

# Augmented Business Intelligence

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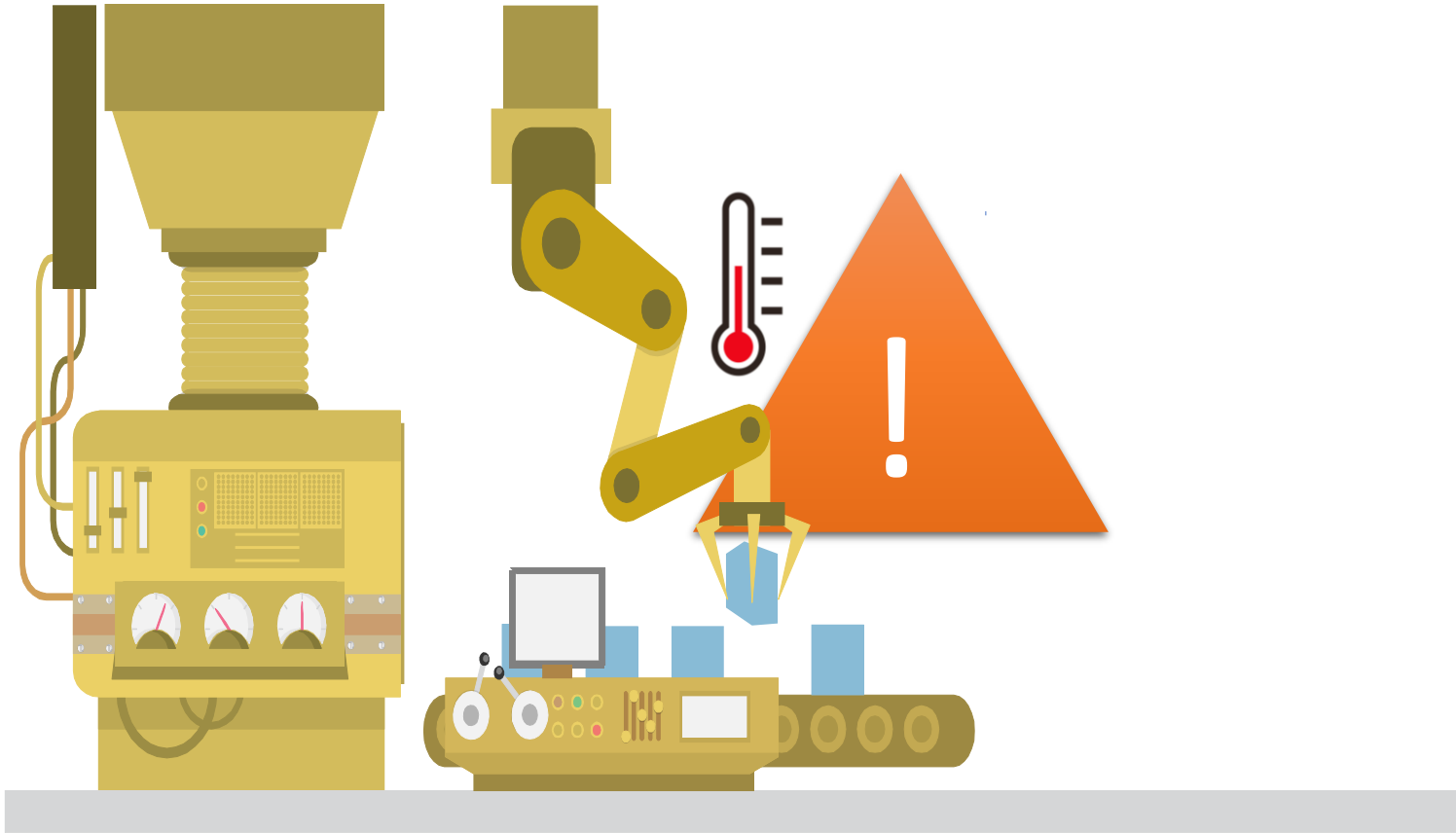
Matteo Francia, Matteo Golfarelli, Stefano Rizzi

*DISI – University of Bologna*

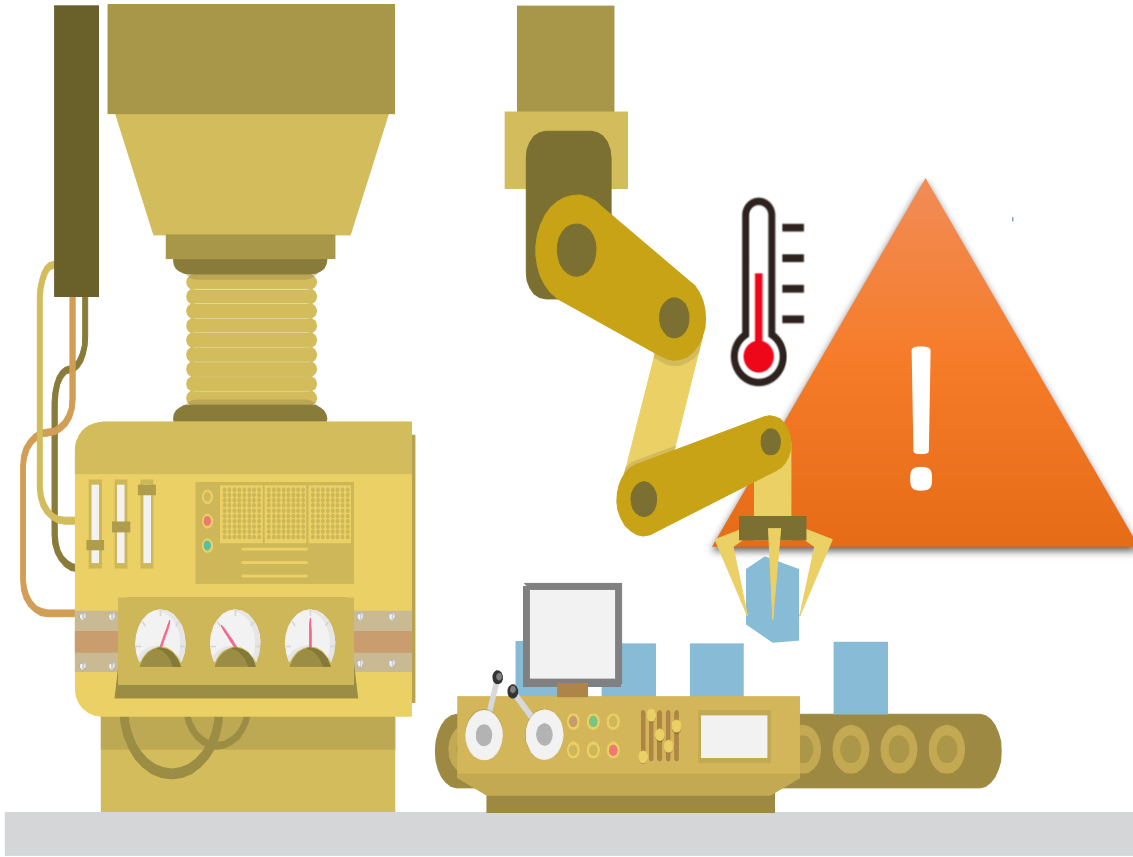
{m.francia, matteo.golfarelli, stefano.rizzi} @unibo.it



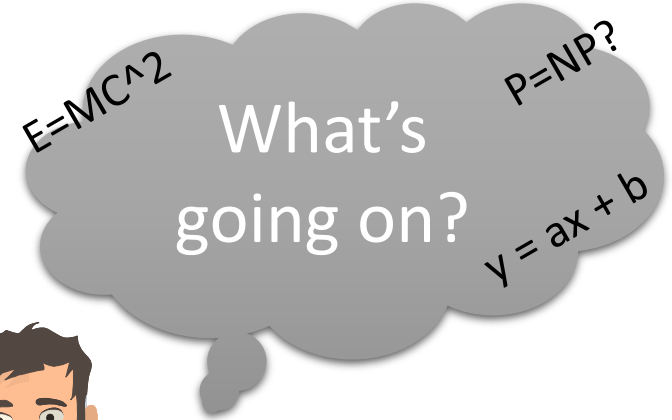
# Application scope



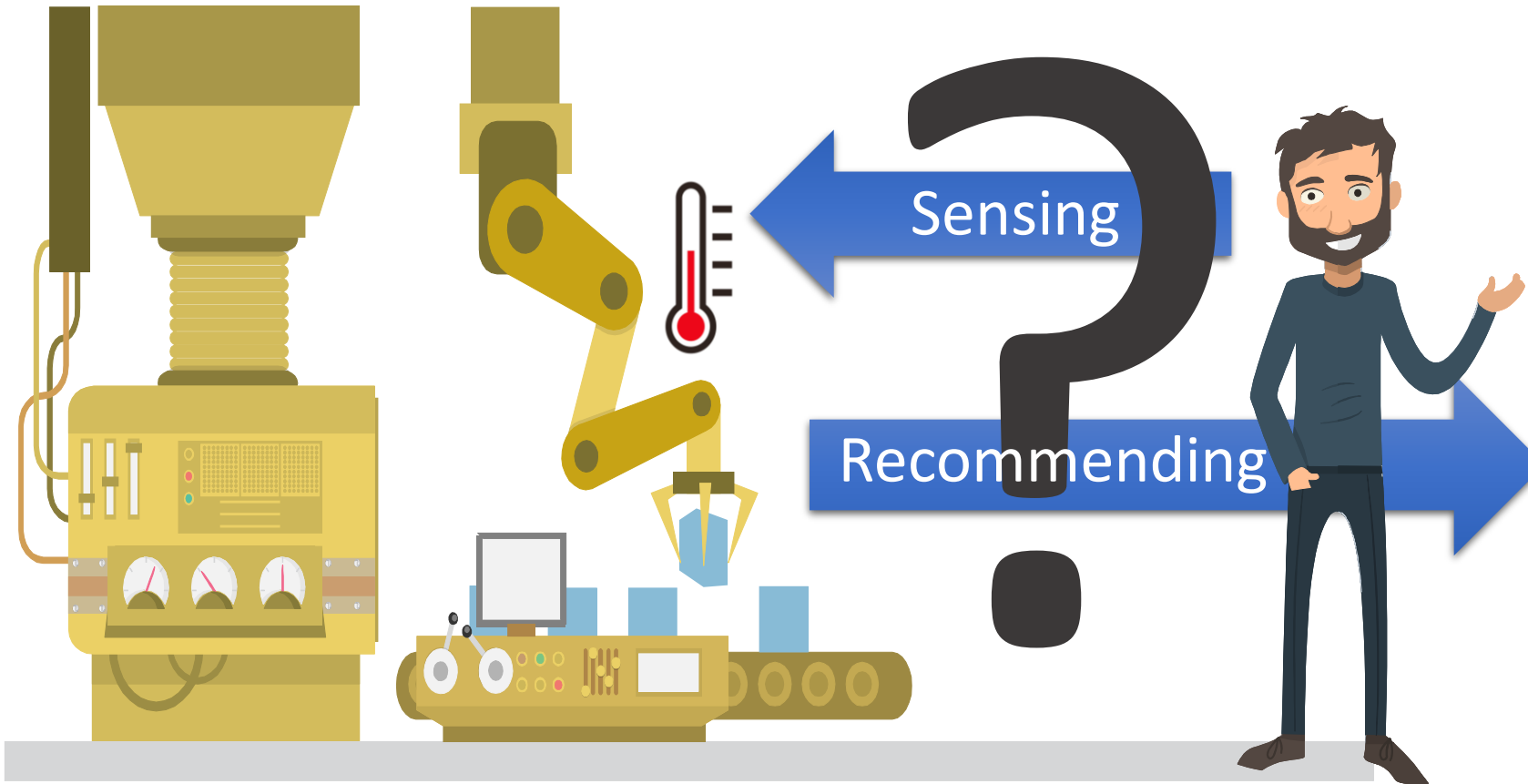
# Application scope



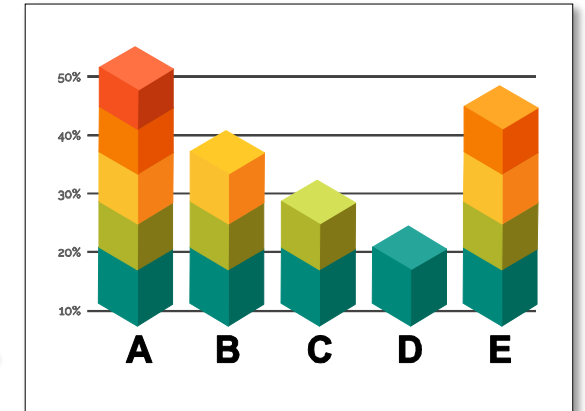
Inspector



# Application scope



## Analytical report



We have data!

- Internet of Things
- Digital twin [1]

# Augmented Business Intelligence

A-BI: a 3D-marriage

- Augmented Reality
- Business Intelligence
- Recommendation

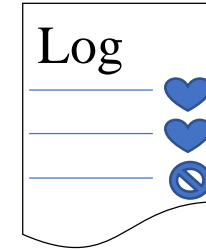
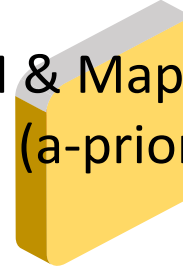
# A-BI: Overview

Data Sources



Augmented Reality  
(real-time)

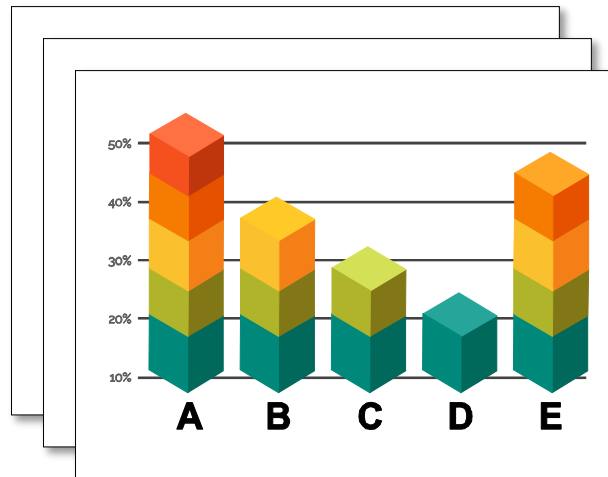
DM & Mappings  
(a-priori)



Query Log  
(user exp.)

## Augmented Business Intelligence

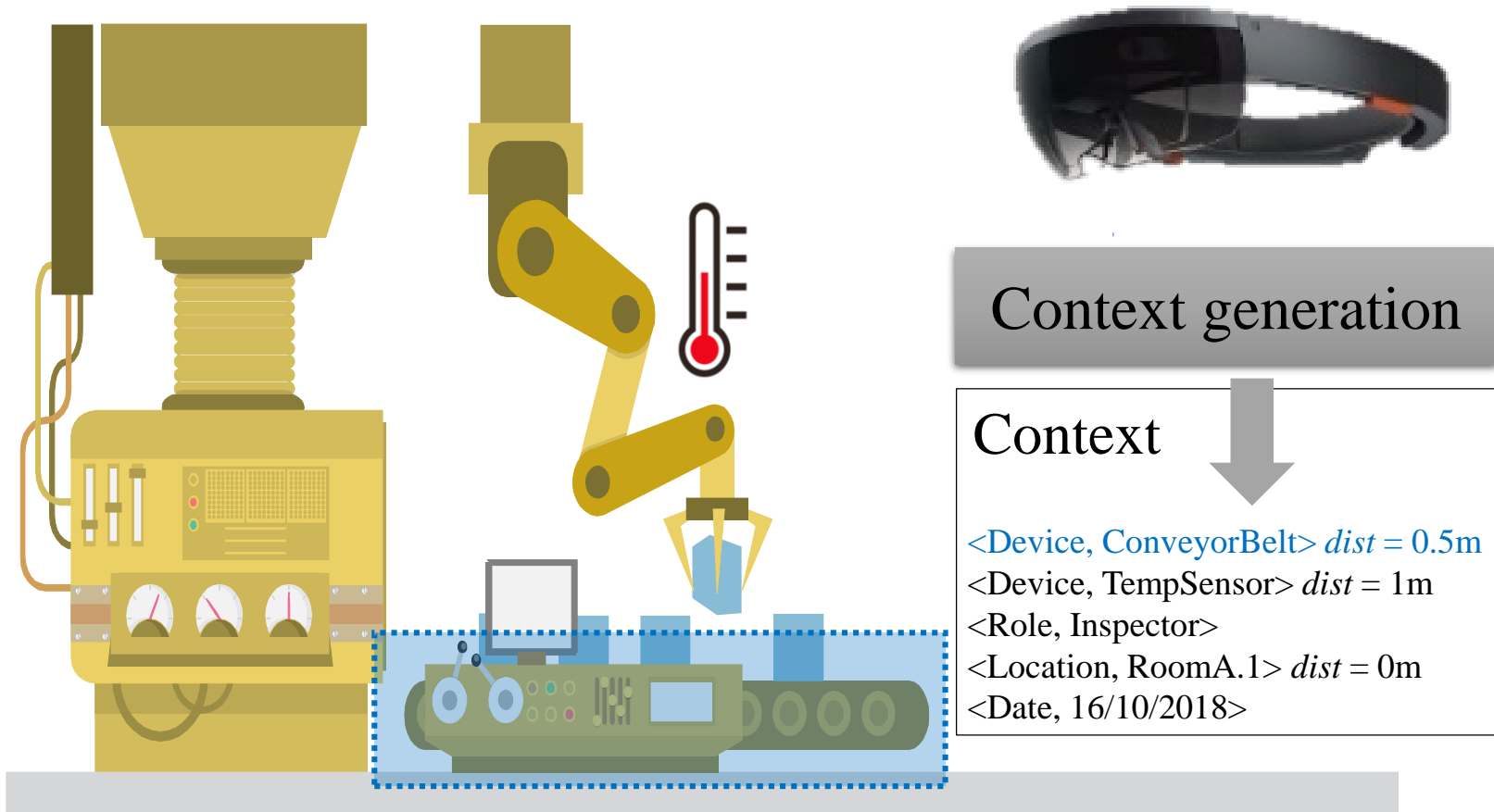
Outputs



OLAP reports



# A-BI: Augmented Reality



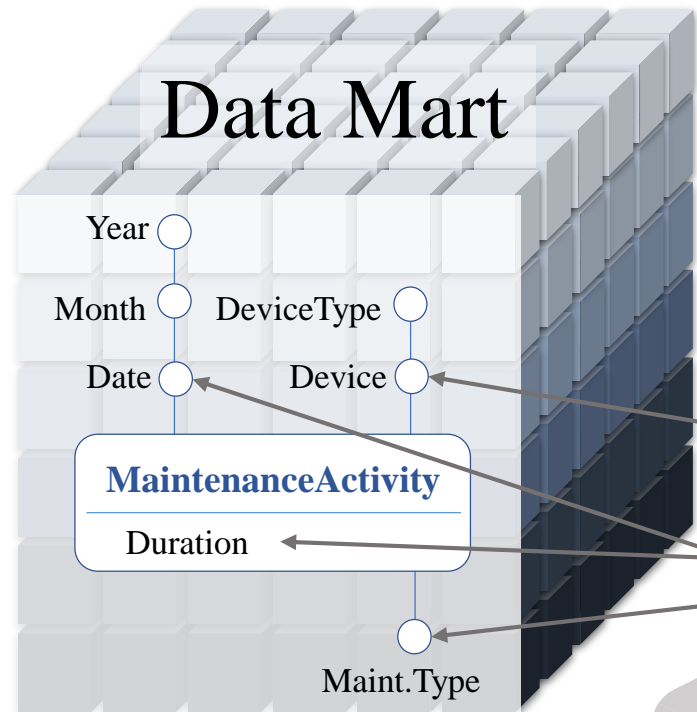
- Sensing augmented environments [2]
  - Real-time information
- Interaction
  - Engagement [3]
- Constrained visualization
  - i.e., cardinality constraint

# A-BI: Business Intelligence

- Data dictionary
  - What do we recognize?
  - Context: subset of data dictionary entries

- Mappings to md-elements
  - A-priori interest

- OLAP
  - Report generation



Context generation

Context

↓

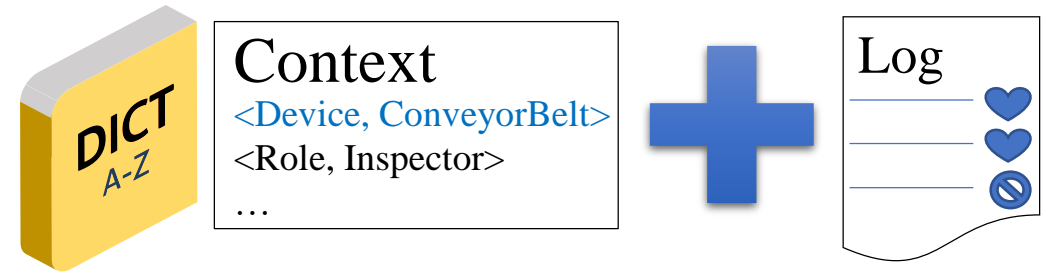
<Device, ConveyorBelt> *dist* = 0.5m  
<Device, TempSensor> *dist* = 1m  
<Role, Inspector>  
<Location, RoomA.1> *dist* = 0m  
<Date, 16/10/2018>



# A-BI: Recommendation

## 1. Get the context

- Context  $T$  over data dictionary
- Follow (a-priori) mappings...
- ... Project  $T$  to image  $I$  of md-elements



## 2. Add the log $L$

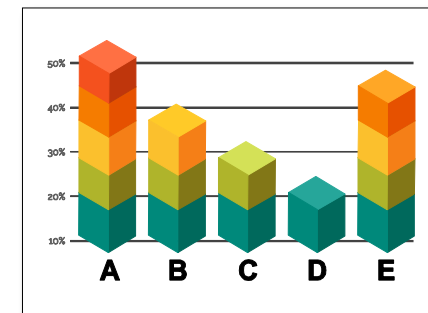
- Get queries with positive feedback from similar contexts
- Enrich  $I$  to  $I^*$  with «unperceived» elements from  $T$



## 3. Get the queries

Directly translate  $I^*$  into a well formed query

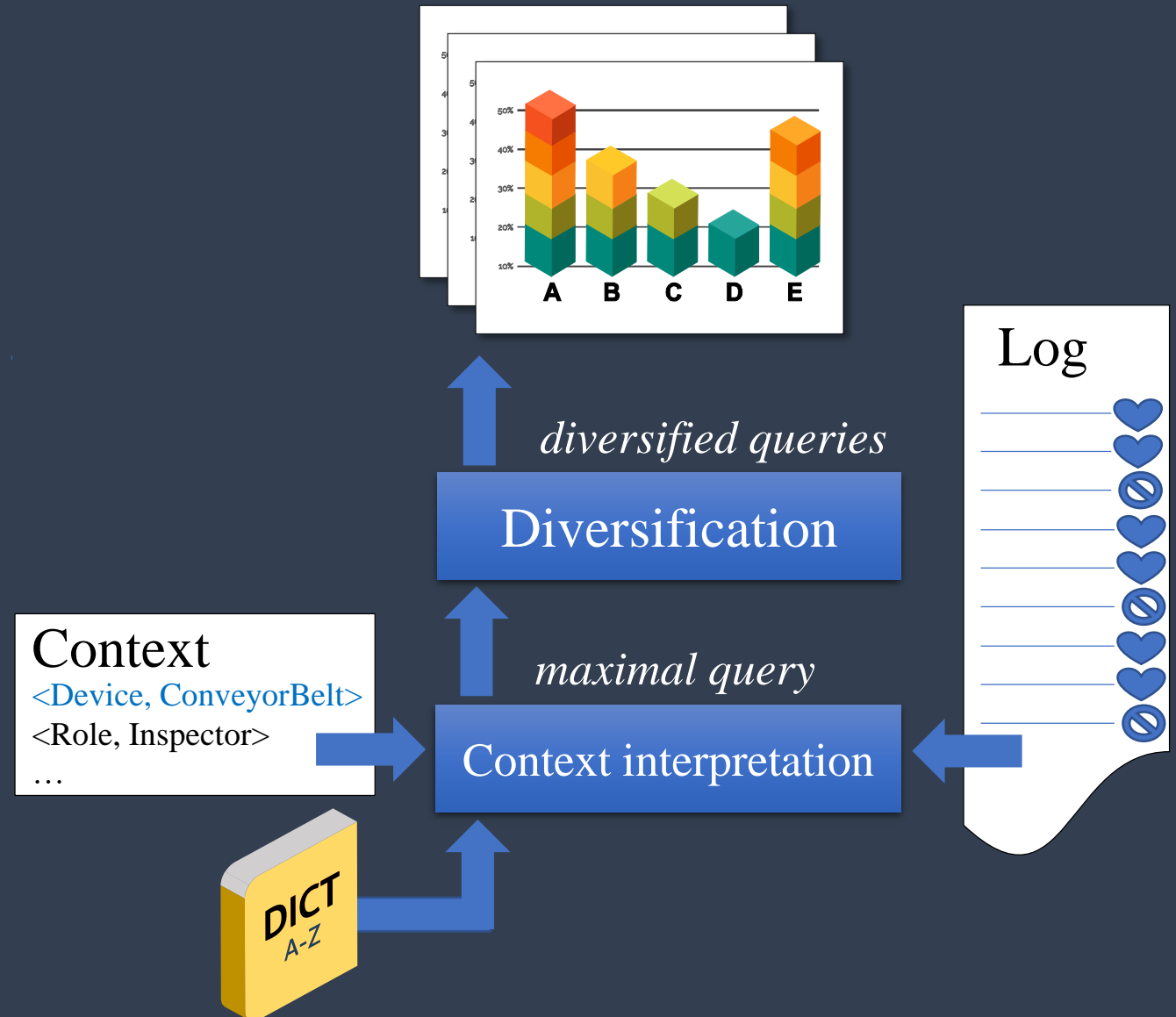
- High cardinality  $I^*$  = hardly interpretable «monster query»
- Single query, no diversification



# A-BI

A two-step approach:

- Context interpretation
- Diversification



# A-BI: Context Interpretation

- md-element relevance

- Context weight
- Mapping weight
- Relevance over log

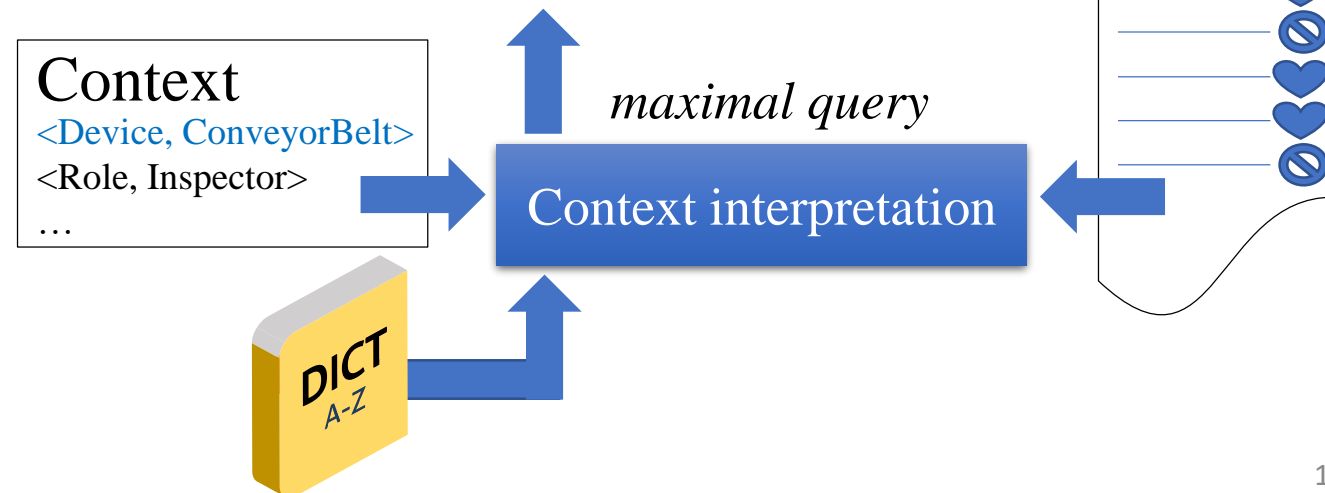
$$rel_T(e) = \begin{cases} \sum_{d \in T | e \in \mu(d)} w(T, d) \cdot w(d, e), & \text{if } L_T(e) = \emptyset \\ \sum_{d \in T | e \in \mu(d)} w(T, d) \cdot w(d, e) \cdot \rho_T(L, e), & \text{otherwise} \end{cases}$$

- Query relevance

$$rel_T(q) = \frac{\sum_{e \in q} rel_T(e)}{\sum_{e \in I_\mu^*(T)} rel_T(e)}$$

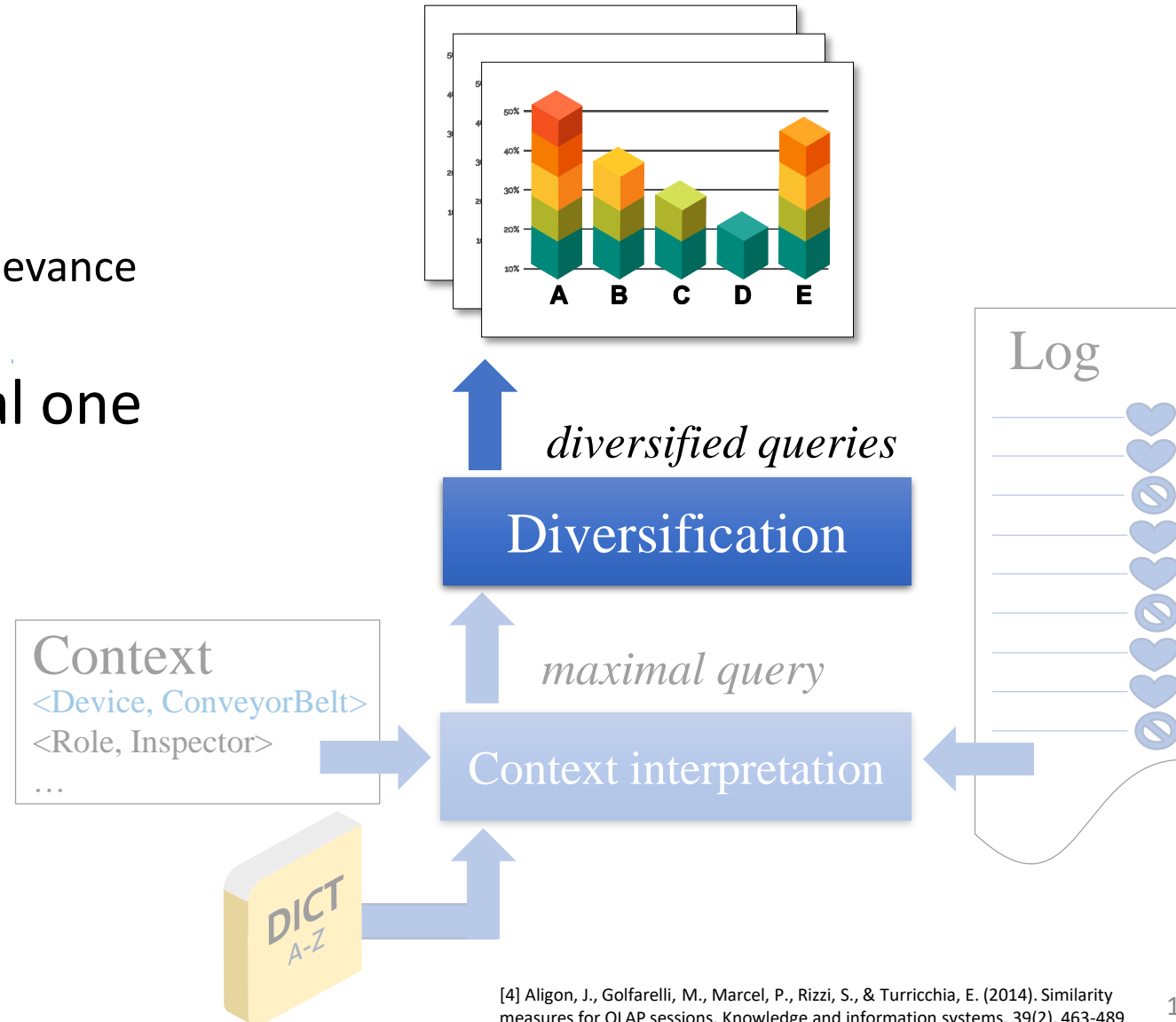
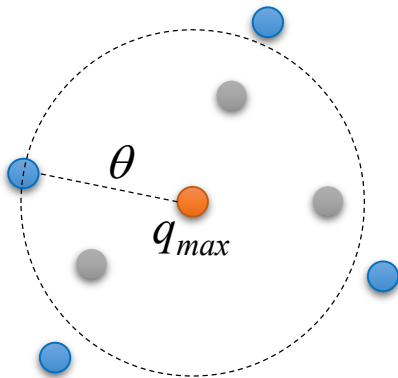
- Maximal query

- Most relevant query enforcing cardinality constraint
- = Knapsack Problem
  - Draw most relevant DM-elements
  - s.t. query cardinality is below threshold



# A-BI: Diversification

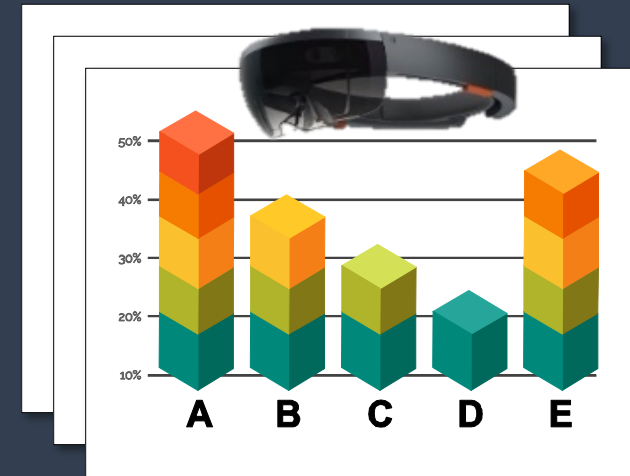
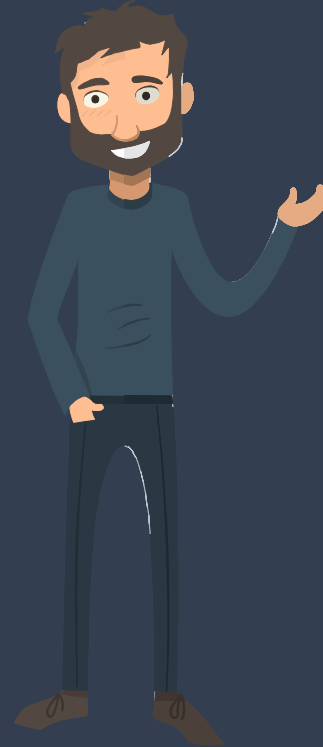
- Diversification
  - Different flavors of same information
  - = Top-N queries maximizing diversity and relevance
- Generate queries from the maximal one
  - Operators: rollup/drill/slice
  - Query similarity  $sim$  (with  $div = 1 - sim$ ) [4]
  - $\theta$  = amount of diversification



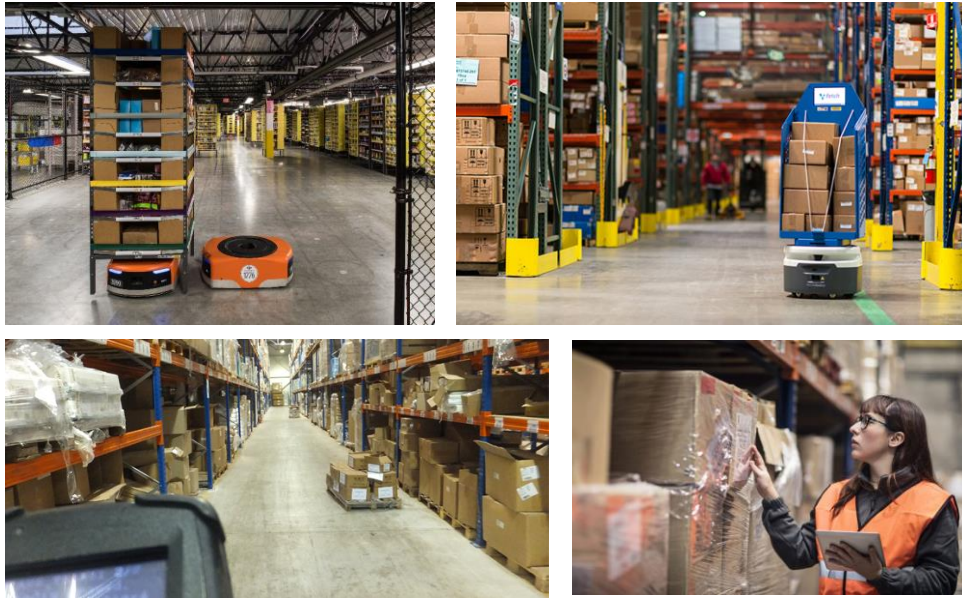
[4] Aligon, J., Golfarelli, M., Marcel, P., Rizzi, S., & Turricchia, E. (2014). Similarity measures for OLAP sessions. Knowledge and information systems, 39(2), 463-489

# Evaluation

- Effectiveness
- (Near) Real-time



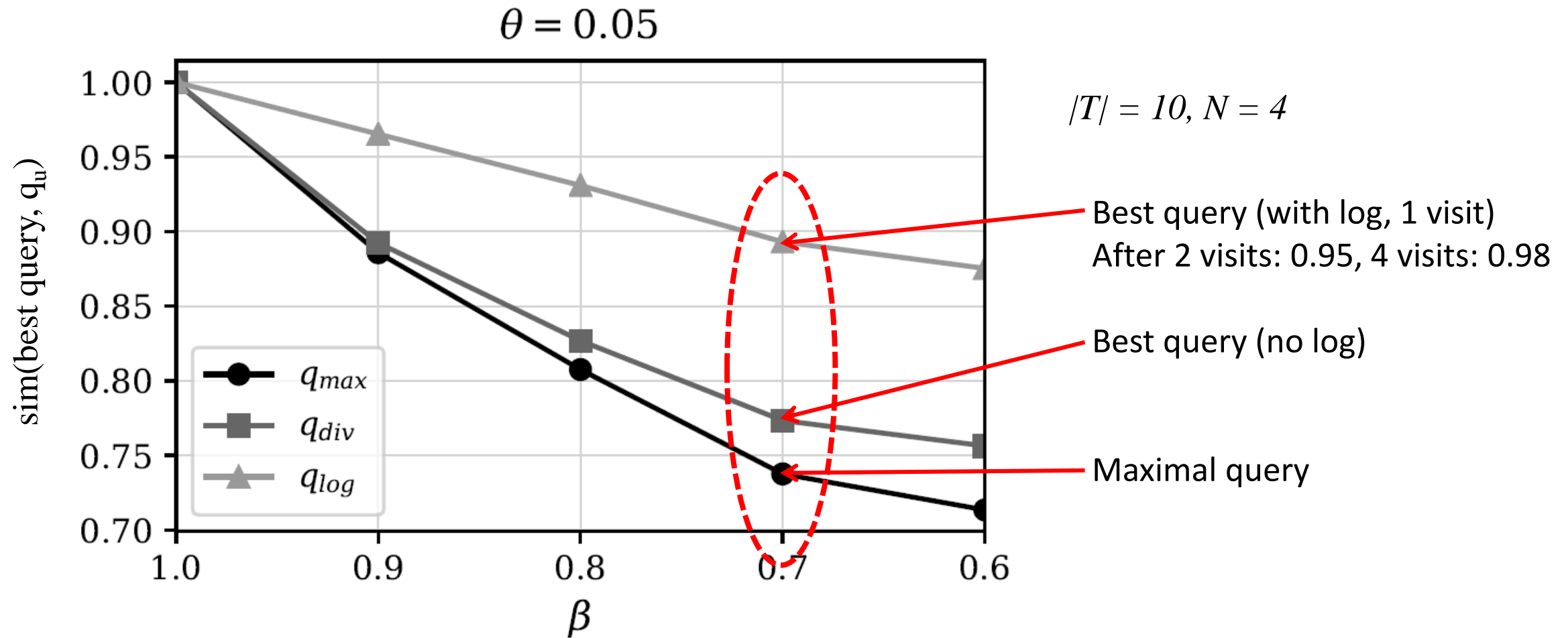
# A-BI: Test setup



Examples of context seeds

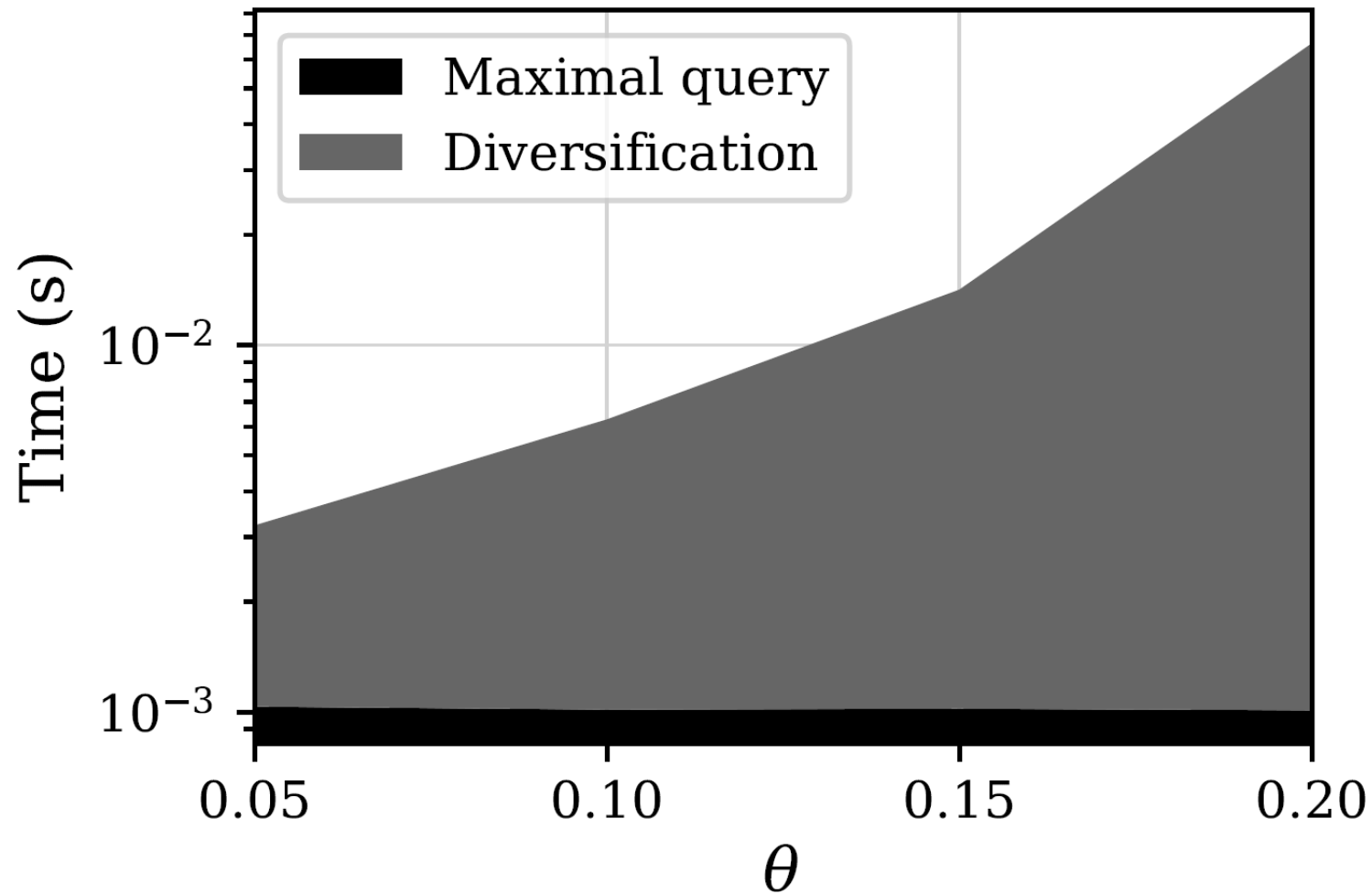
- Cube
  - 5 linear hierarchies, 5 levels each
  - Maximum cardinality  $10^9$
  - Dictionary with one entry for md-element
  - One-to-one mappings (entry  $\rightarrow$  md-element)
  - Random context and mapping weights
- Simulate user moving through a factory
  - In 10 different rooms (i.e., 10 context seeds)
  - 5 to 15 recognized entities
- Simulate multiple visits to rooms
  - Generate seed variations

# A-BI: Effectiveness



$\beta = \text{target similarity between } q_u \text{ and } q_{max}$

# A-BI: Efficiency



- Time required to recommend a query set
- Query execution is then demanded to DW system

$\theta$  = diversity threshold





# Is A-BI out of reach?

Object recognition (YOLO [5])  
Egocentric computer vision [6]

# Work in progress: relevance of groups

- Up to now
  - Relevance of single md-elements
  - Recommendation address all the elements together



## Proposal: Relevance is about groups of md-elements

- Element  $a$  is relevant with  $b$  but not with  $c$
- Definition of group relevance  $rel$
- Review formulation to provide recommendation related to groups
- Given  $rel(\{\text{Maint.Type}\}) = 1$  and  $rel(\{\text{Duration}\}) = 1$ 
  - $rel(\{\text{Maint.Type}, \text{Duration}\}) = 2.5$
  - $rel(\{\text{Device}, \text{Month}\}) > rel(\{\text{Device}, \text{Date}\})$

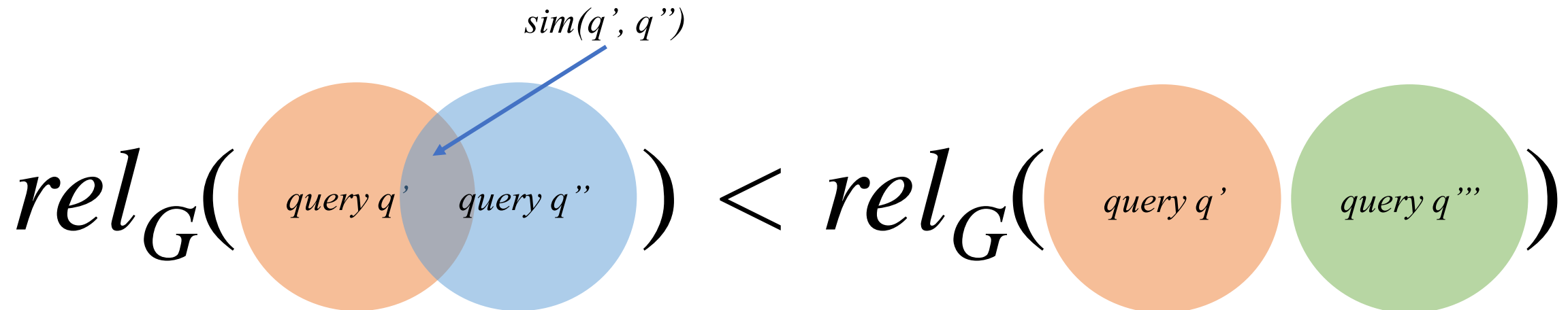
# Work in progress: query generation

- Up to now
  - Recommendation as a two-step approach



## Proposal: Optimal formulation for query generation

- Single-step formulation inspired by mutual information
- Minimize amount of information about one query obtained through other query(s)
- Definition and maximization of global relevance  $rel_G$ 
  - Overlapping queries (i.e., similar queries)  $\rightarrow$  high mutual information



# Conclusion

## Augmented Business Intelligence

- Recommendation of multi-dimensional analytic reports
  - Based on augmented (real) environments
  - Under near-real-time and visualization constraints
- 
- Vision
    - Analytics in Health-care
    - Conversational BI

# Thanks