



**POLITECNICO
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In collaboration with



IoT Notifications: from disruption to benefit

Architectures for the future of notifications in the IoT

Supervisor(s)



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Presenter

Teodoro Montanaro

Research GOAL

Investigate the intelligence component in **Internet of Things (IoT)** architectures and applications: study, define, and prototype **intelligent distributed architectures** that may extract additional value and intelligent behaviors to some significant sample problems, representative of future IoT scenarios.



The **distribution and customization of notifications** in the IoT domain has been treated as an example of possible future IoT scenarios.

Notification Context: sample scenario



Source: <https://iot.do/windstream-research-future-connected-home-community-2015-04>

Notification Context: sample scenario

Date: 9th September 2018



Source: <https://iot.do/windstream-research-future-connected-home-community-2015-04>

Notification Context: sample scenario

Date: 9th September 2018

Time: 19.00



Source: <https://iot.do/windstream-research-future-connected-home-community-2015-04>

Notification Context: sample scenario

Date: 9th September 2018

Time: 19.00

5 people:

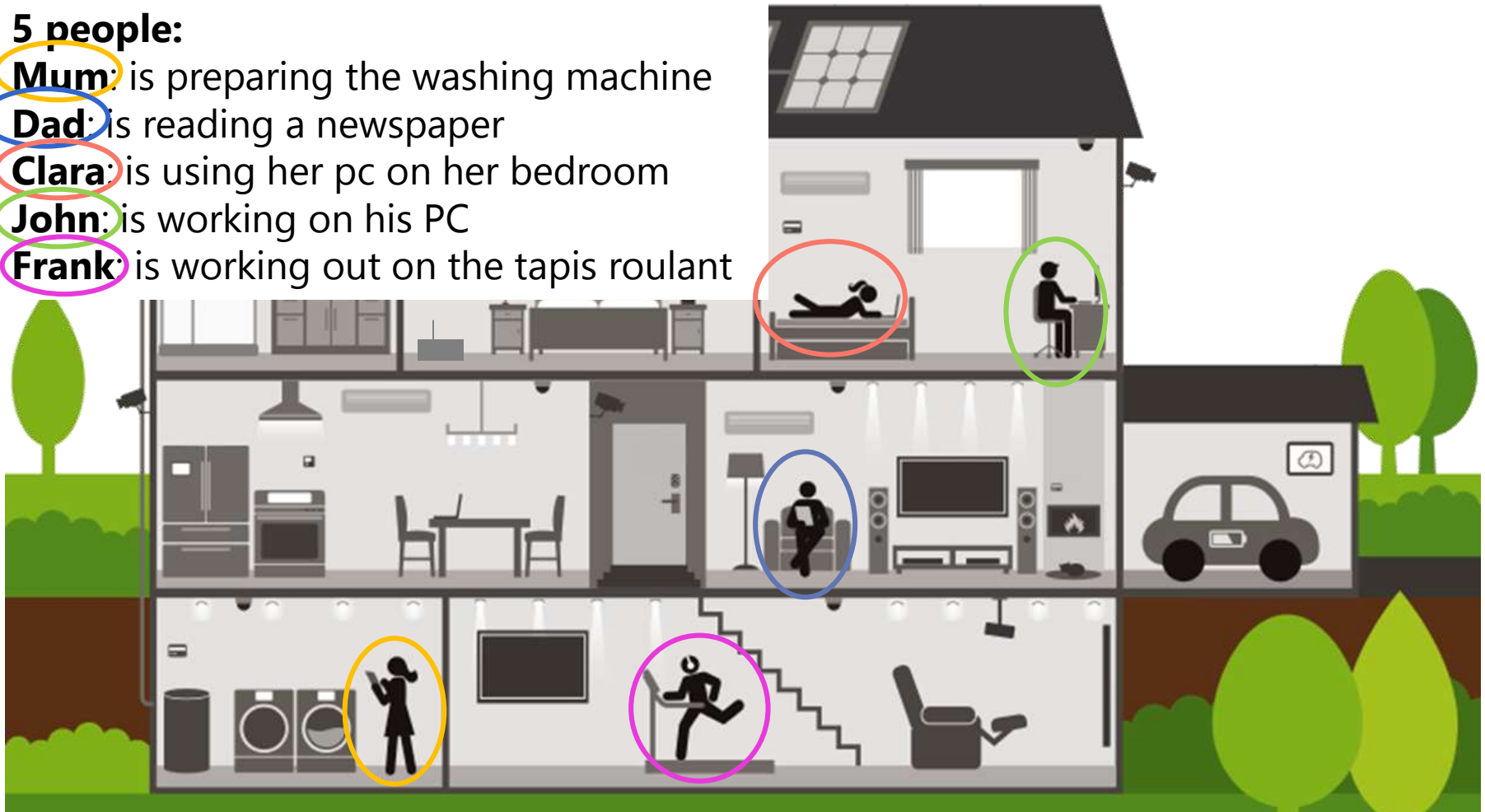
Mum: is preparing the washing machine

Dad: is reading a newspaper

Clara: is using her pc on her bedroom

John: is working on his PC

Frank: is working out on the tapis roulant



Source: <https://iot.do/windstream-research-future-connected-home-community-2015-04>

Notification Context: sample scenario

Date: 9th September 2018

Time: 19.00

5 people:
Mum
Dad
Clara
John
Frank



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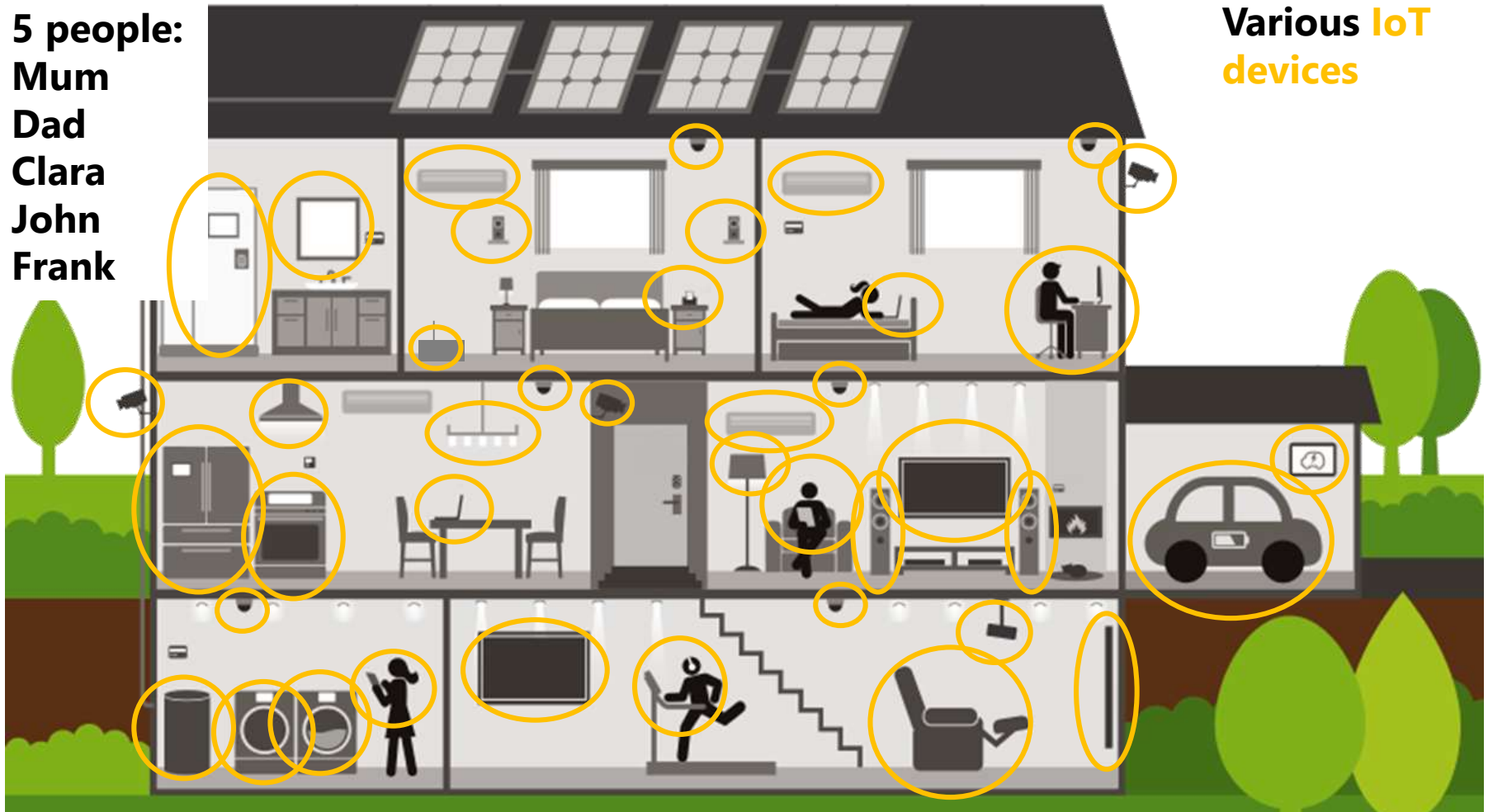
Notification Context: sample scenario

Date: 9th September 2018

Time: 19.00

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Various IoT devices



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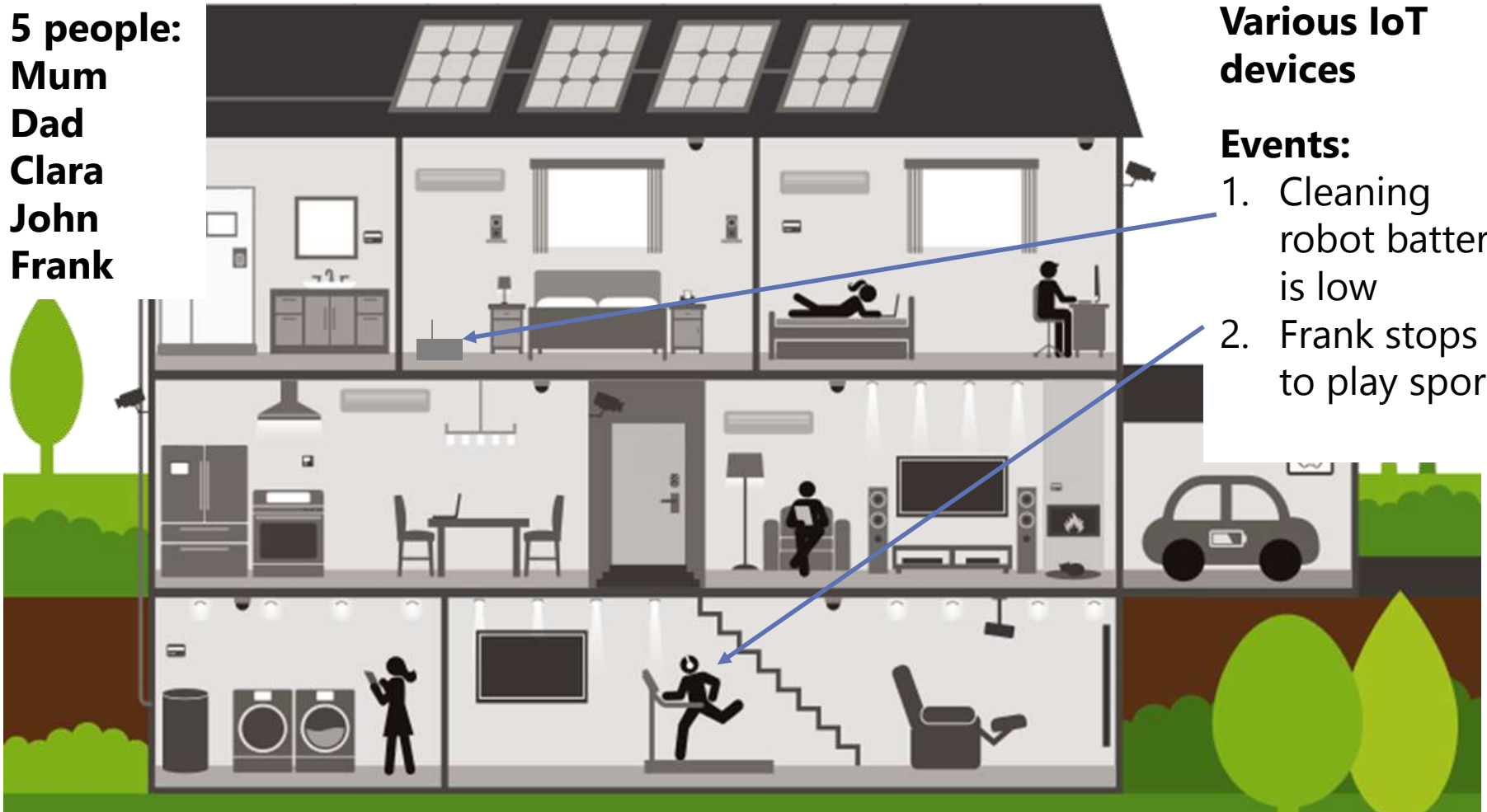
Time: 19.00

5 people:
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Dad
Clara
John
Frank

Various IoT devices

Events:

1. Cleaning robot battery is low
2. Frank stops to play sport



Source: <https://iot.do/windstream-research-future-connected-home-community-2015-04>

Notification Context: sample scenario

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Time: 19.00

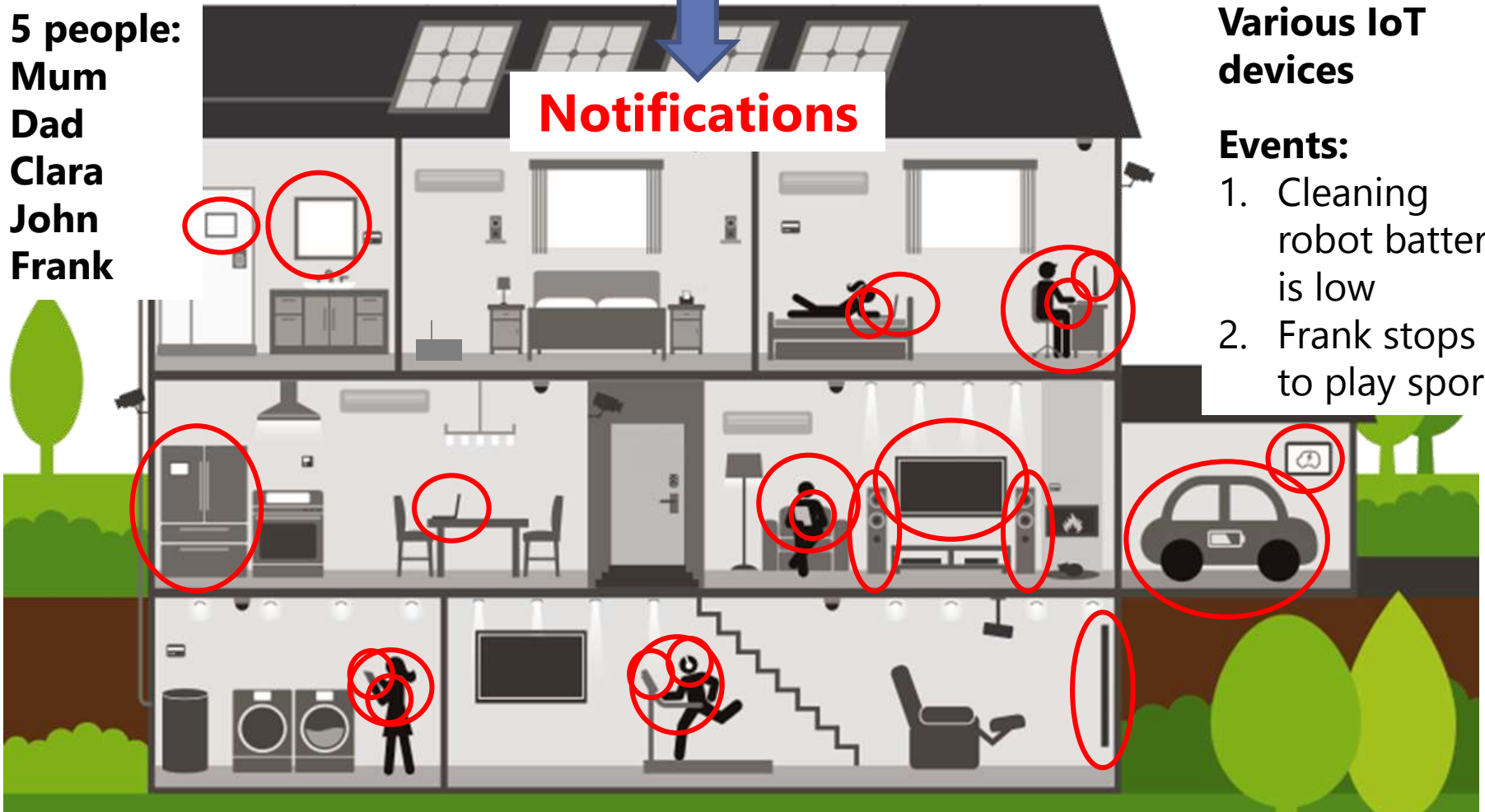
5 people:
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Frank

Notifications

Various IoT devices

Events:

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2. Frank stops to play sport



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Notification Context: sample scenario

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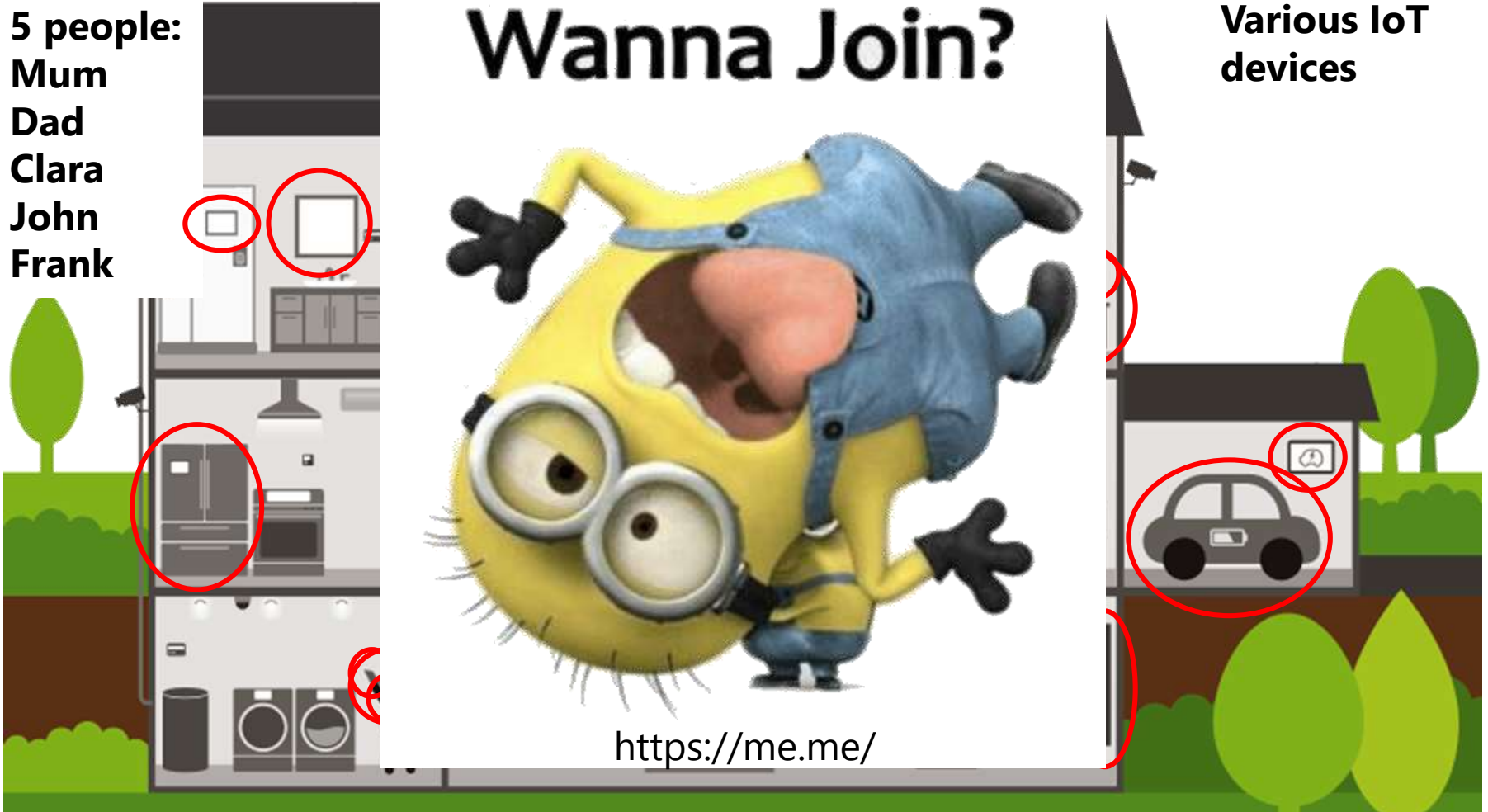
Going Crazy...

Time: 19.00

Wanna Join?

Various IoT devices

5 people:
Mum
Dad
Clara
John
Frank



Source: <https://iot.do/windstream-research-future-connected-home-community-2015-04>

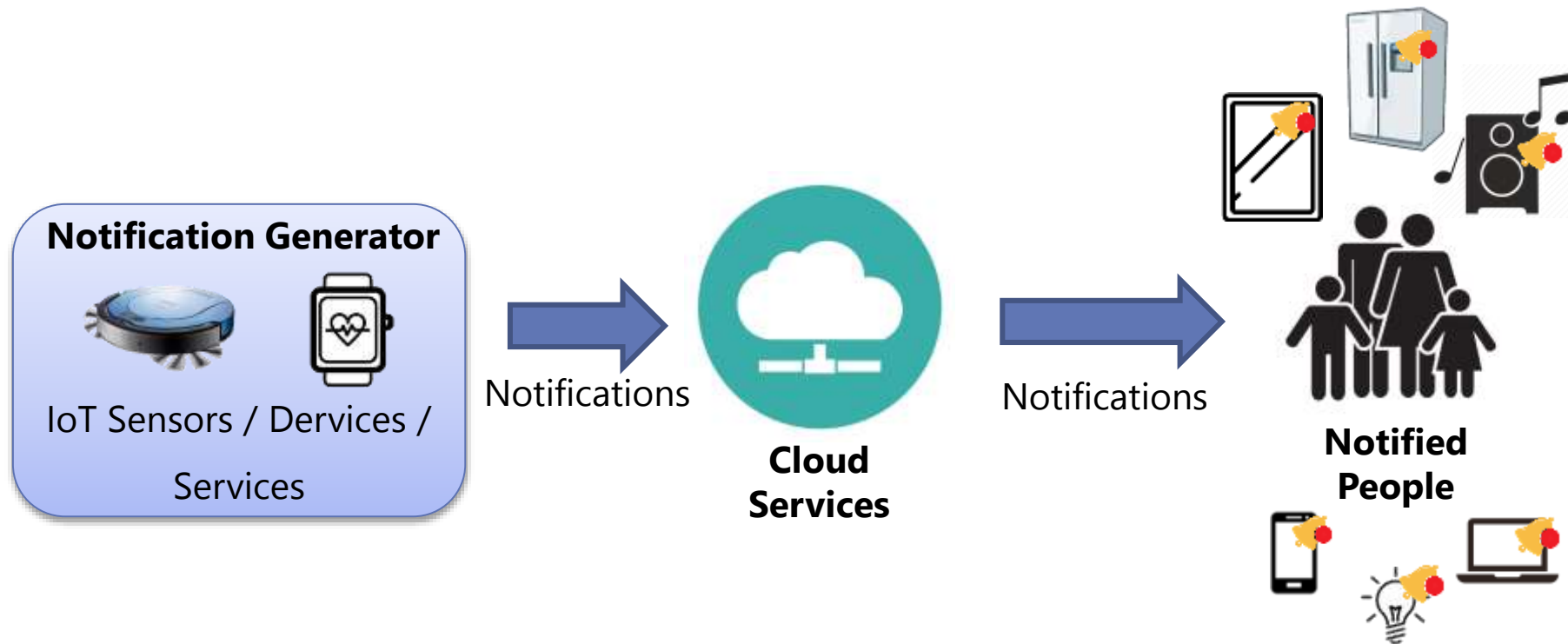
Main problem

Notifications could be disruptive:

- **Wrong moment**
- **Wrong device** on which the notification is shown
- **Wrong modality** (e.g., vibration instead of sound)
- **Wrong person(s)**
- **Repetitive** notifications
- Too many **simultaneous** notifications
- ...

Notification Context: sample scenario

Simplified version (used as a reference)

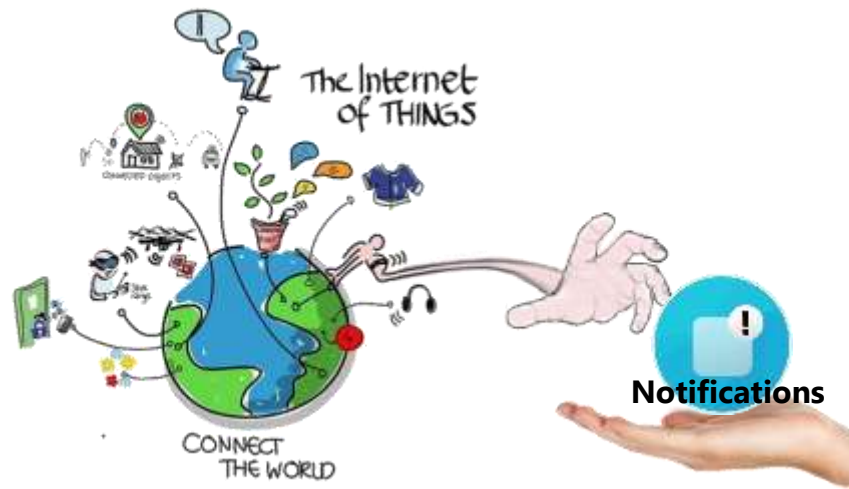


Main Research GOAL

Design and develop new **IoT architectures** to

a) **enhance** the **effect** of **IoT notifications** on **users experience**

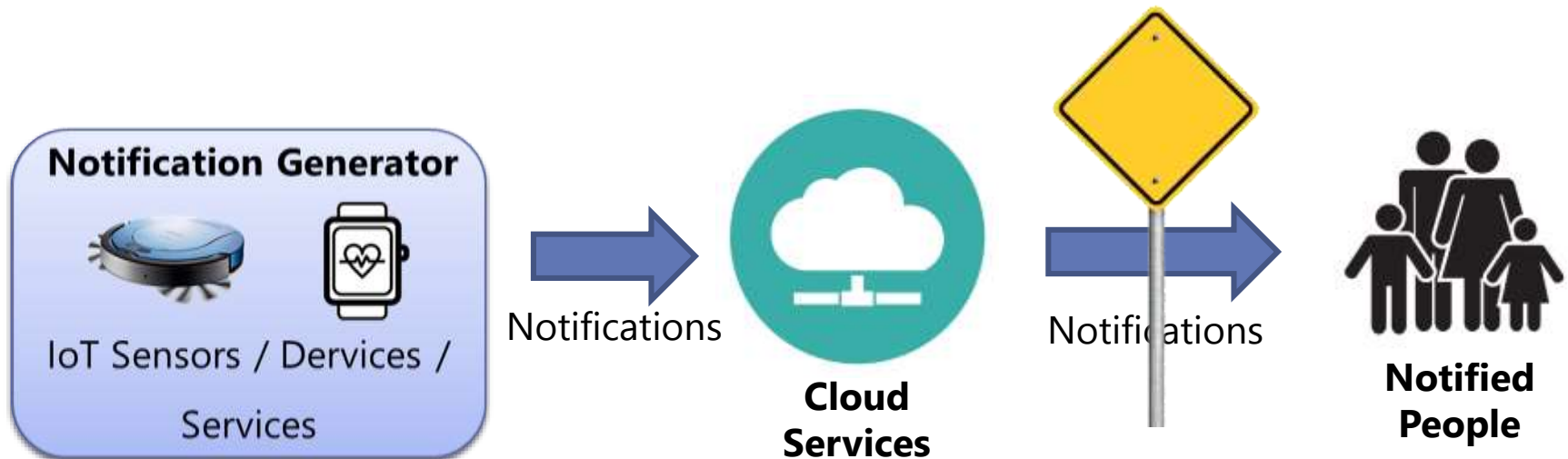
b) allow **developers** to **effectively exploit** the **notifications** improving their services, tools and applications.



Proposed solutions

Two different approaches are possible

1. At the **distribution level**: notifications are intercepted as soon as they arrive on the IoT devices and then systems decide if, when, and how to show them.

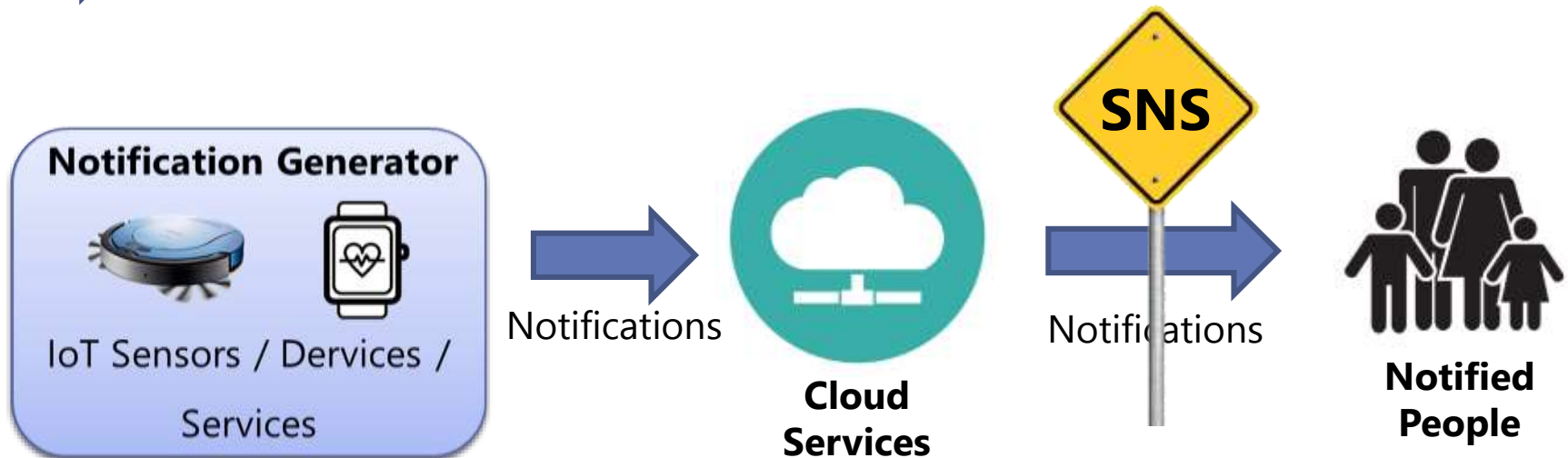


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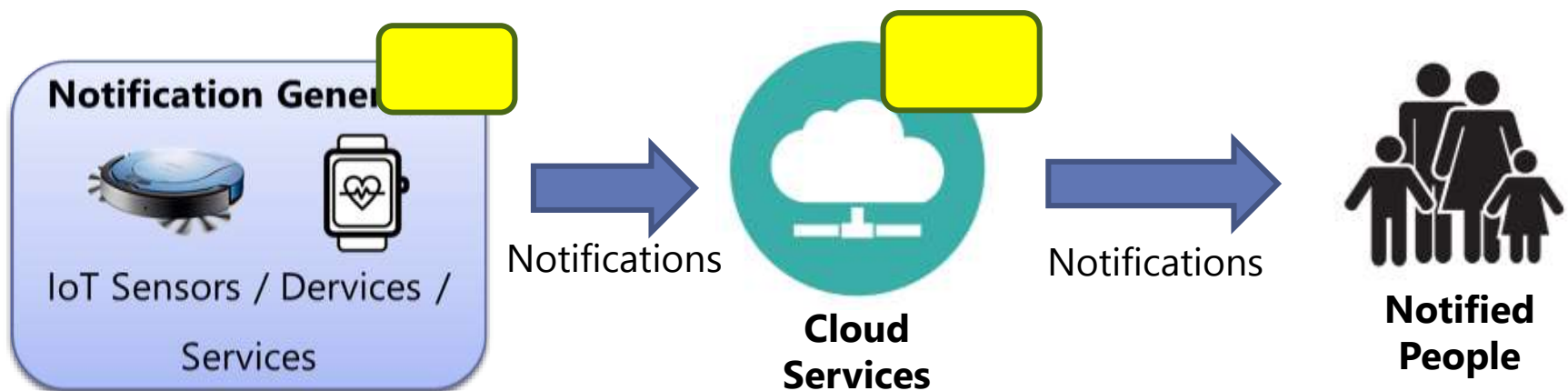
➔ Solution: **Smart Notification System (SNS)**



Proposed solutions

Two different approaches are possible

2. At the **design level**: notifications are designed with the aim of reducing user disruption.

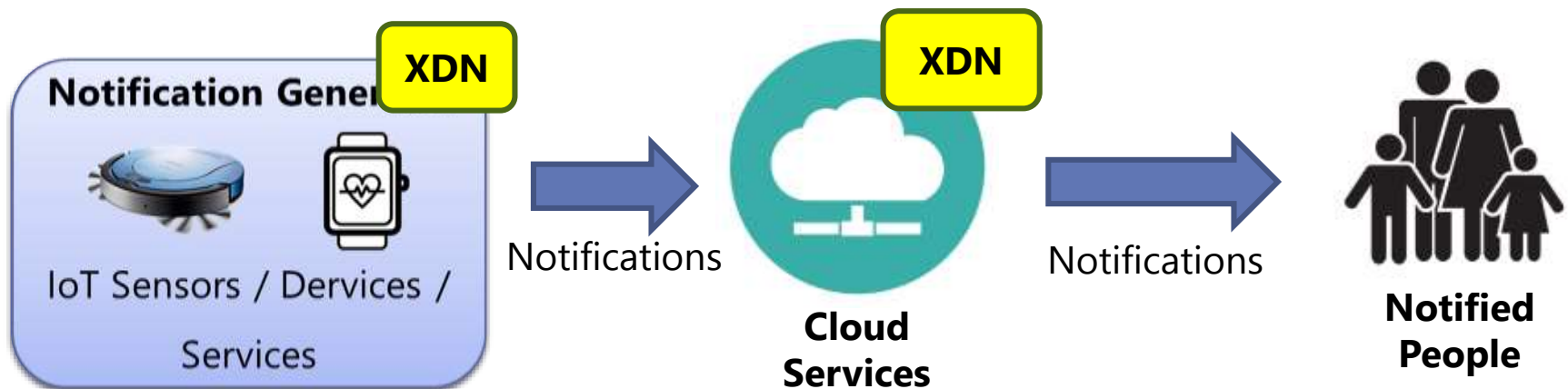


Proposed solutions

Two different approaches are possible

2. At the **design level**: notifications are designed with the aim of reducing user disruption.

➔ Solution : **XDN (Cross Device Notifications) framework**



Smart Notification System (SNS)

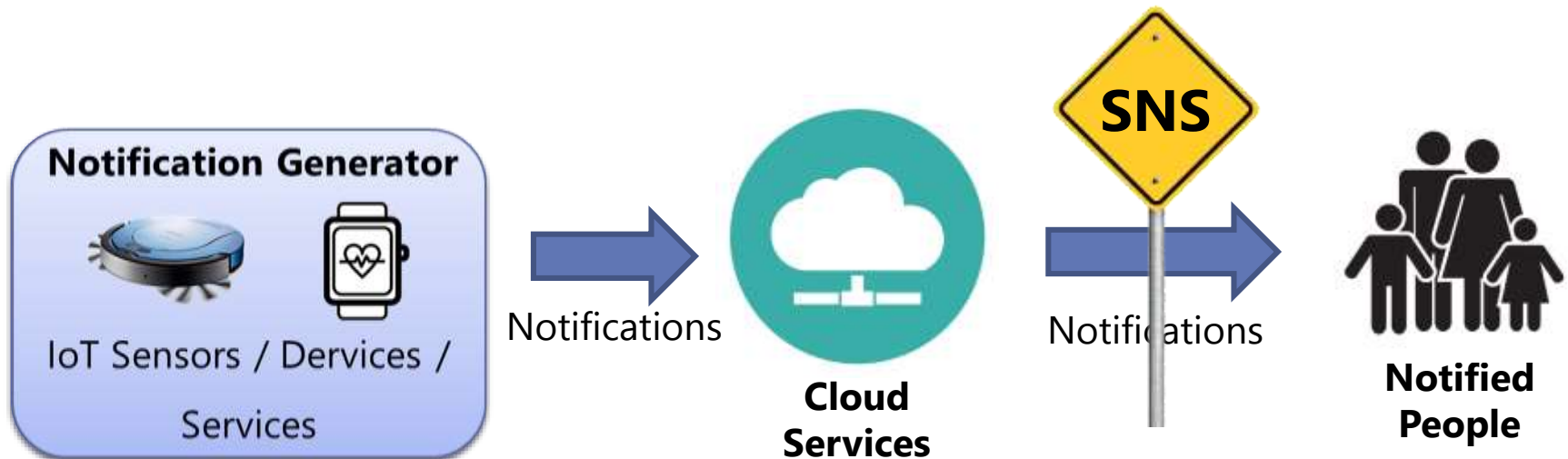
SNS

Smart Notification System (SNS): a modular architecture to deal with notifications at the distribution level.

It uses **machine learning algorithms** to manage **incoming notifications** according to **context awareness** and **users habits**.

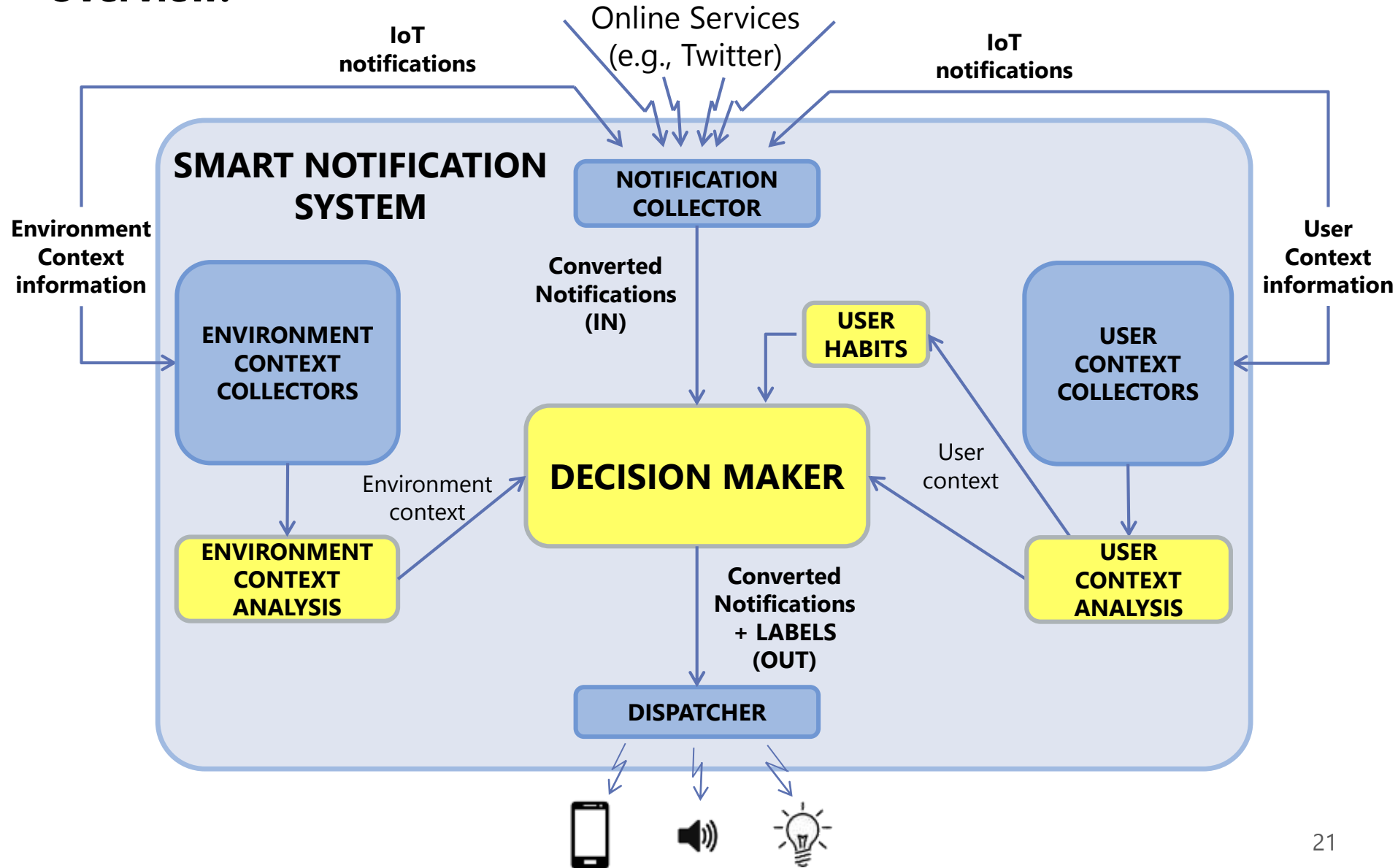
Our contributions:

1. **Architecture** design
2. **Prototypes** implementation of different architectural components

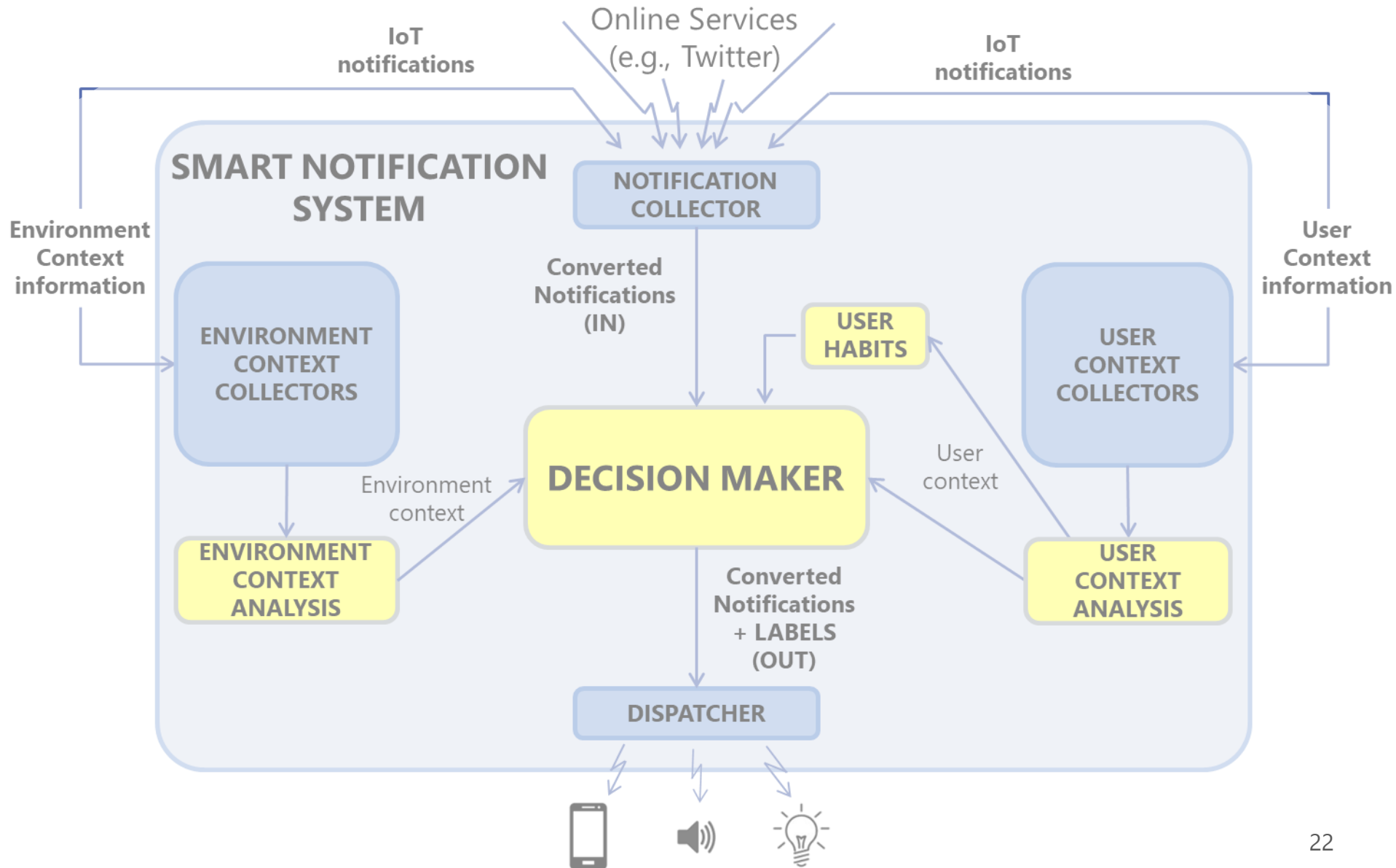


SNS: Architecture

Overview:

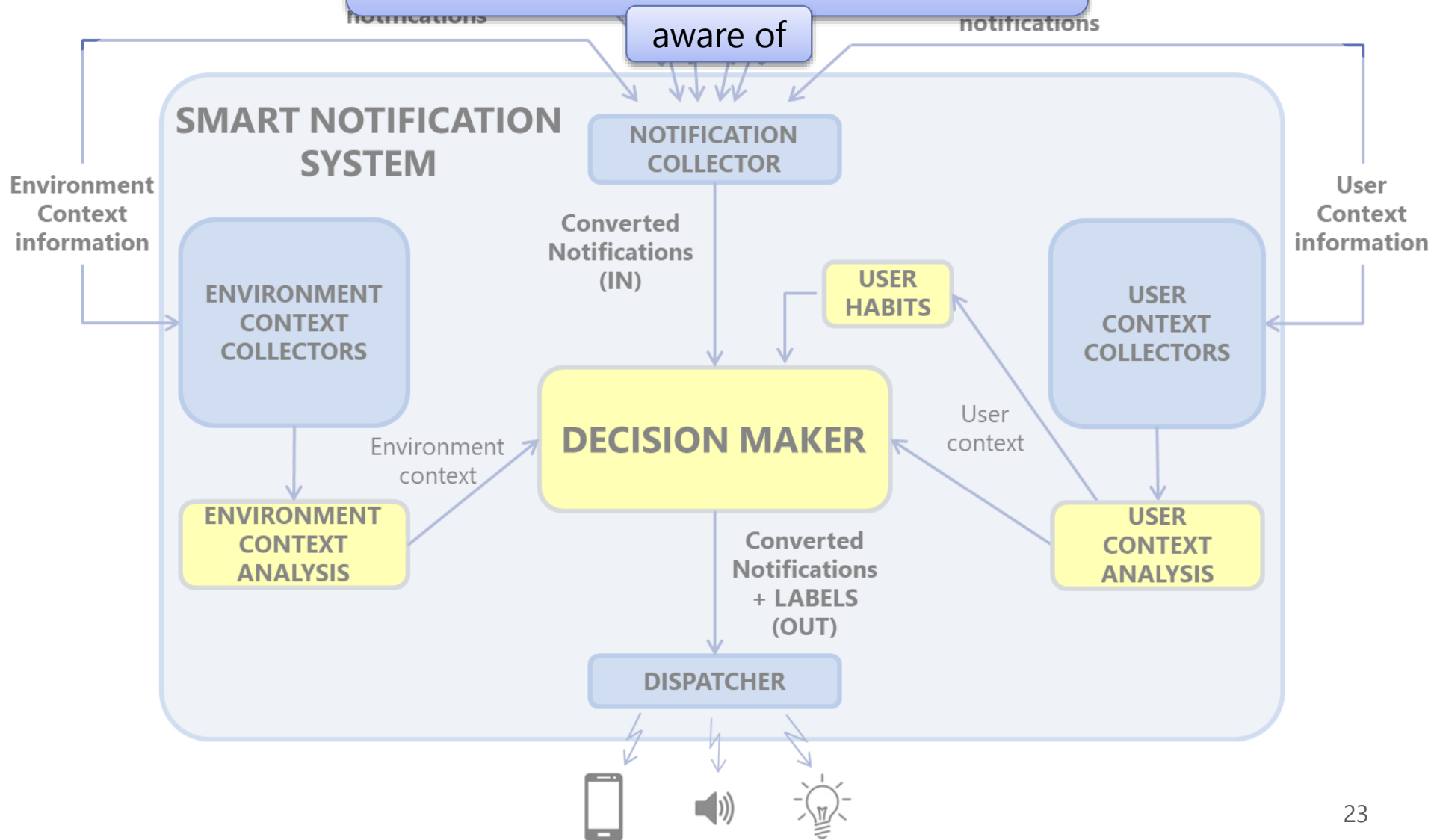


SNS: Architecture



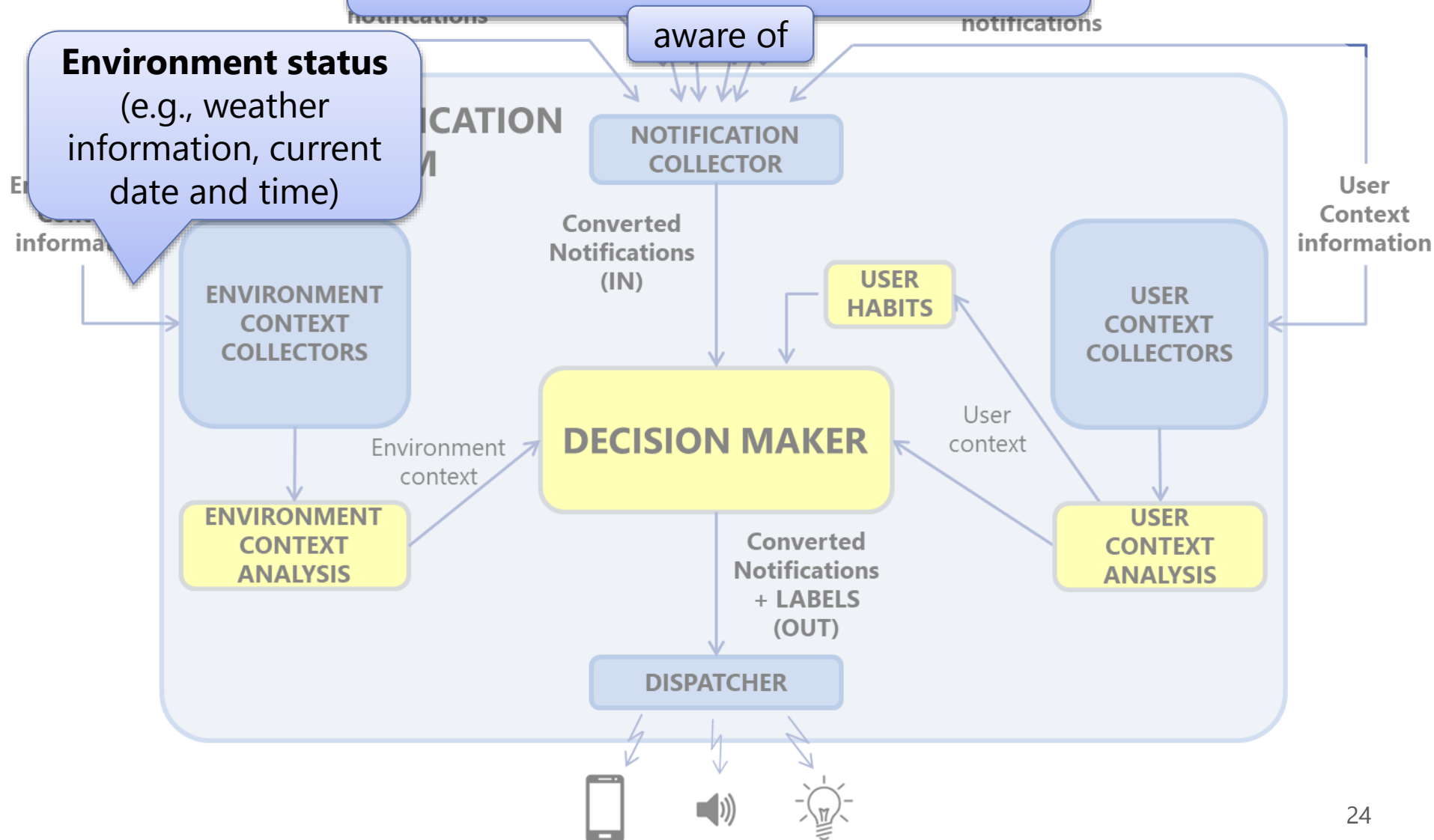
SNS: Architecture

A modular architecture



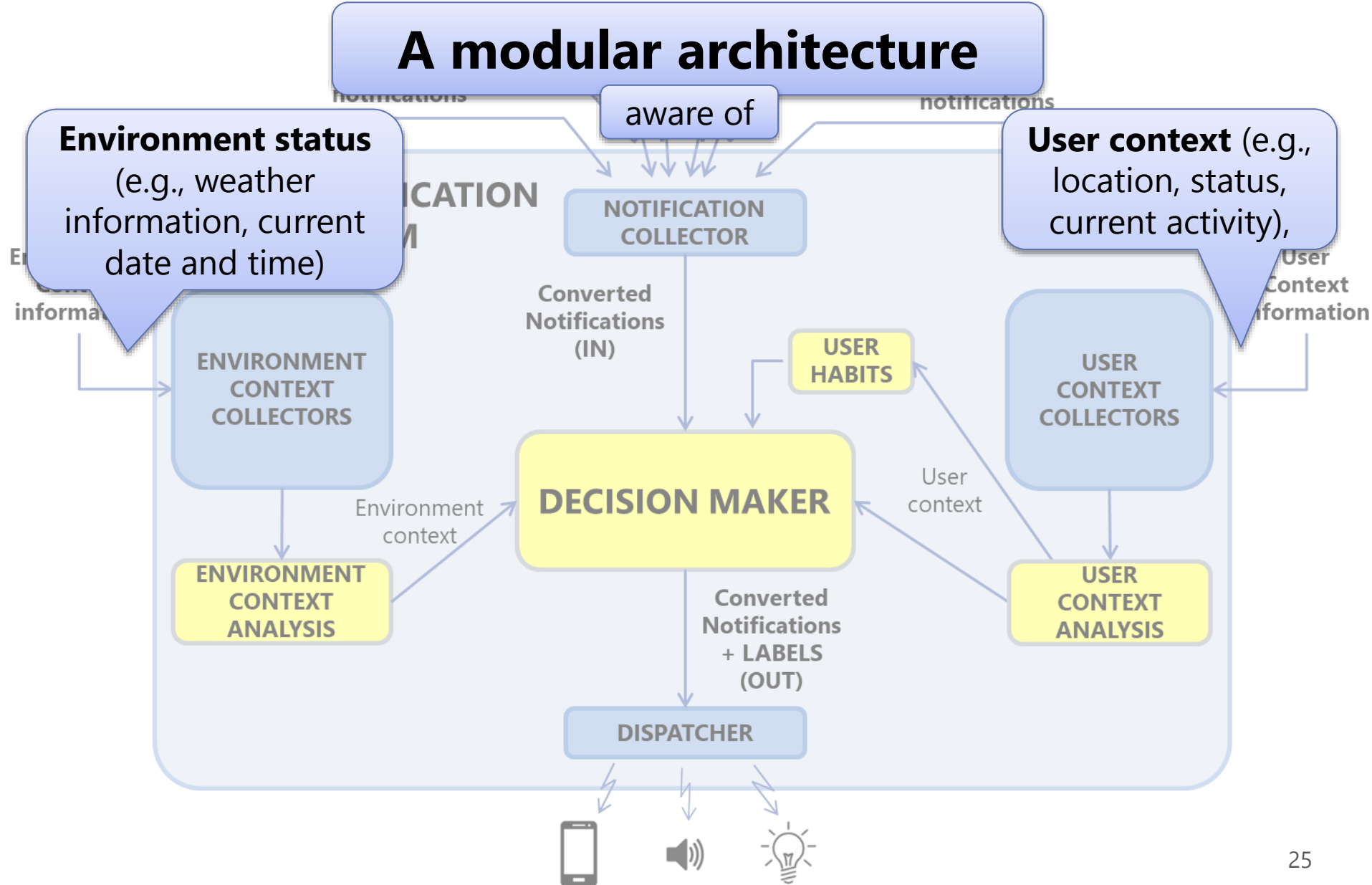
SNS: Architecture

A modular architecture



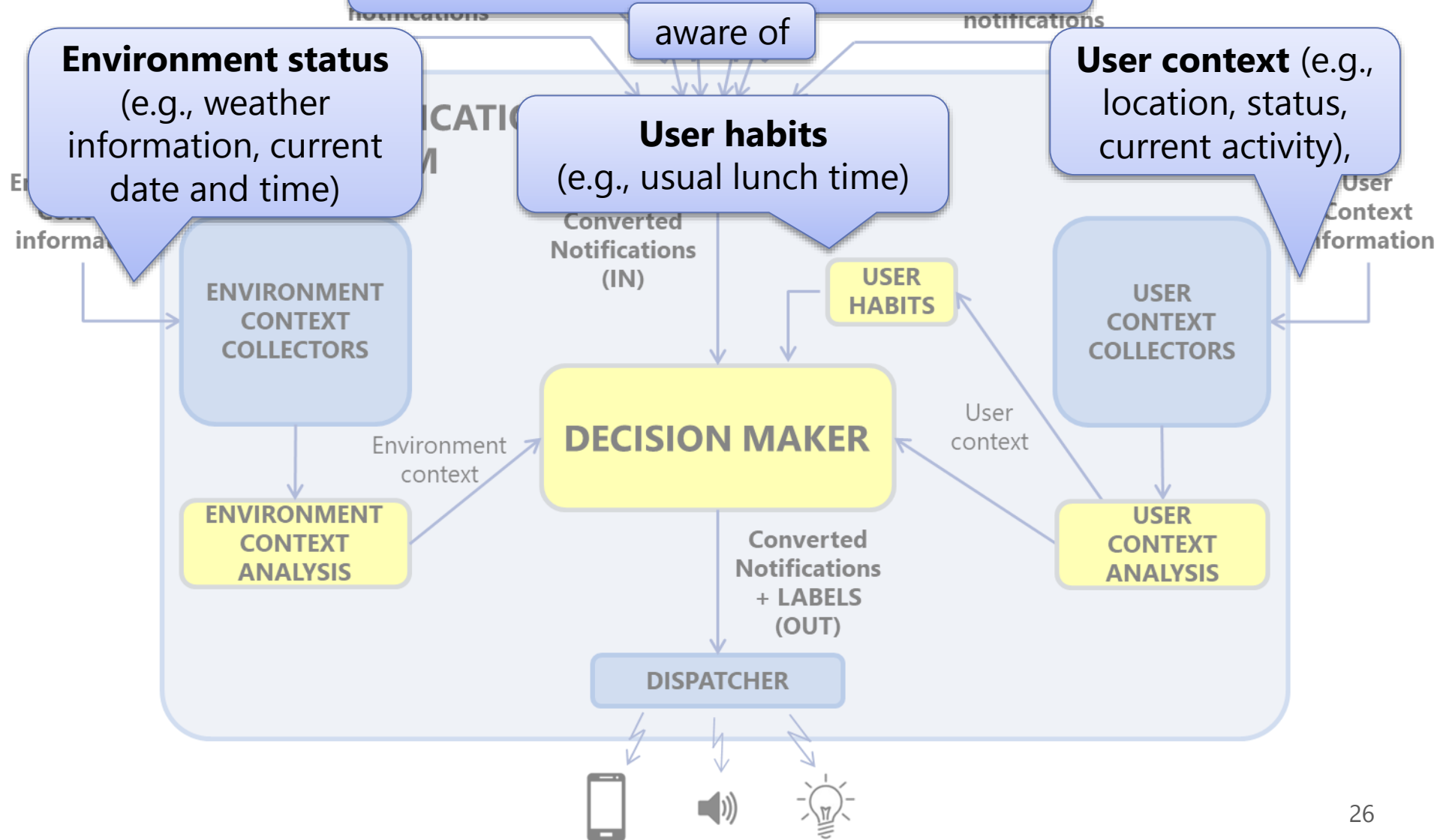
SNS: Architecture

A modular architecture



SNS: Architecture

A modular architecture



SNS: Architecture

A modular architecture

Environment status
(e.g., weather information, current date and time)

User context (e.g., location, status, current activity),

User habits
(e.g., usual lunch time)

Decision maker: makes decisions on **who** should receive the notification, **best moment, best devices** and **best modalities (including actuation)** to present notifications.

ENVIRONMENT
CONTEXT
COLLECTORS

USER
CONTEXT
COLLECTORS

DECISION MAKER

USER
HABITS

USER
CONTEXT
ANALYSIS

DISPATCHER

aware of

Converted
Notifications
(IN)

Converted
Notifications
+ LABELS
(OUT)

User
context

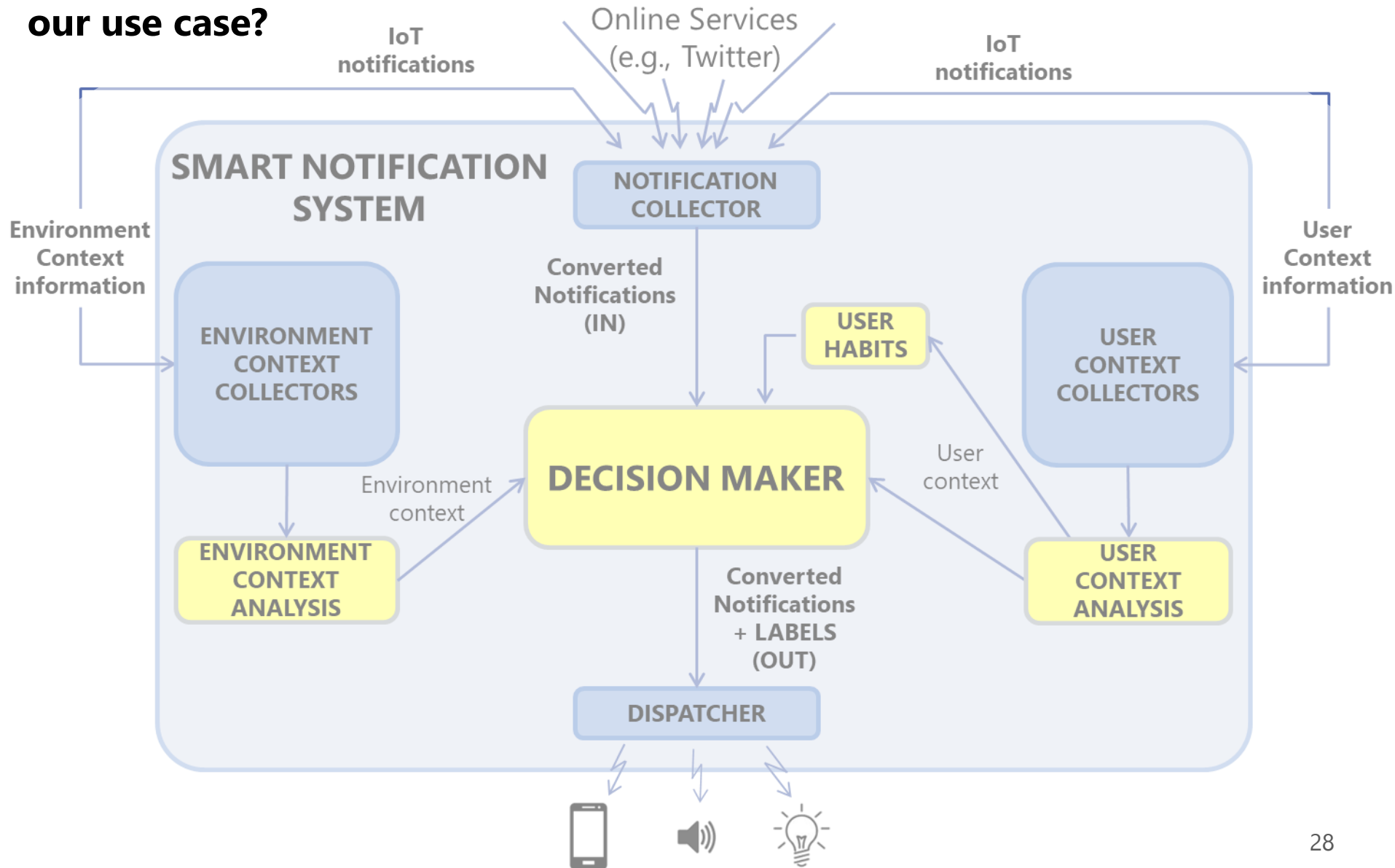
Environment
context

User
Context
Information



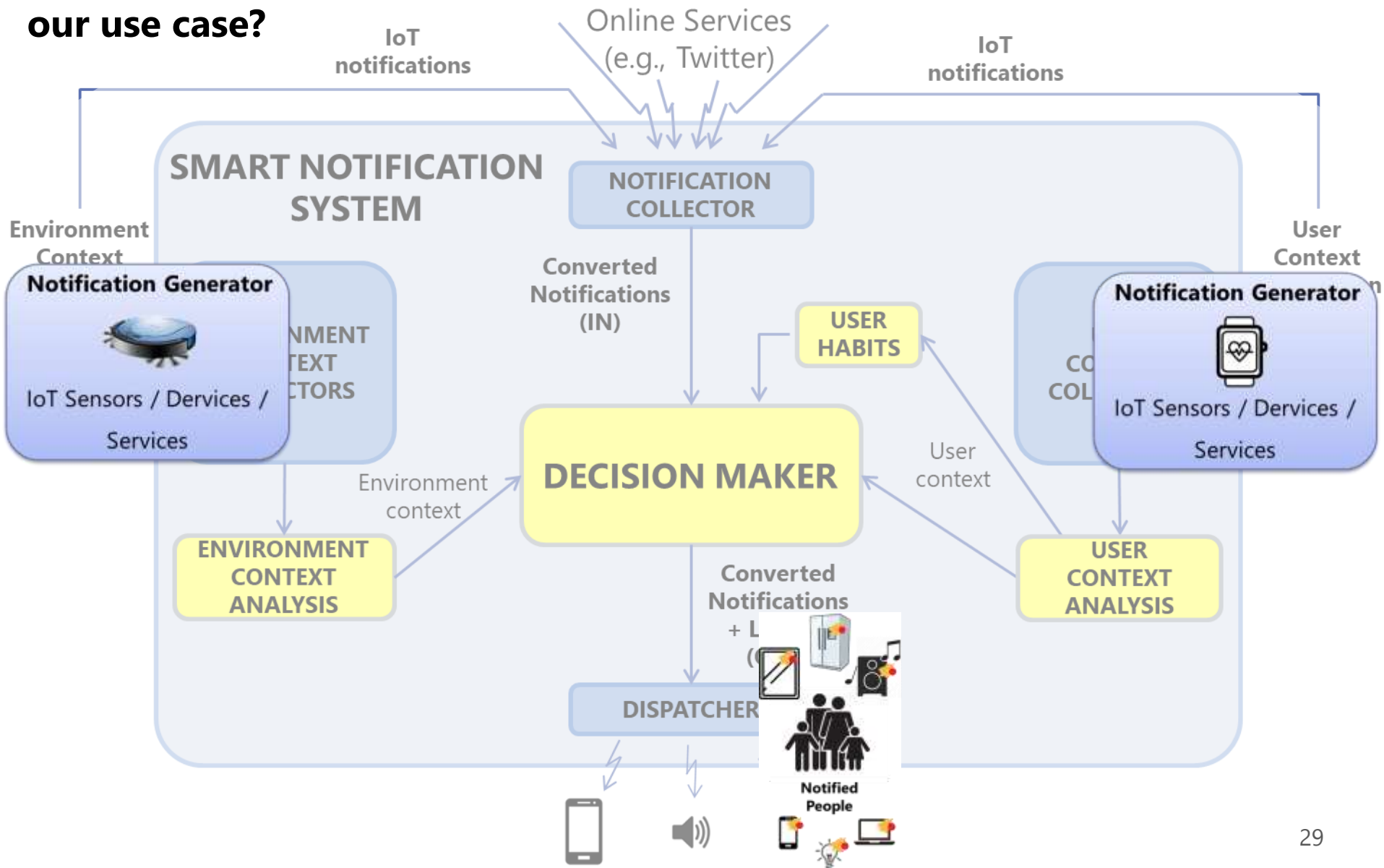
SNS: Architecture

How can we map our use case?



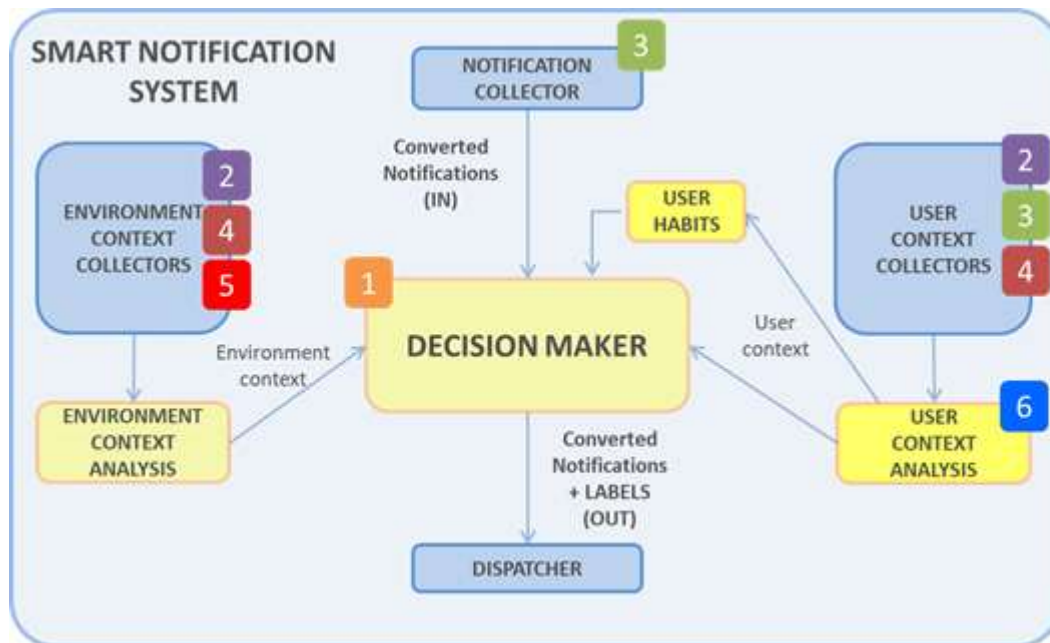
SNS: Architecture

How can we map our use case?



SNS: Prototypes

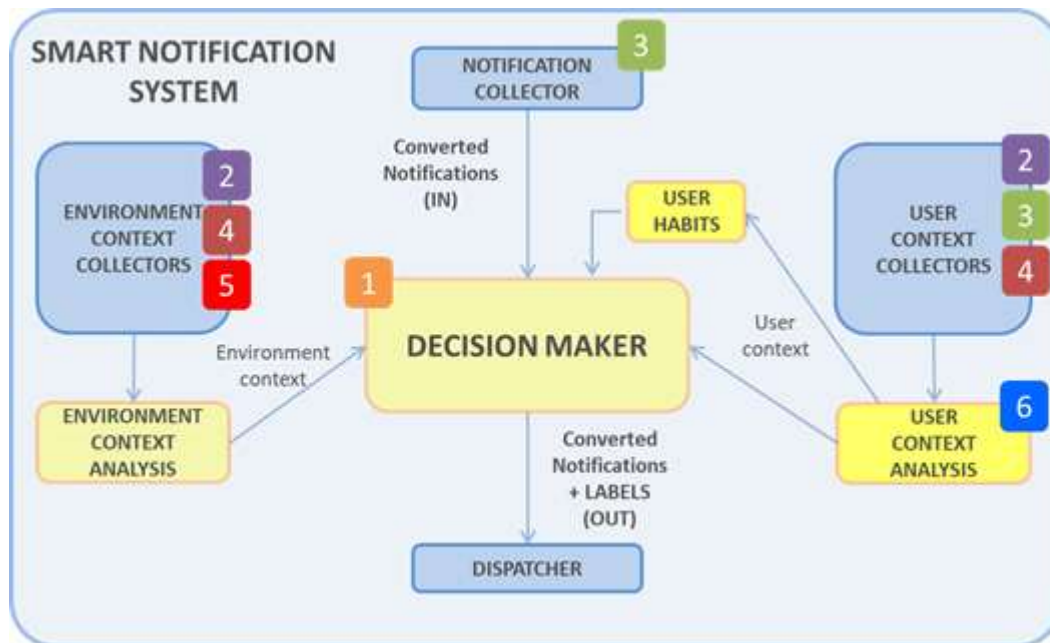
1. The **Decision Maker** contribution:
 - a) Decision maker prototype
2. The **Collectors** group of contributions:
 - a) IoT Collector server
 - b) Mobile Collector
 - c) SmartHome Collector
 - d) SmartCity Collector
3. The **Context Analysis** group of contributions:
 - a) Location Estimator



- 1 Decision maker prototype
- 2 IoT Collector server
- 3 Mobile Collector
- 4 SmartHome Collector
- 5 SmartCity Collector
- 6 Location Estimator

SNS: Prototypes

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SNS: 1. Decision Maker Prototype

Objective: demonstrate that Machine Learning algorithms **can be adopted** to the IoT notifications domain

Contribution: Preliminary version of the **Decision maker module**

Context Information to be used by the ML algorithm:

User context	Current activity
	Current location
Environment context	Current timestamp
	Owner
Available IoT devices information	Current status (e.g., on, off, standby)

Table 2.7 Simplified version of Context information: for the prototype

Notification information to be used by the ML algorithm:

Information about incoming notification	Sender
	Receiver
	Type of notification
	Timestamp of receipt
Assigned labels to outgoing notifications	Target devices

Table 2.8 Simplified version of Notifications information: for the prototype

SNS: 1. Decision Maker Prototype

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User context	Current activity
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Available IoT devices information	Current status (e.g., on, off, standby)

Table 2.7 Simplified version of Context information: for the prototype

Used dataset

 **Reality Mining Dataset**

 **Synthetic information**

Notification information to be used by the ML algorithm:

Information about incoming notification	Sender
	Receiver
	Type of notification
	Timestamp of receipt
Assigned labels to outgoing notifications	Target devices

Table 2.8 Simplified version of Notifications information: for the prototype

SNS: 1. Decision Maker Prototype

Objective: demonstrate that Machine Learning algorithms **can be adopted** to the IoT notifications domain

Contribution: Preliminary version of the **Decision maker module**

Tests:

- **3** different **machine learning algorithms** adopted over an existing dataset (MIT): **Support Vector Machine, Gaussian Naïve Bayes** and **Decision Trees**.

ML Algorithm	Percentage of correct predictions with unrelated data			Percentage of correct predictions with related data		
	Accuracy %	Precision %	Recall %	Accuracy %	Precision %	Recall %
Support Vector Machine	81.60	99.89	82.40	96.10	84.32	96.90
Gaussian Naive Bayes	51.30	99.80	51.30	83.40	95.25	83.40
Decision Trees	99.90	97.06	99.90	93.90	92.76	93.90

Table 2.10 Percentage of correct predictions obtained with used algorithms

Used dataset



+ Synthetic information

Used tools



SNS: 1. Decision Maker Prototype

Objective: demonstrate that Machine Learning algorithms **can be adopted** to the IoT notification system

Contribution:

Tests:

- 3 different datasets (MIT)

Main outcome

- The **three algorithms behave as expected:**
 - **DT works better** than the others due to the **programmatic** approach used to **generate synthetic information**
 - **Almost all the algorithms** obtain an **high** level of **accuracy, precision and recall**
- ML is **promising technique** to enhance the effect of IoT notifications on users experience

set
trees.

dataset
Reality Mining Dataset
synthetic information

tools



Decision
Trees

99.90

97.06

99.90

93.90

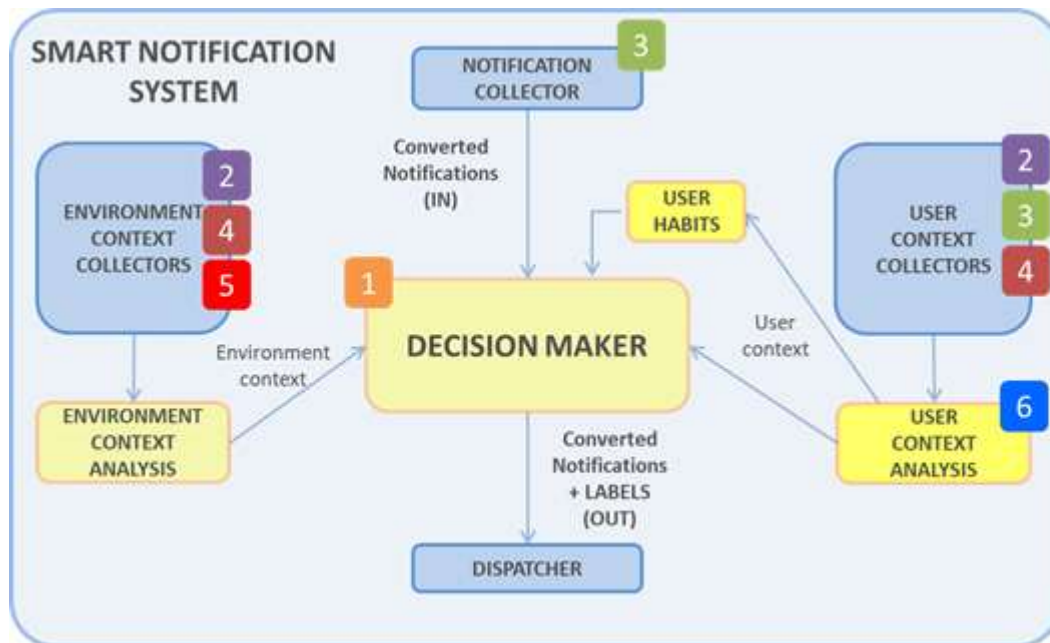
92.76

93.90

Table 2.10 Percentage of correct predictions obtained with used algorithms

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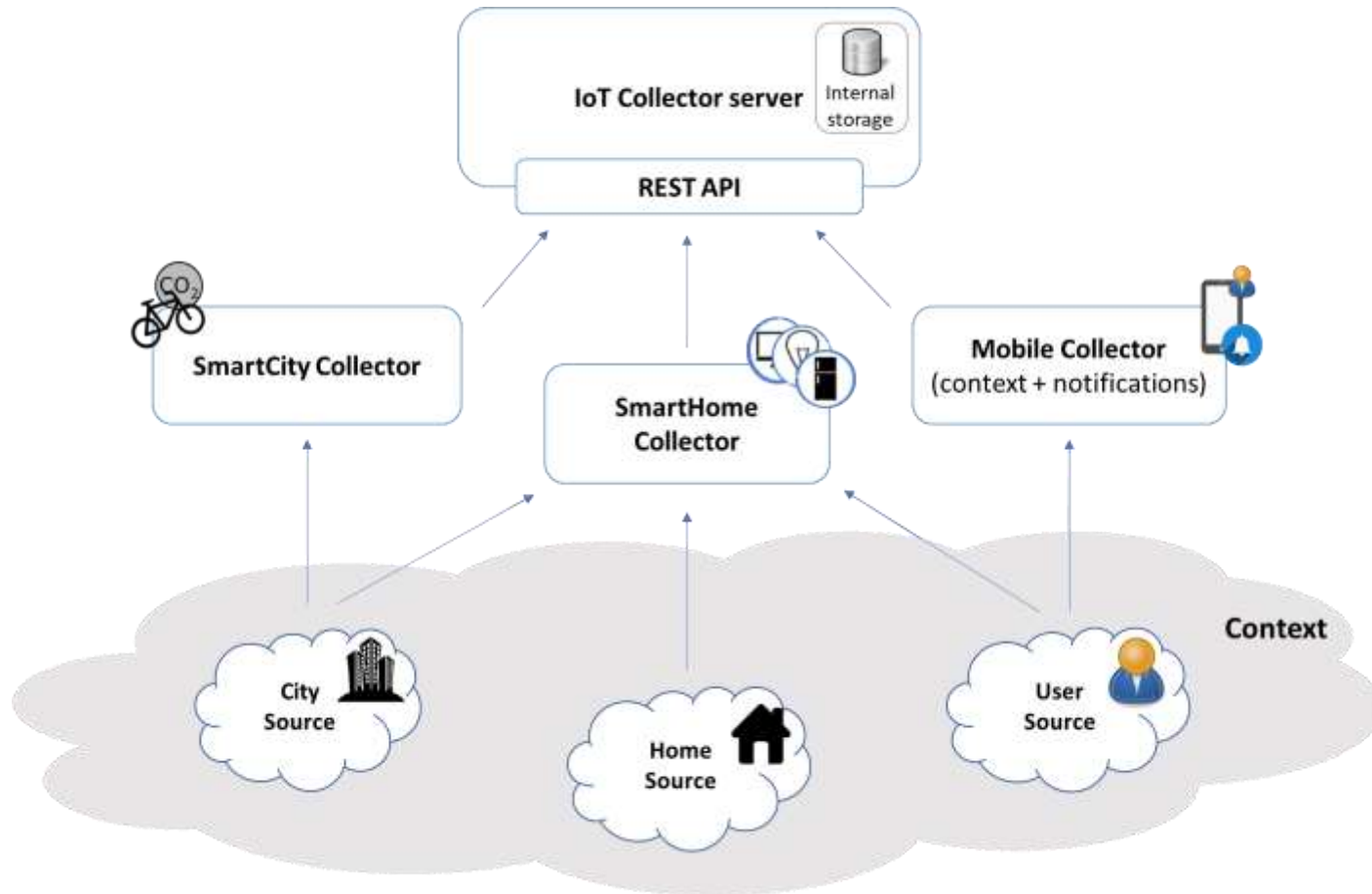


- 1 Decision maker prototype
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- 6 Location Estimator

SNS: 2. Collectors

Aims:

1. collect real data
2. validate the Machine Learning approach used in the Decision Maker Prototype

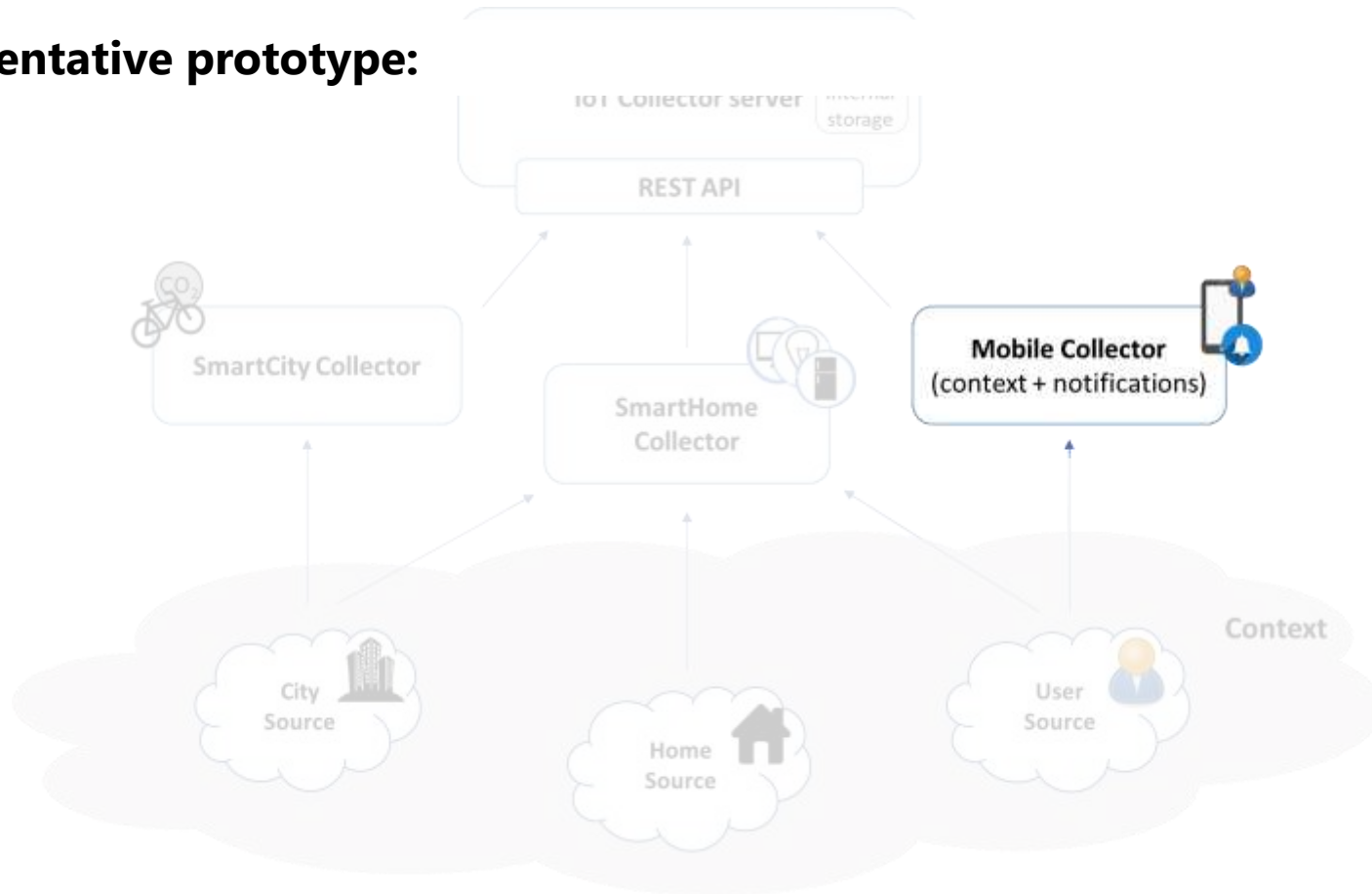


SNS: 2.b Mobile Collector

Aims:

1. collect real data
2. validate the Machine Learning approach used in the Decision Maker Prototype

Representative prototype:



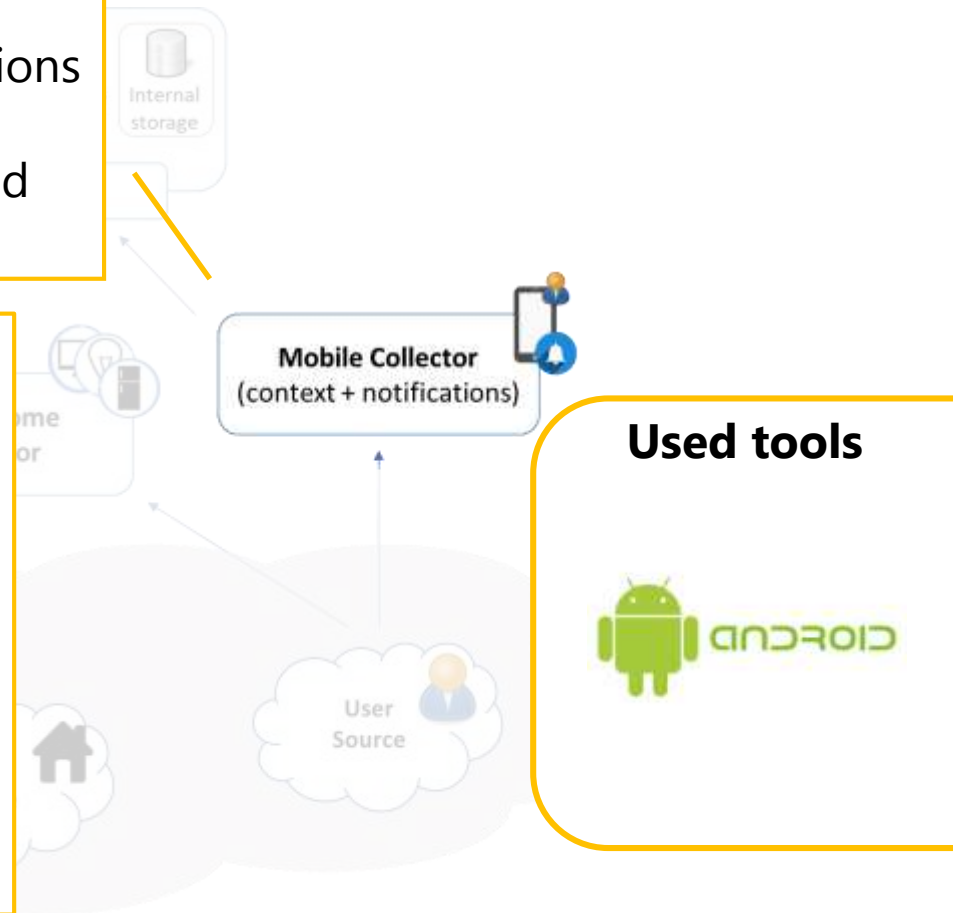
SNS: 2.b Mobile Collector

Objective 1: collect real user data

- Collect user context information (e.g., location and activity);
- Collect all the mobile and IoT notifications received on user smartphone;
- Collect the user reaction to the received notifications.

29 people (5 females and 24 males) used the app for **78 days**

- users receives an average of 247 notifications a day
- users are almost always in the same 3 or 4 places
- users receive most of the notifications from non-important contacts than from important ones.



SNS: 2.b Mobile Collector

Objective 2: validate the Machine Learning approach used in the Decision Maker Prototype

Input features:

- Notification type (mobile, IoT)
- Generating service (e.g., Telegram)
- Ringtone mode
- Notification sender
- Sender-Receiver FAMILY relationship
- Sender-Receiver FRIEND relationship
- Sender-Receiver WORK relationship
- Date and time of receipt (day of week, day of month, month, time)
- User location (Lon/Lat)
- Activity (IN_VEHICLE, ON_BICYCLE, ON_FOOT, RUNNING, STILL, TILTING, UNKNOWN, WALKING)
- Battery level
- Battery status (charging or not charging).
- Connection type (Wifi, network, NoConn)
- Wifi SSID

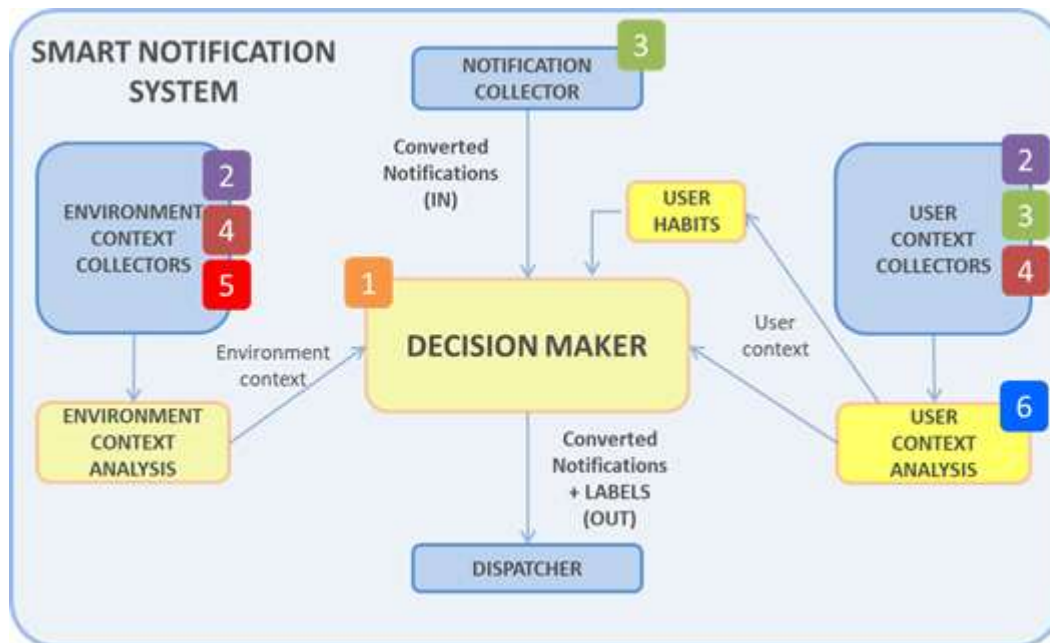
Label: annoying or appreciated notification
(14 users for 15 days)

User	% of appreciated notifications over all	Naïve Bayes			J48 (Decision Trees)		
		accuracy	precision	recall	accuracy	precision	recall
User 1	62%	75%	75%	75%	69%	69%	69%
User 2	70%	94%	94%	94%	99%	99%	99%
User 3	96%	92%	91%	92%	95%	91%	95%
User 4	63%	60%	63%	60%	68%	68%	68%
User 5	81%	92%	94%	92%	92%	91%	92%
User 6	65%	58%	56%	58%	58%	53%	58%
User 7	54%	78%	78%	78%	83%	83%	83%
User 8	72%	72%	78%	72%	70%	65%	70%
User 9	57%	67%	68%	67%	60%	59%	60%
User 10	77%	72%	75%	72%	70%	64%	70%
User 11	66%	70%	70%	70%	74%	81%	74%
User 12	61%	82%	82%	82%	91%	92%	91%
User 13	53%	78%	82%	78%	88%	88%	88%
User 14	87%	74%	76%	74%	89%	80%	89%
Mean		76%	77%	76%	79%	77%	79%
Standard deviation		11%	11%	11%	13%	13%	13%

Table 2.17 Preliminary results with real data

SNS: Prototypes

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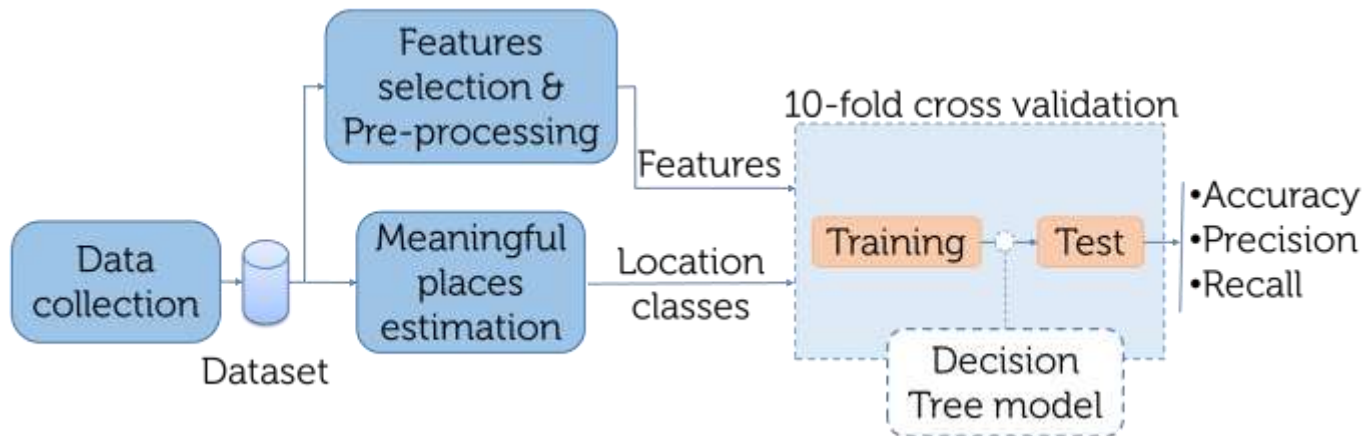
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SNS: 3. Context Analysis: location estimation

Proposal: demonstrate possibility of inferring user location without energy-hungry methods (e.g., GPS)

People usually spend 85% of their time staying in a few places.

The proposed solution uses Decision Trees as Machine Learning supervised classification algorithm to establish user presence in the **two most attended meaningful places**



Model that describes the estimation process performed for each user

SNS: 3. Context Analysis: location estimation

Proposal: demonstrate possibility of inferring user location without energy-hungry methods (e.g., GPS)

Tests:

- 10-fold cross validation through the Weka workbench
- **user presence in a meaningful place** was **estimated** every time a new notification is received.
- **Input features:** combination of Feature Classes (A-AB-ABC-ABCD-ABCDE-BC-...)

Category	Feature	Feature Class
Time information	Time	A
	Month	
	Day	
	Day of the week	
Notification information	Type	B
	Generating service	
	Sender-receiver relationship	
Device state	Battery level	C
	Charging state	D
	Ringtone mode	D
User information	Current activity	E
	Absolute location	-

TABLE 1: Collected data

Results:

- Most important features (that mainly influence decision) are related to **time**
- **“Current activity” (E)** (i.e., the only feature that consumes extra energy), is **not necessary**
- **Accuracy > 75%** in almost all tests

XDN Framework

XDN: Motivation

Main Problem: Overwhelming notifications

Second approach

- At the **design level:** notifications are designed with the aim of reducing user disruption

XDN: Motivation

Main Problem: Overwhelming notifications

Second approach

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Developers:

- **define their strategies** to let their software, then, influence users' behaviors with respect to notifications
- **exploiting** the **advantages** of the **cross-device** approach

XDN: Motivation

Main Problem: Overwhelming notifications

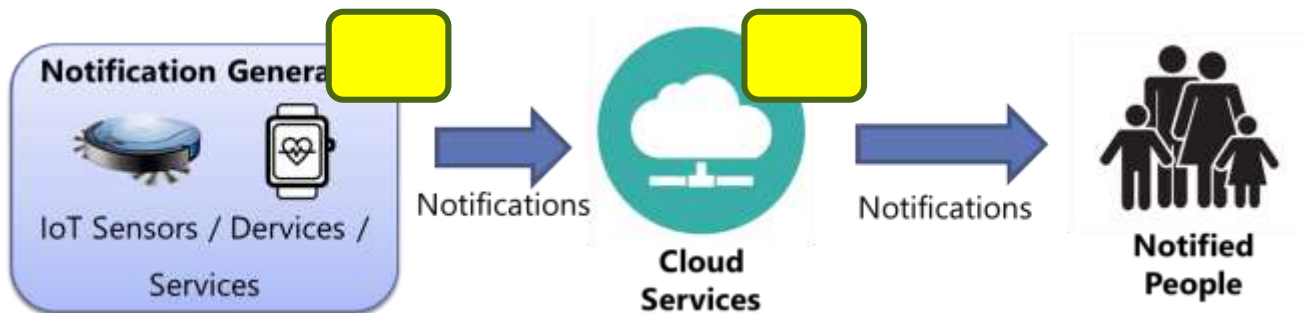
Second approach

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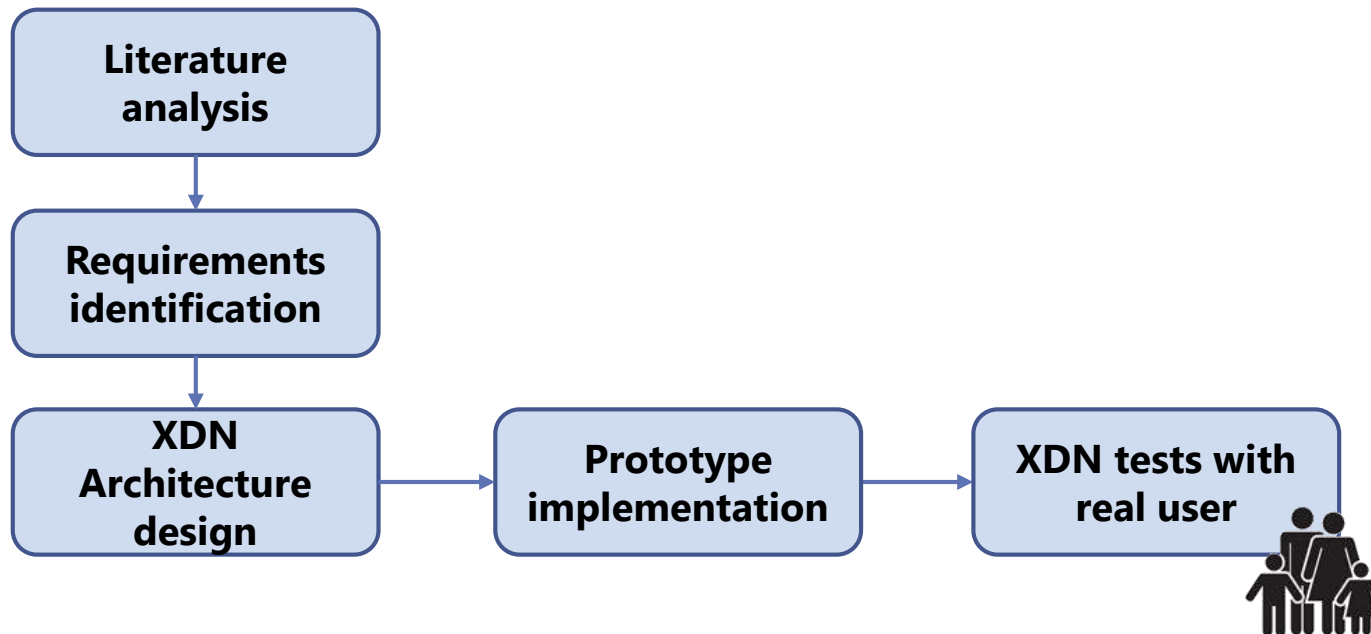


XDN: Our Proposal

Main Problem: Overwhelming notifications

Second approach

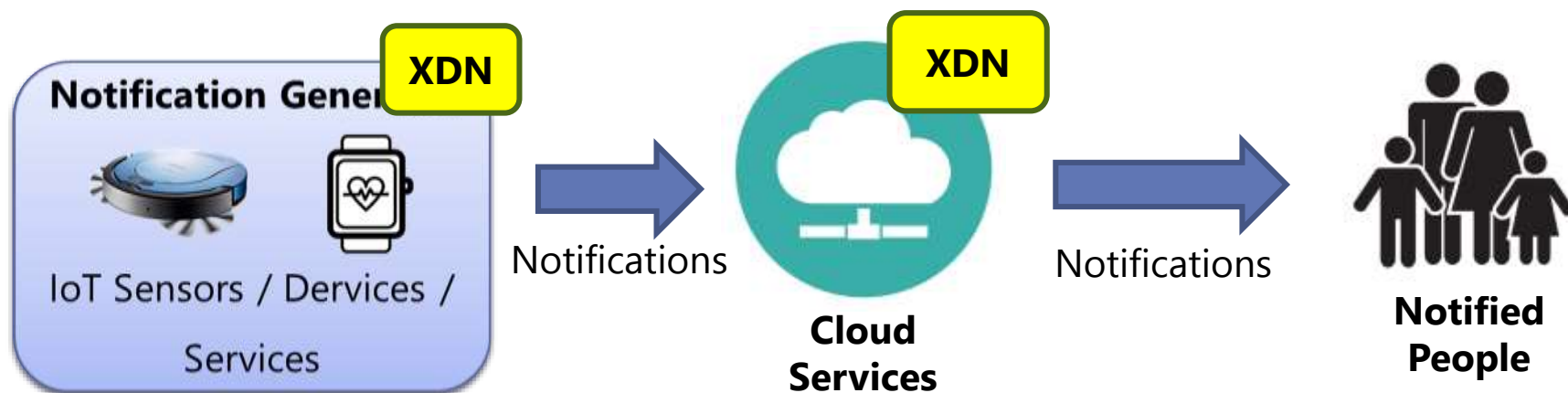
- At the **design level**: notifications are designed with the aim of reducing user disruption



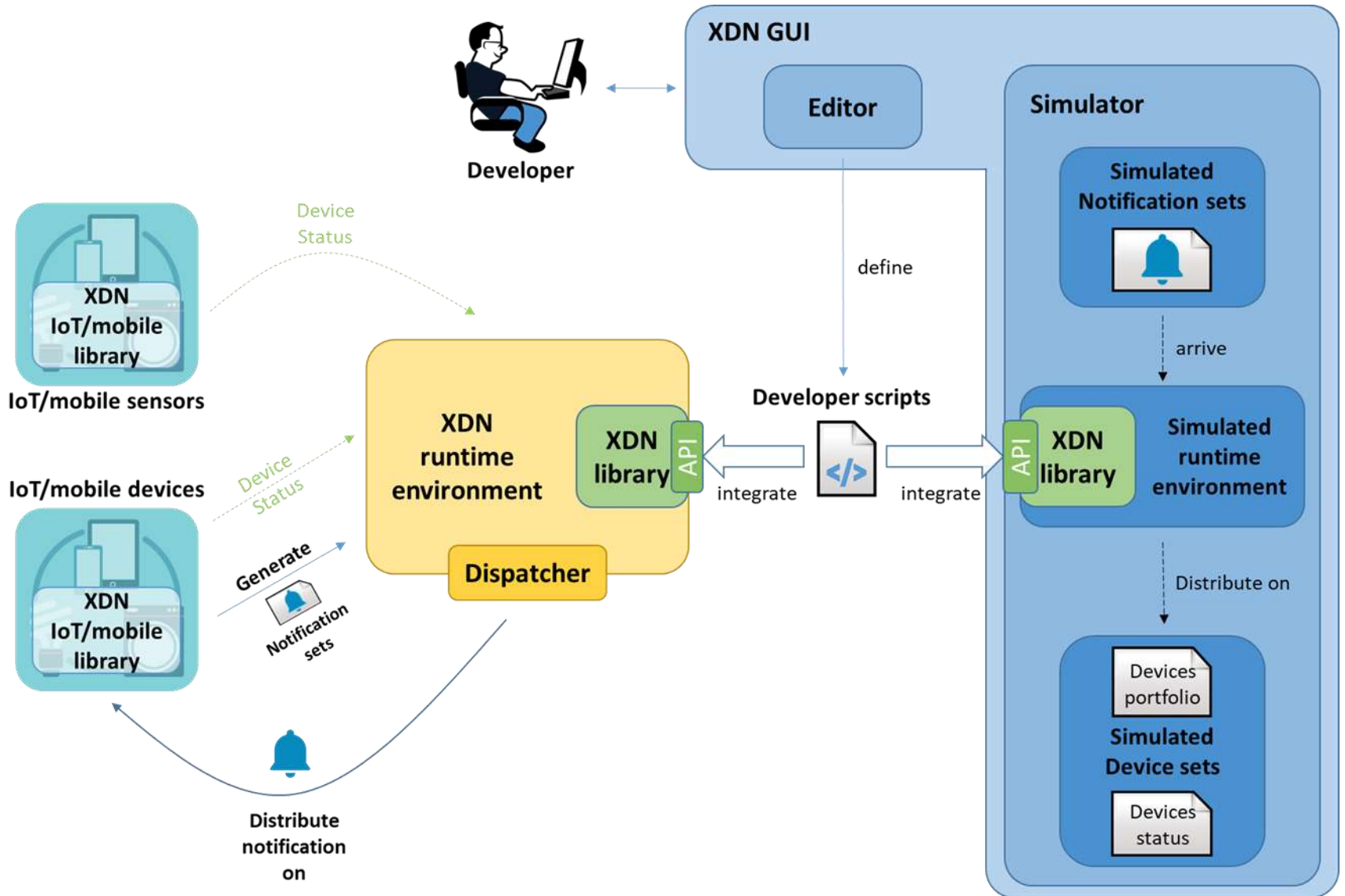
XDN: Our Proposal

XDN (Cross Device Notifications), a framework to assist **developers** in:

- a) **personalizing notifications** to differentiate important and unimportant ones
- b) **designing, implementing, and testing cross-device notifications strategies** to inform users without causing too much disruption and involving both mobile and IoT devices.



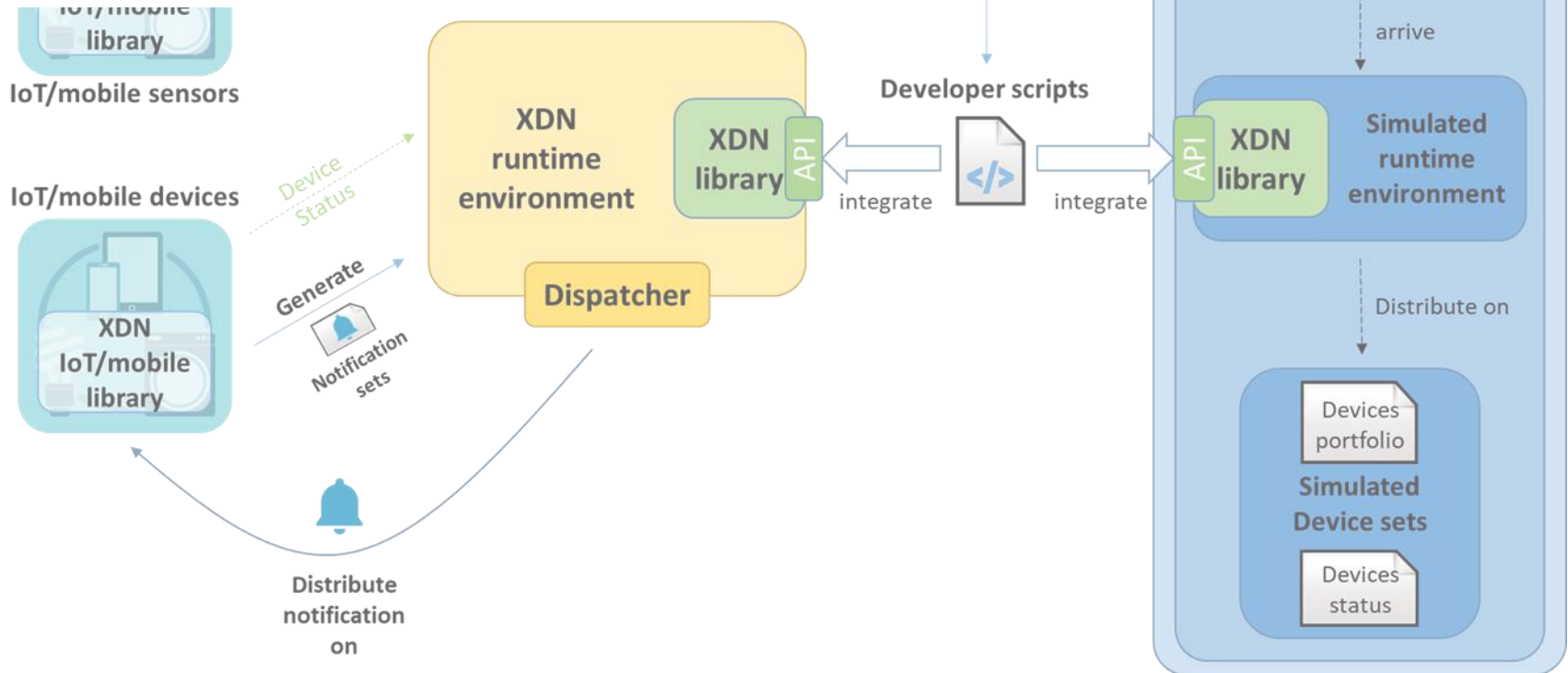
XDN: Architecture



XDN: Architecture

4 main components:

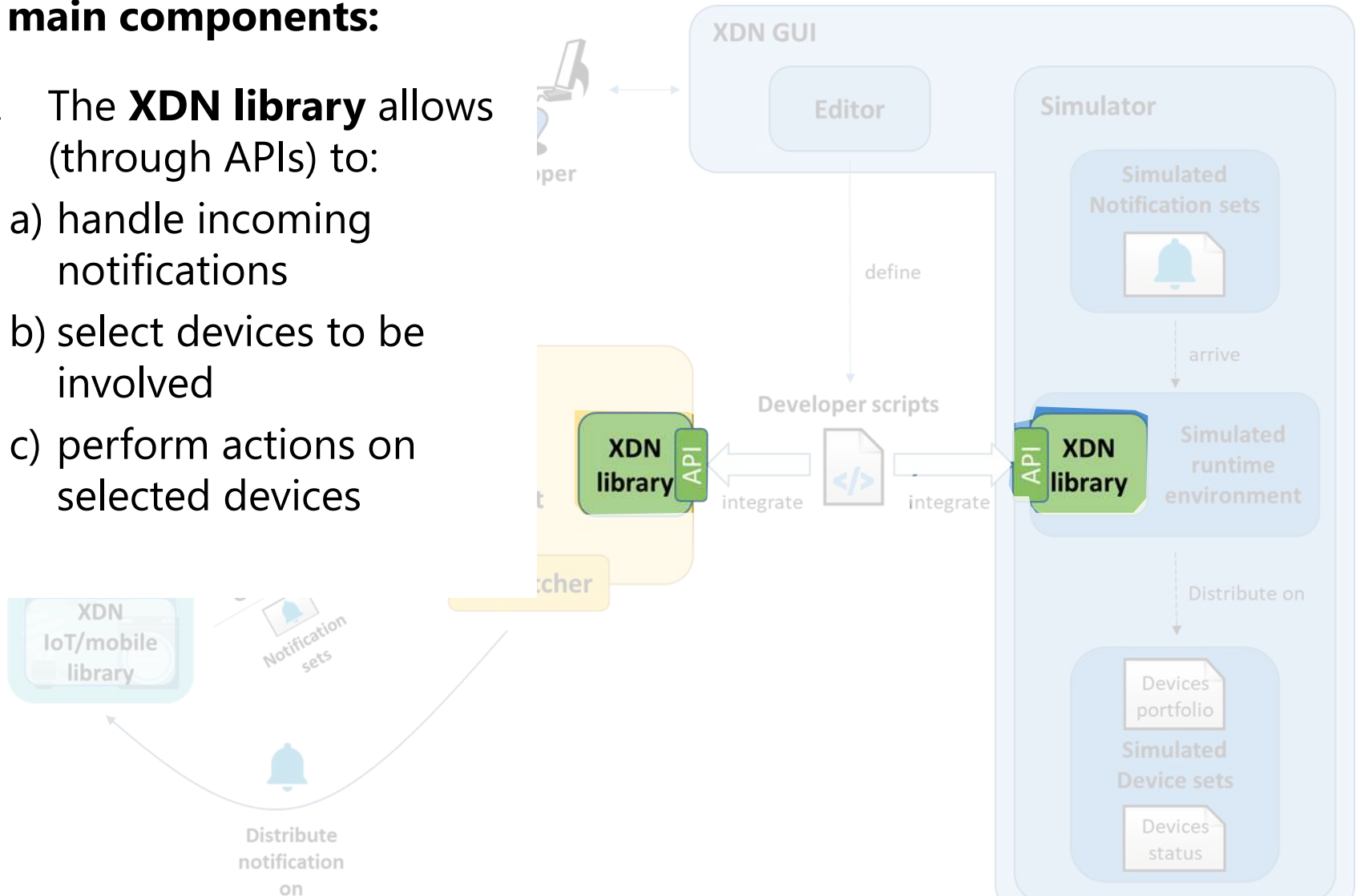
1. The **XDN library**
2. The **XDN GUI**
3. The **XDN Runtime Environment**
4. The **XDN IoT/Mobile library**



XDN: Architecture

4 main components:

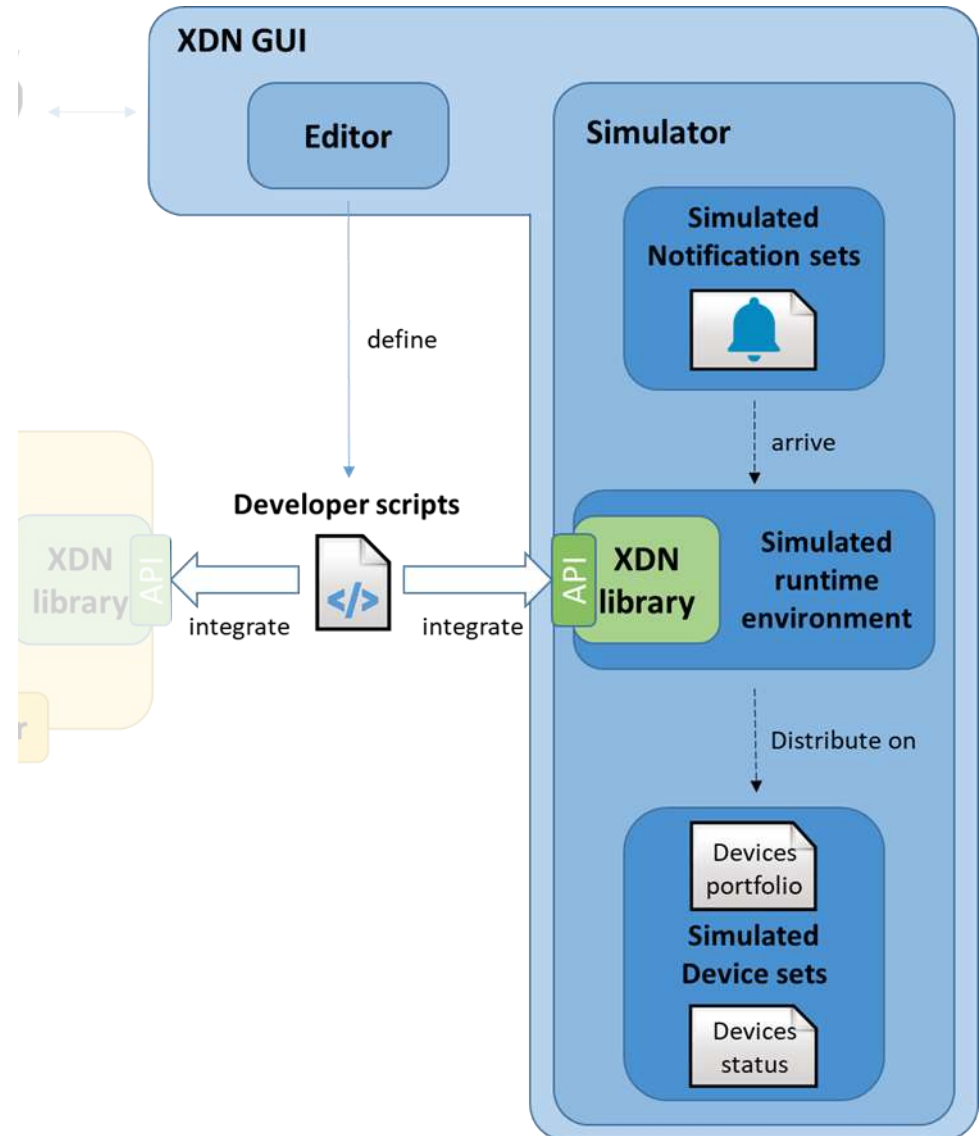
1. The **XDN library** allows (through APIs) to:
 - a) handle incoming notifications
 - b) select devices to be involved
 - c) perform actions on selected devices



XDN: Architecture

4 main components:

2. The **XDN GUI** allows developers to **explore and evaluate different design alternatives** by providing:
 - a) an **IDE** to implement and test developed notification strategies
 - b) a **simulator** to simulate the behavior of the devices

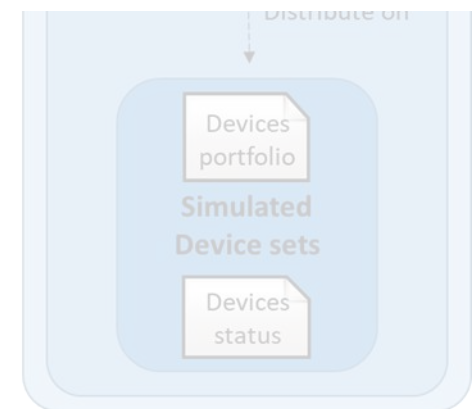
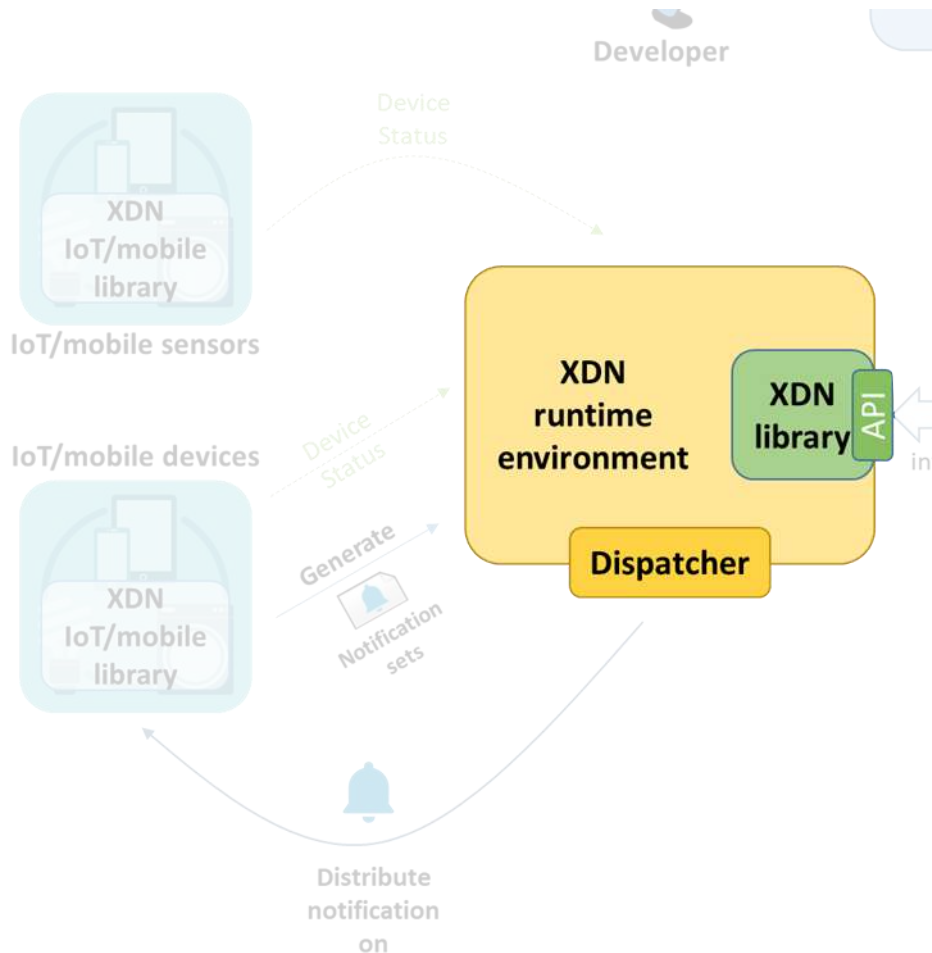


XDN: Architecture

4 main components:

3. The **XDN Runtime Environment** is run on a server to:

- accept device registration requests;
- accept update requests
- accept new notifications
- customize and dispatch the notifications



XDN: Architecture

4 main components:

4. The **XDN IoT/Mobile library** to be integrated in the IoT/mobile applications to:



IoT/mobile sensors

- generate notifications compatible with XDN
- send the generated notifications to the XDN runtime environment;
- receive commands from the XDN runtime environment (in JSON)
- execute the received commands.

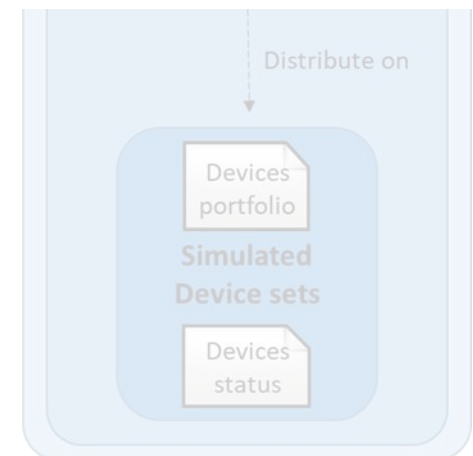
IoT/mobile devices



Generate
Notification sets

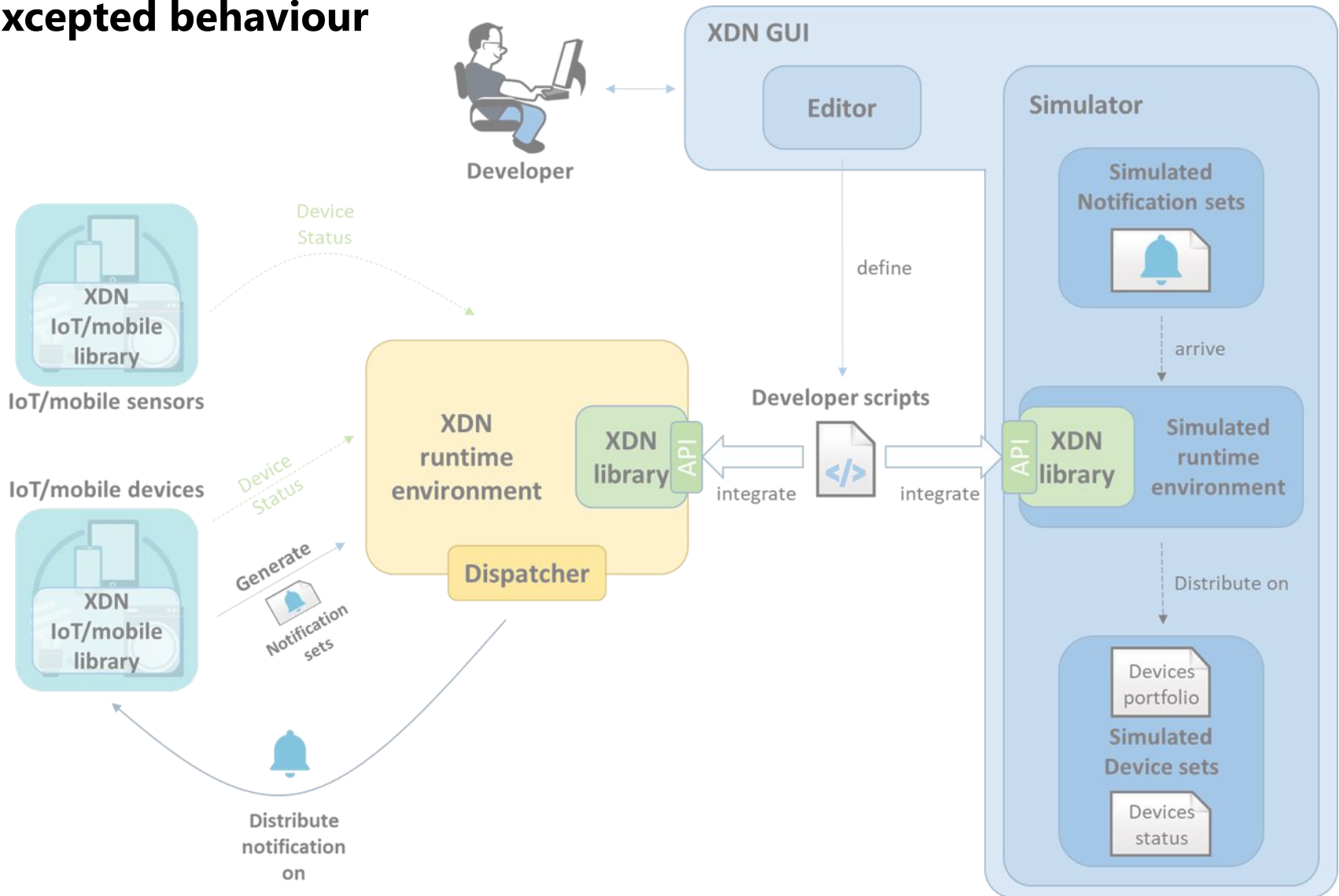
Dispatcher

Distribute
notification
on



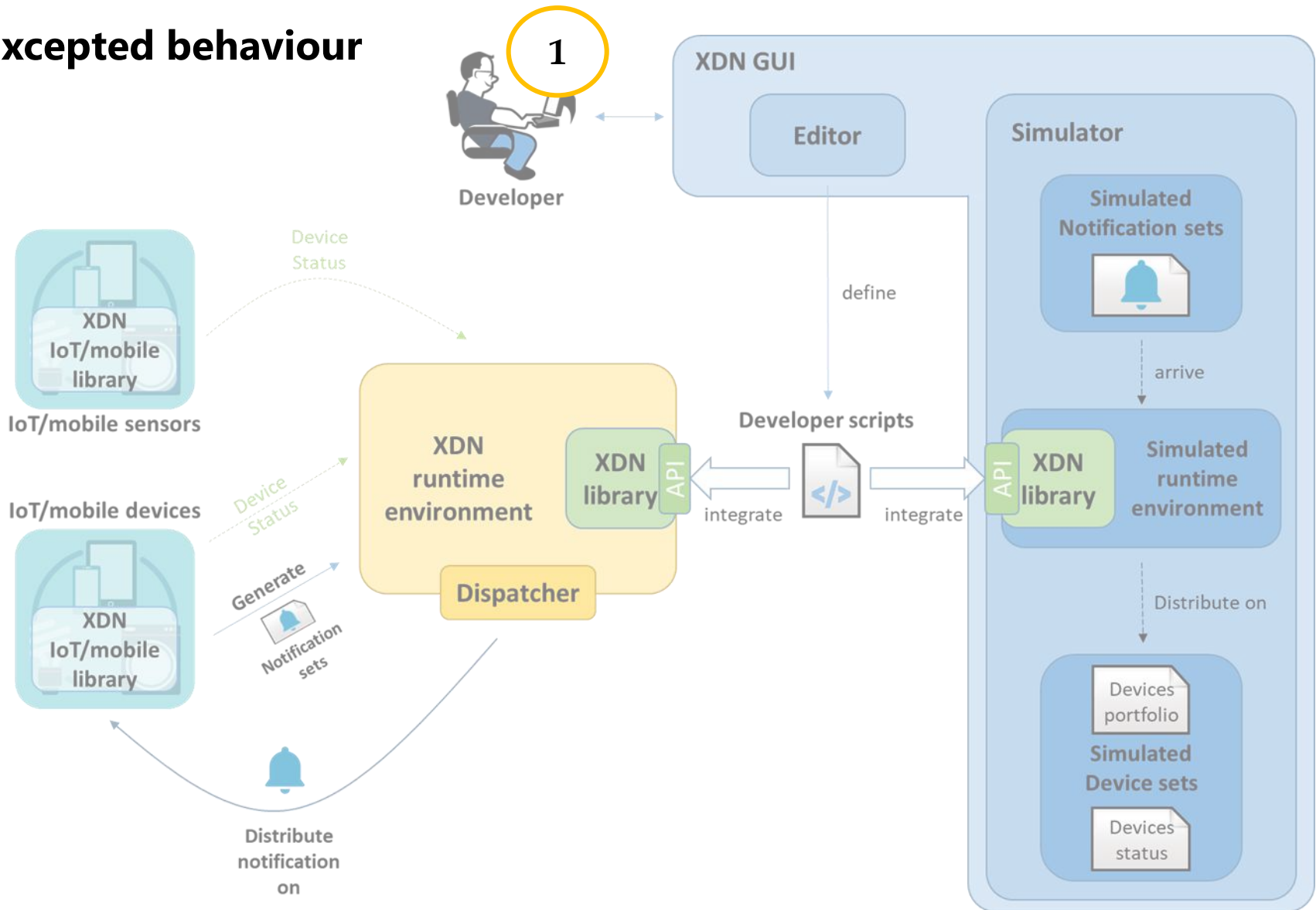
XDN: Architecture

Excepted behaviour



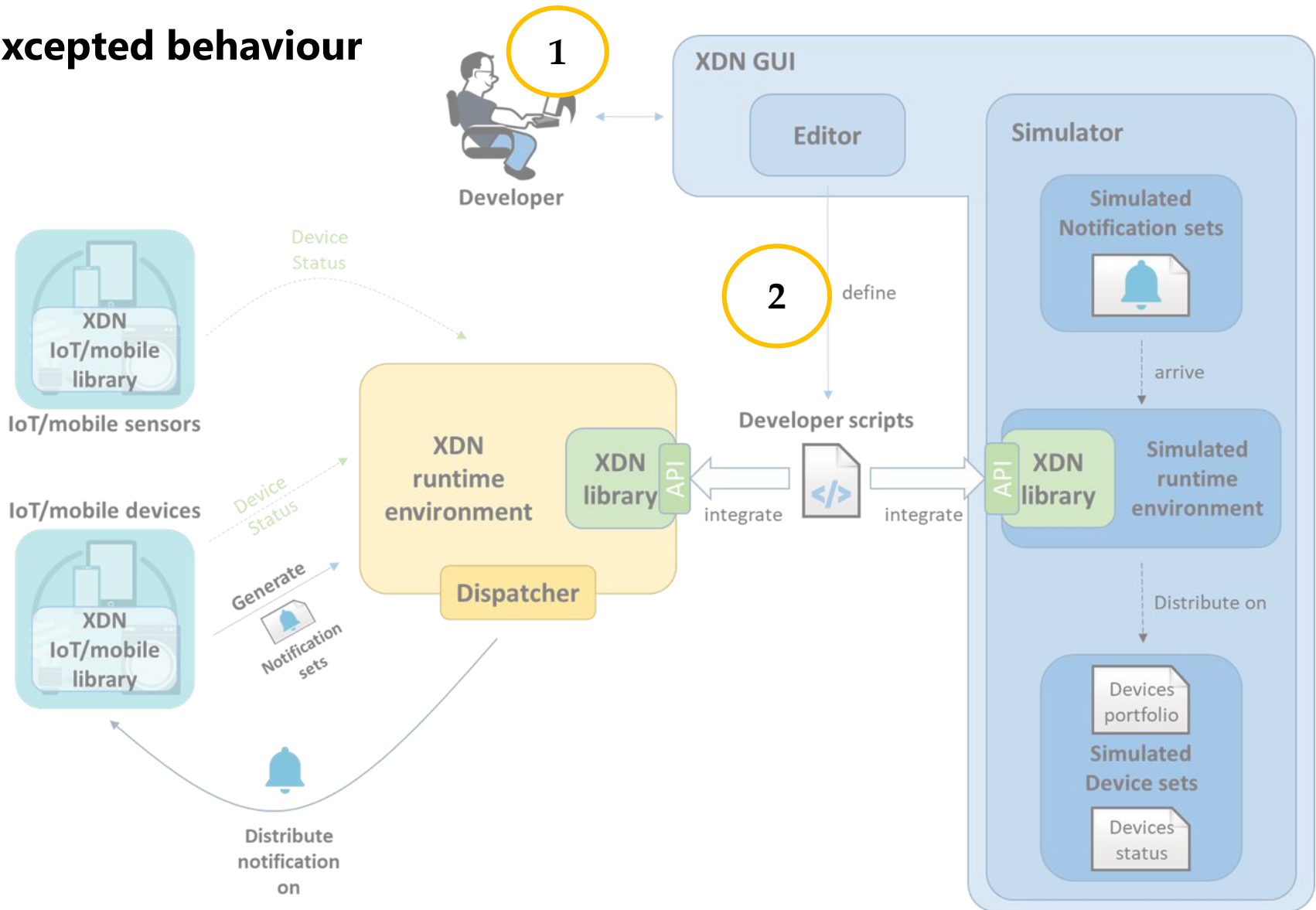
XDN: Architecture

Excepted behaviour



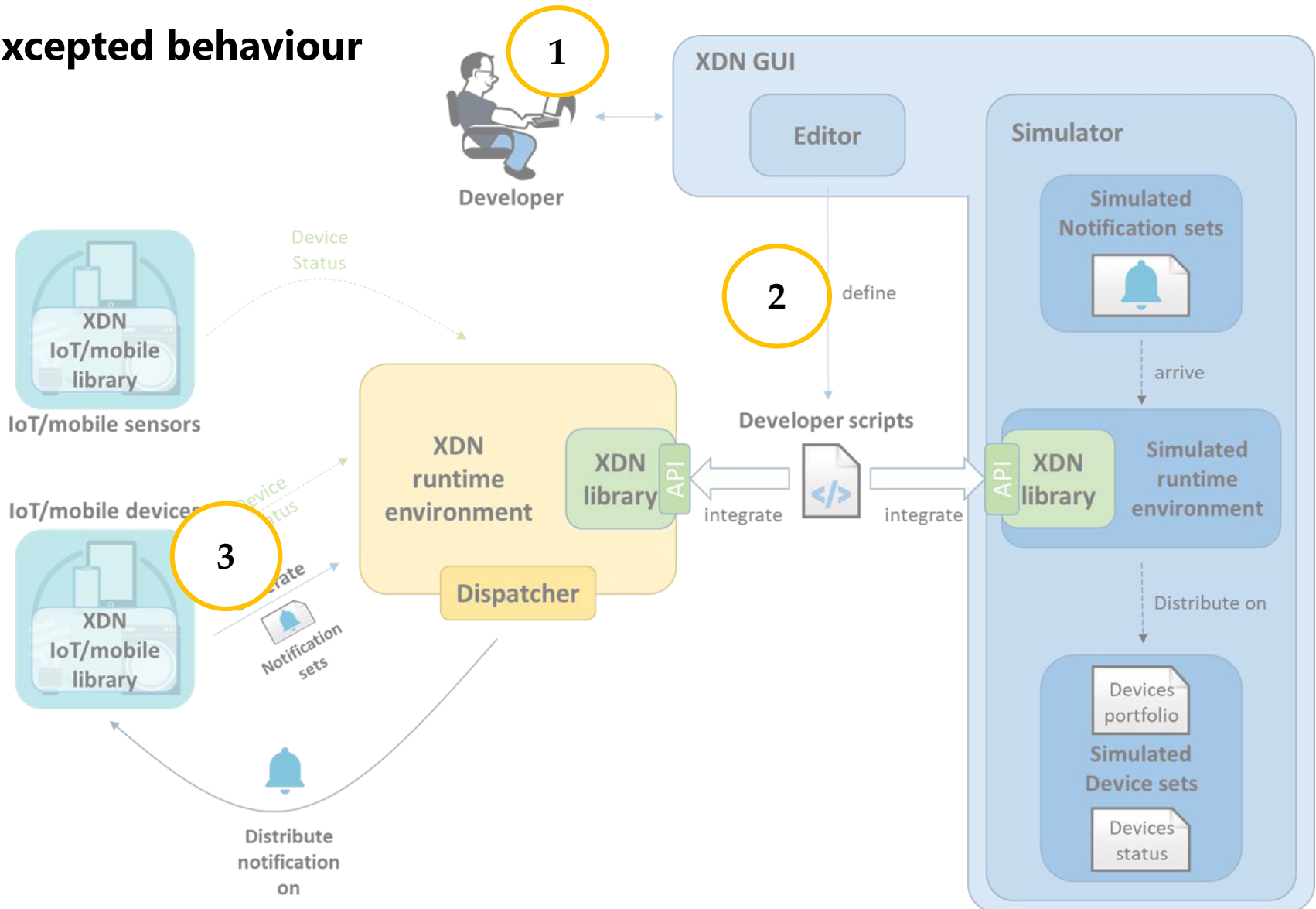
XDN: Architecture

Excepted behaviour



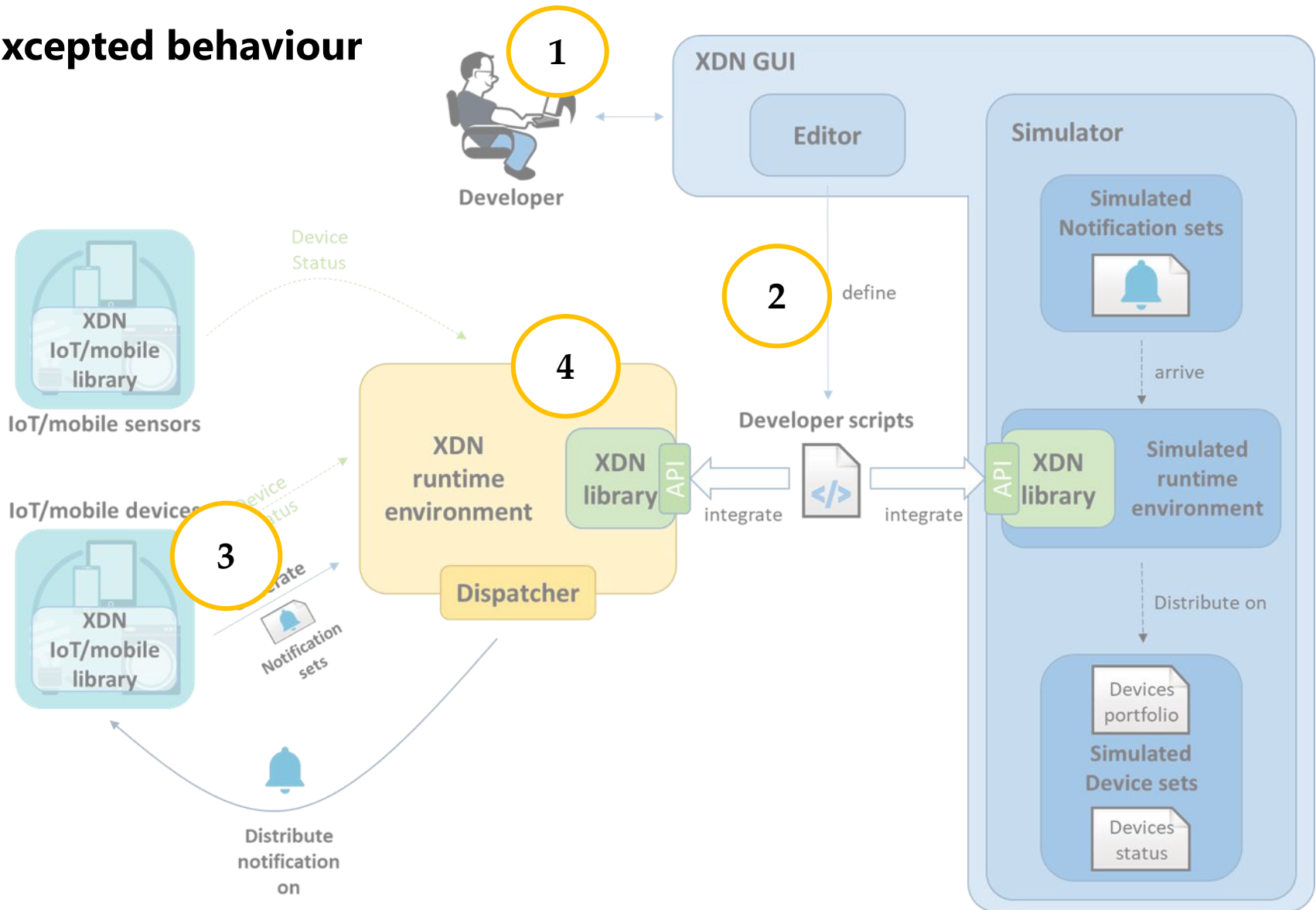
XDN: Architecture

Excepted behaviour



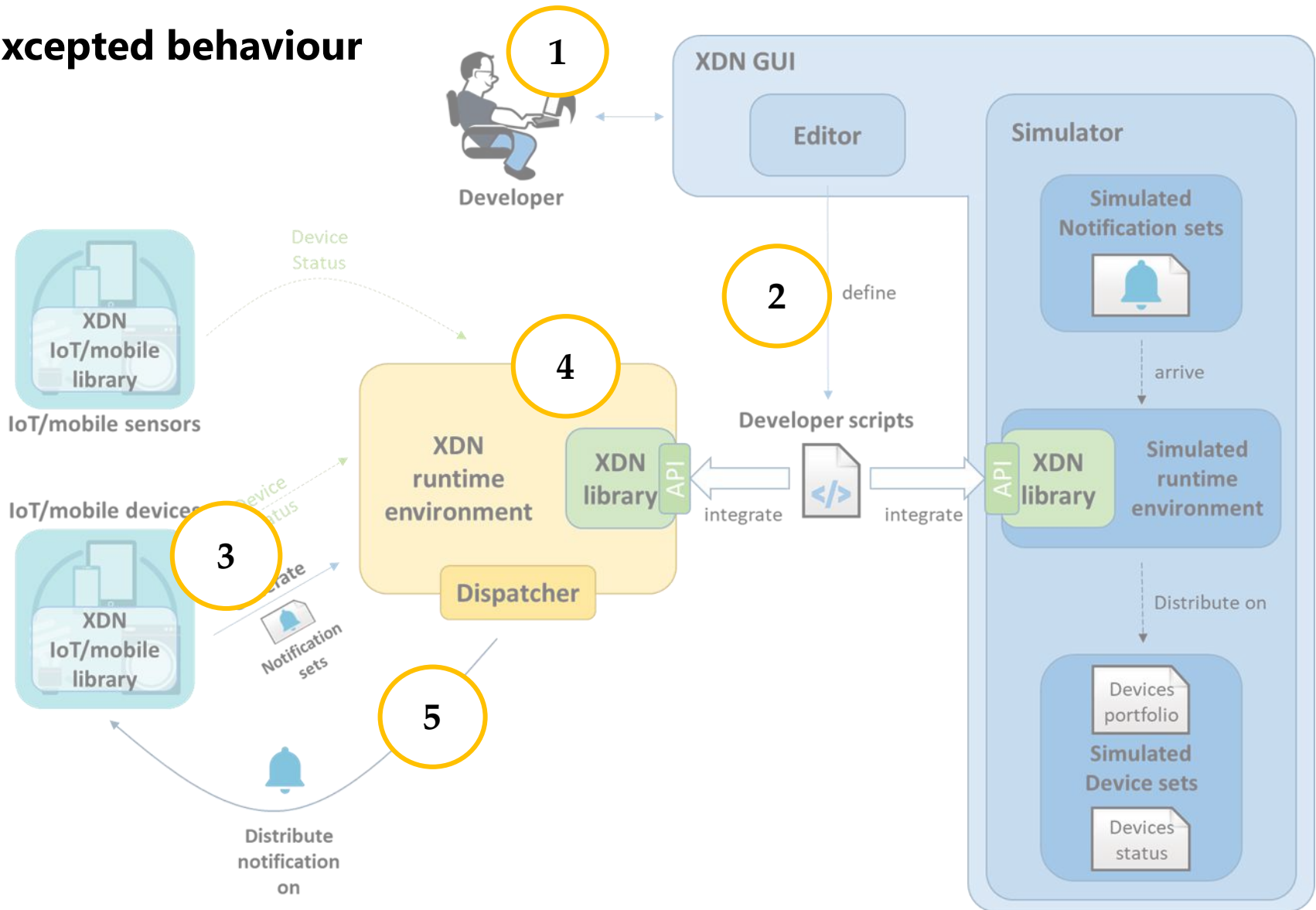
XDN: Architecture

Excepted behaviour



XDN: Architecture

Excepted behaviour



XDN: first prototype

2 components were developed:

1. The **XDN library (API)**
2. The **XDN GUI**

Tests with 12 volunteers (11 males and 1 female)

Aims:

- **demonstrate the fulfillment of all the requirements**
- **collect a feedback** about APIs and GUI

Each **user tasks**:

- modify an existing notification strategy
- develop a new notification strategy respecting some given requirements

Volunteers' main requirement:

- Previous experience with JavaScript

Used tools



XDN: first prototype

Results:

7 participants over 12 were able to complete all the tasks in the required time.

User feedback: **survey** (from 0 to 5)

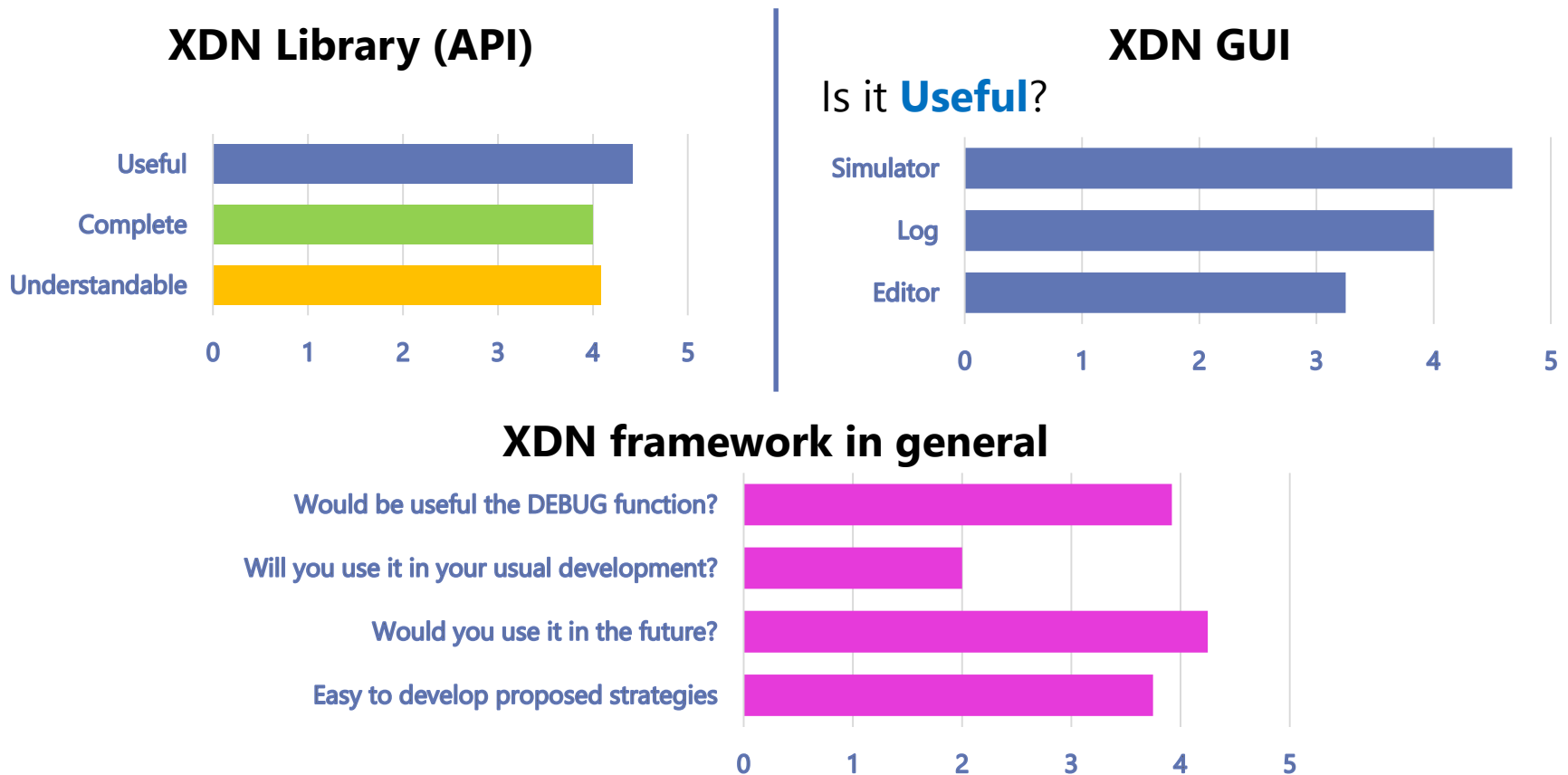


Table 3.6 - Final survey proposed to user

XDN: first prototype

Results:

7 participants over 12 were able to complete all the tasks in the required time

XDN Main outcome

- **Efficacy** of the proposed solution to **enhance developers** that want to design, develop and test their own notification strategies

Easy to develop proposed strategies

0 1 2 3 4 5

Table 3.6 - Final survey proposed to user

Thesis Conclusions

Main Problem: Overwhelming notifications

Our proposals:

1. **SNS** that acts at the **distribution** level and fosters **ML algorithms** (autonomous system that directly influences **end-users**)
2. **XDN** that acts at the **design** level and fosters **cross-device approach** (framework for **developers**)

Main outcome:

- **Feasibility** of the proposed approaches was demonstrated
- **Efficacy** of the **proposed solutions** to **enhance**
 - **user experience** with notifications
 - **developers** support in designing, developing and testing their own notification strategies also exploiting the **cross-device approach**
- **Efficacy** of the **user-centered design** methodology in notification domain

Publications during the Ph.D.

2018

- Corno, F and De Russis, L. and Marcelli, A. and Montanaro, T. . **An Unsupervised and Non-Invasive Model for Predicting Network Resource Demands.** In *IEEE Internet of Things Journal*.
- Cagliero, L. and De Russis, L. and Farinetti, L and Montanaro, T. . **Improving the effectiveness of SQL learning practice: a data-driven approach.** In *2018 IEEE 42nd Annual Computer Software and Applications Conference (COMPSAC)*.

2017

- Corno F.; De Russis L.; Montanaro T.- **XDN: Cross-Device Framework for Custom Notifications Management** - In: *The 9th ACM SIGCHI Symposium on Engineering Interactive Computing Systems, Lisbon (Portugal), June 26-29, 2017. (In Press)*
- Corno, Fulvio; Montanaro, Teodoro; Migliore, Carmelo; Castrogiovanni, Pino - **SmartBike: an IoT Crowd Sensing Platform for Monitoring City Air Pollution.** In: *INTERNATIONAL JOURNAL OF ELECTRICAL AND COMPUTER ENGINEERING (IJECE, ISSN: 2088-8708, a SCOPUS indexed Journal - Q2), vol. 7 n. 6. (In Press)*



Publications during the Ph.D.

2016

- Corno F., De Russis L., Montanaro T. - **Estimate User Meaningful Places through Low-Energy Mobile Sensing**. In: SMC 2016: IEEE International Conference on Systems, Man, and Cybernetics, Budapest, 9-12 October, 2016.
- Ghajargar M., Zenezini G., and Montanaro T. - **Home delivery services: innovations and emerging needs**. In: 8th IFAC Conference on Manufacturing Modelling, Management and Control MIM 2016, Troyes, France, 28—30 June 2016. pp. 1371-1376



2015

- Corno F.; De Russis L.; Montanaro T.; Castrogiovanni P. - **IoT Meets Exhibition Areas: a Modular Architecture to Improve Proximity Interactions**. In: FiCloud 2015: The 3rd International Conference on Future Internet of Things and Cloud, Roma, 24-26 August, 2015. pp. 293-300
- Corno, Fulvio; De Russis, Luigi; Montanaro, Teodoro - **A Context and User Aware Smart Notification System**. In: IEEE 2nd World Forum on Internet of Things (WF-IoT), Milan, Italy, 14-16 December 2015. pp. 645-651
- Montanaro, Teodoro (2015) - **SWARM Joint Open Lab Politecnico Di Torino, Italy**. In: CROSSROADS, vol. 22 n. 2, pp. 70-71. - ISSN 1528-4972





**THANK YOU
FOR YOUR
ATTENTION**

Summary of contributions

Main Problem: Overwhelming notifications

Our proposals:

- 1. SNS** acts at the **distribution** level and fosters **ML algorithms** (autonomous system that directly influences **end-users**)
 - Decision Maker
 - Collectors
 - Context Analysis
- 2. XDN** acts at the **design** level and fosters **cross-device approach** (framework for **developers**)
 - XDN library
 - XDN GUI
 - XDN Runtime Environment
 - XDN IoT/Mobile library



**THANK YOU
FOR YOUR
ATTENTION**