

Context Aware Computing for The Internet of Things

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ABOUT THE STUDY PAPER



Perera, C., Zaslavsky, A., Christen, P., & Georgakopoulos, D. (2014). Context aware computing for the internet of things: A survey. *IEEE Communications Surveys and Tutorials*, 16(1), 414–454. <http://doi.org/10.1109/SURV.2013.042313.00197>

The survey brings an evaluation of 50 projects that represented the majority of research and solutions proposed in the context-aware computing field conducted over years 2001 and 2011.

DEFINITIONS

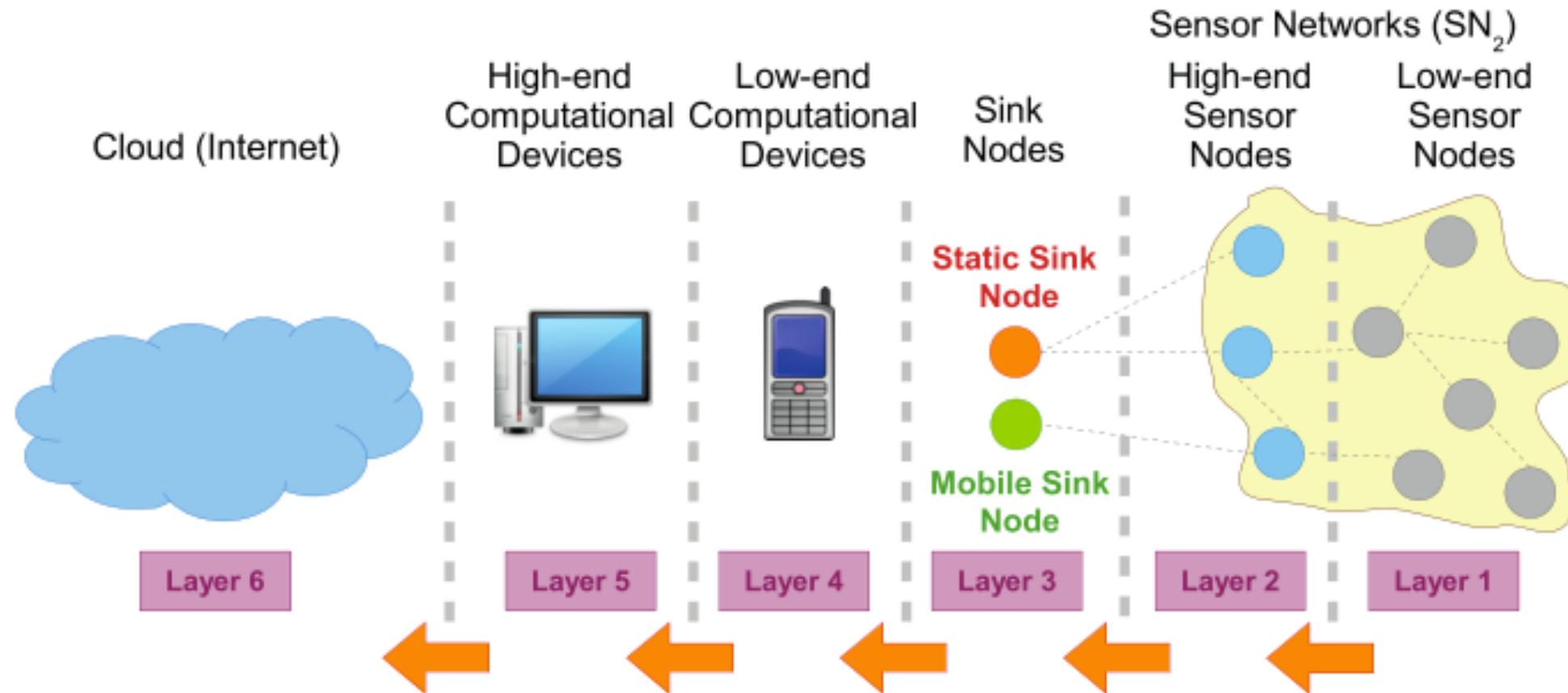
INTERNET OF THINGS



The Internet of Things allows people and things to be connected anytime, anywhere, with anything and anyone, ideally using any path/network and any service (*Vermesan et al. [7]*)

DEFINITIONS

LAYERED STRUCTURE OF A SENSOR NETWORK



DEFINITIONS

WHAT IS CONTEXT? *IT IS HARD TO TELL.*

Various attempts to explain it:

- Through synonyms (*Pereira et al. [1]*)
 - “Circumstance, situation, phase, position, posture, attitude, place, point; terms; regime; footing, standing, status, occasion, surroundings, environment, location, dependence.”
- Through enumeration of examples
 - location, identities of nearby people and objects, and changes to those objects (*Schilit and Theimer [3]*)
 - people around the user, time of day, season, temperature (*Brown et al. [4]*)
- 5 W’s (Who, What, Where, When, Why) (*Abowd and Mynatt [5]*)

DEFINITIONS

CONTEXT & ENTITY



- **Context**

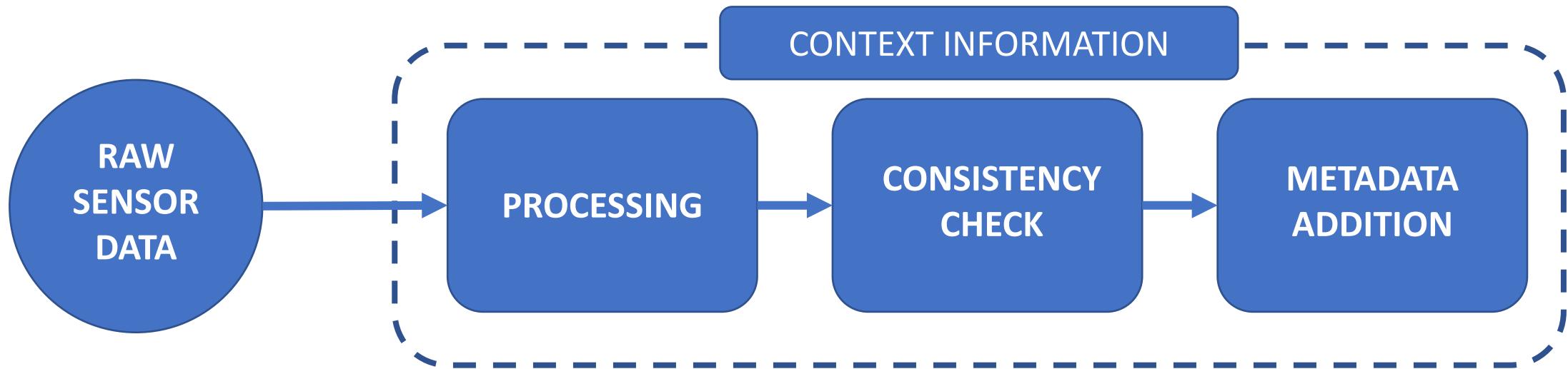
*Any **information** that can be used to **characterize the situation** of an entity (Dey et al. [2]).*

- **Entity**

*A person, place, or object that is considered **relevant to the interaction** between a **user** and an **application**, including the user and applications themselves (Dey et al. [2]).*

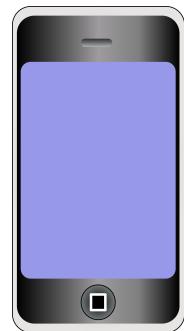
DEFINITIONS

CONTEXT INFORMATION

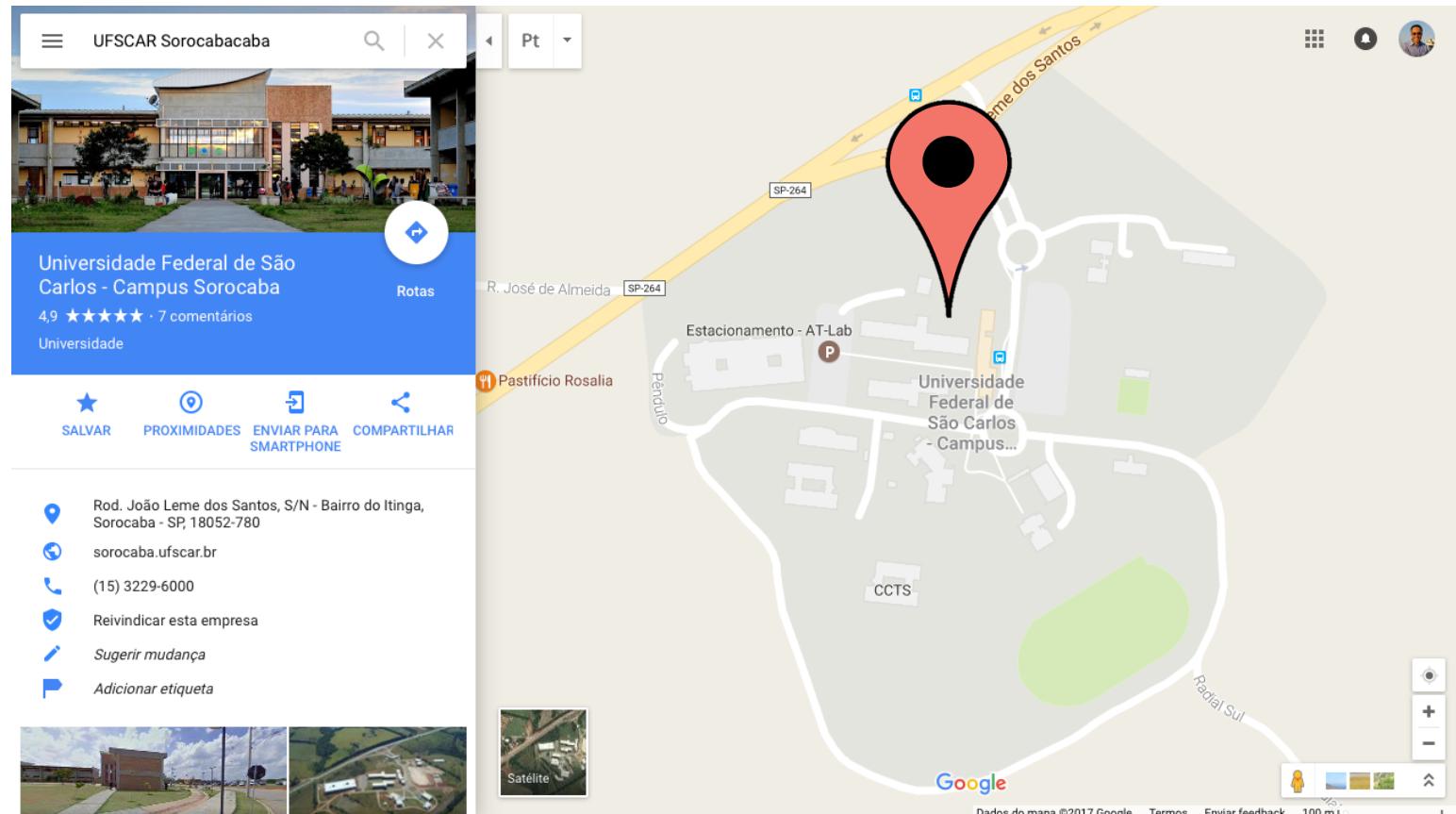


DEFINITIONS

CONTEXT & ENTITY EXAMPLE – GEO-LOCALIZATION



GPS SENSOR
(RAW DATA)



GEOGRAPHICAL LOCATION (CONTEXT INFORMATION)

DEFINITIONS

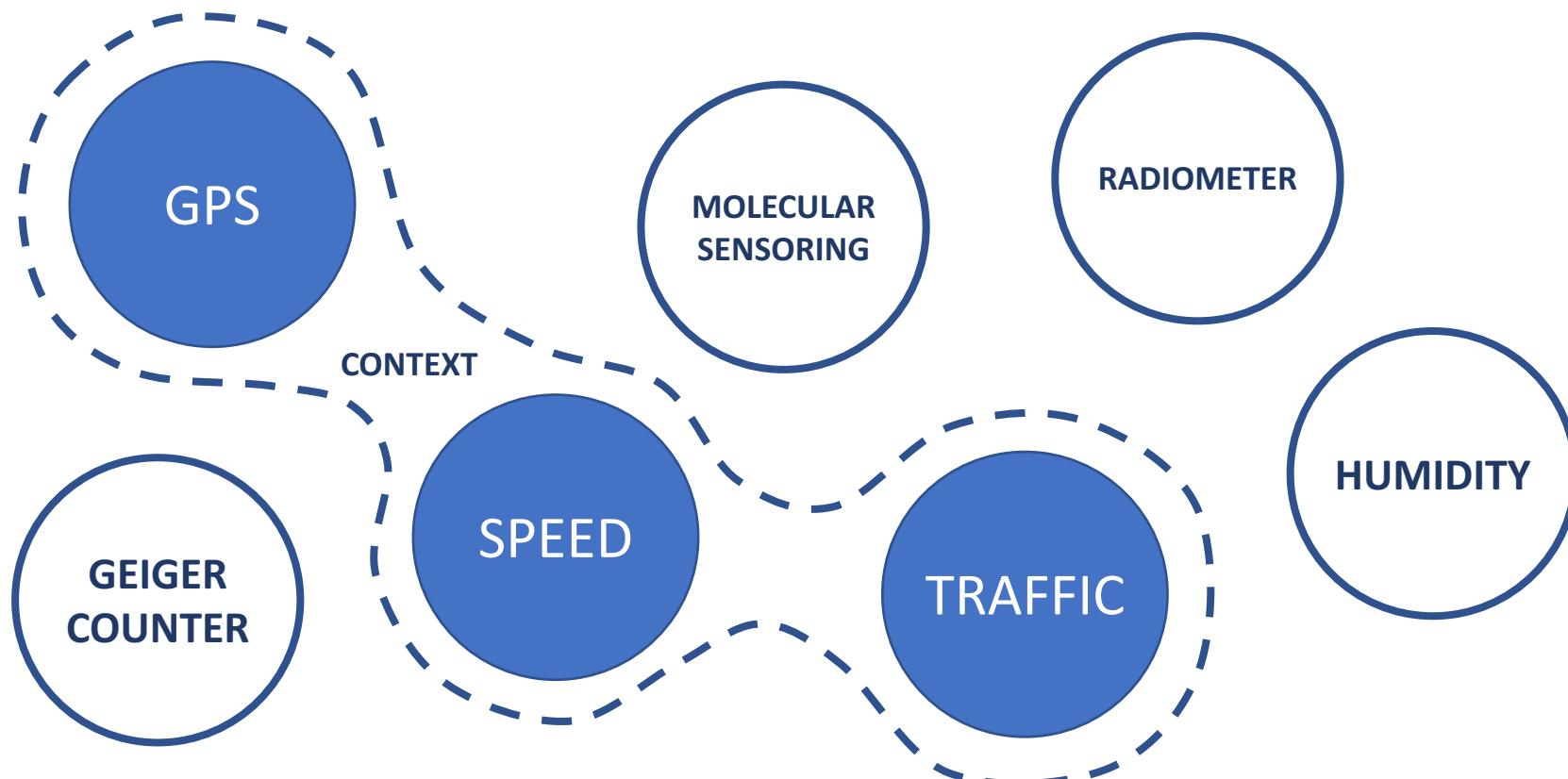
CONTEXT AND RAW DATA



The raw data values produced by sensors can be considered as data. If this data can be used to generate context information, we identify these data as context (*Pereira et al. [1]*).

DEFINITIONS

CONTEXT AND RAW DATA EXAMPLE – GEO-LOCALIZATION



DEFINITIONS

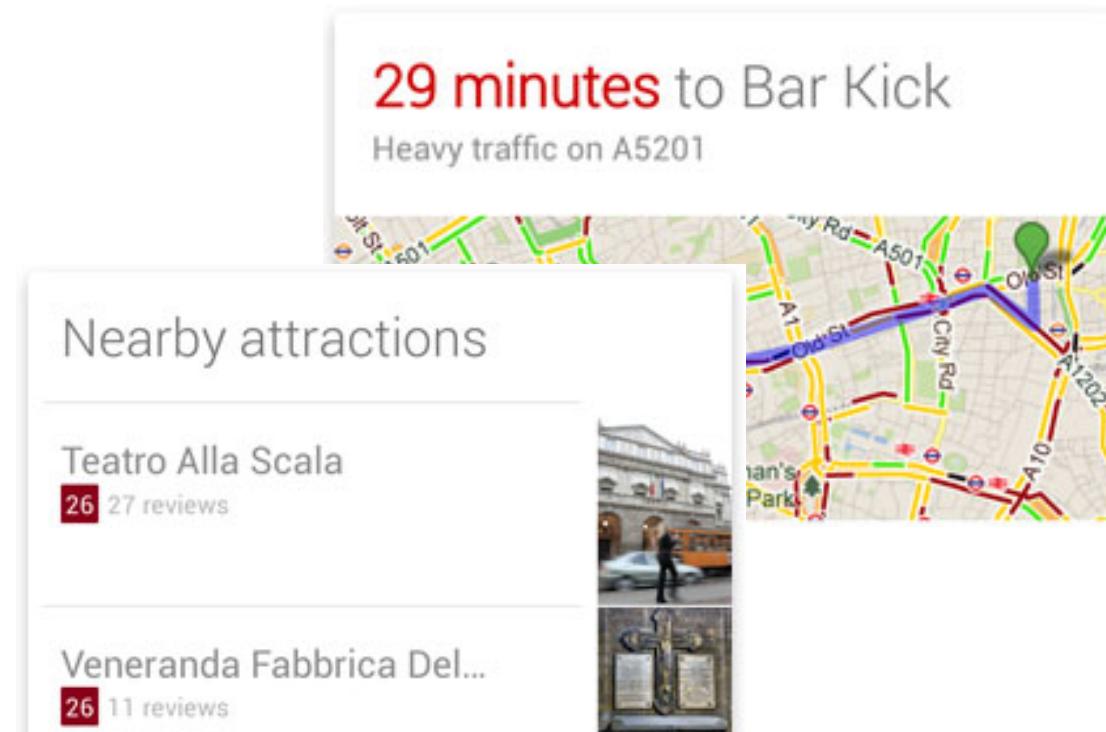
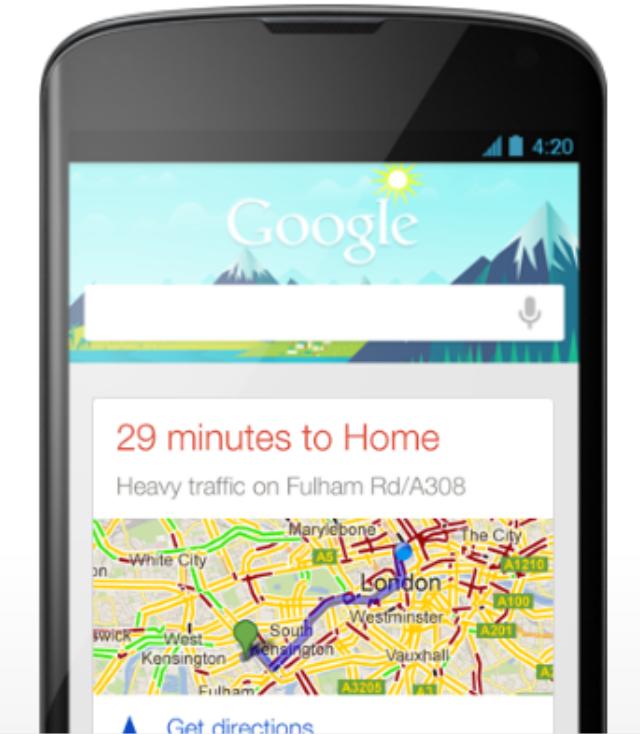
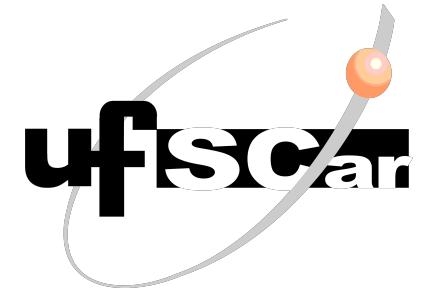
CONTEXT AWARENESS



A system is context-aware if it uses context to provide relevant information and/or services to the user, where relevancy depends on the user's task (*Pereira et al. [1]*).

DEFINITIONS

CONTEXT AWARENESS



*Pictures from Google®
page*

DEFINITIONS

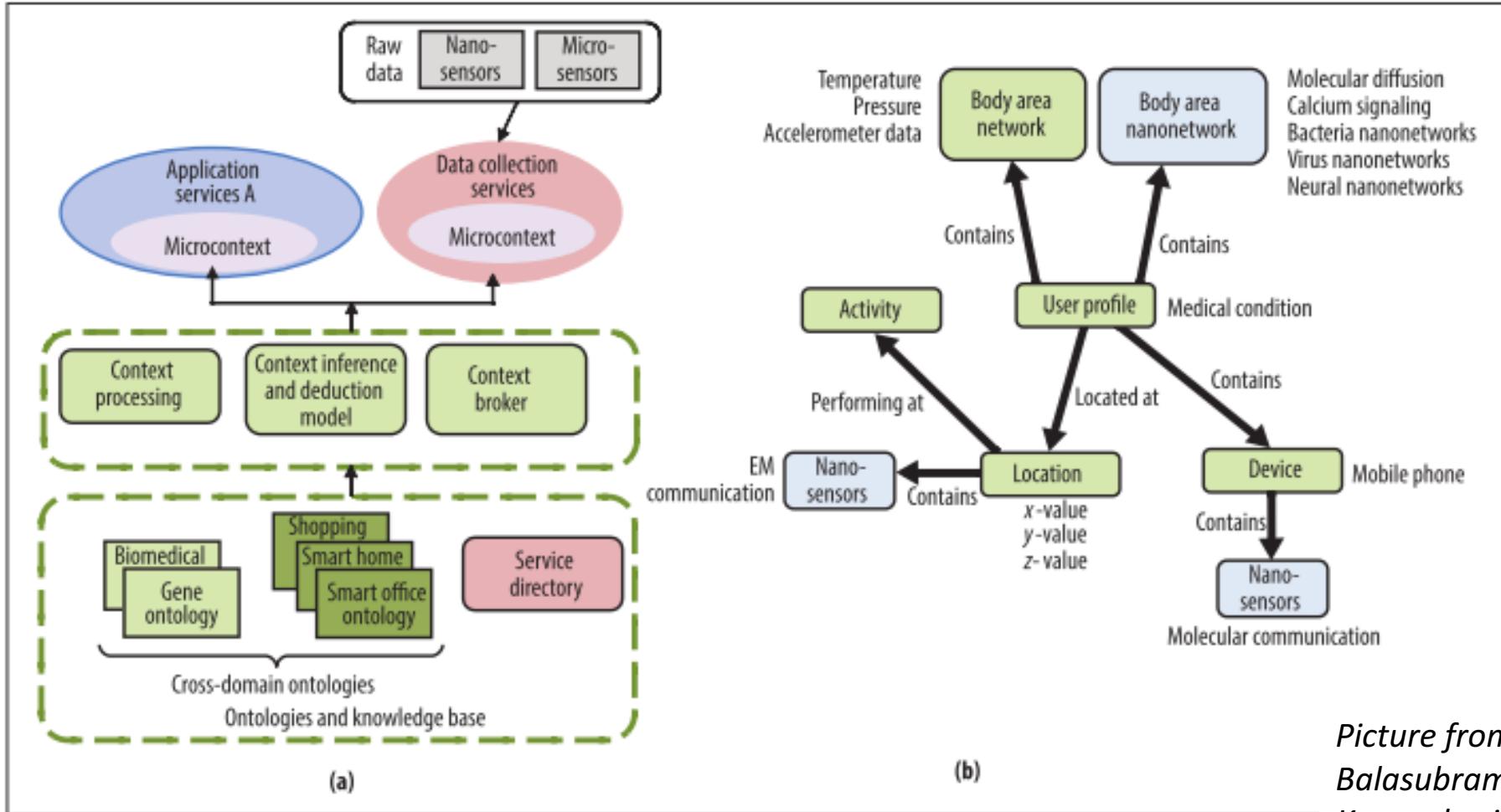
CONTEXT MODEL



Identifies a concrete subset of the context that is realistically *obtainable* from sensors, applications and users and able to be *used* in the execution of the task (*Pereira et al. [1]*).

DEFINITIONS

CONTEXT MODEL



Picture from
Balasubramaniam and
Kangasharju [6]

DEFINITIONS

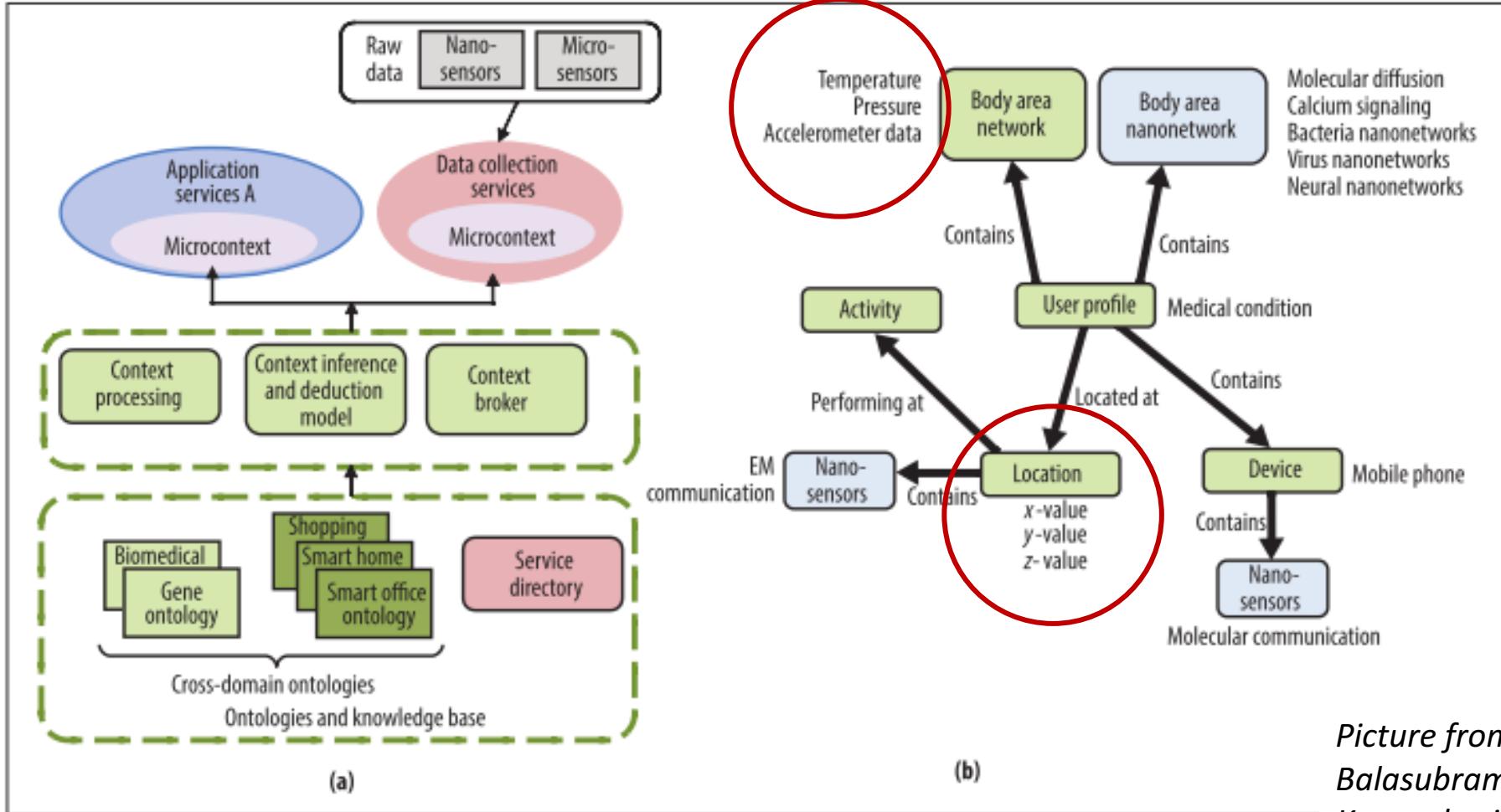
CONTEXT ATTRIBUTE



It is an element of the context model describing the context. A context attribute has an identifier, a type and a value, and optionally a collection of properties describing specific characteristics (*Pereira et al. [1]*).

DEFINITIONS

CONTEXT ATTRIBUTE



Picture from
Balasubramaniam and
Kangasharju [6]

DEFINITIONS

CONTEXT-AWARE FEATURES



- **Presentation**

Context can be used to decide what information and services need to be presented to the user (Pereira et al. [1]).

- **Execution**

Automatic execution of services (Pereira et al. [1]).

- **Tagging**

Context needs to be tagged together with the sensor data to be processed and understood later (Pereira et al. [1]).

DEFINITIONS

CONTEXT-AWARE FEATURES



- Presentation
- Execution
- Tagging

DEFINITIONS

CONTEXT CATEGORIZATION

- **Primary**

Any information retrieved without using existing context and without performing any kind of sensor data fusion operations.

- **Secondary**

Any information that can be computed using primary context.

DEFINITIONS

CONTEXT CATEGORIZATION



Categories of Context (Operational Perspective)				
	Primary	Secondary		
Location	Location data from GPS sensor (e.g. longitude and latitude)	Distance of two sensors computed using GPS values Image of a map retrieved from map service provider	Primary	Secondary
Identity	Identify user based on RFID tag	Retrieve friend list from users Facebook profile Identify a face of a person using facial recognition system		
Time	Read time from a clock	Calculate the season based on the weather information Predict the time based on the current activity and calendar		
Activity	Identify opening door activity from a door sensor	Predict the user activity based on the user calendar Find the user activity based on mobile phone sensors such as GPS, gyroscope, accelerometer		

The diagram illustrates the relationship between operational and conceptual contexts. On the left, a vertical stack of four boxes represents conceptual categories: Location, Identity, Time, and Activity. To the right of each category is a horizontal table with two columns: Primary (blue) and Secondary (orange). Red dashed arrows point from the conceptual categories to their corresponding operational definitions. Below the tables are icons representing the operational concepts: a location pin, a person icon with a signal, a clock, and a door sensor.

Categories of Context (Conceptual Perspective)

Categories of Context (Operational Perspective)

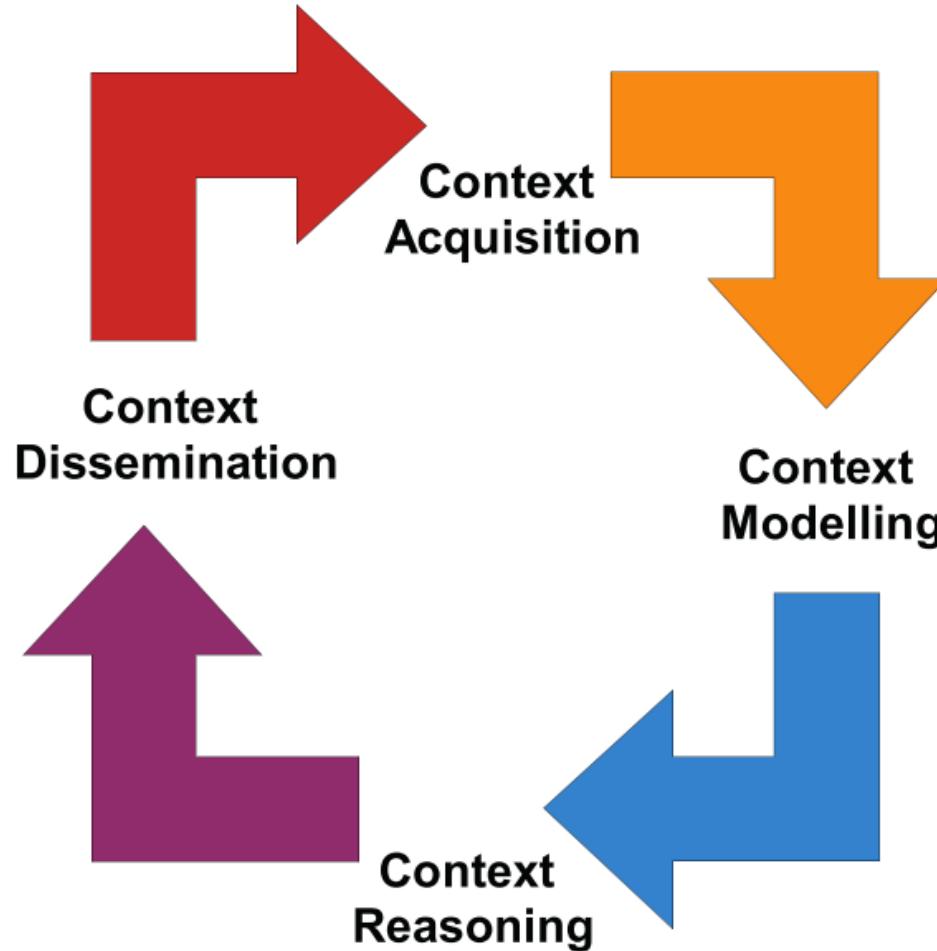
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Icons and Arrows:

- Location:** A red location pin icon above a blue box labeled "Primary". A red dashed arrow points from the conceptual "Location" box to the operational "Primary" box.
- Identity:** An icon of a person with concentric circles around them above a blue box labeled "Primary". A red dashed arrow points from the conceptual "Identity" box to the operational "Primary" box.
- Time:** A clock icon above a blue box labeled "Primary". A red dashed arrow points from the conceptual "Time" box to the operational "Primary" box.
- Activity:** An icon of a door with a person walking through it above a blue box labeled "Primary". A red dashed arrow points from the conceptual "Activity" box to the operational "Primary" box.
- Secondary:** Icons for "Secondary" categories are shown in orange boxes: a map image for Location, a Facebook logo for Identity, weather icons for Time, and a mobile phone icon for Activity.

DEFINITIONS

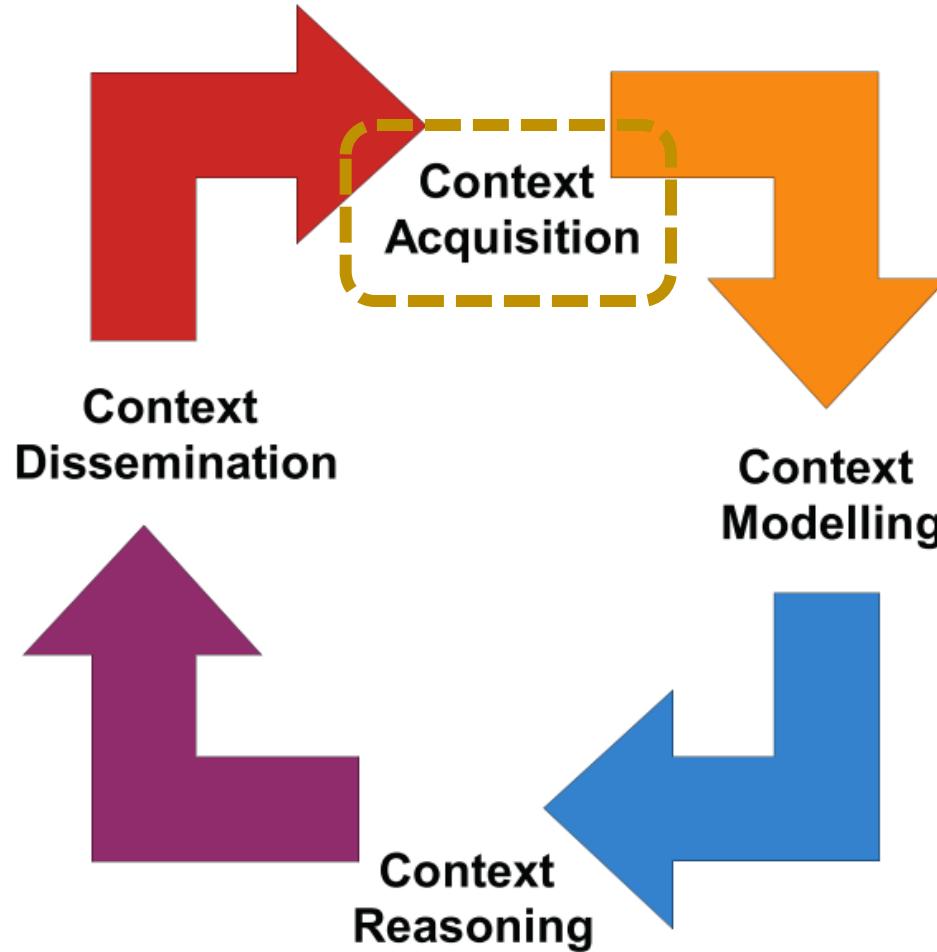
LIFE CYCLE



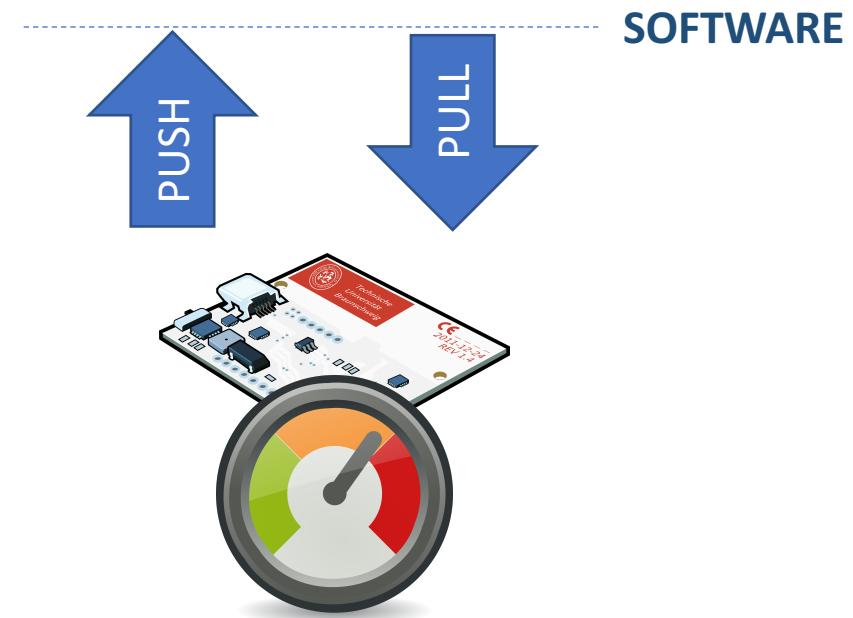
A data life cycle shows **how data moves from phase to phase in software systems**
...explaining... where the data is generated and where the data is consumed. (Pereira et al. [1]).

DEFINITIONS

CONTEXT LIFE CYCLE – CONTEXT ACQUISITION

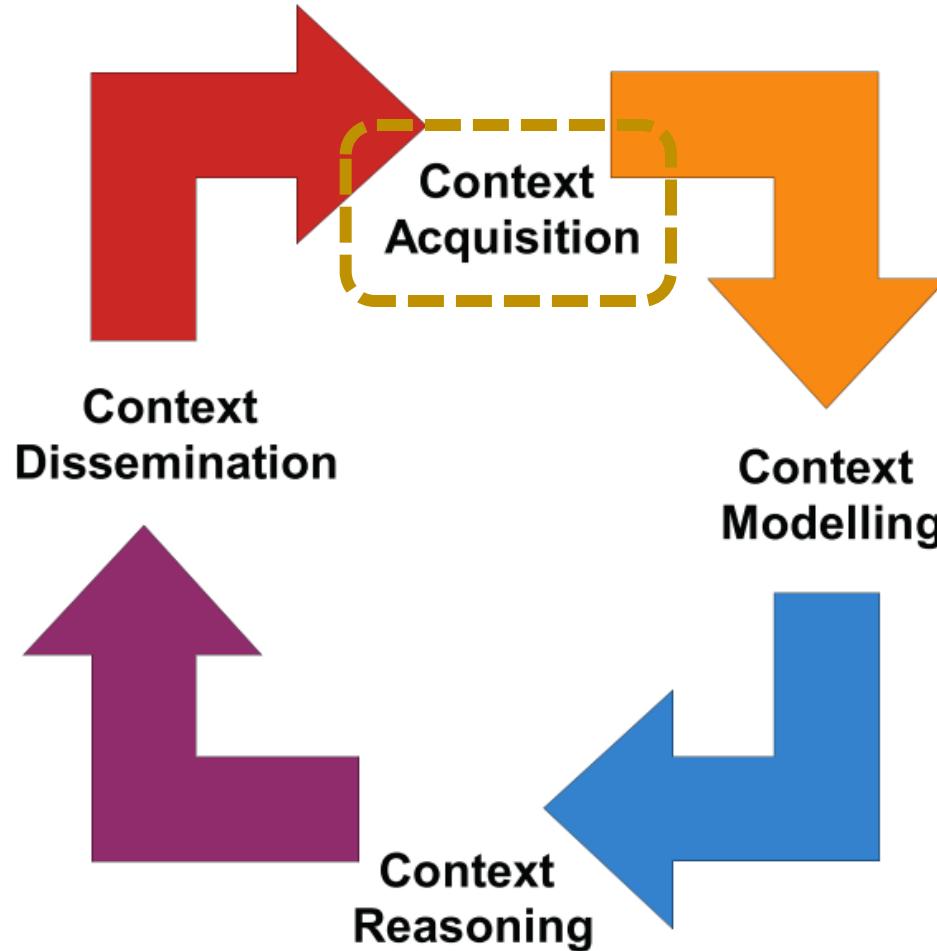


- Based on responsibility:

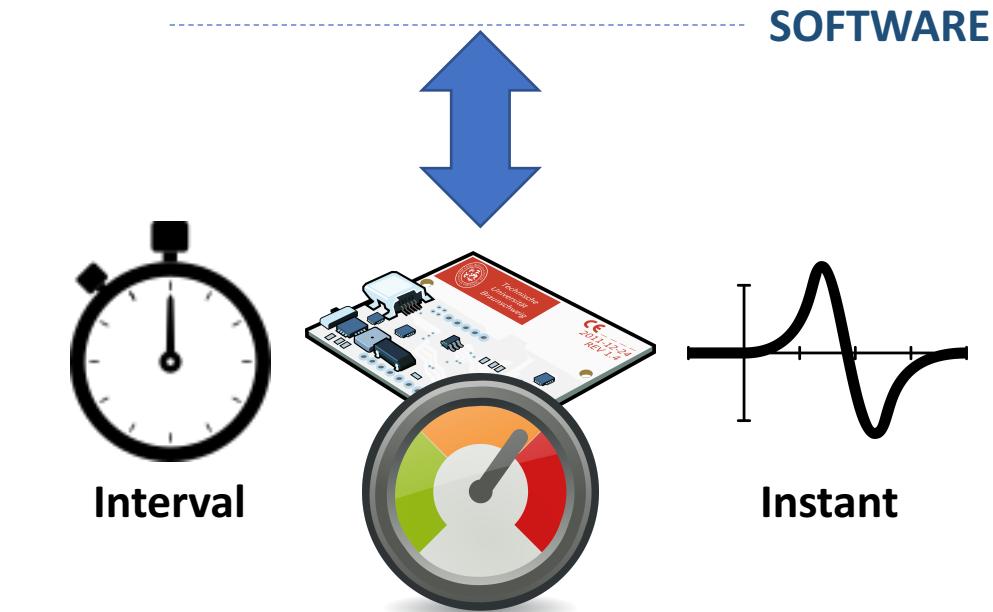


DEFINITIONS

CONTEXT LIFE CYCLE – CONTEXT ACQUISITION

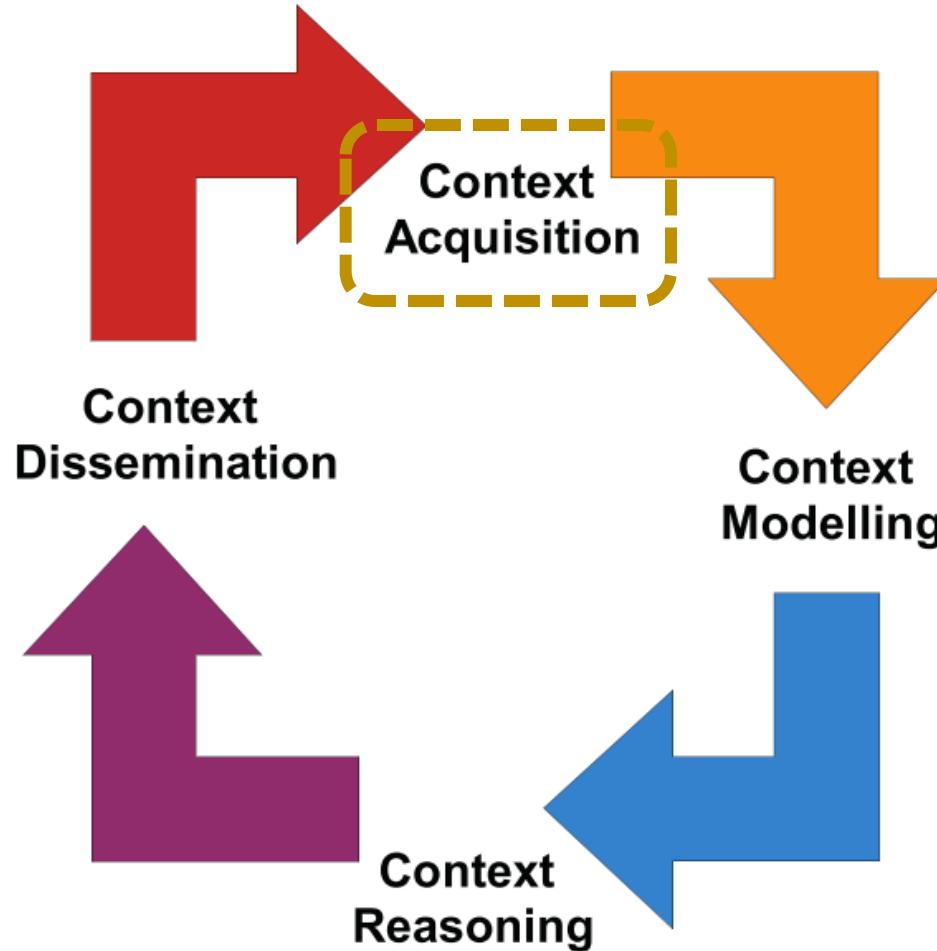


- Based on frequency:

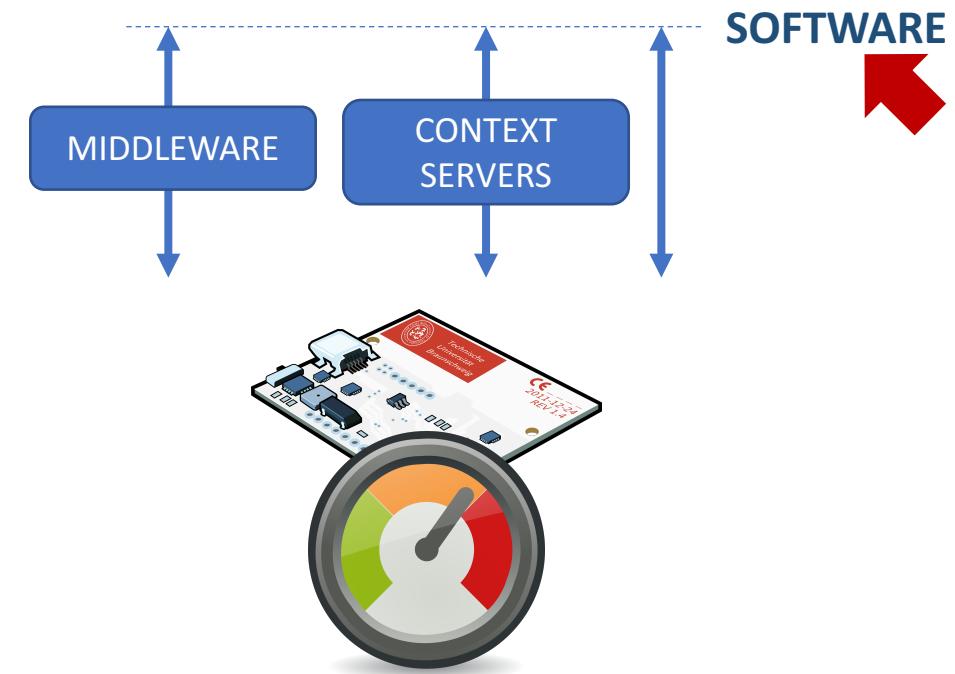


DEFINITIONS

CONTEXT LIFE CYCLE – CONTEXT ACQUISITION

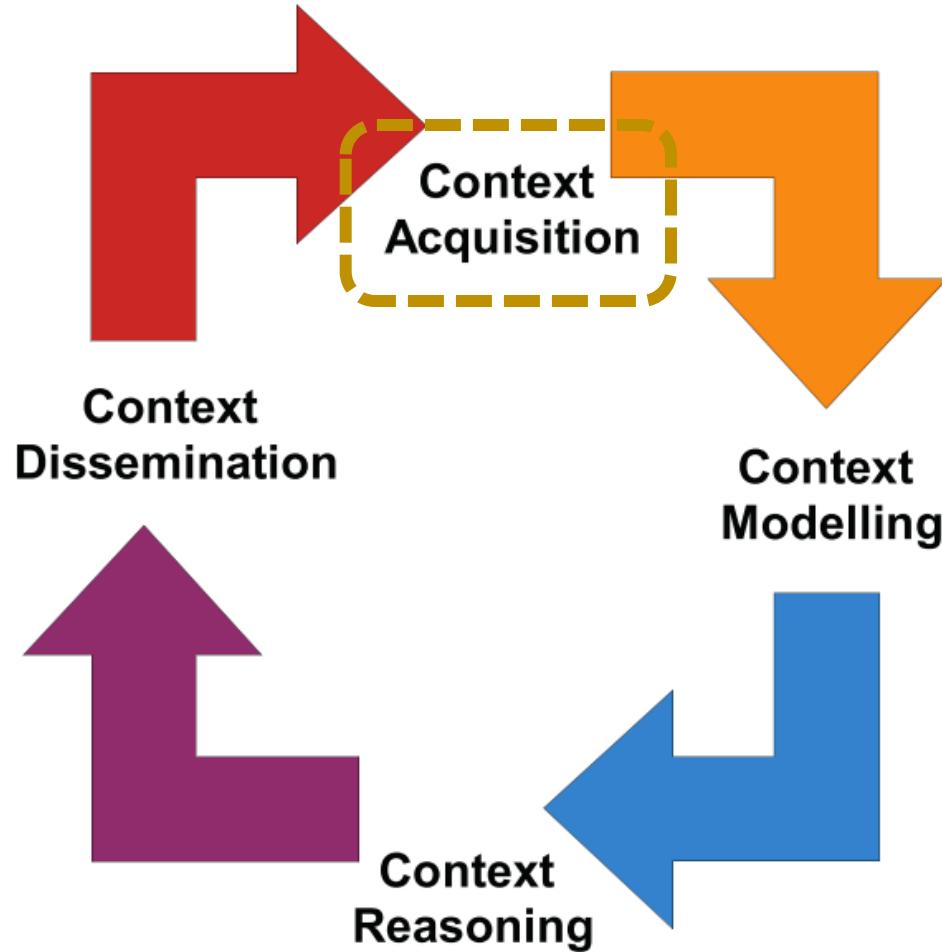


- Based on source:

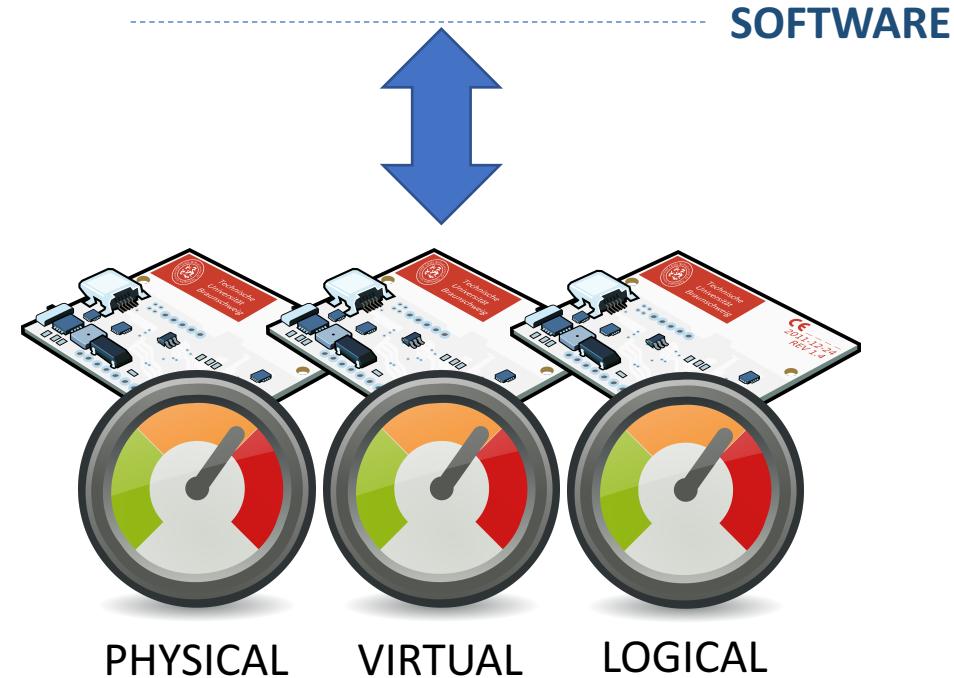


DEFINITIONS

CONTEXT LIFE CYCLE – CONTEXT ACQUISITION

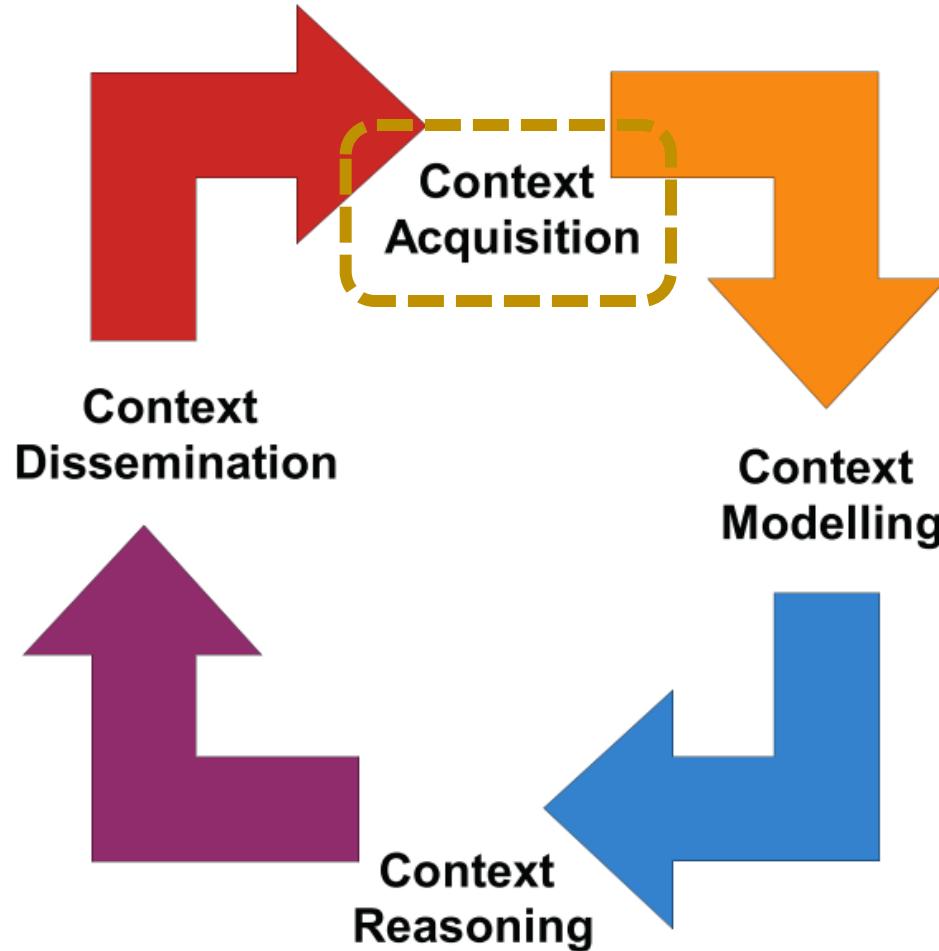


- Based on sensor type:



DEFINITIONS

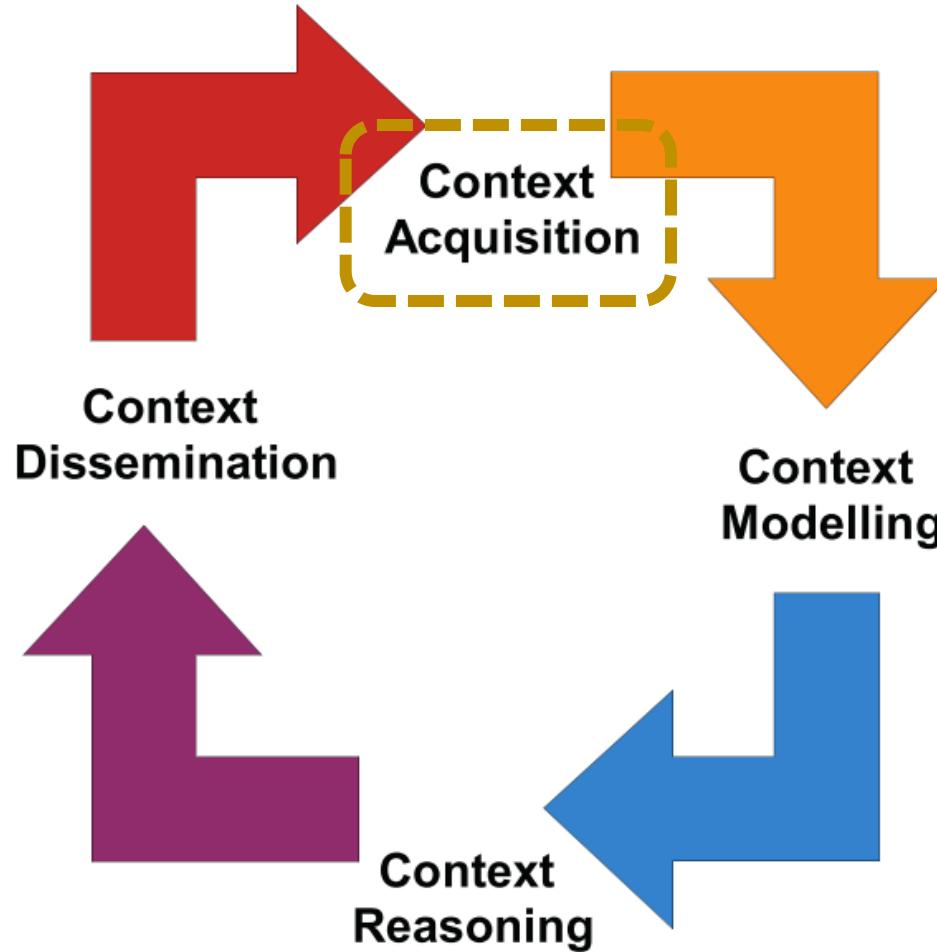
CONTEXT LIFE CYCLE – CONTEXT ACQUISITION



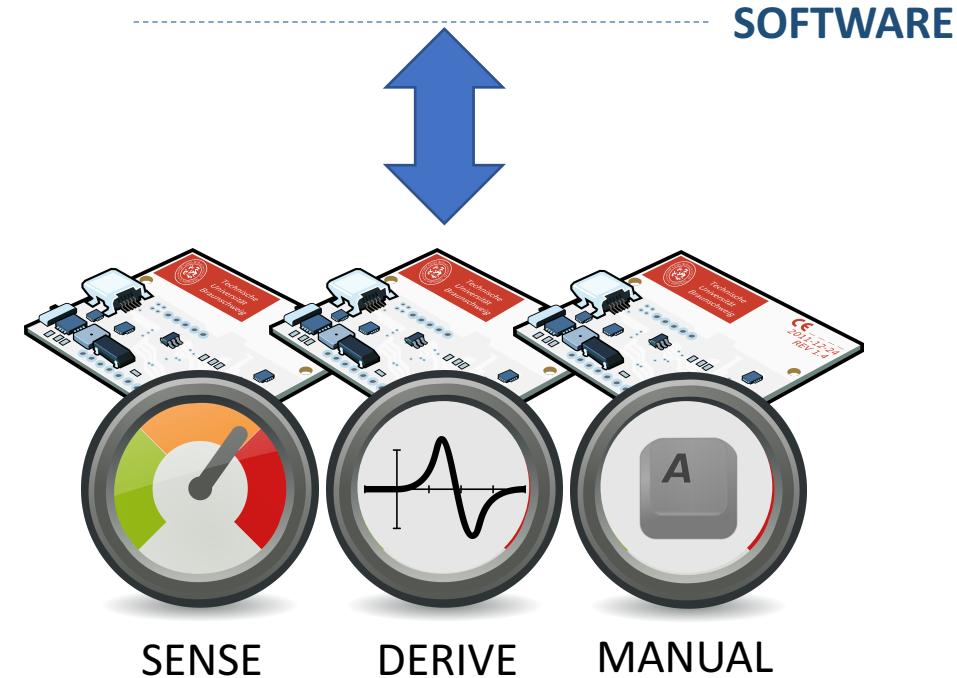
- Based on sensor type:
 - Physical
the actual sensor
 - Virtual
twitter statuses, calendar
 - Logical
physical + virtual
weather information web server
(physical temperature sensor + virtual sensors such as maps, calendars,...)

DEFINITIONS

CONTEXT LIFE CYCLE – CONTEXT ACQUISITION

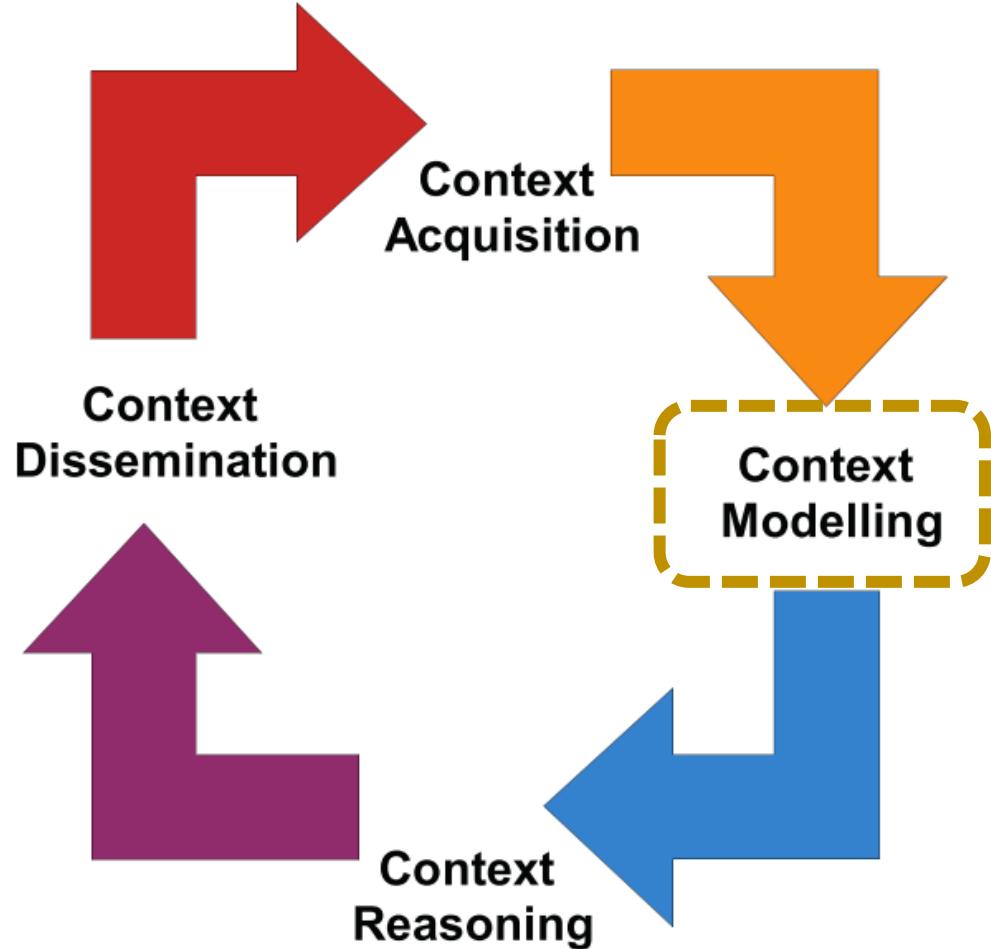


- Based on acquisition process:



DEFINITIONS

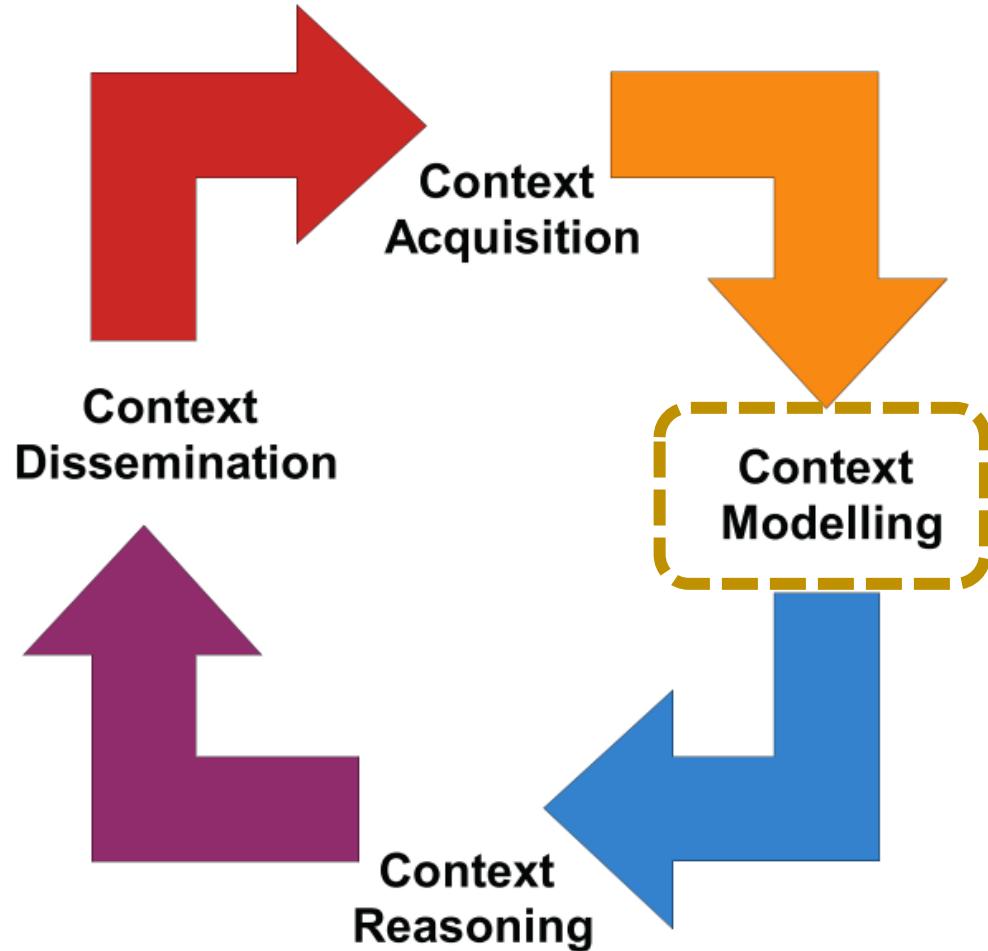
CONTEXT LIFE CYCLE – CONTEXT MODELLING



Define context information in terms of attributes, validate the attributes and then merge these attributes.

DEFINITIONS

CONTEXT LIFE CYCLE – CONTEXT MODELLING

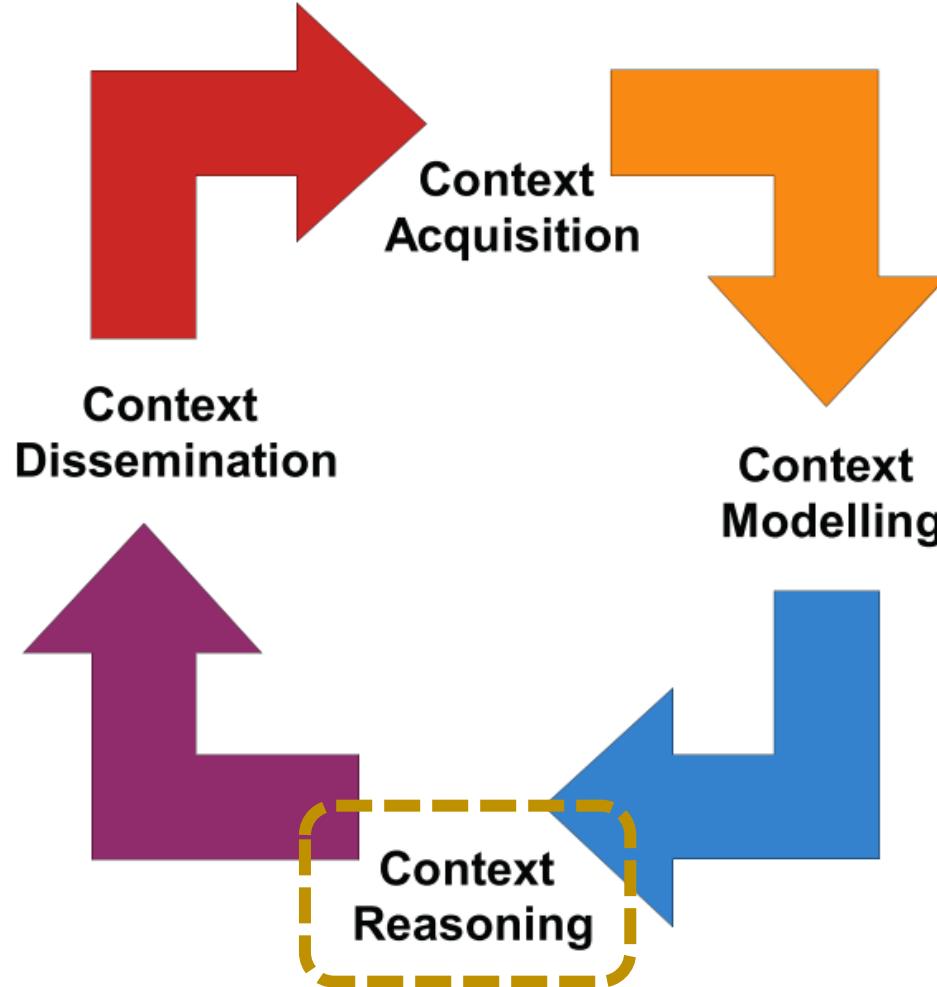


Modeling techniques:

- key-value
- markup scheme (XML)
- graphical (database)
- object based
- logic based
- ontology based

DEFINITIONS

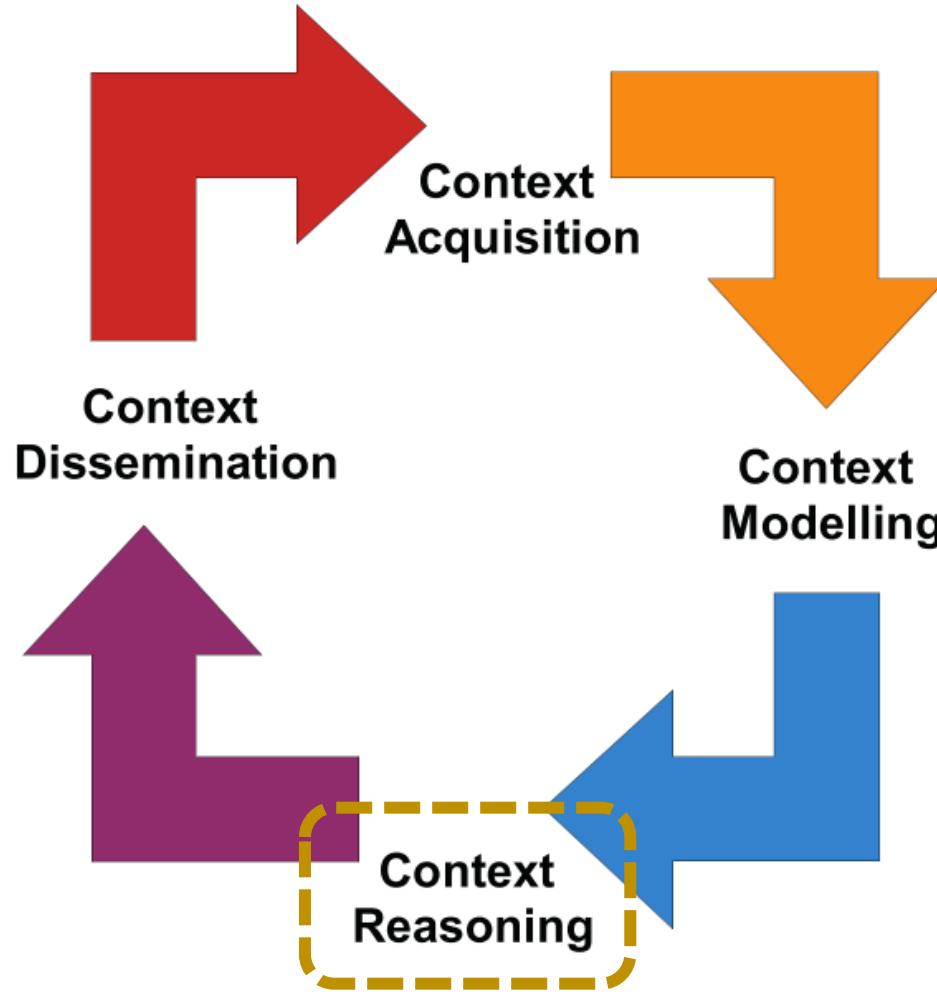
CONTEXT LIFE CYCLE – CONTEXT REASONING



Context reasoning can be defined as a method of **deducing new knowledge**, and understanding better, based on the available context (*Bikakis et al. [8]*).

DEFINITIONS

CONTEXT LIFE CYCLE – CONTEXT REASONING

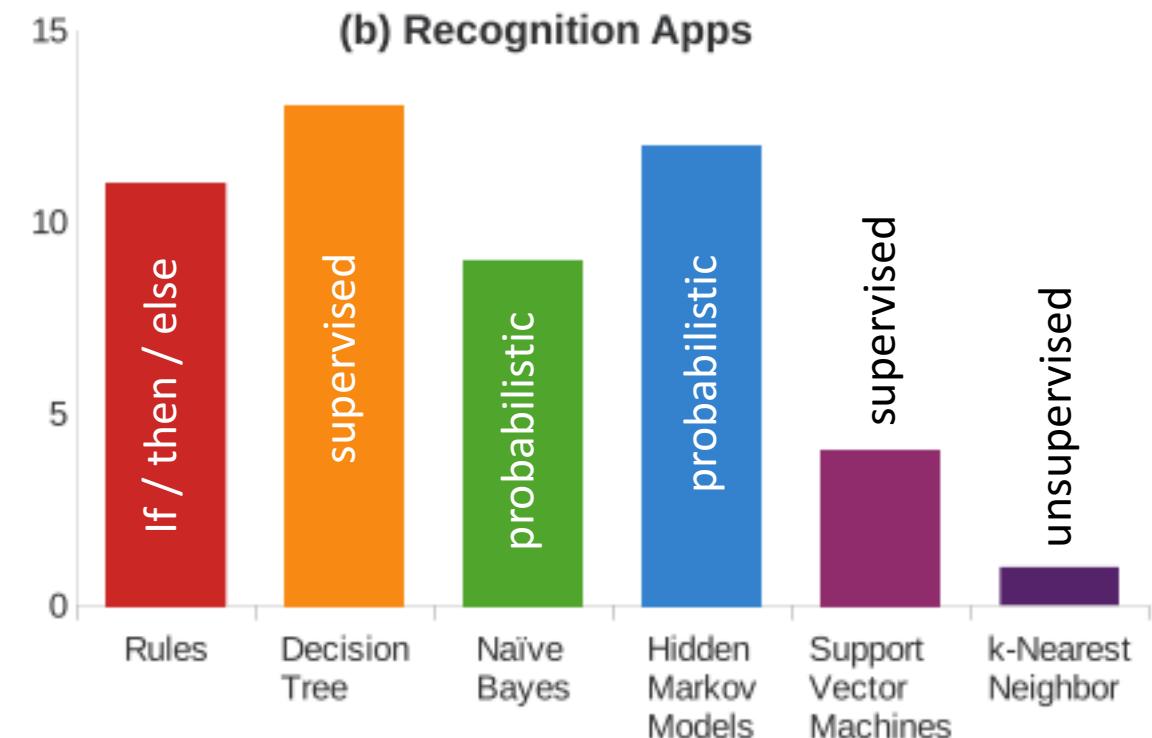
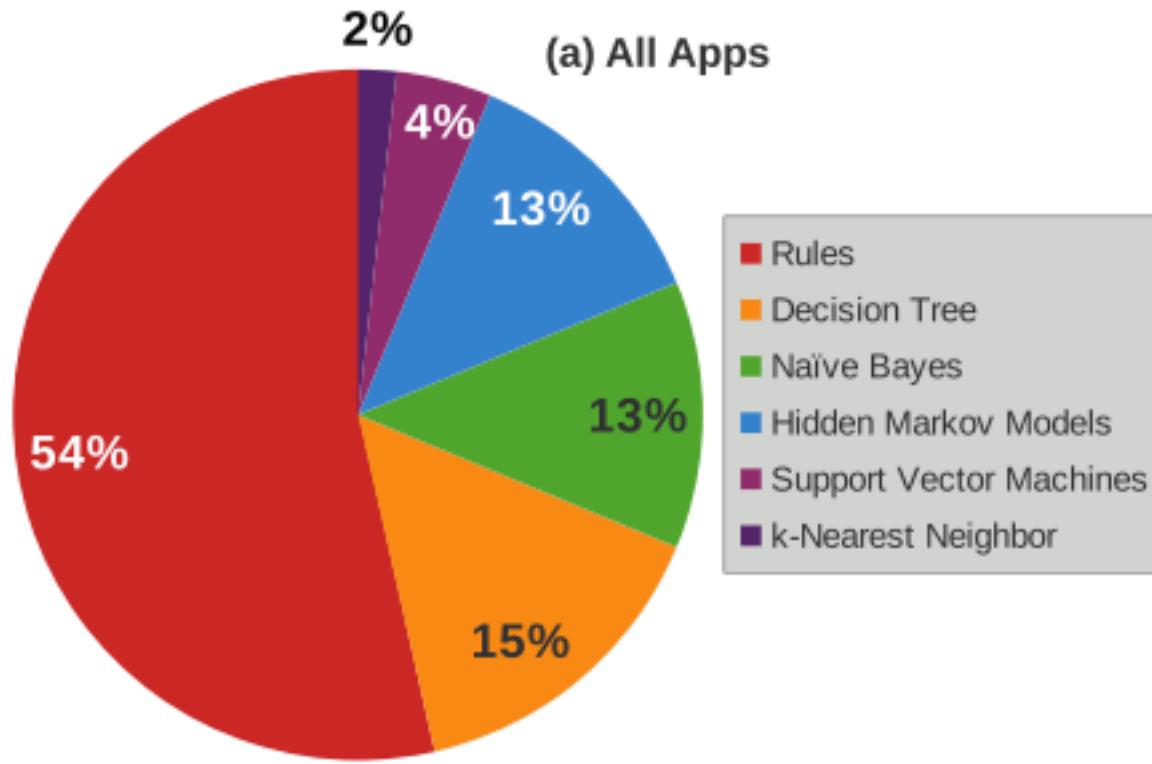


categories:

- supervised learning,
- unsupervised learning,
- rules,
- fuzzy logic,
- ontological reasoning
- probabilistic reasoning

DEFINITIONS

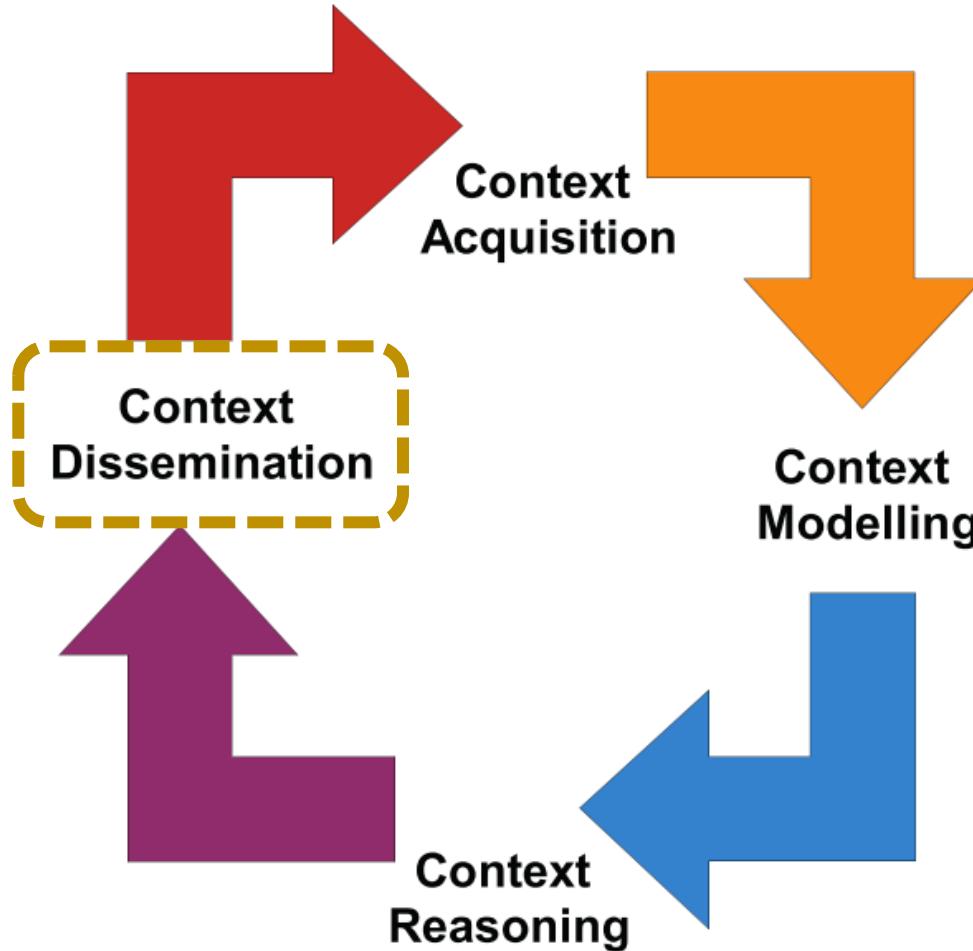
CONTEXT LIFE CYCLE – CONTEXT REASONING



(a) Counts of model types used in 109 of 114 reviewed context-aware applications and (b) counts for 50 recognition applications (*Pereira et al. [1]*).

DEFINITIONS

CONTEXT LIFE CYCLE – CONTEXT DISTRIBUTION



Commonly used methods

- **Query**
Context consumer makes the request
- **Subscription**
Context consumer subscribes

USE CASE #1

WANTED DRUG DEALER



Red car with license plate
XXX-1234 wanted for drug
dealing

USE CASE #1

WANTED DRUG DEALER



License Plate Recognition system located many times the car, but nothing was found after police inspection.

USE CASE #1

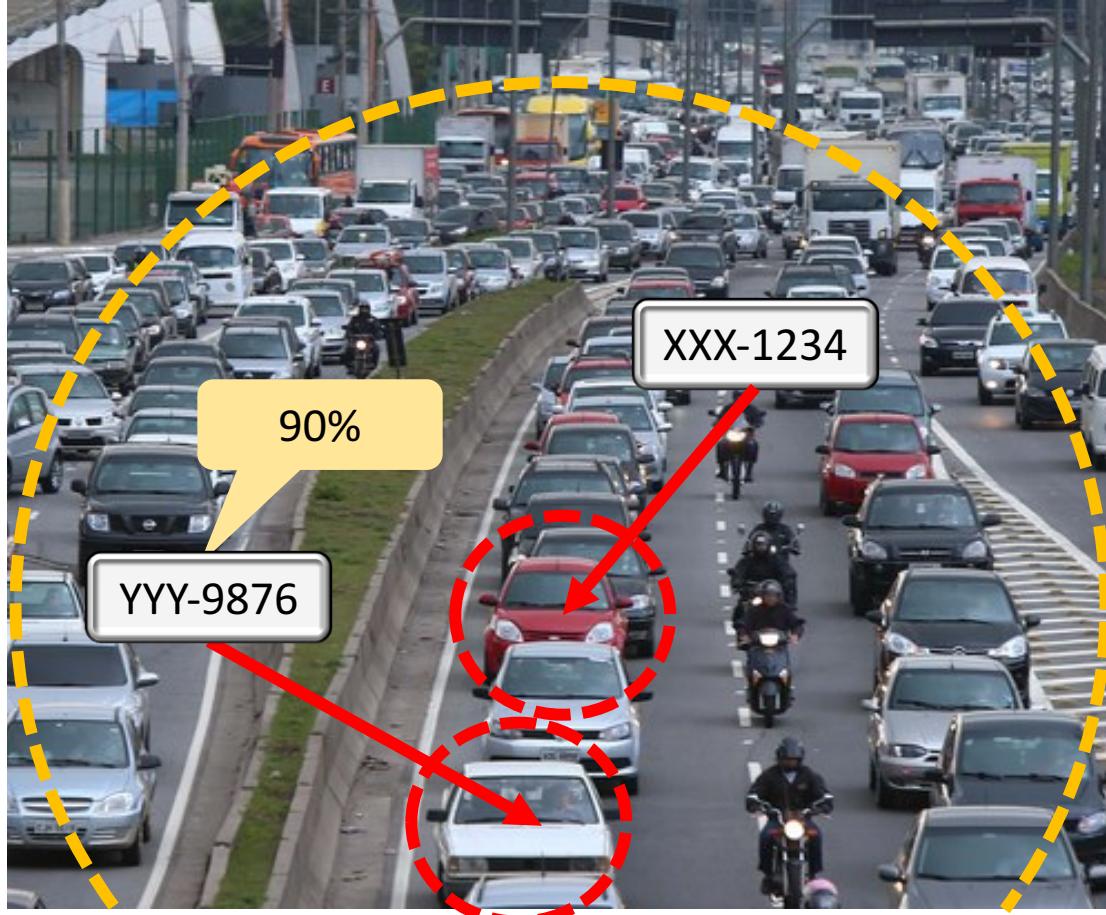
WANTED DRUG DEALER



Using context aware technologies, the system identified and stored in a database the closest cars to the one wanted.

USE CASE #1

WANTED DRUG DEALER



In 90% of the cases, a white car with license plate YYY-9876 was close to the wanted red one.
After police inspections, the white car was identified as the drug transport.

Microsoft Domain Aware System (DAS),
aka Detecta in Brazil

USE CASE #2

CRIME INVESTIGATION



A woman died by asphyxiation in her apartment.

The husband was arrested for the crime.

USE CASE #2

CRIME INVESTIGATION



Police started investigations and after entering the apartment address in the system, it was found that neighbors called the gas company complaining about gas leakage several times.

USE CASE #2

CRIME INVESTIGATION

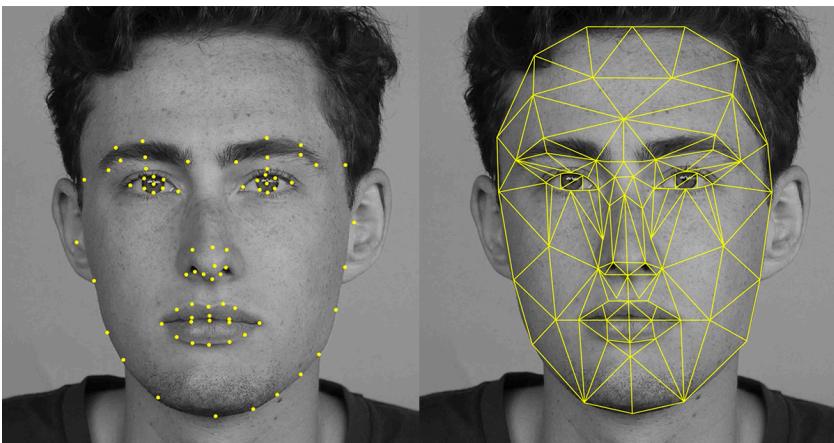
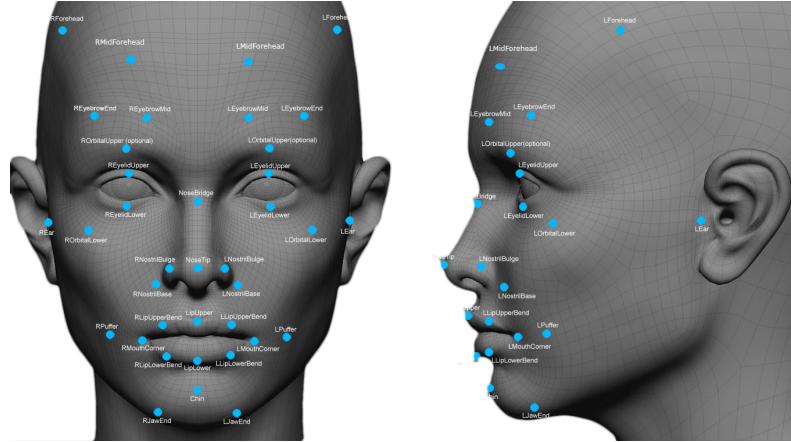


By using context aware concepts the husband was acquitted of the crime.

Microsoft Domain Aware System (DAS),
aka Detecta in Brazil

USE CASE #3

FACIAL RECOGNITION

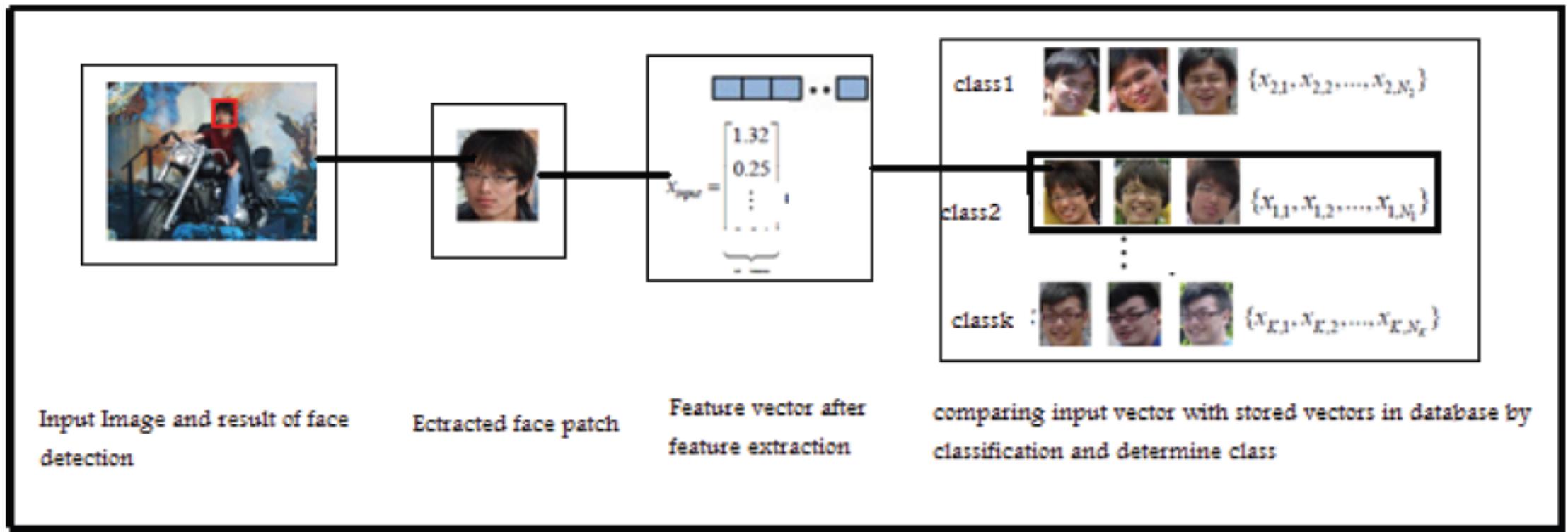


A good Facial Recognition System depends on many variables, not only the face itself, but the surrounding environment.

All these data must be considered in a context model.

USE CASE #3

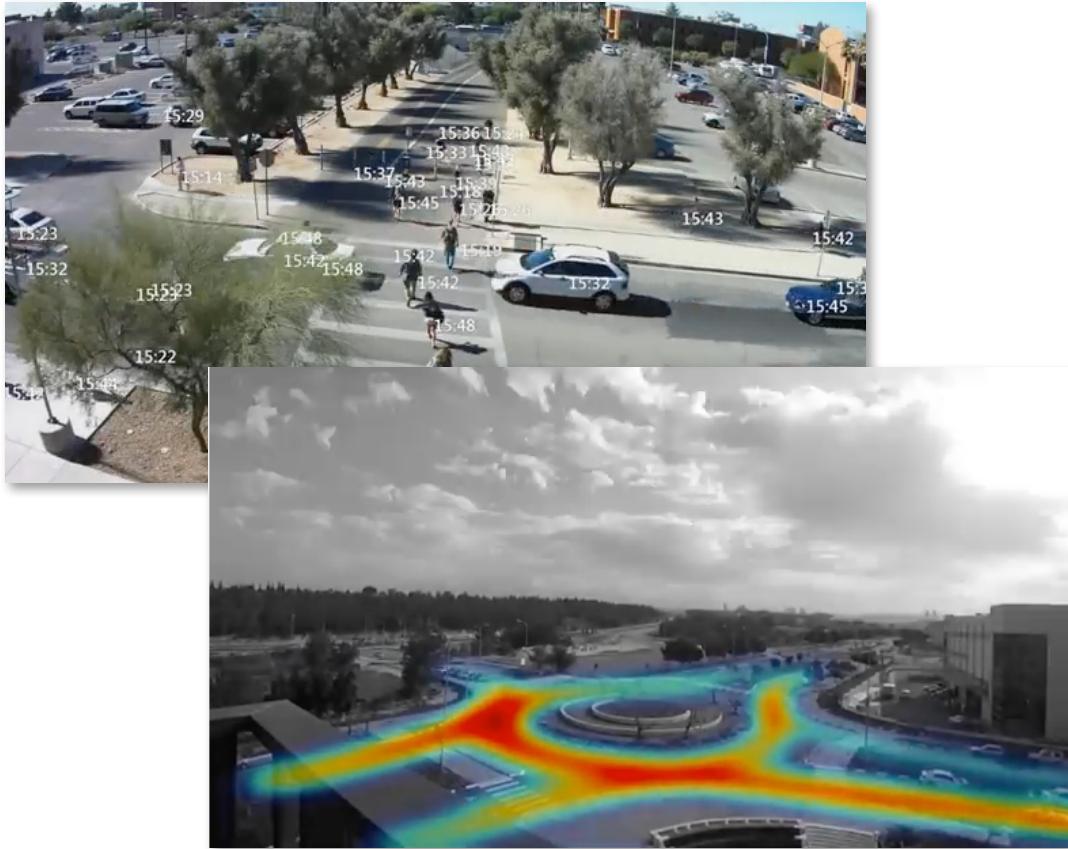
FACIAL RECOGNITION



Picture from
Sundaram and Mani [9]

USE CASE #4

VIDEO ANALYSIS



- The system automatically tags a context by colors, sizes, directions, etc.
 - Context presentation is implemented in order to make easier to the user to take an action.

USE CASE #4

VIDEO ANALYSIS



<https://www.youtube.com/watch?v=rY1qmXda0Oo>

- Entity:
cars
- Context:
time stamp, direction, license plate, speed,...
- Context-awareness:
BriefCam system
- Context model:
time stamp, direction, license plate.
- Context attribute:
attribute id, type = license plate, value = XXX-1234

CONCLUSION

The IoT has gained significant attention over the last few years, with the advances in sensor hardware technology and cheap materials. Sensors are expected to be attached to all the objects around us with minimum human intervention. Understanding sensor data is one of the main challenges that the IoT faces. **The results clearly show the importance of context awareness in the IoT paradigm (Pereira et al. [1]).**

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