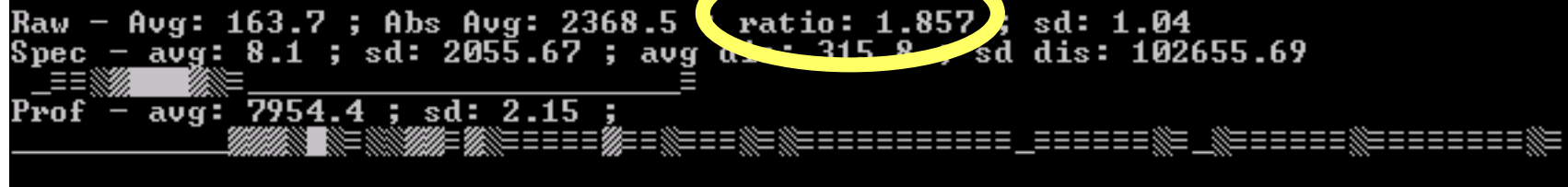


- ratio = Direction changes / Zero crossings

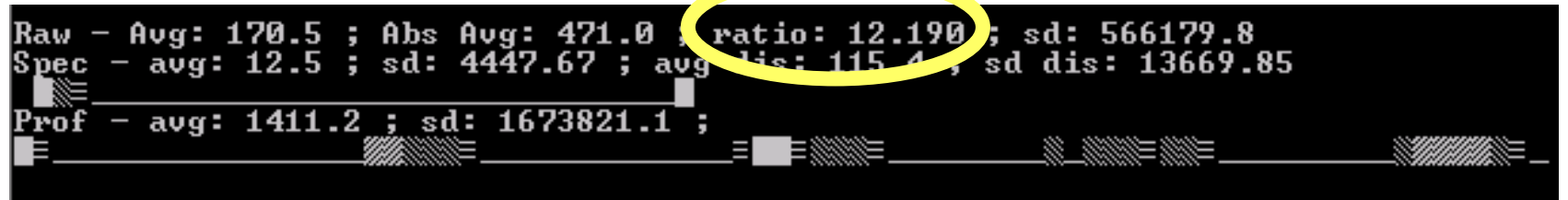
# Processing

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whistling



speech

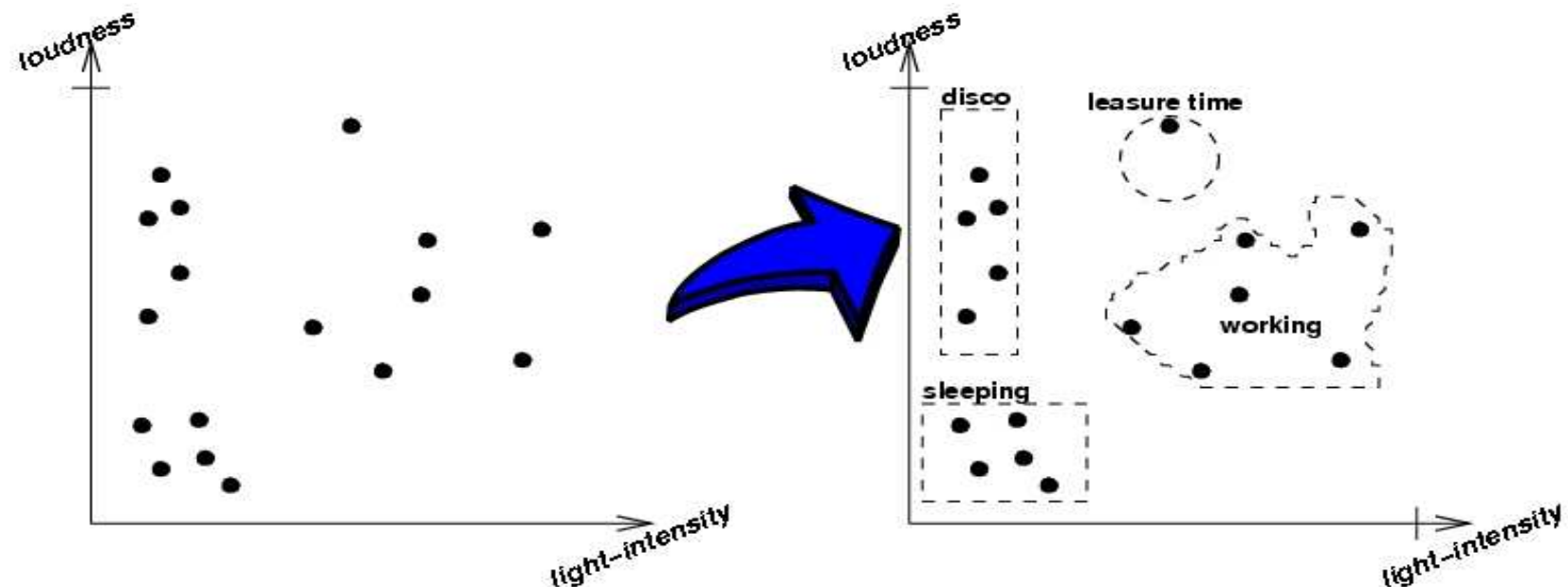


- Distinct ratio zero crossings / direction changes

# Context recognition

## From features to contexts

- Measure available data on features
- Probably with regard to probability distribution
  - Measured value always approximation of actual value
- Context reasoning by appropriate method
  - Syntactical (rule based ; e.g. RuleML);
  - Statistical: HMM, NN, SOM, SVM, Bayes Nets ...

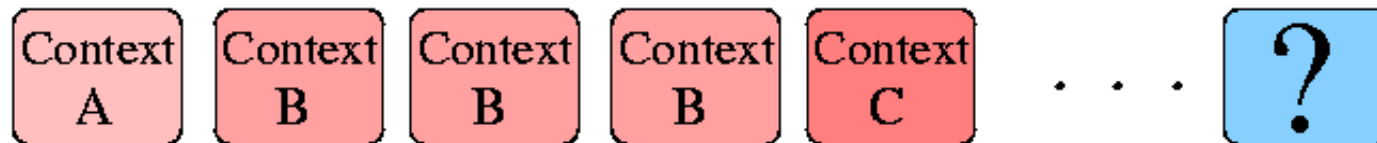


# What is context prediction?

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## Informal descriptions:

- „Context Prediction [...] aims at inferring future contexts from past (observed contexts).“ [Mayr04]
- „In Kombination mit verschiedensten bekannten Informationen soll aus dem augenblicklichen Kontext heraus der nächste Kontext vorhergesagt werden.“ [Petz05]



## Literature:

- [Mayr04] Mayrhofer, R.M, An Architecture for Context Prediction, PhD-Thesis, 2004.
- [Petz05] Petzold, J, Zustandsprädiktoren zur Kontextvorhersage in ubiquitären Systemen, PhD-Thesis, 2005.

# What is context prediction?

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## Formal definition of context prediction:

- Let  $k, n, i \in \mathbb{N}$  and  $t_i$  describe any interval in time. Furthermore, let  $T$  be a context time series. Given a probabilistic process  $\pi(t)$  that describes the context evolution at time  $t_i$ , context prediction is the task of learning and applying a prediction function  $f_{t_i} : T_{t_{i-k+1}, t_i} \rightarrow T_{t_{i+1}, t_{i+n}}$  that approximates  $\pi(t)$ .

# What is context prediction?

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## Context prediction is a search problem:

- A search problem  $\Pi$  is described by

1. the set of valid inputs  $\Lambda_{\Pi}$
2. for  $I \in \Lambda_{\Pi}$  the set  $\Omega_{\Pi}(I)$  of solutions

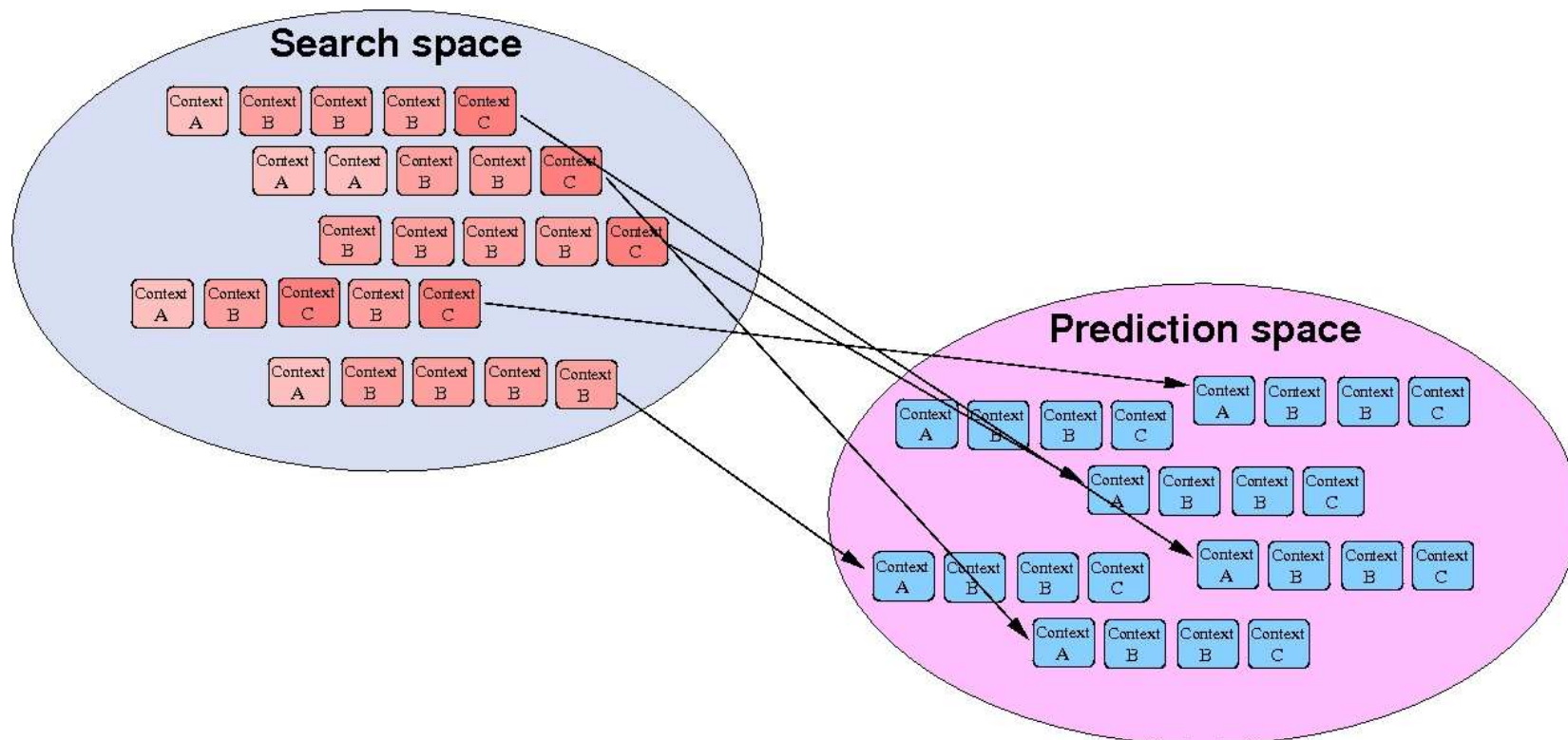
An algorithm solves the search problem  $\Pi$  if it calculates for

$I \in \Lambda_{\Pi}$  an element  $\Omega_{\Pi}(I)$  if  $\Omega_{\Pi}(I) \neq \emptyset$  and rejects otherwise.

# What is context prediction?

## Context prediction is a search problem:

- Context prediction is mainly to find the correct mapping between search space and prediction space



# What is context prediction?

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## Possible distance metrics between contexts:

- Table-Look-up for non-numeric context types
  - Alternatively, non-numeric contexts might be mapped onto numeric context types
- Various approaches for numeric context types:
  - One-dimensional: simple difference between values
  - Multi-dimensional:
    - Euclidic distance between input vectors
    - RMSE

$$RMSE = \sqrt{\frac{\sum_{i=1}^n (p_i - d_i)^2}{n}}$$

- BIAS

$$BIAS = \frac{\sum_{i=1}^n |p_i - d_i|}{n}$$

- Further approaches feasible

# What is context prediction?

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## Requirements:

- In order for context prediction to be feasible, the input sequence has to be predictable in any sense:
  - Periodic patterns
  - Trends
  - Repetitions of typical patterns
  - ...
- Problem:
  - Mood is part of definition of context but hardly accessible by sensors

# What is context prediction?

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## Input sequence typically predictable in UbiComp:

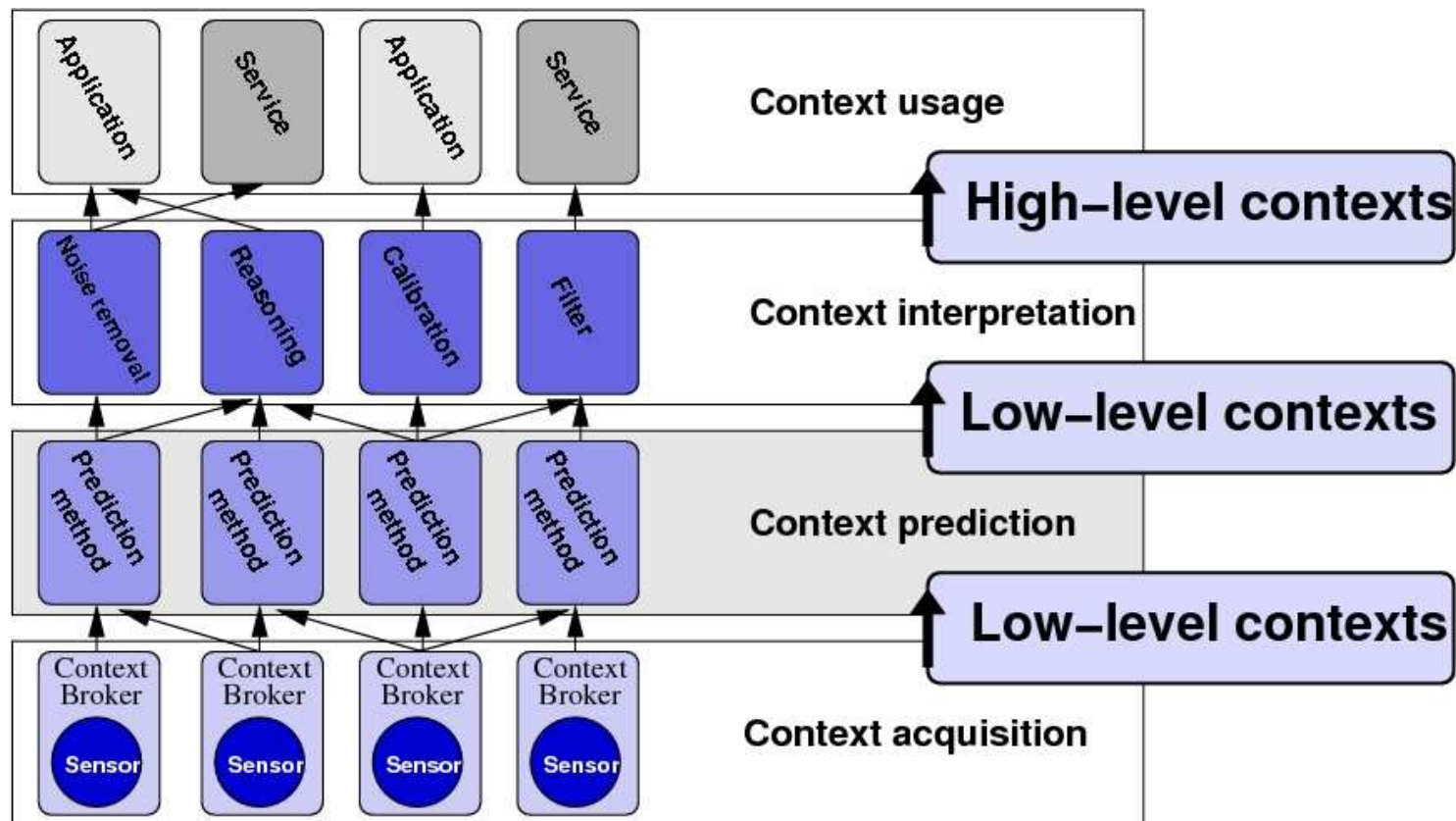
- Human behaviour patterns are reproducible [Ande01]
  - Cognitive psychology: 'script' describes actions and circumstances that characterise specific context or context pattern.
  - Scripts similar for groups of individuals; small alterations between different cultures or societies
- "Behaviour consists of patterns in time" [Magn04]
- Typical behaviours in team-sport games like soccer [JBGB03]
- It is possible to recognise the software programmer of a piece of programming code based on her programming style [Krsu94].

## Literature:

- [Ande01] Anderson, J.R., Cognitive psychology and its implications, Spectrum, 2001.
- [Magn04] Magnusson, M.S., Repeated patterns in behaviour and other biological phenomena, In: Oller, K., Gabriel, U.: Evolution of Communication systems: A comprehensive approach, MIT Press, 2004.
- [JBGB03] Jonsson, G.K., Bjarkadottir, S.H., Gislason, B., Borrie, A., Magnusson, M.S., Detection of real time patterns in sports: Interactions in football, L'ethologie applique aujourd'hui, 2003.
- [Krsu94] Krsul, I., Authorship analysis: Identifying the author of a program, 1994

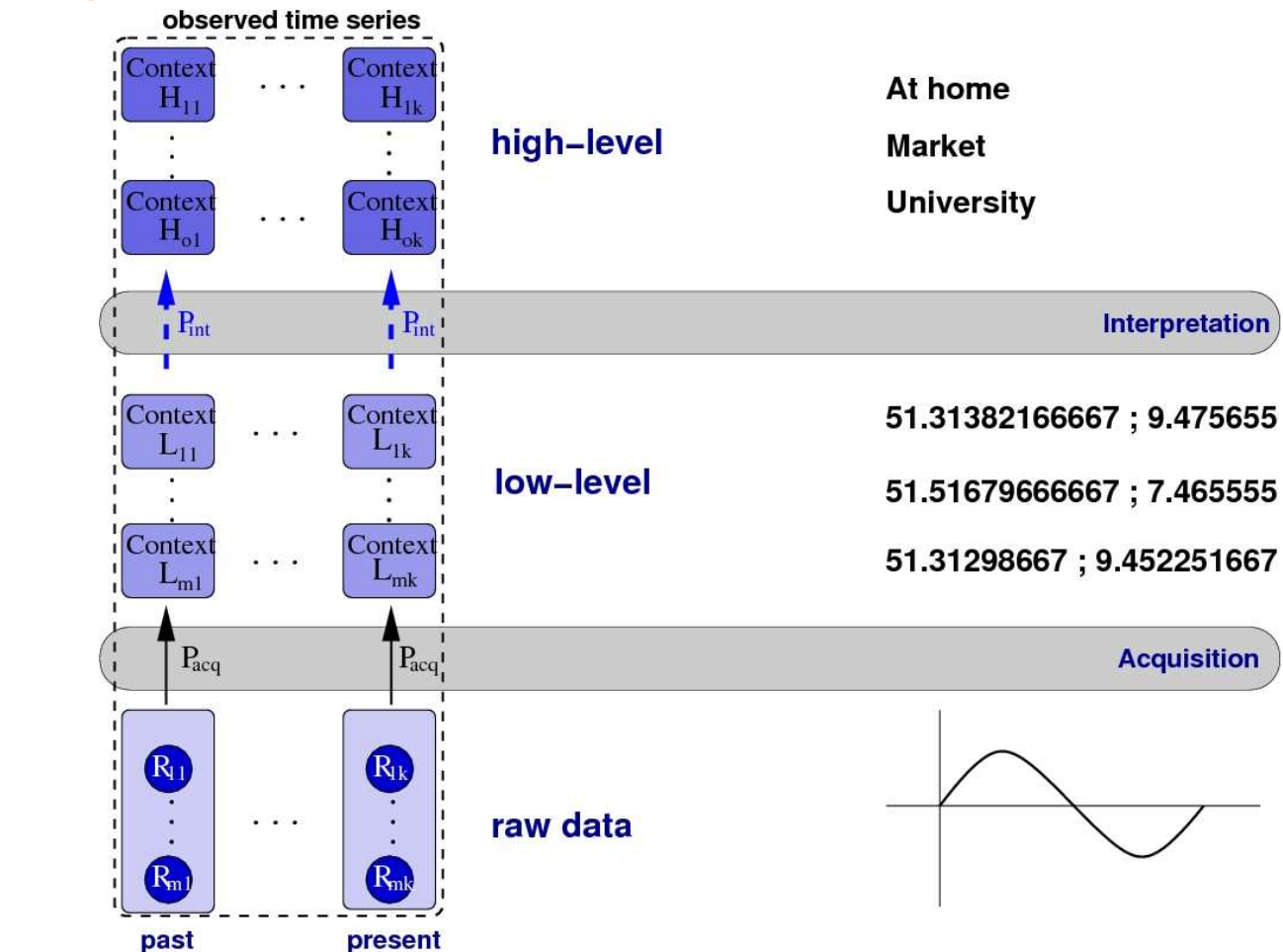
# Context prediction architectures

## Context prediction architectures:



# Context abstraction levels

## High-level and low-level context prediction:



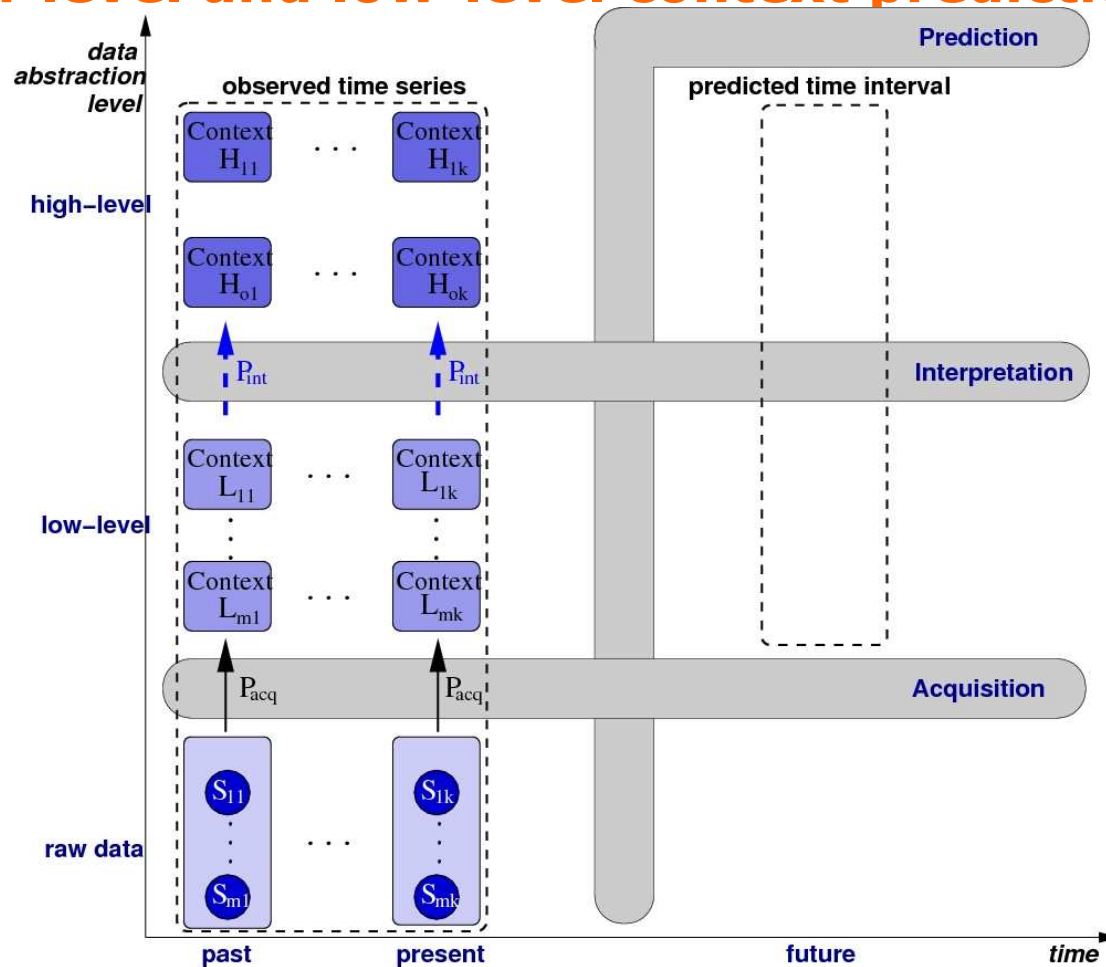
# Context abstraction levels

## Example: Location prediction:



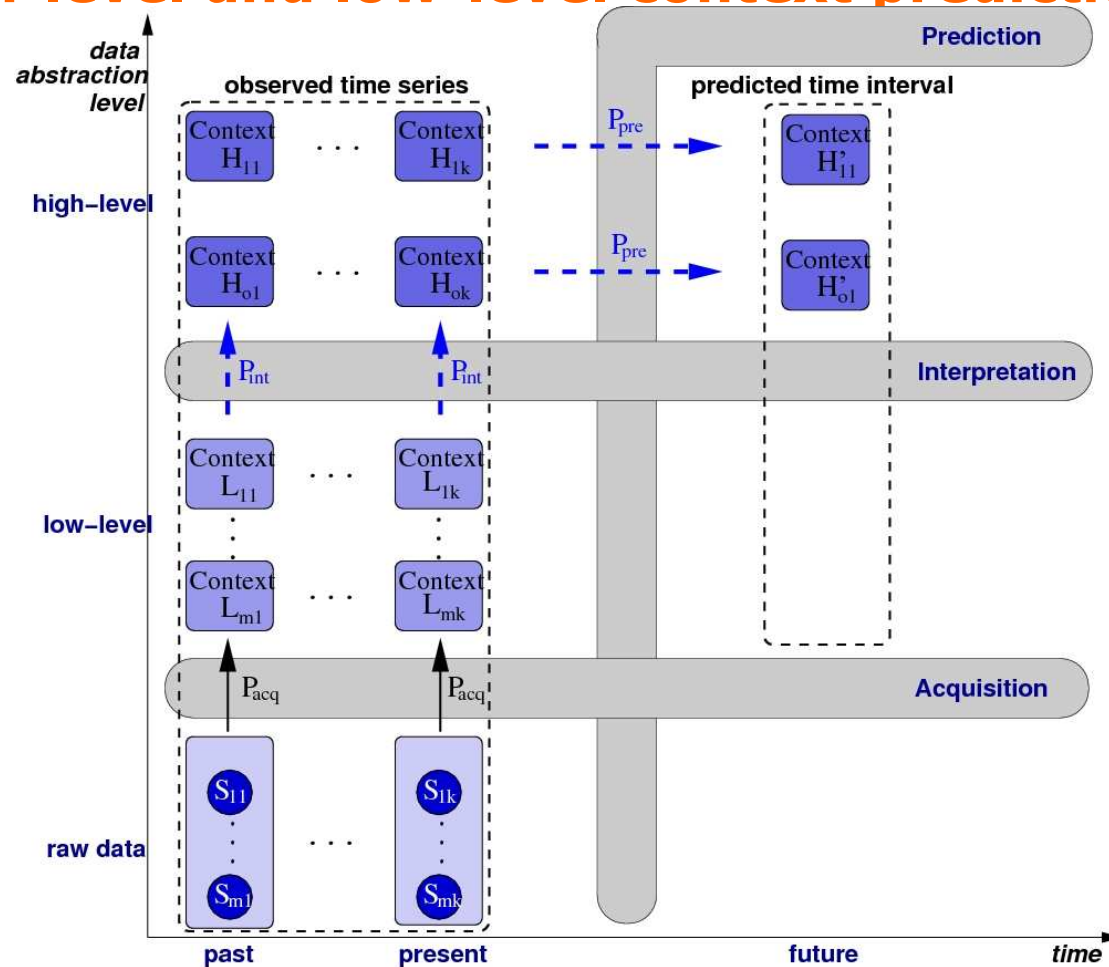
# Implication of abstraction levels

## High-level and low-level context prediction:



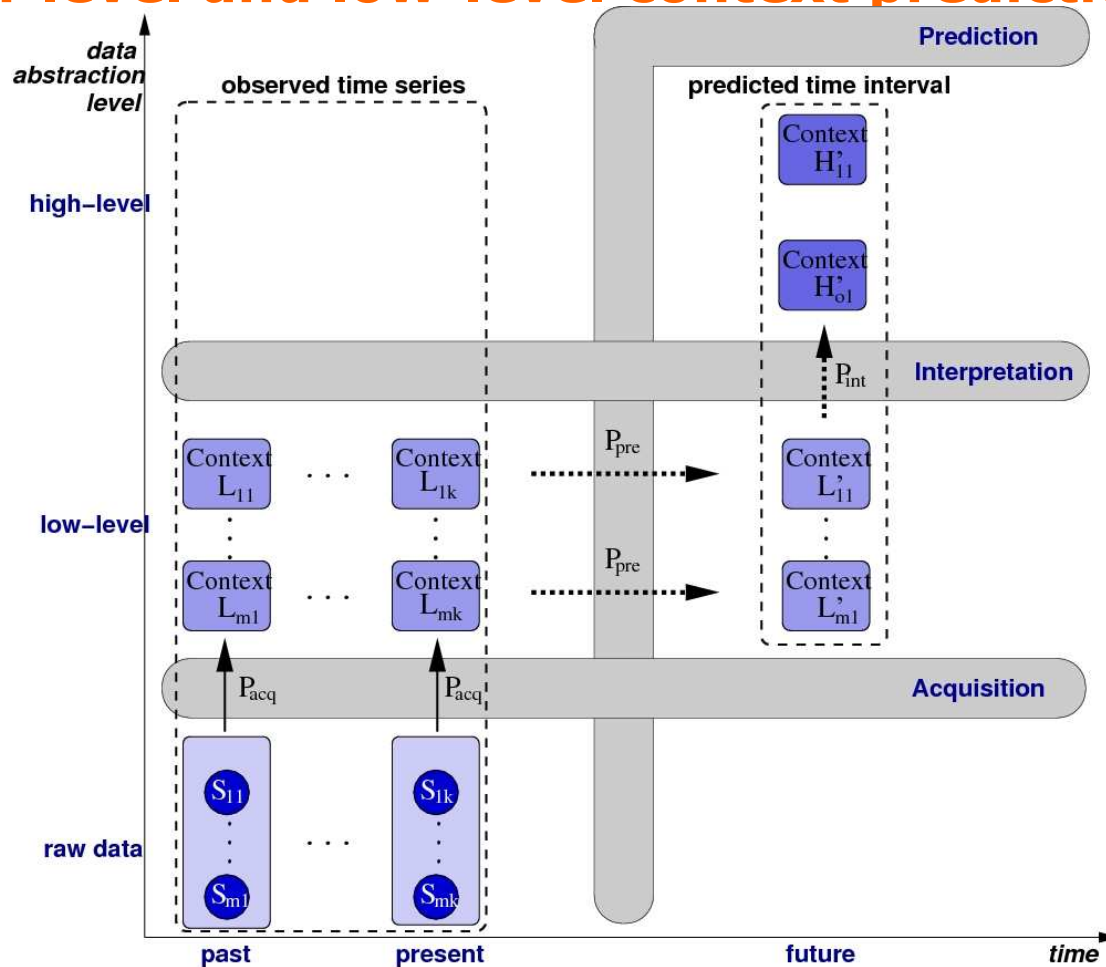
# Implication of abstraction levels

## High-level and low-level context prediction:



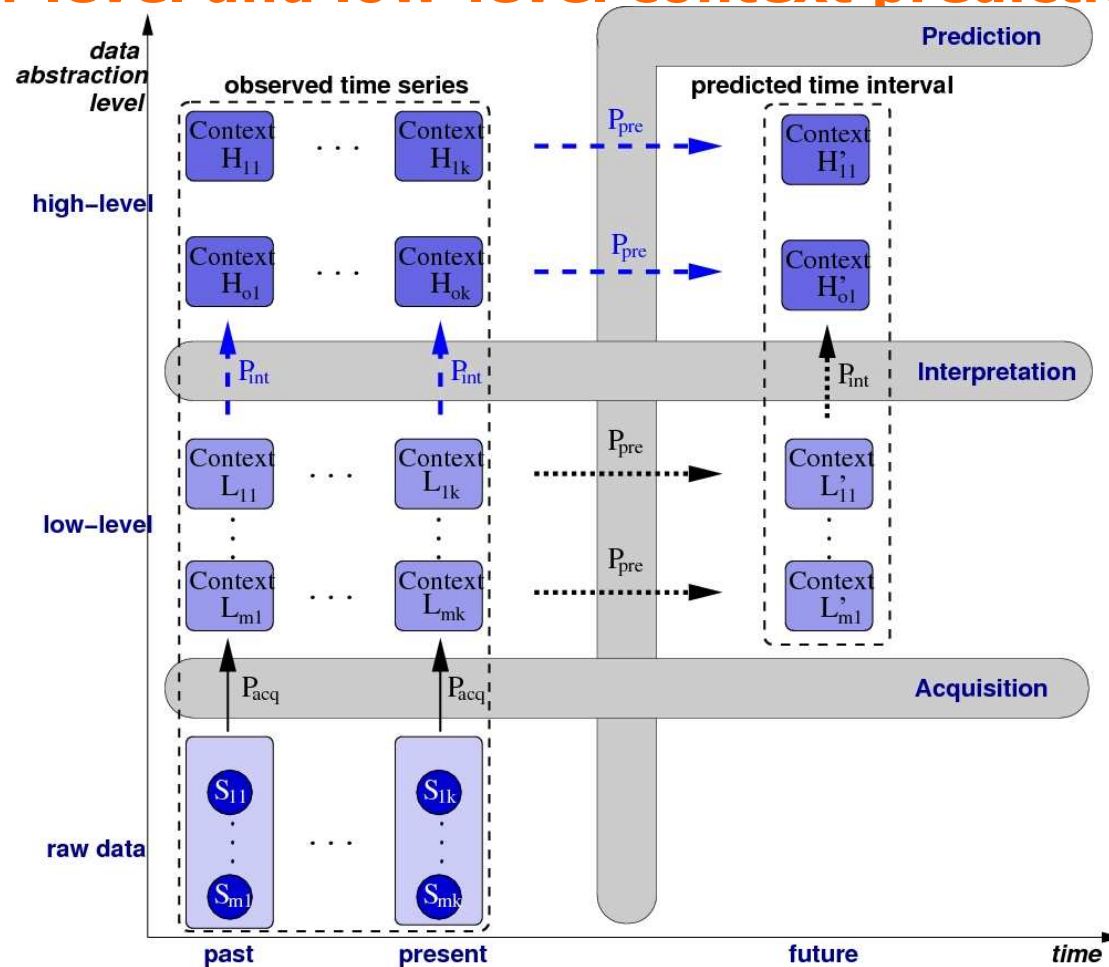
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# Implication of abstraction levels

## High-level and low-level context prediction:



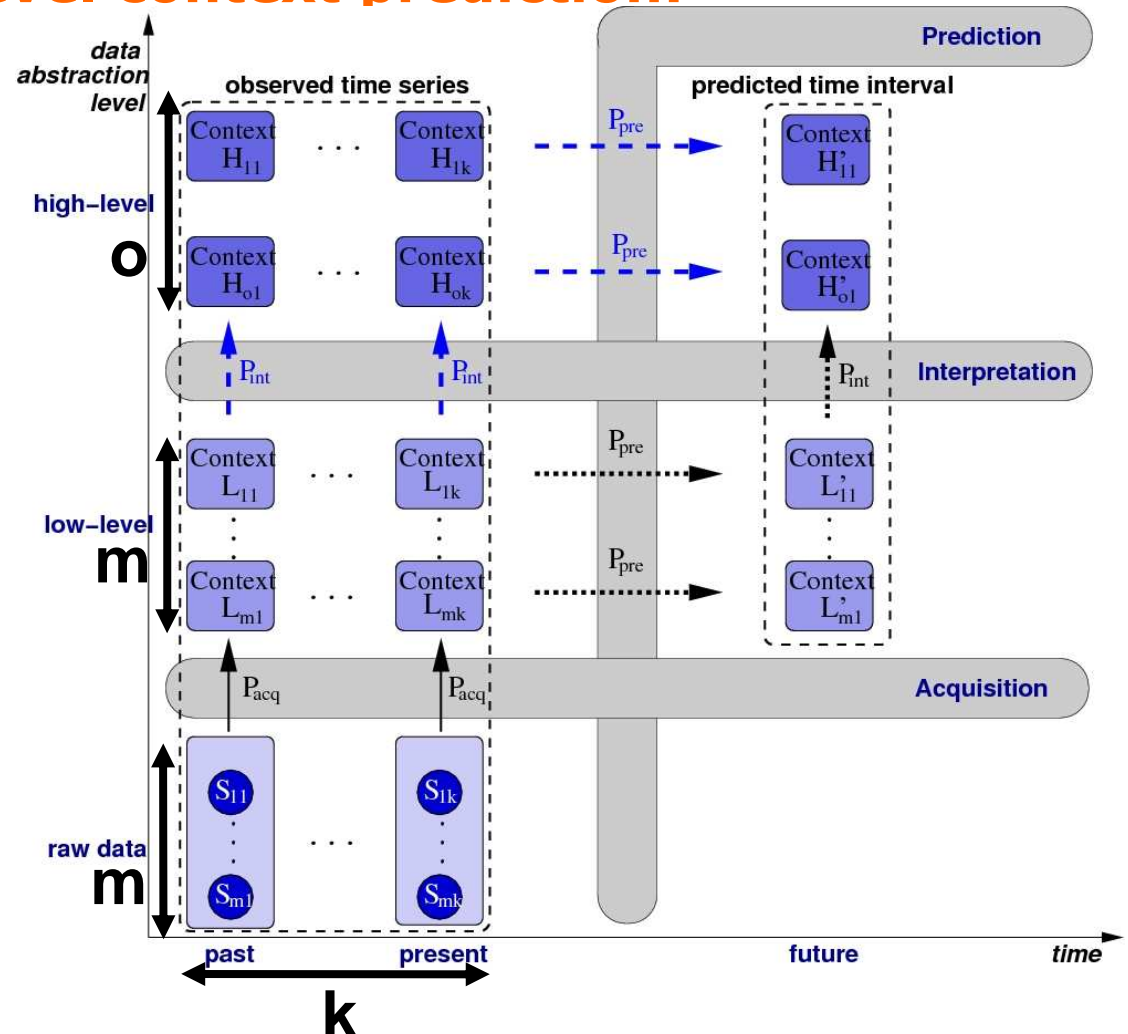
# Implication of abstraction levels

## High-level and low-level context prediction:

$$P_{acq}^{km} \cdot P_{int}^{ko} \cdot P_{pre}^o$$

$$P_{acq}^{km} \cdot P_{pre}^m \cdot P_{int}^o$$

k: # of input time intervals  
m: context sources per interval  
o: high-level contexts per interval  
 $P_{acq}$ : Probability: No acquisition error  
 $P_{pre}$ : Probability: No prediction error  
 $P_{int}$ : Probability: No interpretation error



# Implication of abstraction levels

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**Exact probability estimation:**

$$P_{hl}(i) = (P_{acq}^m P_{int}^o + P_{cor}^{int})^k P_{pre}^o(i) + \left(1 - (P_{acq}^m P_{int}^o + P_{cor}^{int})^k\right) \frac{1 - P_{pre}^o(i)}{v_h^o - 1}.$$

# Implication of abstraction levels

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$$P_{ul}(i) = (P_{acq}^k P_{pre}(i) + P_{cor}^{pre})^m P_{int}^o + \left(1 - (P_{acq}^k P_{pre}(i) + P_{cor}^{pre})^m\right) \frac{1 - P_{int}^o}{v_h^o - 1}.$$

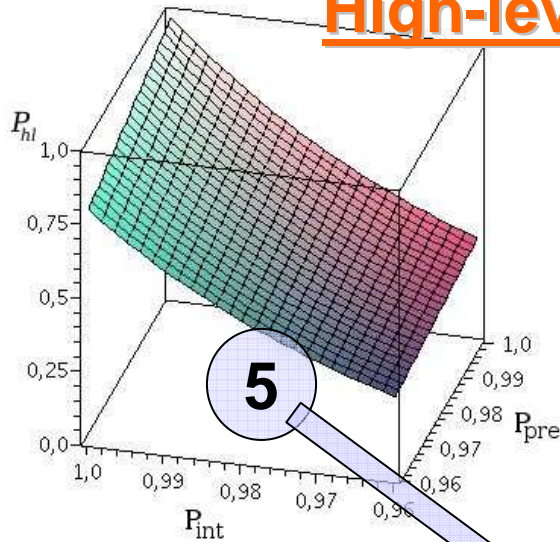
# Implication of abstraction levels

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**Study: When to utilise which prediction scheme**

# Implication of abstraction levels

High-level accuracy - Increasing parameters



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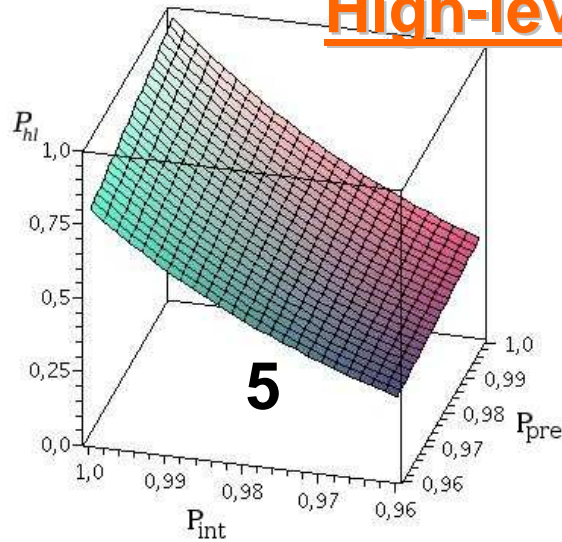
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## Parameters:

<u>k</u> :	# of input time intervals	= 5
<u>m</u> :	context sources per interval	= 5
<u>o</u> :	high-level contexts per interval	= 5
<u>P<sub>acq</sub></u> :	Probability: No acquisition error	= 0.99
<u>P<sub>pre</sub></u> :	Probability: No prediction error	= 0.9
<u>P<sub>int</sub></u> :	Probability: No interpretation er	= 0.9

# Implication of abstraction levels

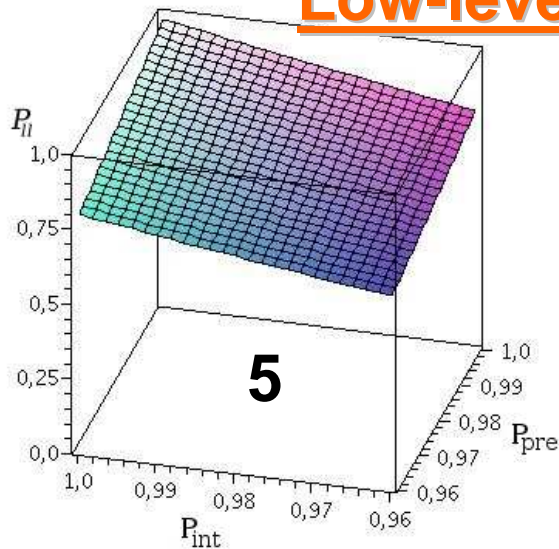
High-level accuracy - Increasing parameters



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Low-level accuracy - Increasing parameters

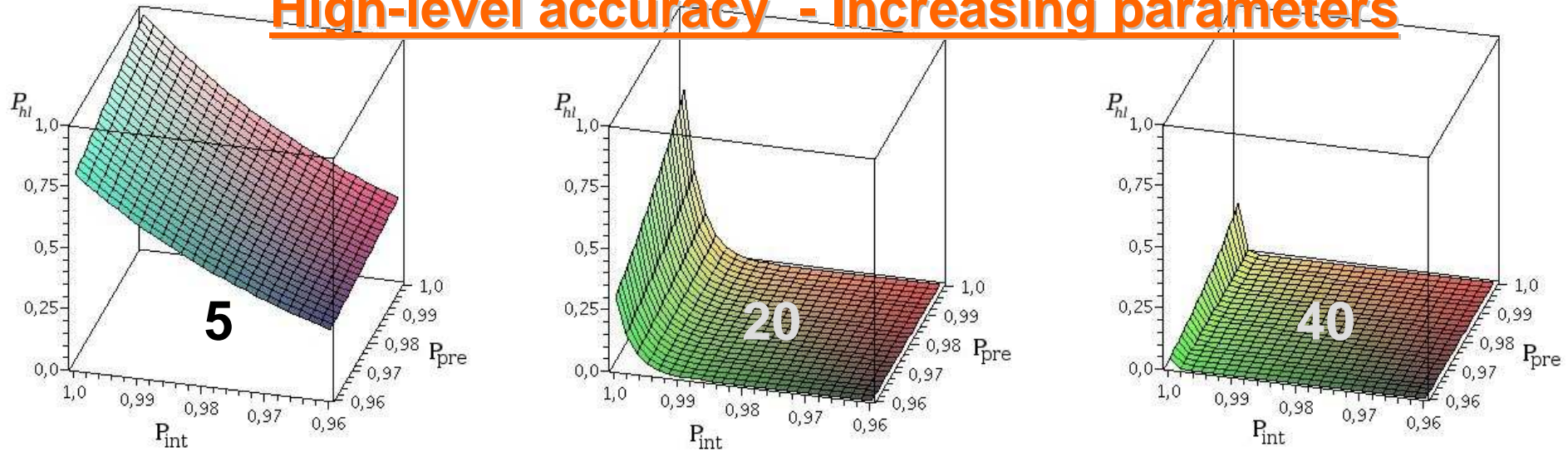


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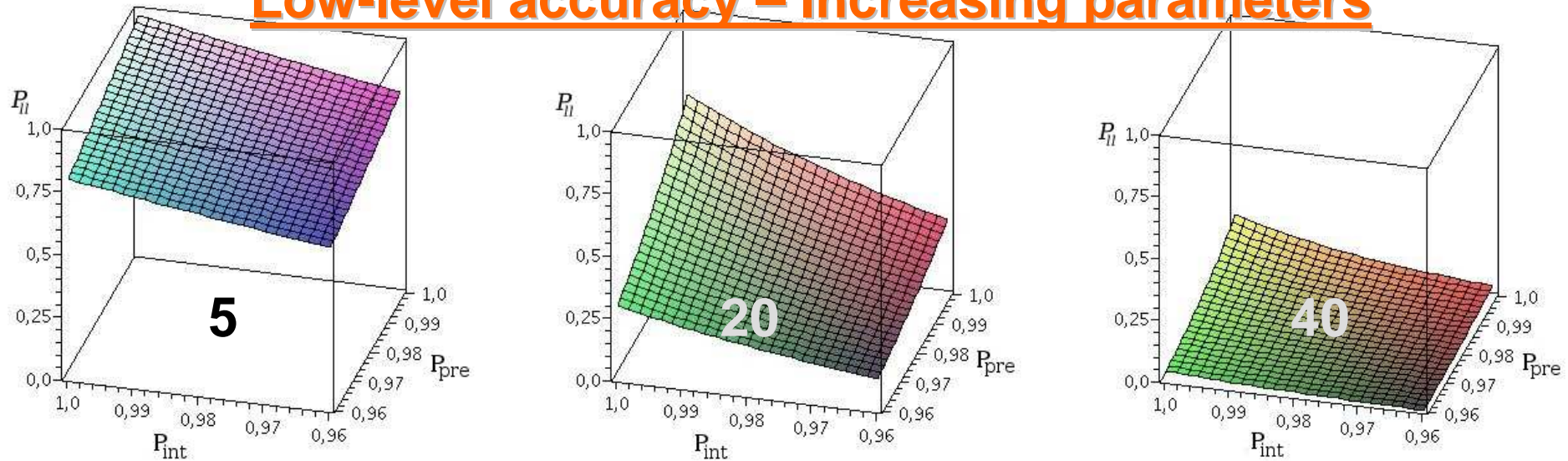
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# Implication of abstraction levels

## High-level accuracy - Increasing parameters



## Low-level accuracy - Increasing parameters



# Implication of abstraction levels

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## Conclusion – Context abstraction levels:

- Analyse scenario in advance to implementation
- Environmental parameters and error probabilities of context processing modules impact prediction accuracy
- Prediction accuracy differs with context abstraction level of input data.