



*Be smart in a SNAP!*

HERIT-DATA Meeting  
15-07-2020 on line



<https://herit-data.interreg-med.eu/>

SCALABLE SMART ANALYTIC APPLICATION BUILDER FOR SENTIENT CITIES




UNIVERSITÀ  
DEGLI STUDI  
FIRENZE

DINFO  
DIPARTIMENTO DI  
TECNOLOGIA DELL'INFORMAZIONE

DISIT  
DISTRIBUTED SYSTEMS  
AND INFRASTRUCTURE  
TECHNOLOGIES LAB

- 14:00-15:00
  - overview DISIT activity (herit data organization on snap4city)
  - demo of Twitter Vigilance
  - comparison with former tool of Almaviva
  - Integration of Twitter Vigilance with Dashboard and IOT App in Node-red
- 15:00-16:00
  - Acquired data from the cities
- 16:00-16:30
  - exploitation of data from IOT App and Dashboard
  - data analytic on accessible data vs COVID lockdown


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# Work Done

- **Organizations on Snap4City have been created**
  - Knowledge base and maps areas are active
  - According to your login on left side you see KB and Twitter Vigilance, etc.
- **Twitter Vigilance Channels have been set up,**
  - accounts are available (please ask), only 1 or 2 have been provided
  - Keywords have been setup and Tweets collected
  - NLP, SA has been activated so far in Italian and English
- **Example IOT Application** collecting data from Twitter Vigilance for processing them for Dashboard has been created
- **Data Collection**
  - Template for collecting data have been distributed
  - Only few data have been provided
  - Some of them have been started processing, collecting: Florence, Dubrovnik??

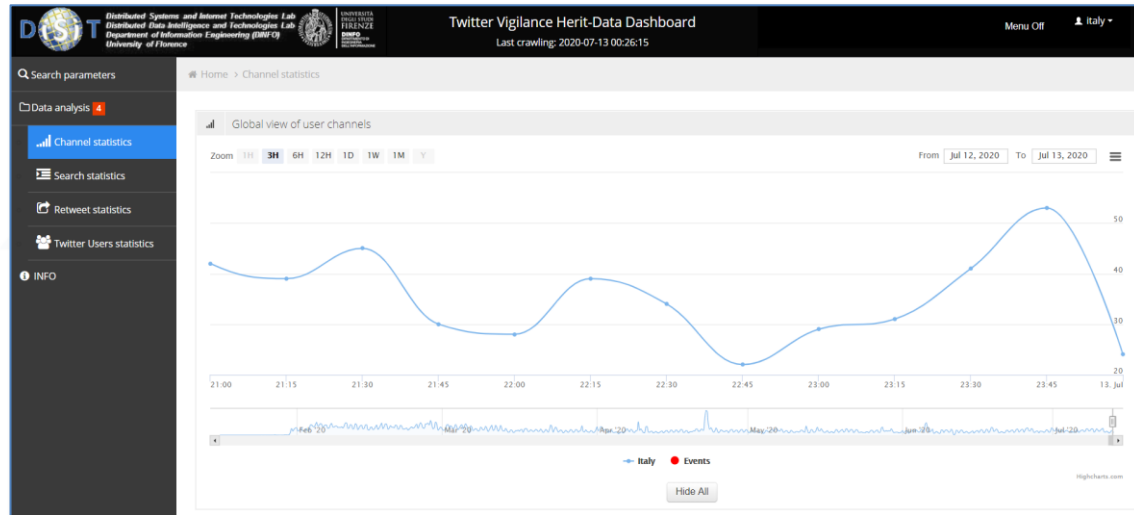




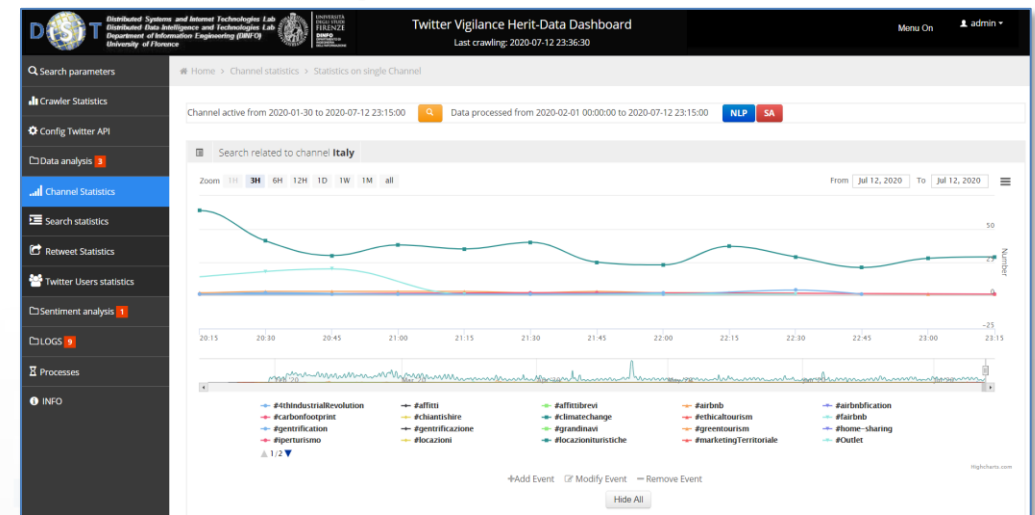
- 14:00-15:00
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# Twitter Vigilance Herit-Data: Channel & Key-Search

- Twitter Vigilance for Herit-Data: <https://rttvhd.snap4city.org/>
- Data Acquisition is based on the concept of **Channel**
- A *Channel* is a thematic set of simple and complex **Key-Searches**



Italy



#4thIndustrialRevolution #affitti #affittibrevi #airbnb  
#airbnbfcation #carbonfootprint #chiantishire #climatechange  
#ethicaltourism #fairbnb #gentrification #gentrificazione [...]

- $Channel := \{ Key-Search1, Key-Search2, \dots, Key-SearchN \}$ .
- For a specific channel, the Twitter Vigilance platform collects Twitter posts (tweets and retweets) containing:  
 $Key-Search1 \text{ OR } Key-Search2 \text{ OR } \dots \text{ OR } Key-SearchN$

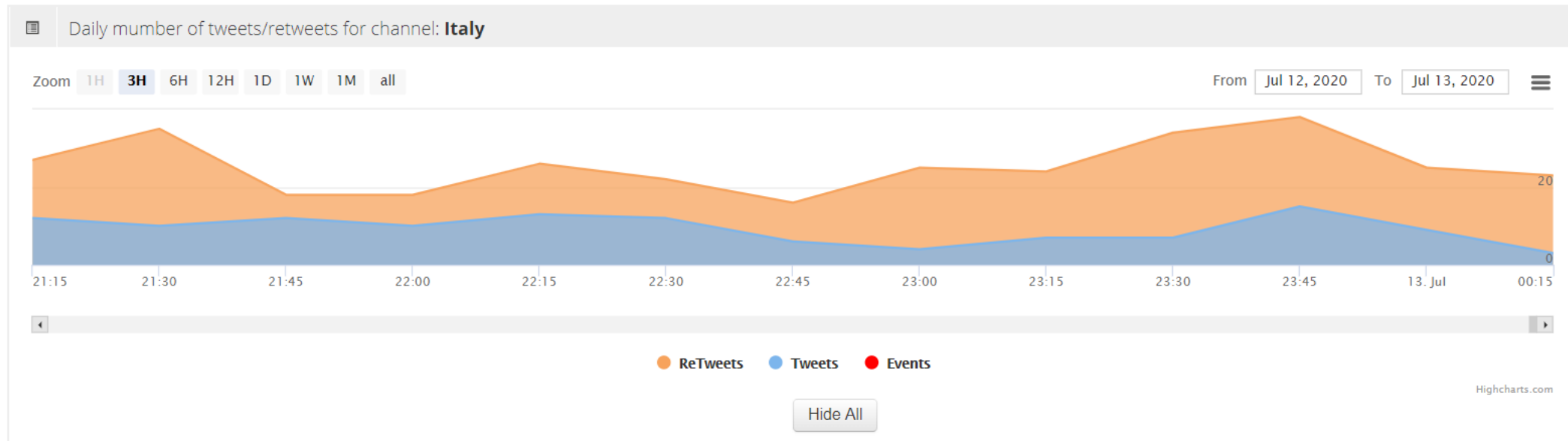
# Twitter Vigilance Herit-Data: *Key-Search*

- **Key-Searches** can be single or groups of terms used to monitor and extract tweets and retweets information through Twitter APIs.
- Each term can be:
  - *keyword*
  - *#<hashtag>*
  - *@<username> citation*
  - *from: @<username>*
- A complex **Key-Search** can combine several terms: string keywords, hashtags, username citations etc.
- Terms in a complex Key-Search are logically combined in AND relation:  
crowded beach Dubrovnik search for: crowded AND beach AND Dubrovnik in the same Tweet message.

➤ Channel / Key-Search Statistics: Total Number of Posts (Tweets + Retweets)



➤ Graphical details: charts, temporal trends of tweets and retweets for Channel / Key-Search



Research list in channel **Italy** research

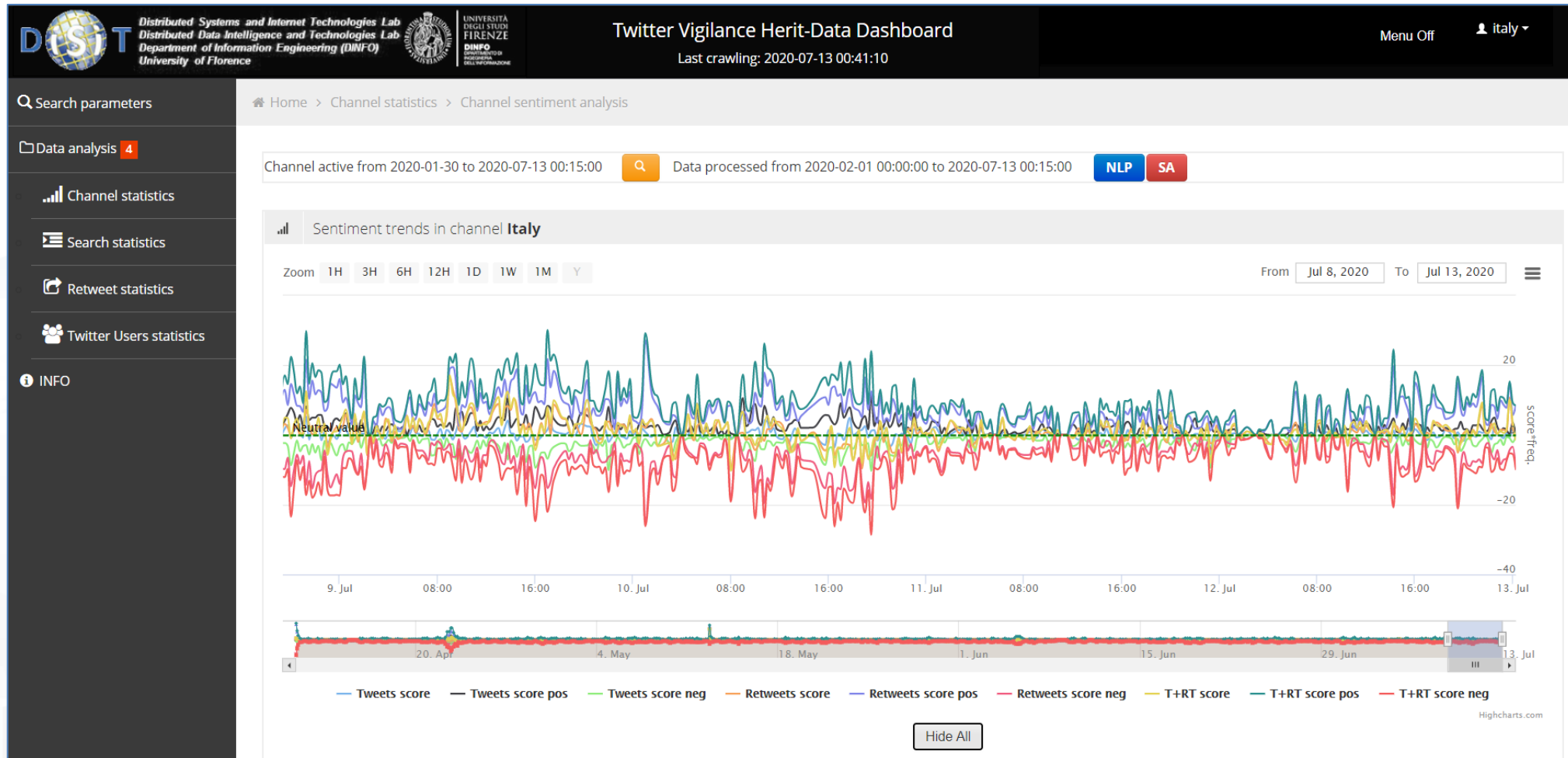
Click research for T and RT details

#4thIndustrialRevolution	#affitti	#affittibrevi	#airbnb	#airbnbification	#carbonfootprint	#chiantishire	#climatechange	#ethicaltourism	#fairbnb	#gentrification	#gentrificazione	#grandinavi
#greentourism	#home-sharing	#iperturismo	#locazioni	#locazionituristiche	#marketingTerritoriale	#Outlet	#overtourism	#responsibletravel	#sharingEconomy	#socialtourism		
#SustainableDevelopmentGoals	#sustainabletourism	#Tourism4SDGs	#turismoEnogastronomico	#turismoEsperenziale	#turismoetico	#turismoSmart	#turismosostenibile	#turismoverde	#voluntourism			

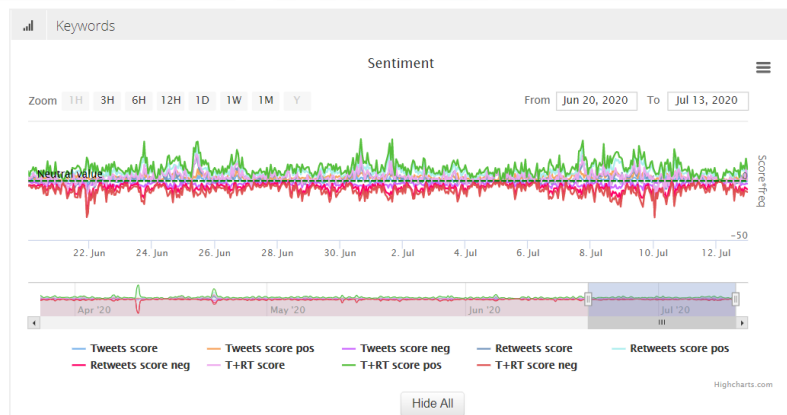
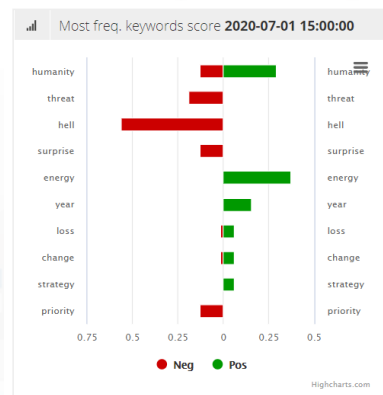
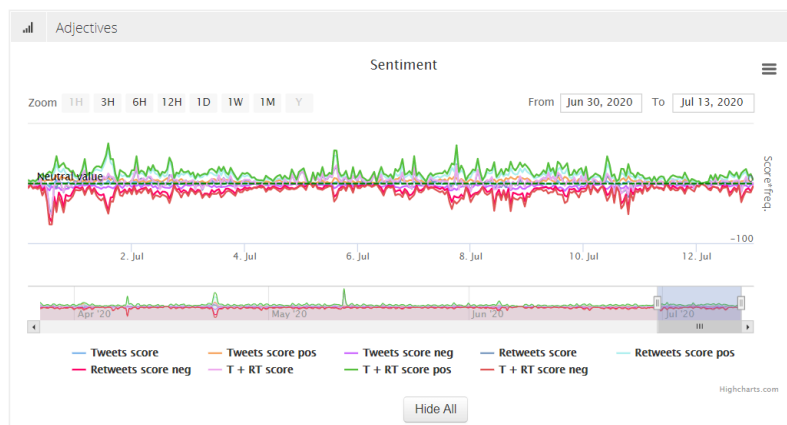
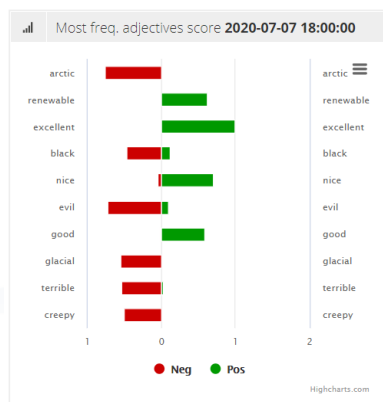




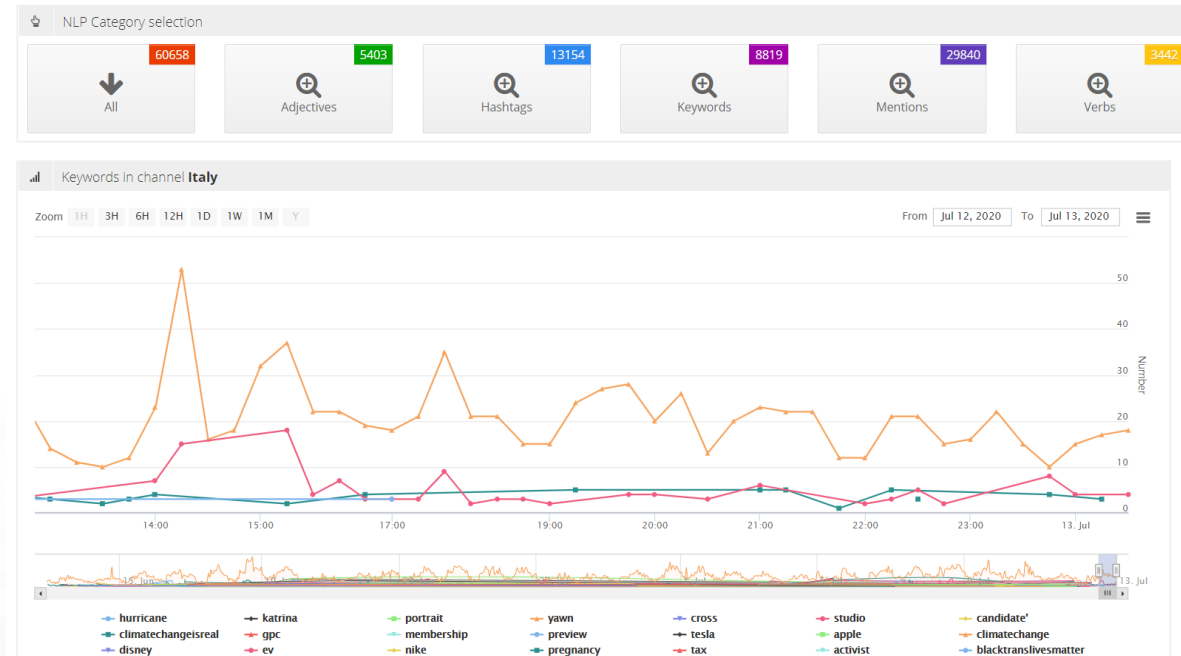
- Natural Language Processing (NLP) and Sentiment Analysis (SA) of collected tweets and retweets for Channel / Key-Search



## ➤ Sentiment Analysis for different Part-Of-Speech: Adjectives, Nouns, Verbs.

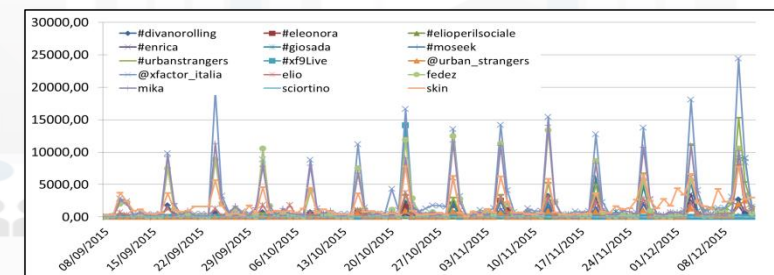
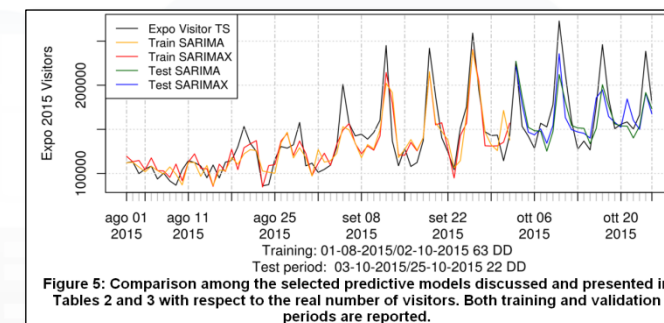
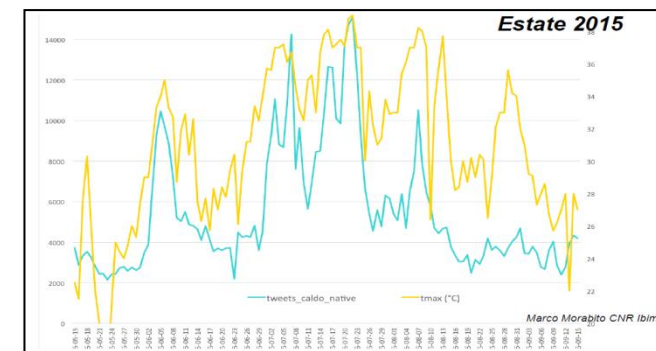


## ➤ NLP Extraction of different Part-Of-Speech: Adjectives, Nouns, Verbs, Hashtags, Usernames.

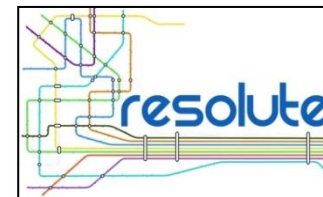


## Prediction/Assessment

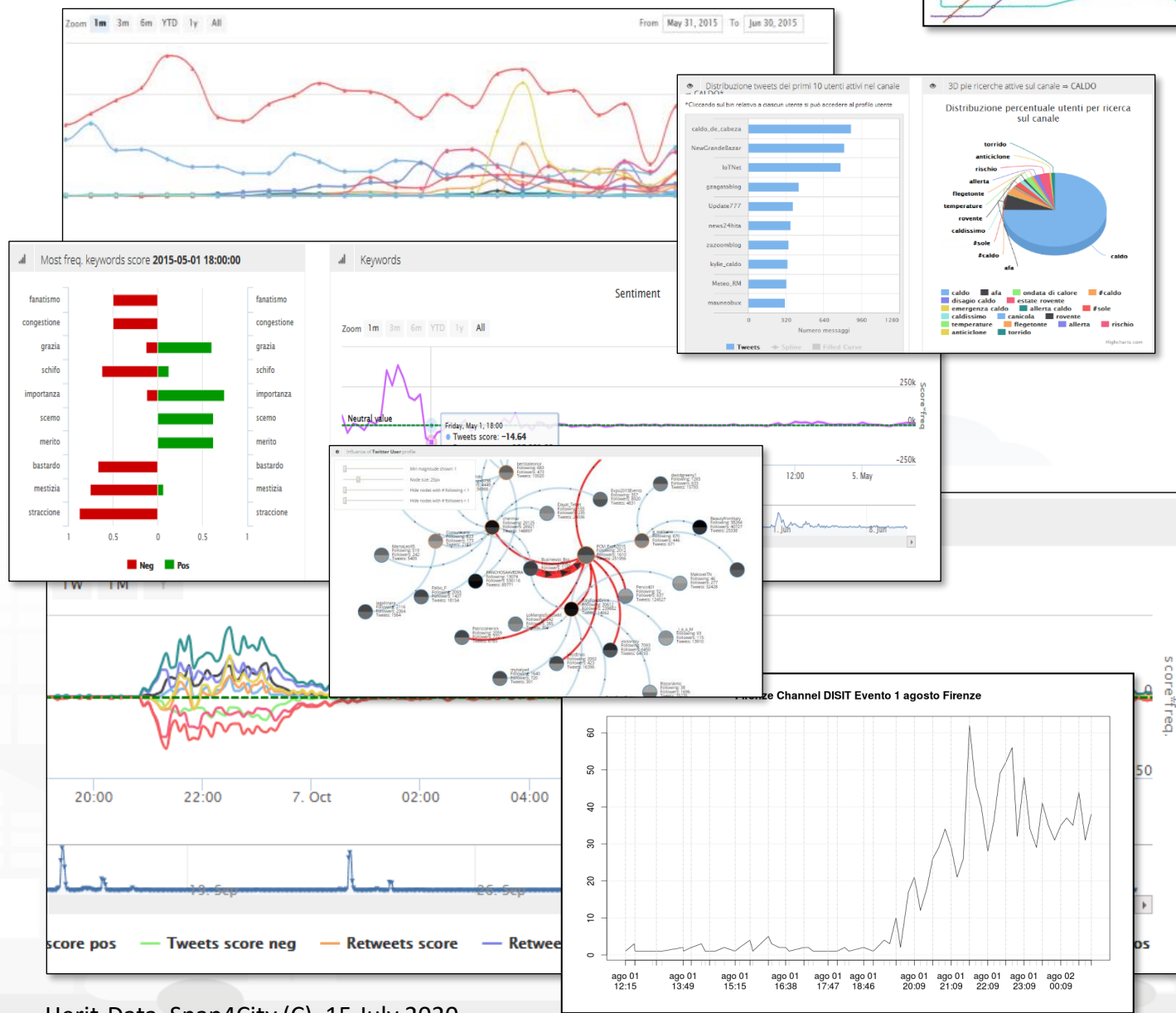
- Football game results as related to the volume of Tweets
- Number of votes on political elections, via sentiment analysis, SA
- Size and inception of contagious diseases
- marketability of consumer goods
- public health seasonal flu
- box-office revenues for movies
- places to be visited, most visited
- number of people in locations like airports
- audience of TV programmes, political TV shows
- weather forecast information
- Appreciation of services



# Twitter Vigilance



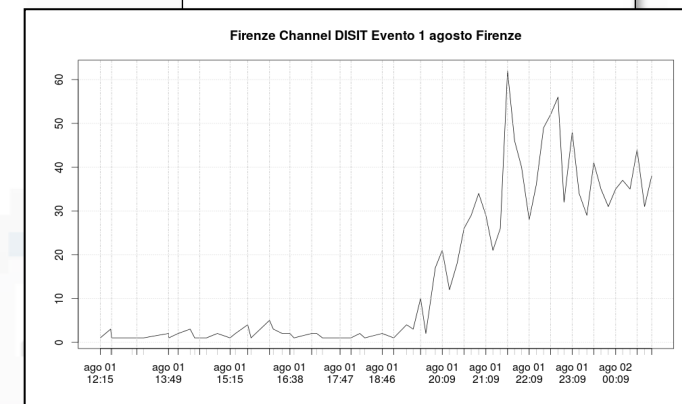
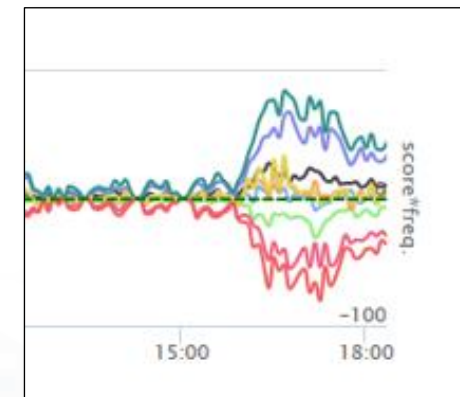
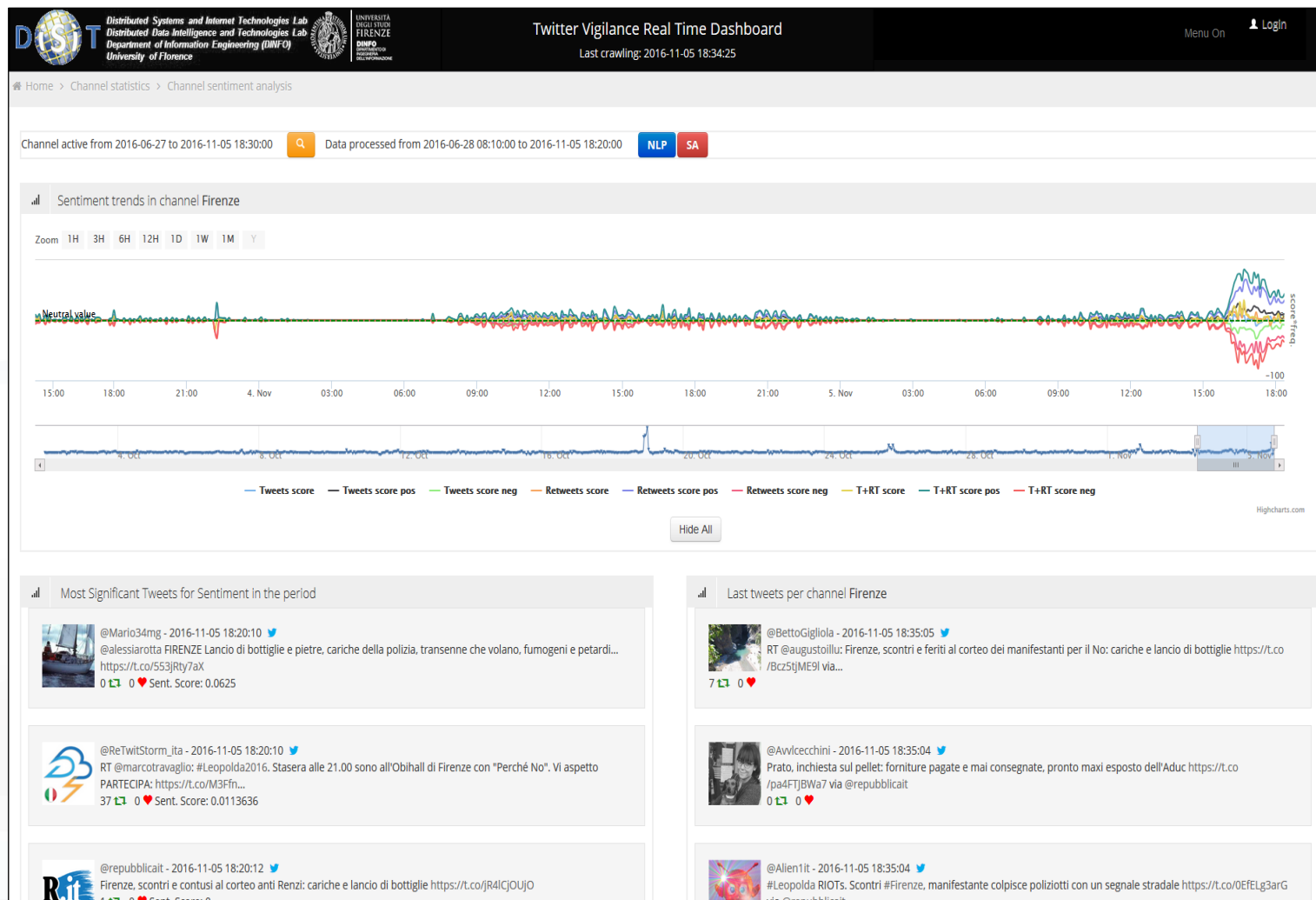
- <http://www.disit.org/tv>
- <http://www.disit.org/rttv>
- Citizens as sensors to
  - Assess sentiment on services, events, ...
  - Response of consumers wrt, ...
  - Early detection of critical conditions
  - Information channel
  - Opinion leaders
  - Communities
  - Formation
  - Predicting volume of visitors for tuning the services





# Twitter Vigilance RT: sentiment analysis

Real time  
Early Warning

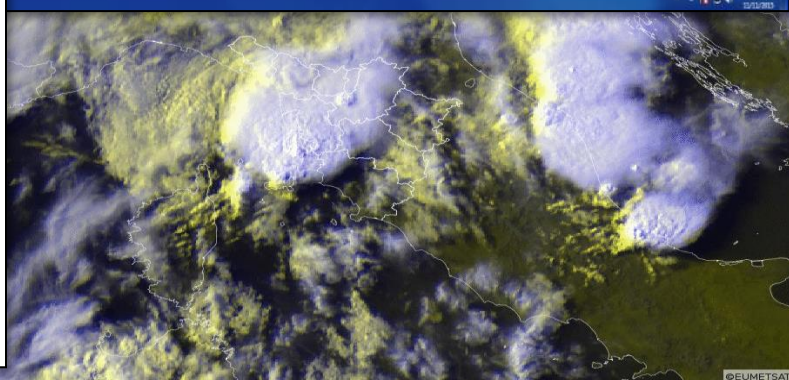
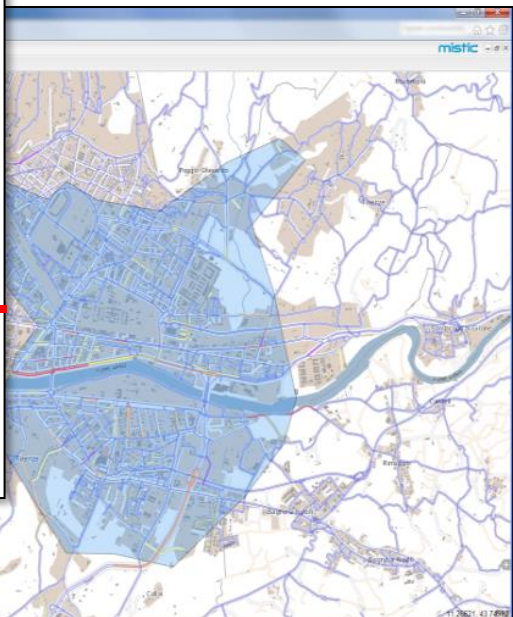
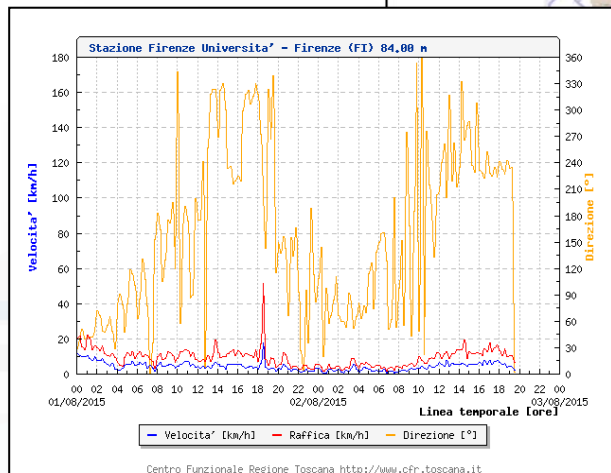
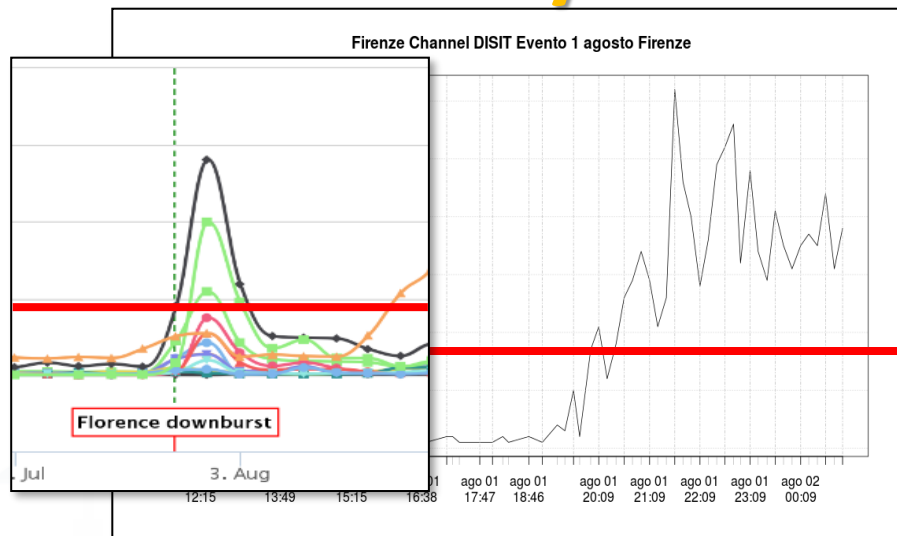




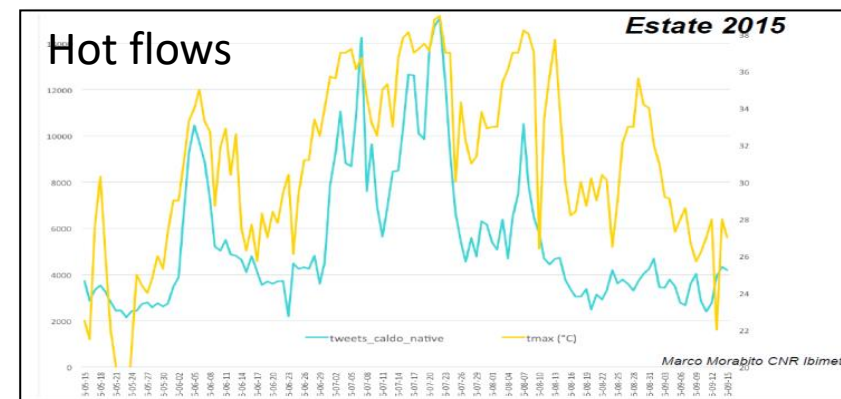


# Twitter Vigilance

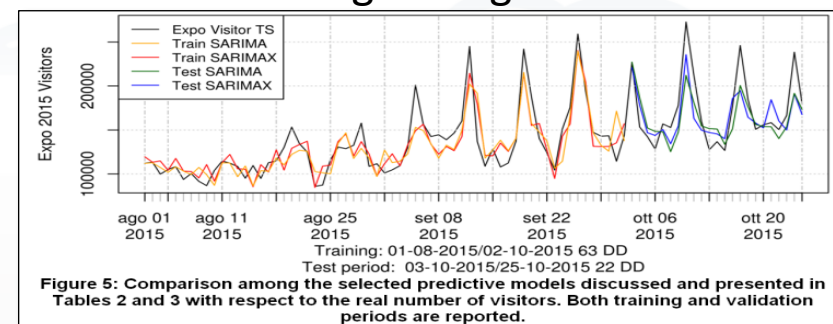
## Early Warning



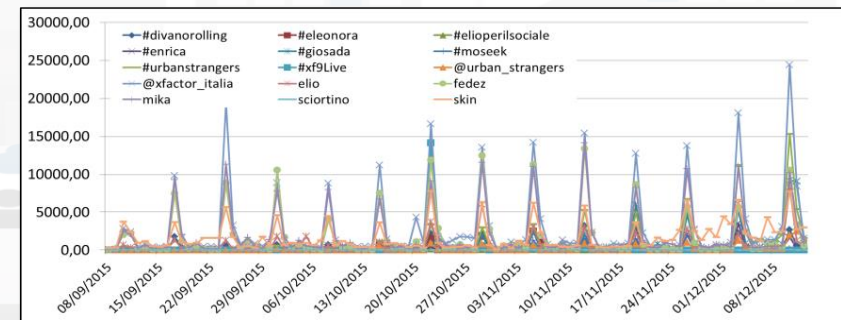
## Predictive models



## Attendance at long lasting events: EXPO2015



## Attendance at recurrent events: TV, football



# Twitter Vigilance Herit-Data: *Some Numbers*

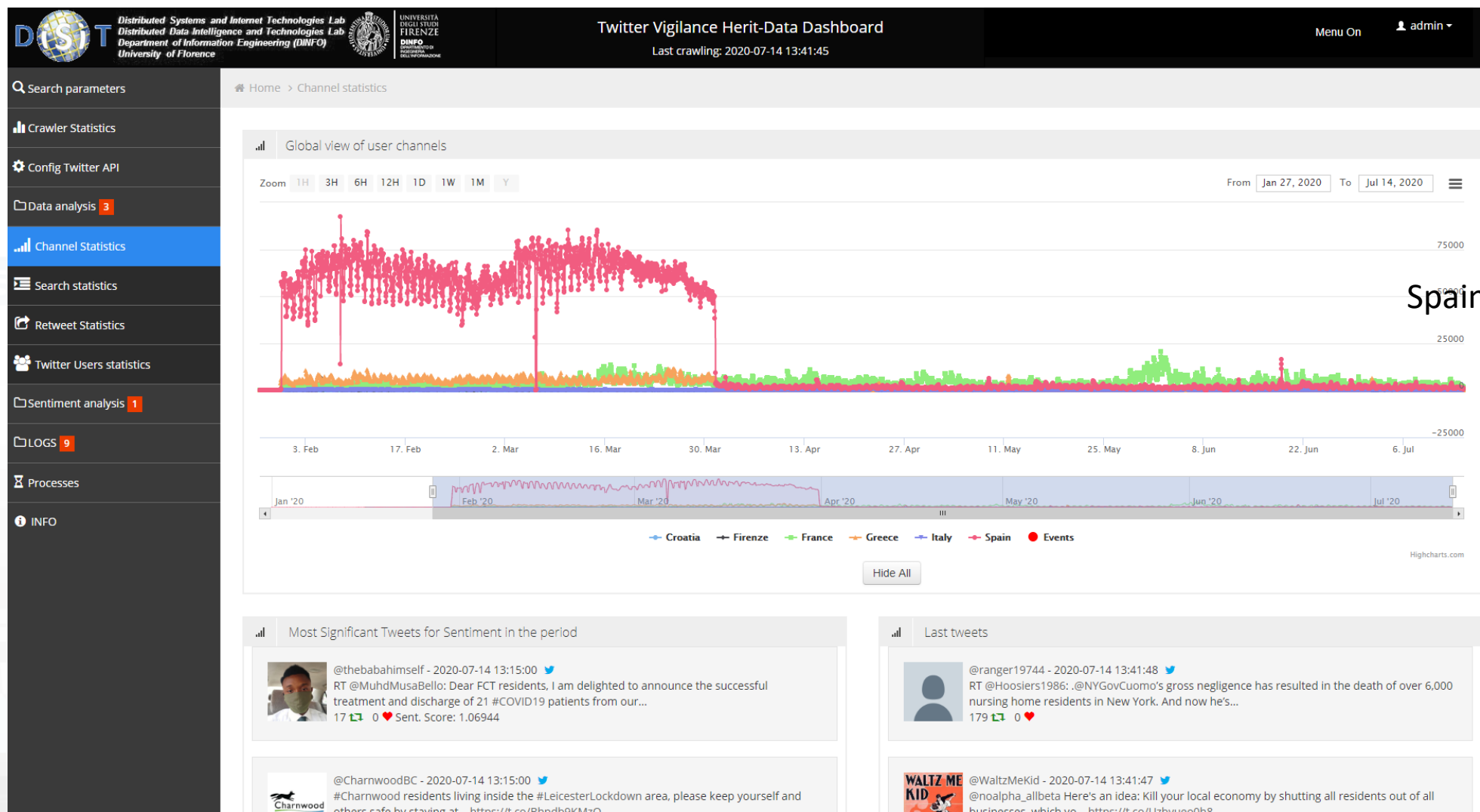
Channel Name	Total Number of Collected TW+RTW	Number of Collected Tweets	Number of Collected Retweets	Twitter Volume Processing Time Range	NLP & Sentiment Analysis Processing Time Range	NLP & Sentiment Analysis Languages
<b><i>Spain</i></b>	94.7 Millions	34.5 Millions	60.2 Millions	From 30-01-2020 to current datetime	From 01-02-2020 to 17-05-2020 and from 05-07-2020 to current datetime	English
<b><i>France</i></b>	15.8 Millions	3.7 Millions	12.1 Millions	From 30-01-2020 to current datetime	From 01-02-2020 to 8-04-2020 and from 05-07-2020 to current datetime	Italian, English
<b><i>Greece</i></b>	9.6 Millions	3 Millions	6.6 Millions	From 30-01-2020 to current datetime	From 01-02-2020 to current datetime	English
<b><i>Italy</i></b>	762 Thousands	264 Thousands	498 Thousands	From 30-01-2020 to current datetime	From 01-02-2020 to current datetime	Italian, English
<b><i>Croatia</i></b>	14 Thousands	5.3 Thousands	8.7 Thousands	From 30-01-2020 to current datetime	From 01-02-2020 to current datetime	English

- For Spain: 87 million of TW taken for too generic keys
- For France: 8.2 Million of TW taken for too generic keys

# Demo Twitter Vigilance

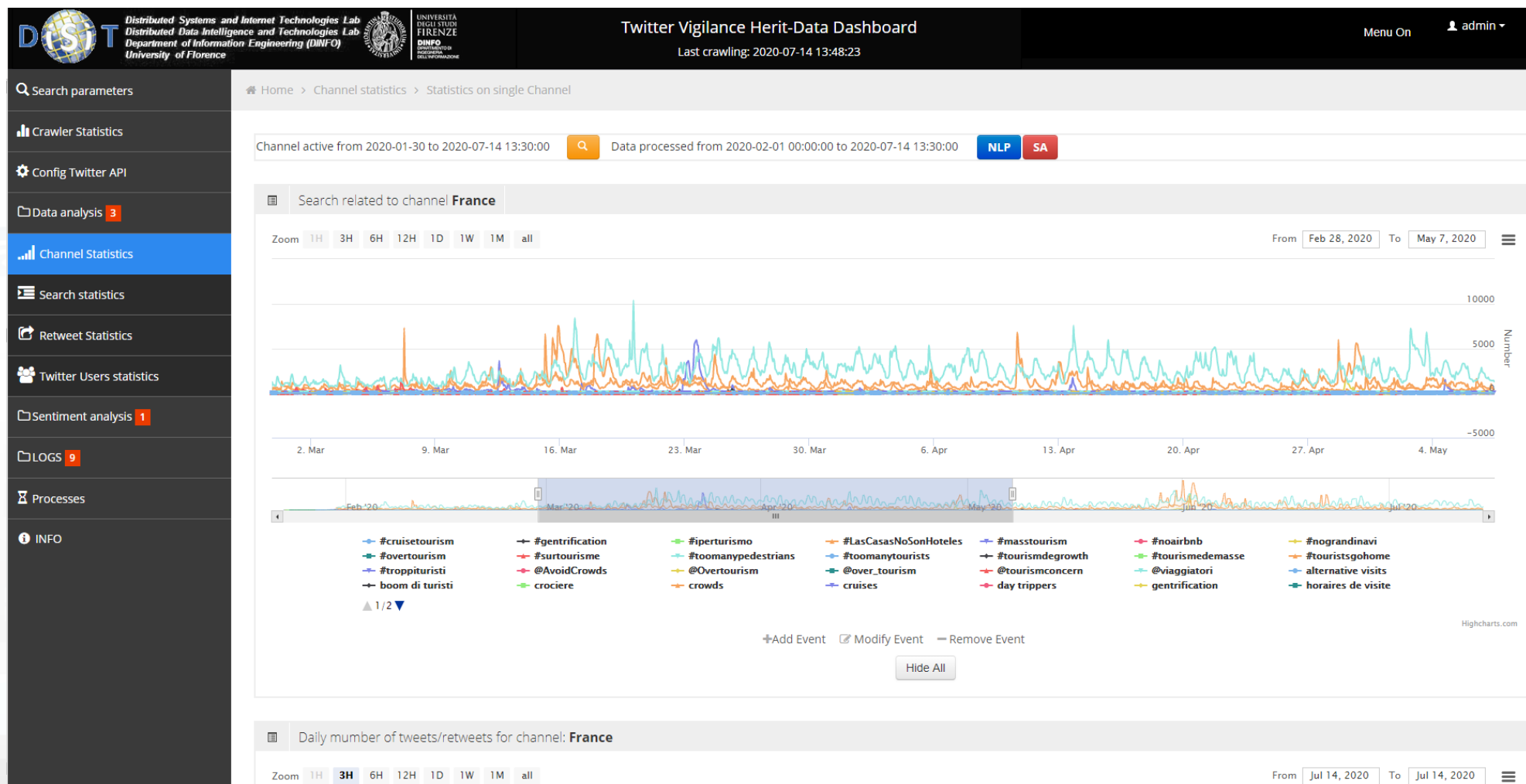
- Show Twitter Vigilance channel (Italia)
  - Trends for the Channel
  - Trends for the Keys
  - Trends for the NLP
  - Trends for the SA

# Too Generic Keywords at the beginning






# Some Keywords increased in Lockdown





## Detected issues

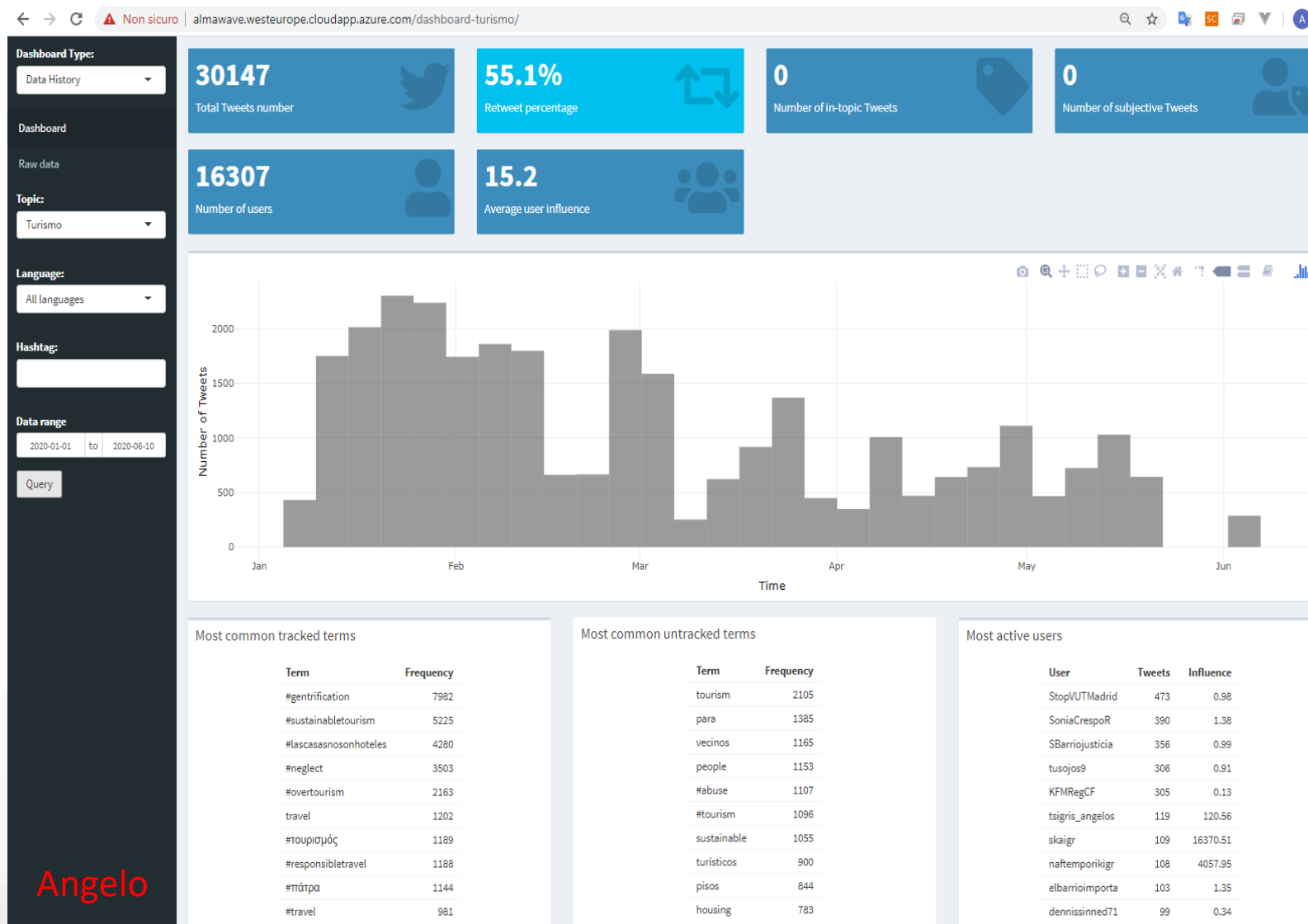
- Increment of some Keywords in Lockdown
- Decrement of some keywords in Lockdown
- Detection of spikes in Lockdown:
  - start, end,
  - in the middle probably for some events that we do not know
- Thus we performed a Data Analytic study as presented later on

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# Almaviva Tool and comparison (1)

- <http://almawave.westeurope.cloudapp.azure.com/poc>
  - final user, no admin privileges (?)
  - Doc 15/01/2019 «Descrizione architettura e addestramento modello sentiment»
- Functionalities: collect tweets of interest in specific topics, visualize in an aggregate way, enable manual SA tagging, create model to generate prediction for the sentiment
  - dashboard-turismo/
  - tagging-turismo/
  - trainingapp/ (not available for turismo?)

# Almaviva Tool and comparison (2)



## Dashboard

- data-history (from-to) / real-time (30-360min)
- topics managements
- several languages
- hashtags, users, keywords?
  - (dynamically?)
- tweet: 30147
  - en:18470
  - es:4542
  - el:2475 (Greek, Modern)
  - fr:1180
  - it:766
  - de:436,
  - nl:206,
  - hr?
- most commons

# Almaviva Tool and comparision (3)

## Tagging

- topic/off-topic
- objective/subjective
- sentiment positive/negative

Tourism Tagging App

Topic: Turismo

Language: fr

Skip tweet

Topic

☒ In Topic

☐ Off Topic

objectivity

☒ subjective

☐ Objective

Sentiment

☒ Positive

☐ Negative

Enter!

1180  
Number of downloaded tweet

NA  
Number of In-Topic tweet

NA  
Number of subjective tweet

NA  
Number of positive tweet

0  
Total number of tagged tweet

NA  
Number of Off-Topic tweet

NA  
Number of ojective tweet

NA  
Number of negative tweet

Angelo



## Dashboard Type:

Data History ▼

Dashboard

Raw data

## Topic:

Turismo ▼

## Language:

All languages ▼

## Hashtag:

#gentrification

## Data range

2020-02-01

to

2020-06-01

Query

4972

Total Tweets number



46.7%

Retweet percentage



0

Number of in-topic Tweets



0

Number of subjective Tweets



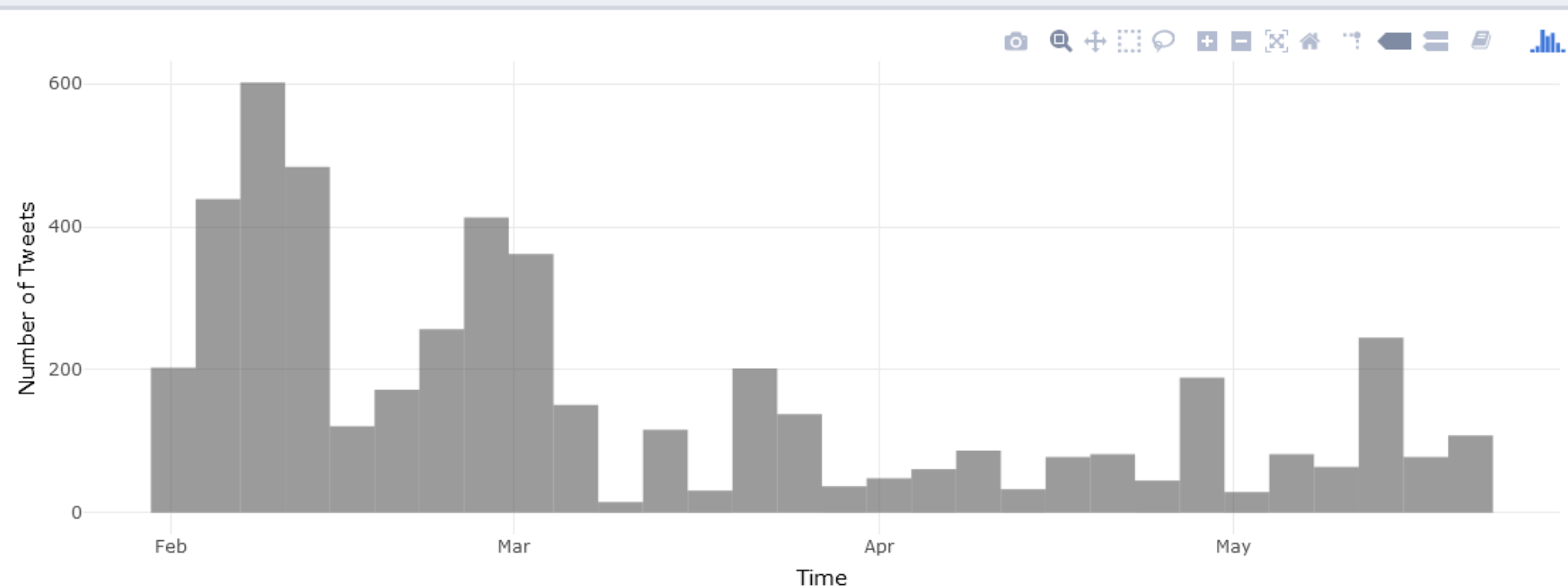
3411

Number of users



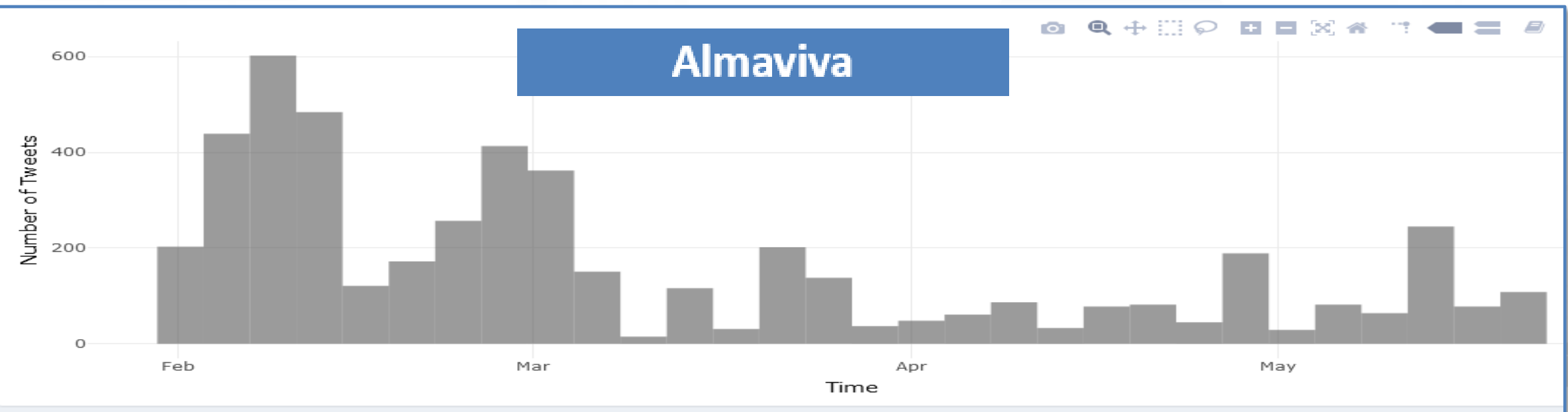
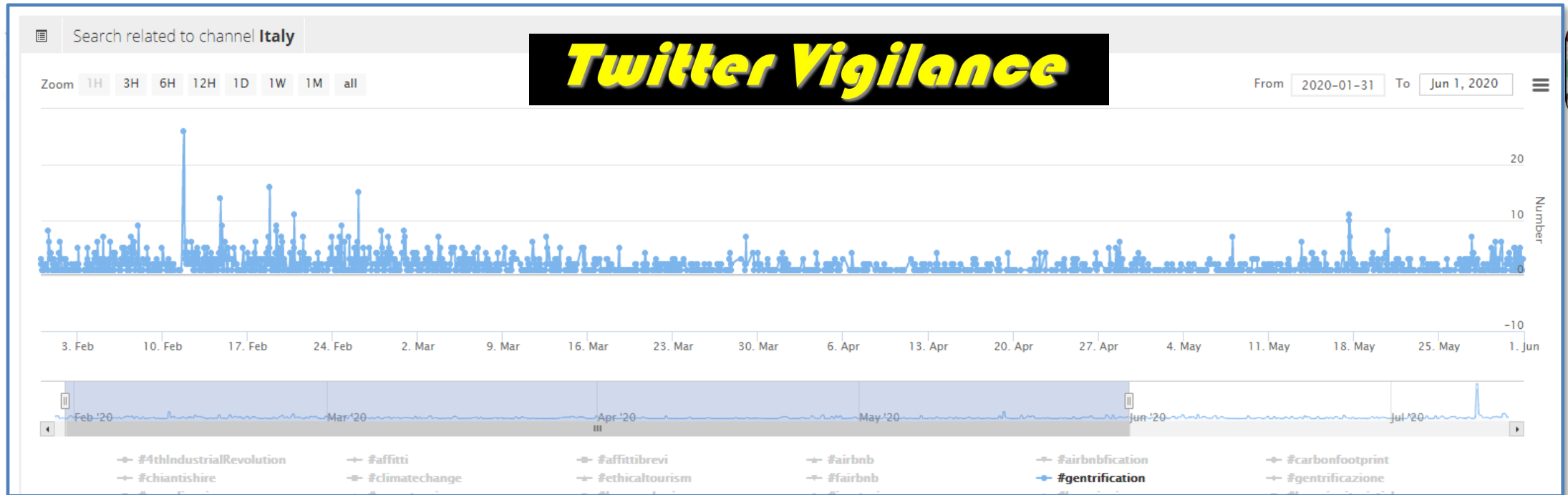
4.67

Average user influence



# Almaviva Tool and comparision (4)

	Almaviva	Twitter Vigilance
Supported languages	all (HR is missing?)	all (SentimentAnalysis currently for EN + IT)
Data collections	hashtags, users, keywords	hashtags, users, keywords
Data organization	topics (turism) (one for all pilots)	channels (Greece, Italy, ...)
Dashboard types	Data history (from-to) / real-time (2sec)	Data history (from-to) / real-time (15min)
Data presentation granularity	5 days (fixed resolution)	15 minutes up to 1 days
Data graph type	Tweets (Tw + ReTW), hashtags	tweets, retweets, hashtags, keywords, SA, citations,...
Type of data	tweet, retweet, quote tweet (?)	tweet, retweet
Export data	png	png, csv, xls, pdf, ...
User management	no	yes, with roles and sharing
NLP	no	adjectives, hashtag, keyword, mentions, verbs
Ratio TW/RTW	yes	yes
Sentiment Analysis	yes, manual	yes, sentiwordnet
Learning/Prediction on SA	yes	Not needed
User influence	yes	yes
Volume	30147	120 million
Connection with Dashboards	no	Yes
API and MicroServices	no	Yes and IOT App integration



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Paolo





# Snap4City: Builder of Sentient Cities Solutions

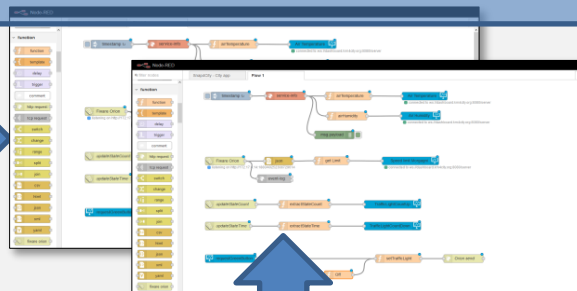
Dashboards with data driven IOT Applications enforcing intelligence

IOT and data World



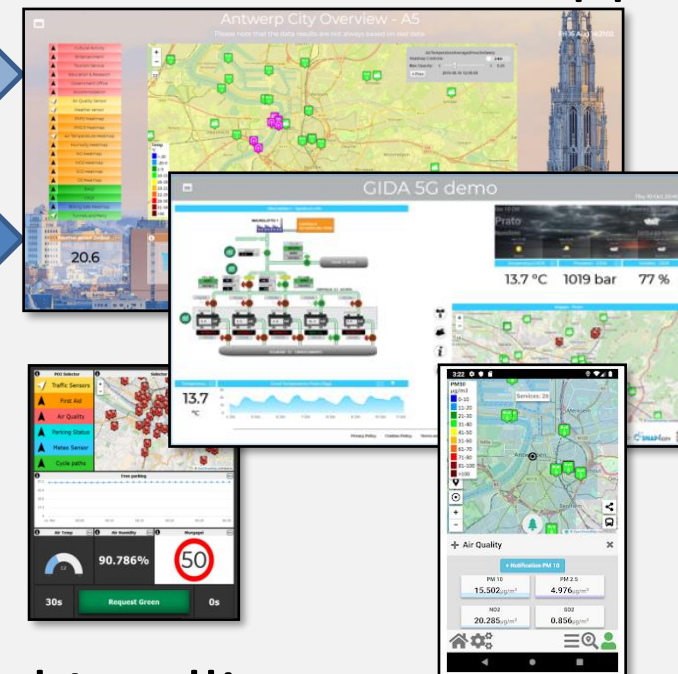
My IOT Devices

IOT Applications



Big Data Analytics, Artificial Intelligence

Dashboards and Apps



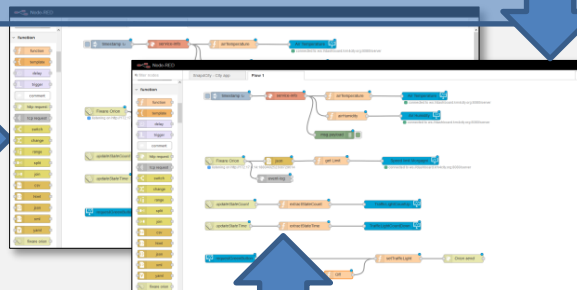
# Twitter Vigilance

IOT and data World



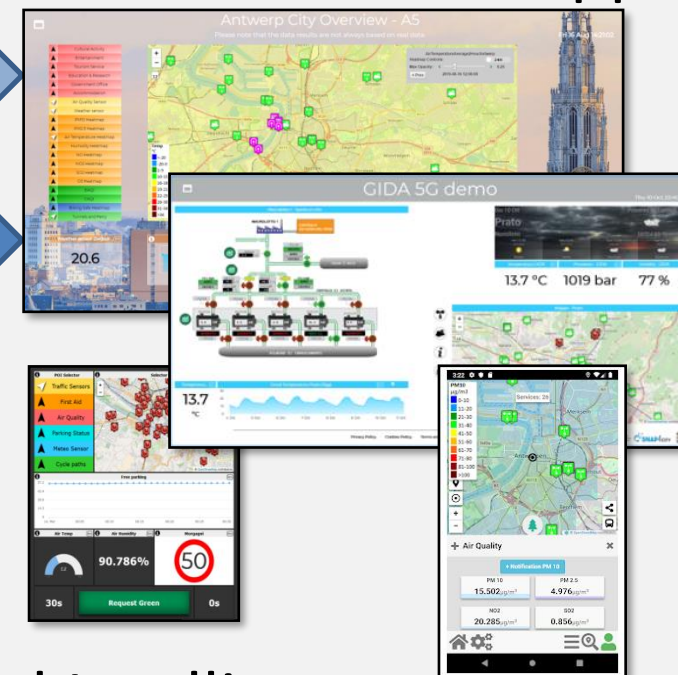
My IOT Devices

IOT Applications



Big Data Analytics, Artificial Intelligence

Dashboards and Apps

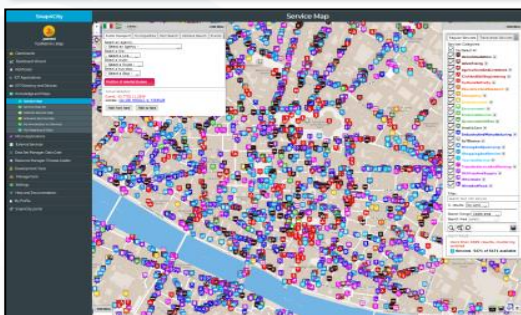
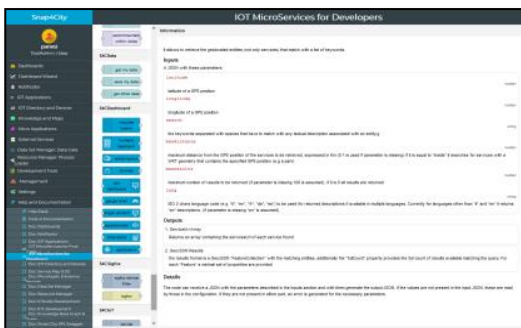




# IOT Applications Development

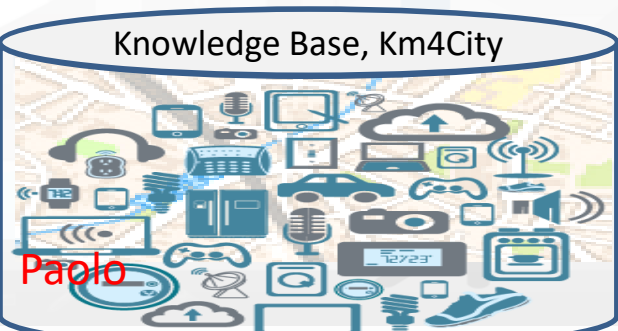
IoT Discovering

MicroServices collections



ServiceMap Discovery

Knowledge Base, Km4City

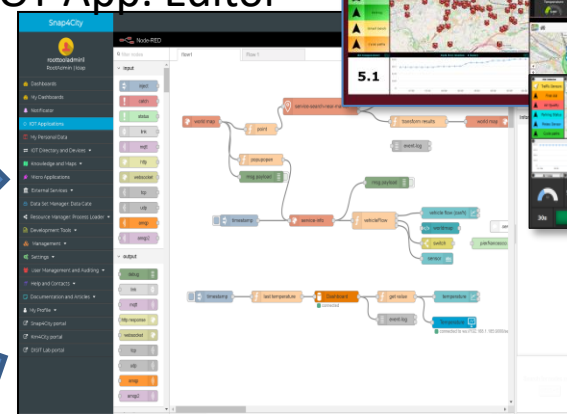


My IOT Applications



Dashboard Collection,  
Editor and Wizard

IOT App. Editor



Sharing/saving  
reusing IOT App

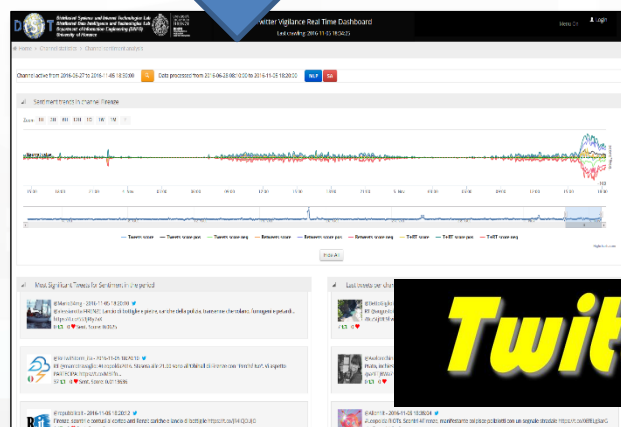
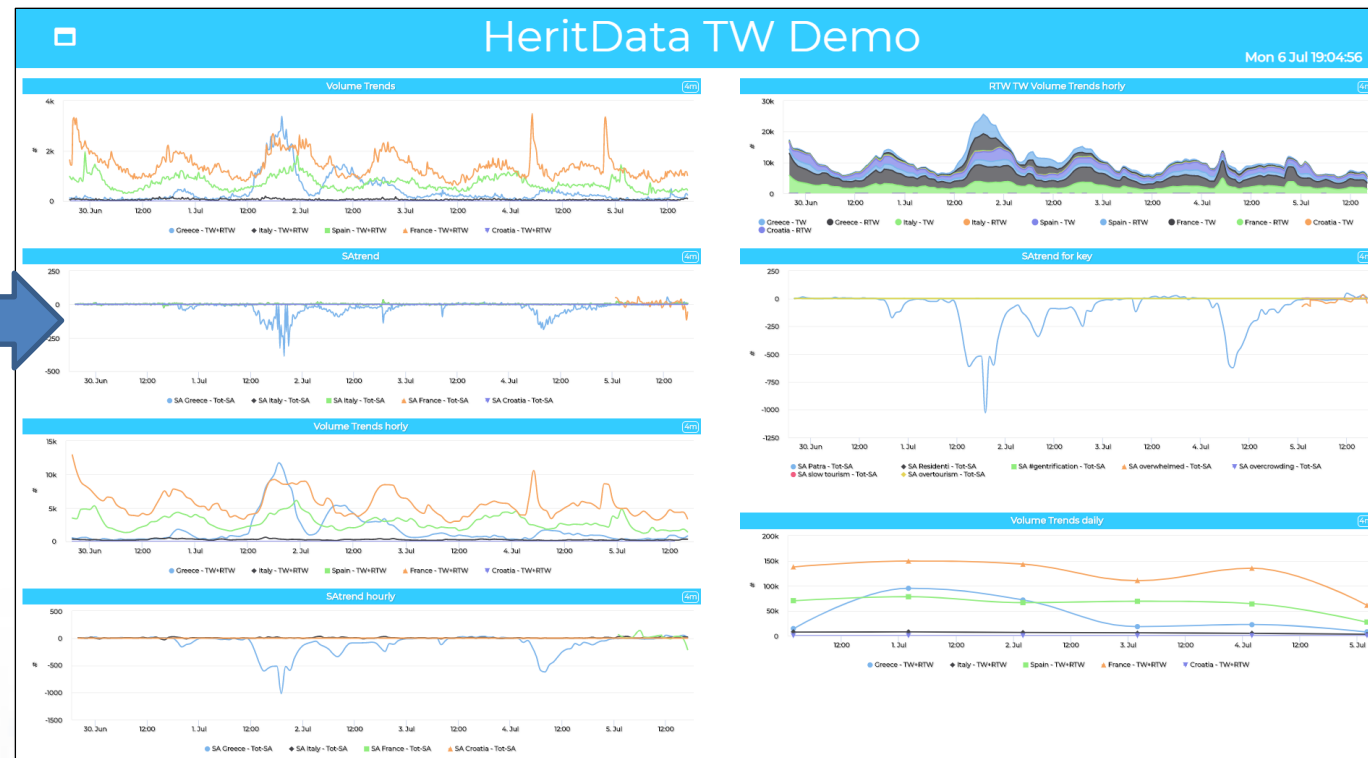
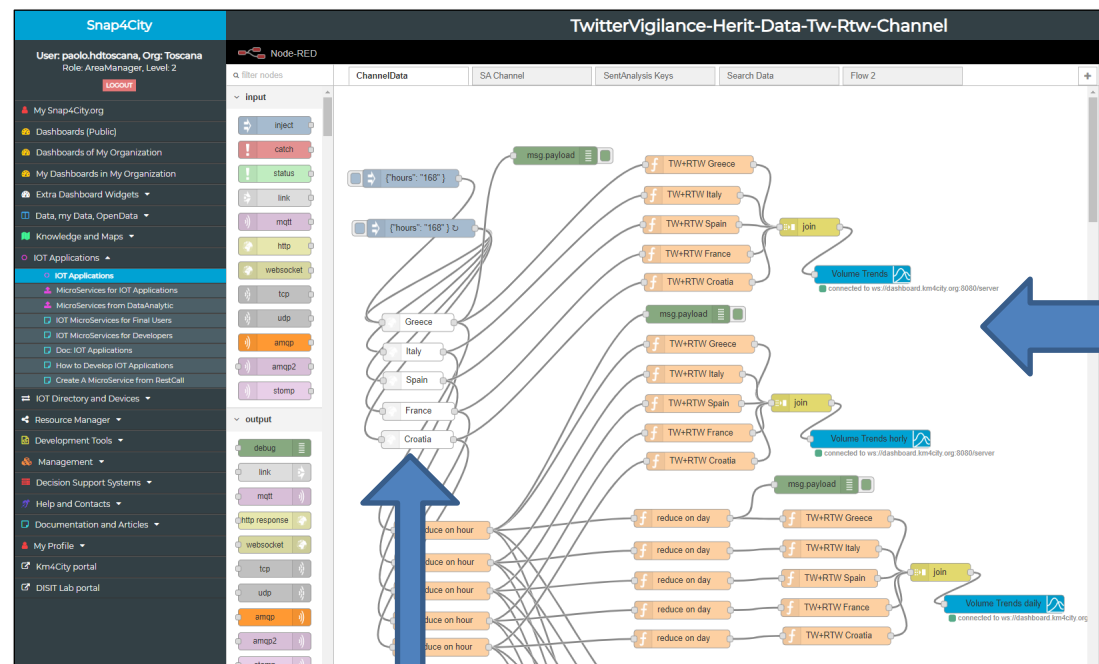


Resource Manager

Generating IOT App  
With Dashboard



# Overview (paolo)



**Twitter Vigilance**

<https://www.snap4city.org/dashboardSmartCity/view/index.php?iddashboard=Mjc3NQ==>



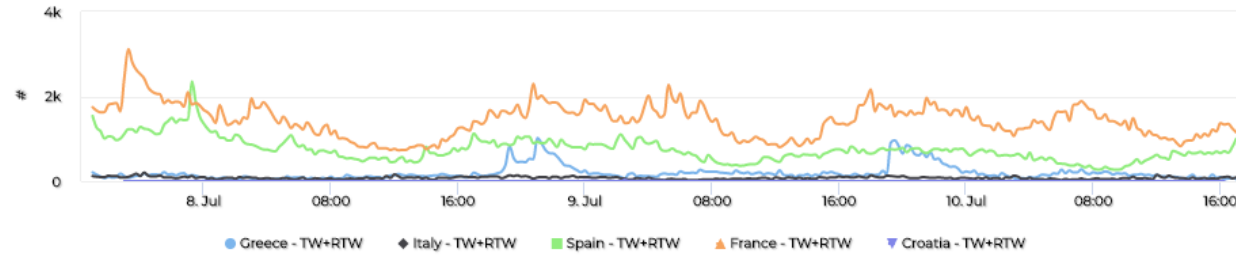


# HeritData TW Demo

Tue 14 Jul 16:54:43

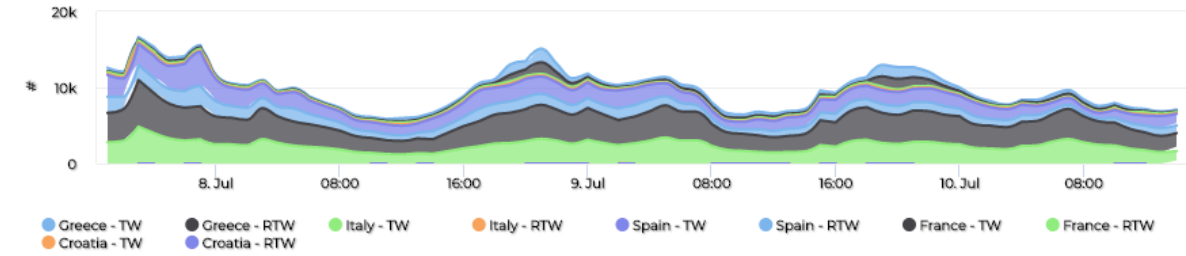
Volume Trends

4m



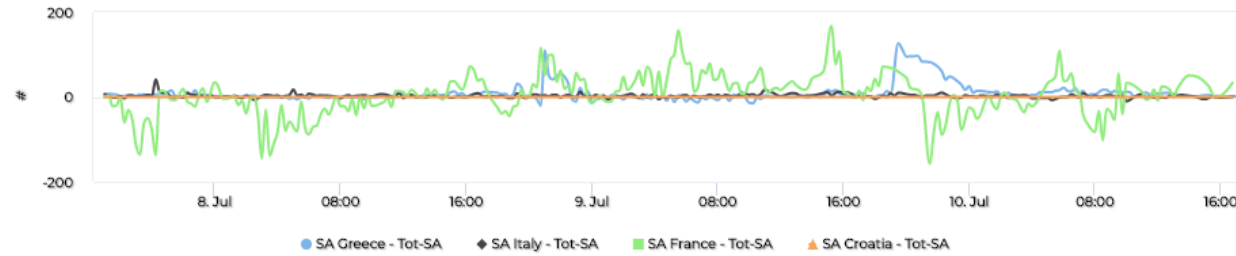
RTW TW Volume Trends hourly

4m



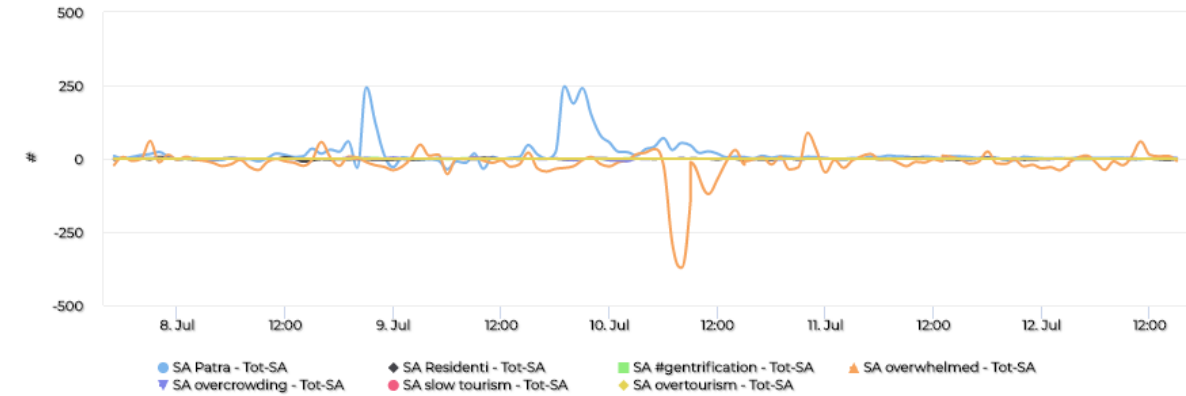
SAtrend

4m



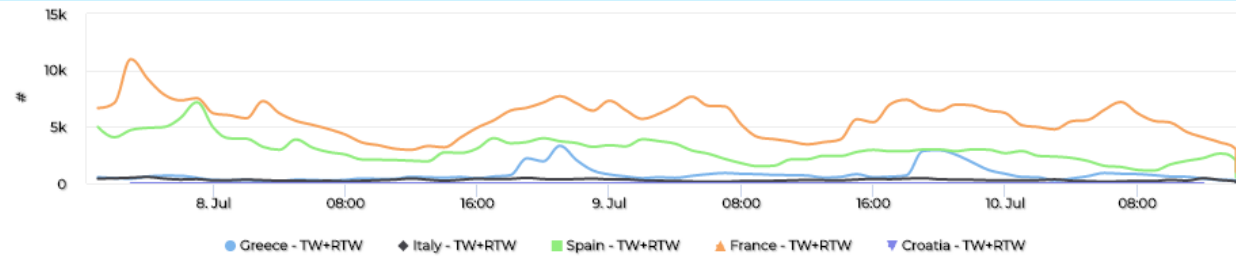
SAtrend for key

4m



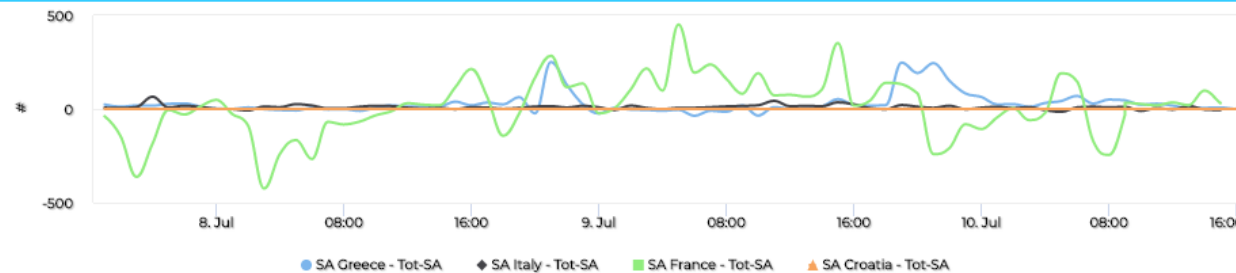
Volume Trends hourly

4m



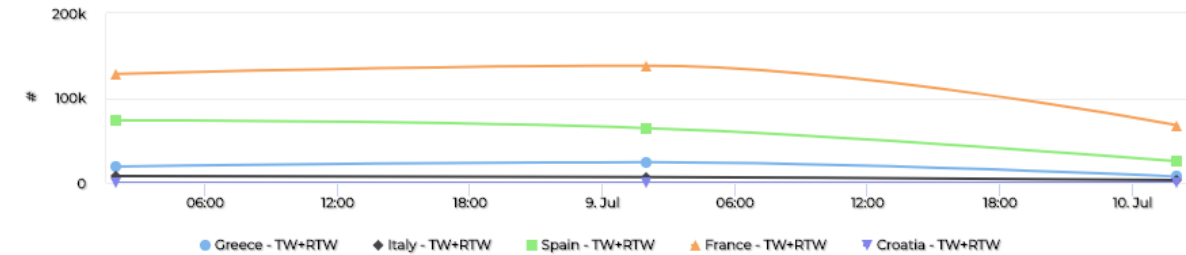
SAtrend hourly

4m



Volume Trends daily

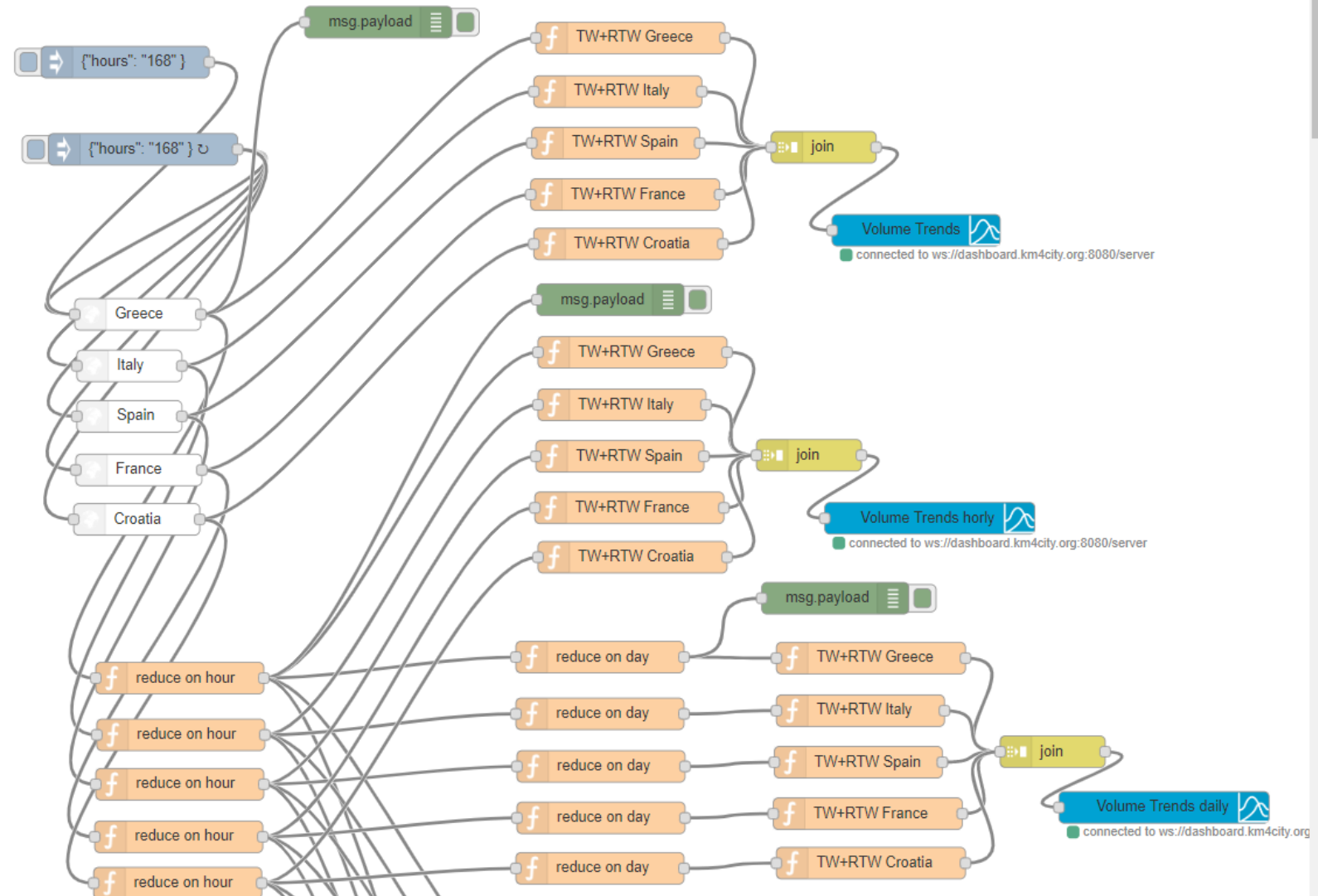
4m




Q filter nodes

Flow 2

4/10/2019



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  - data analytic on accessible data vs COVID lockdown

	Data Table	Data Availability	Data Description (source)	Data ingestion in Snap4City
Dubrovnik	✓	✓	<ul style="list-style-type: none"> <li>Camera position</li> <li>#people every minute</li> </ul> Licence: <b>Attribution-NonCommercial-NoDerivatives 4.0 International</b>	<b>Work in Progress:</b> <ul style="list-style-type: none"> <li>Camera Position</li> <li>Average #people every 15 minutes</li> </ul>
Florence	✓ (partial)	✓	<ul style="list-style-type: none"> <li>Camera position, #people every minute</li> <li>Many other data in place NOW !!!</li> </ul> Licence: <b>Open data</b>	<b>Work Finalised:</b> <ul style="list-style-type: none"> <li>Wifi position ✓</li> <li>Average #people every 15 minutes</li> </ul>
Pont Du Gard	✗	✗	Monitoring the Pont ?? ✗	✗
Mostar	✗	✗	✗	✗
Valencia	✓	Not yet ✗	<ul style="list-style-type: none"> <li>Pax counters position</li> <li>Number of persons in transit between two points</li> <li>Average time spent in sensor location</li> </ul> Licence: <b>private data</b>	✗
WestGreece	✓	Not yet ✗	Building/Site Capacity - Static + dynamic Data Licence: <b>Public as CC: to be decided the specific ones</b>	✗



## 1) Model creation

**IOT Device Models**

**Edit Model - Florence wifi average person**

General Info | IoT Broker | Static Attributes | Values

Florence wifi average person

Name  
Ok

wifiSensor

Device Type  
Ok

Comune Di Firenze

Producer  
Ok

Healthiness Criteria

Automatically generated  
Key Generation

Average Number of person for each wifi point in Florence

Description  
Ok

Sensor

Kind

900

Frequency

Healthiness Value

Edge-Gateway Type

Cancel Confirm

Model name: Florence wifi average person

**Edit Model - Florence wifi average person**

General Info | IoT Broker | Static Attributes | Values

orionToscana-UNIFI

ContextBroker

ngsi

Protocol

json

Format

Service/Tenant  
only ngsi v2/MultiService supports

ServicePath

**Edit Model - Florence wifi average person**

General Info | IoT Broker | Static Attributes | Values

Wifi (TourismService)

Subnature

Locality

FIRENZE

Value

Remove

Region

FI

Value

Remove

Add Attribute

**Edit Model - Florence wifi average person**

General Info | IoT Broker | Static Attributes | Values

meanPeople	float	People Count	Mean number of peo
Value Name	Data Type	Value Type	Value Unit
false	Refresh rate	900	Remove Value
Editable	Healthiness Criteria	Healthiness_Value	
dateObserved	datetime	Timestamp	timestamp in millisec
Value Name	Data Type	Value Type	Value Unit
false	Refresh rate	900	Remove Value
Editable	Healthiness Criteria	Healthiness_Value	

Add Value

Cancel Confirm

**Florence  
Wifi**

**Snap4City**

User: michela\_toscana, Org: Toscana  
Role: ToolAdmin, Level: 3

[Logout](#)

- My Snap4City.org
- Dashboards (Public)
- My Dashboards in All Org.
- Dashboards of My Organization
- My Dashboards in My Organization
- Extra Dashboard Widgets
- Notifier
- Data, my Data, OpenData
- Knowledge and Maps
- IOT Applications
  - MicroServices for IOT Applications
  - MicroServices from DataAnalytic
  - IOT MicroServices for Final Users
  - IOT MicroServices for Developers
  - Doc: IOT Applications
  - How to Develop IOT Applications
  - Create A MicroService from RestCall

**Florence\_wifi**

Node-RED Static flow

Flow 1

Input: inject, catch, status, link, mqtt, http, websocket, top, udp, amqp, amqp2, stomp

Output: debug, link, mqtt, http response, websocket, top

Flow 1 nodes: provider, timestamp, f: wifi\_location\_temp, f: All\_devices\_cleaned, delay 5s, /data/firenze\_wifi, split, every device, iotdirectory-new-device-from-model, msg

iotdirectory-new-device-from-model node properties:

- deviceName: deviceName
- latitude: Latitude
- longitude: Longitude
- k1: 42a68bf-6114-4c0b-84fb-21238e709ef7
- k2: 564cf1f4-3bd4-4acc-be2d-9ea940c24ea1
- Model: Florence wifi average person

**Florence Wifi**

## 2) IoT Devices Creation

BLOCK: 'IoTDirectory-new-device-from-model'  
Model name: Florence wifi average person

**Florence\_wifi**

Edit iotdirectory-new-device-from-model node

node properties:

- deviceName: deviceName
- latitude: Latitude
- longitude: Longitude
- k1: 42a68bf-6114-4c0b-84fb-21238e709ef7
- k2: 564cf1f4-3bd4-4acc-be2d-9ea940c24ea1
- Model: Florence wifi average person

Information:

It allows to create a device from model

Inputs:

A JSON with these parameters:

- deviceName: The name of the device you want to create
- latitude: latitude of a GPS position
- longitude: longitude of a GPS position
- k1: K1 and K2 are the keys necessary to read and write access to the device. They must be different from each other.
- k2: K1 and K2 are the keys necessary to read and write access to the device. They must be different from each other.
- model: The name of the model

Outputs:

Returns the k1 and k2 of device created

Details:

The node can receive a JSON with the parameters described in the Inputs section and with them generate the output JSON. If the values are not present in the input JSON, these are read by those in the configuration. If they are not present in either part, an error is generated for the necessary parameters.

3) Group Creation (more than 200 devices) -> put all the devices in the group and put them as 'public' (or they remain private)

Michela

Herit-Data, Snap4City (C), 15 July 2020

**Snap4City**

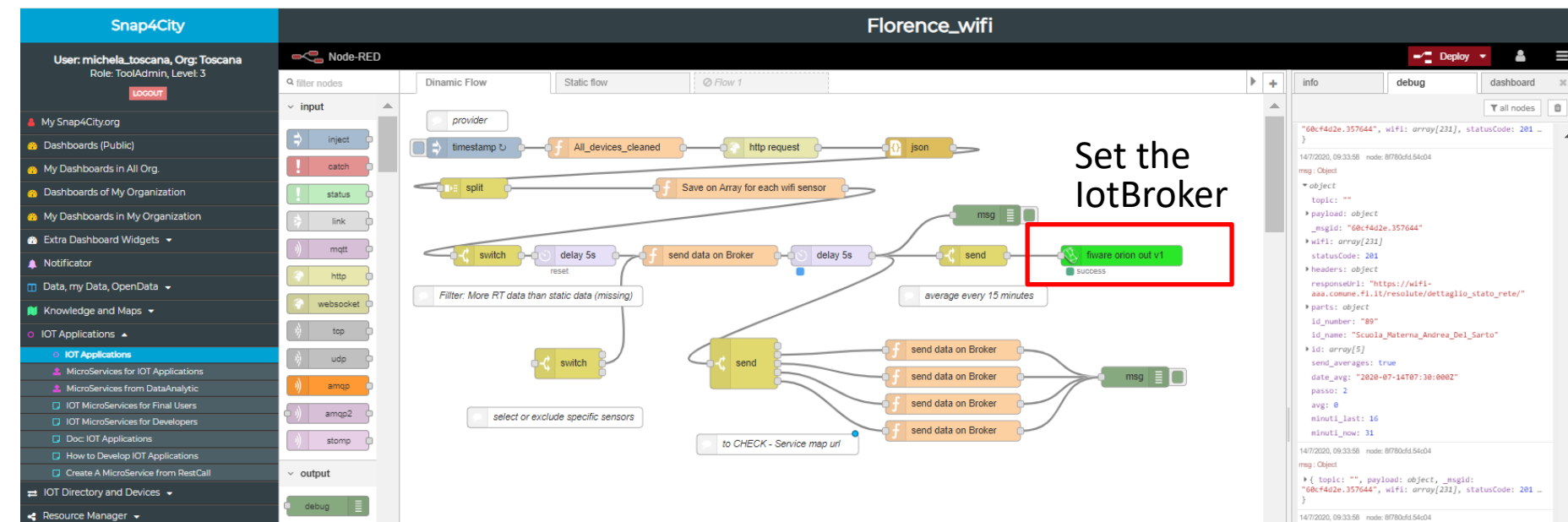
User: michela\_toscana, Org: Toscana  
Role: ToolAdmin, Level: 3

[Logout](#)

My Groups of Entities

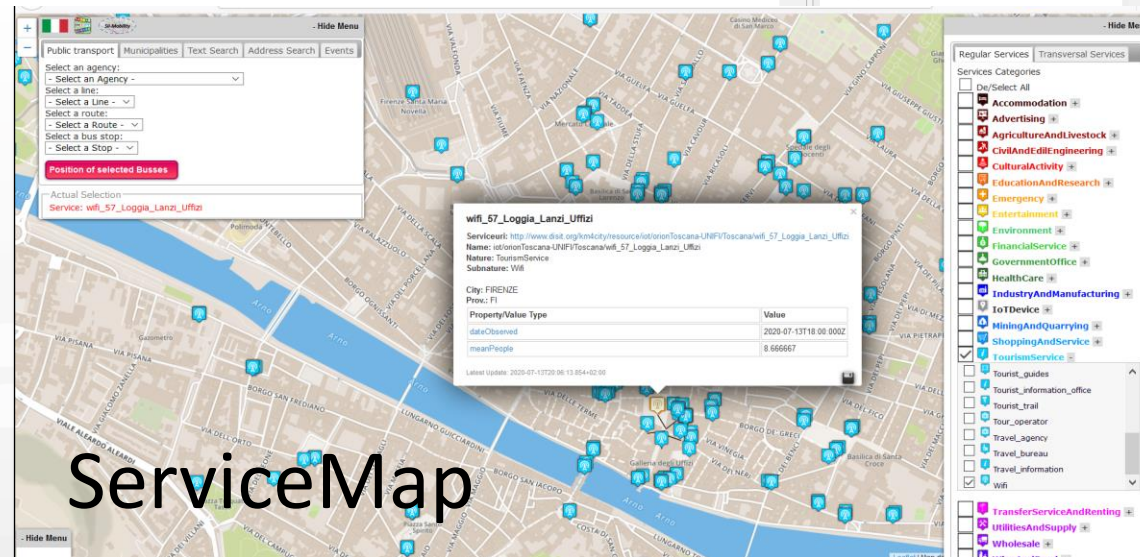
Device Group ID 20 Name Florence\_Wifi Description Wifi averages

No.	Username	Element ID	Element Type	Element Name	Added
104	michela_toscana	Toscana.orionToscana-UNIF: wifi_2_Parcheggio_Porta_al_Prato_Leopolda	IOT Device	wifi_2_Parcheggio_Porta_al_Prato_Leopolda	6/7/2020, 18:36:17
105	michela_toscana	Toscana.orionToscana-UNIF: wifi_6_Parcheggio_Santa_Maria_Novella	IOT Device	wifi_6_Parcheggio_Santa_Maria_Novella	6/7/2020, 18:36:17
106	michela_toscana	Toscana.orionToscana-UNIF: wifi_5_Parcheggio_Beccaria	IOT Device	wifi_5_Parcheggio_Beccaria	6/7/2020, 18:36:17
107	michela_toscana	Toscana.orionToscana-UNIF: wifi_9_Ospedale_Pediatrico_Meyer	IOT Device	wifi_9_Ospedale_Pediatrico_Meyer	6/7/2020, 18:36:17
108	michela_toscana	Toscana.orionToscana-UNIF: wifi_0_Parcheggio_Europa	IOT Device	wifi_0_Parcheggio_Europa	6/7/2020, 18:36:17
109	michela_toscana	Toscana.orionToscana-UNIF: wifi_4_Parcheggio_San_Lorenzo_Mercato_Centrale	IOT Device	wifi_4_Parcheggio_San_Lorenzo_Mercato_Centrale	6/7/2020, 18:36:17
110	michela_toscana	Toscana.orionToscana-UNIF: wifi_7_Parcheggio_S_Ambrogio	IOT Device	wifi_7_Parcheggio_S_Ambrogio	6/7/2020, 18:36:17



4) Send RT data to the IoTDevices

5) Verify RT Data via Snap4City API or via ServiceMap

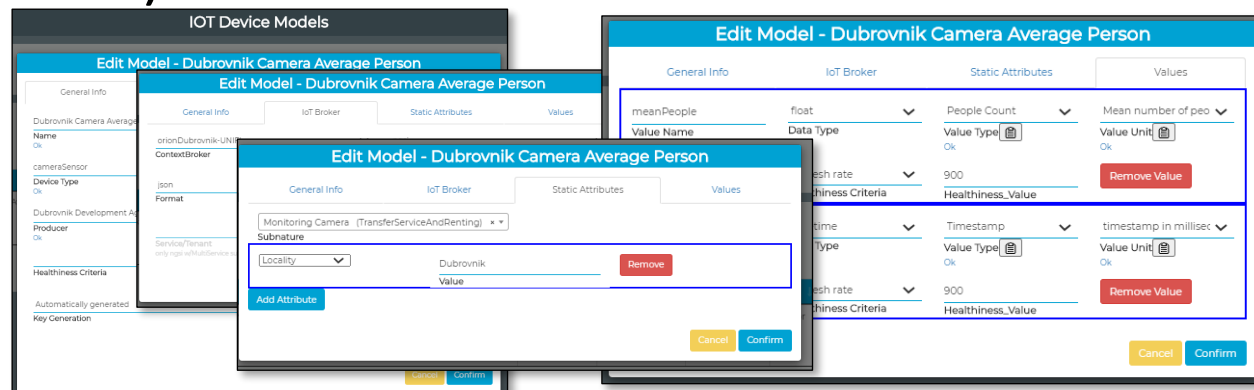


## Snap4City API

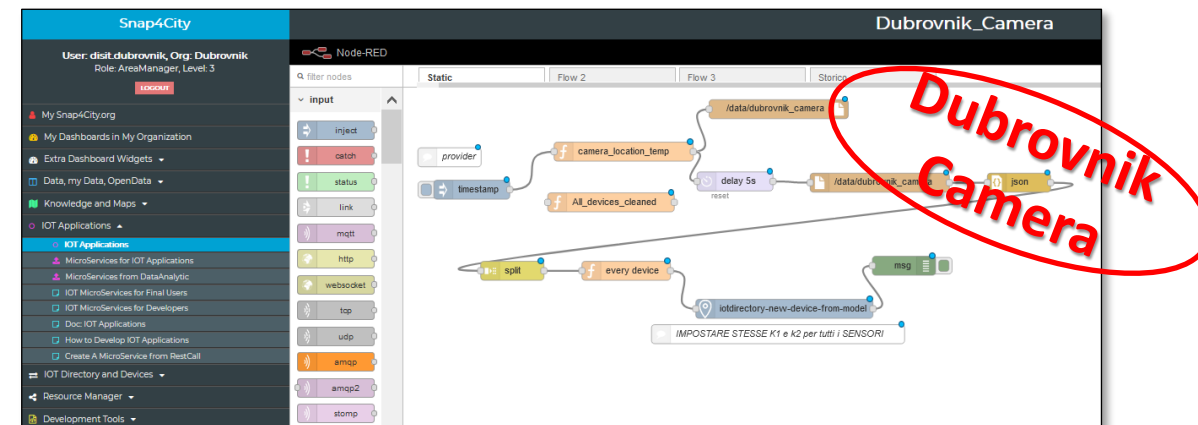
JSON	Dati non elaborati	Header
Salva	Copia	Comprimi tutto
Espandi tutto		
Filtro JSON		
<b>Service:</b> type: "FeatureCollection" features: [...]		
<b>realtime:</b> head: vars: 0: "measuredTime" 1: "dateObserved" 2: "meanPeople" results: bindings: 0: measuredTime: "2020-07-13T19:49:26.780+02:00" dateObserved: "2020-07-13T17:45:00Z" meanPeople: "0" 1: measuredTime: "2020-07-13T19:40:43.168+02:00" dateObserved: "2020-07-13T17:30:00Z" meanPeople: "0" 2: measuredTime: "2020-07-13T19:20:31.181+02:00" dateObserved: "2020-07-13T17:15:00Z" meanPeople: "0" 3: measuredTime: "2020-07-13T19:01:21.564+02:00" dateObserved: "2020-07-13T17:00:00Z" meanPeople: "0" 4: measuredTime: "2020-07-13T19:01:21.564+02:00" dateObserved: "2020-07-13T17:00:00Z" meanPeople: "0"		



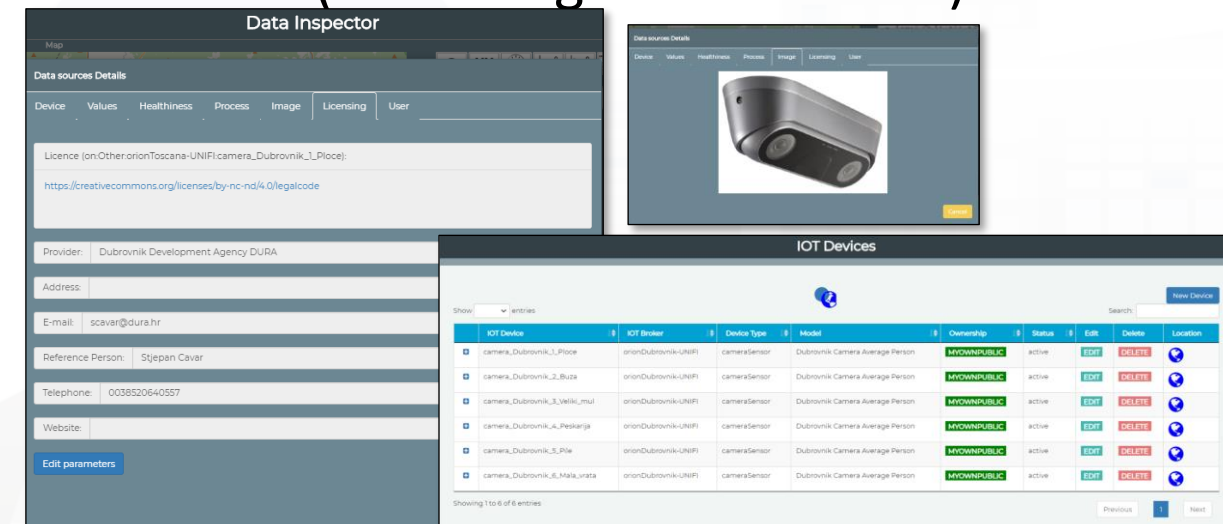
## 1) IoTModel



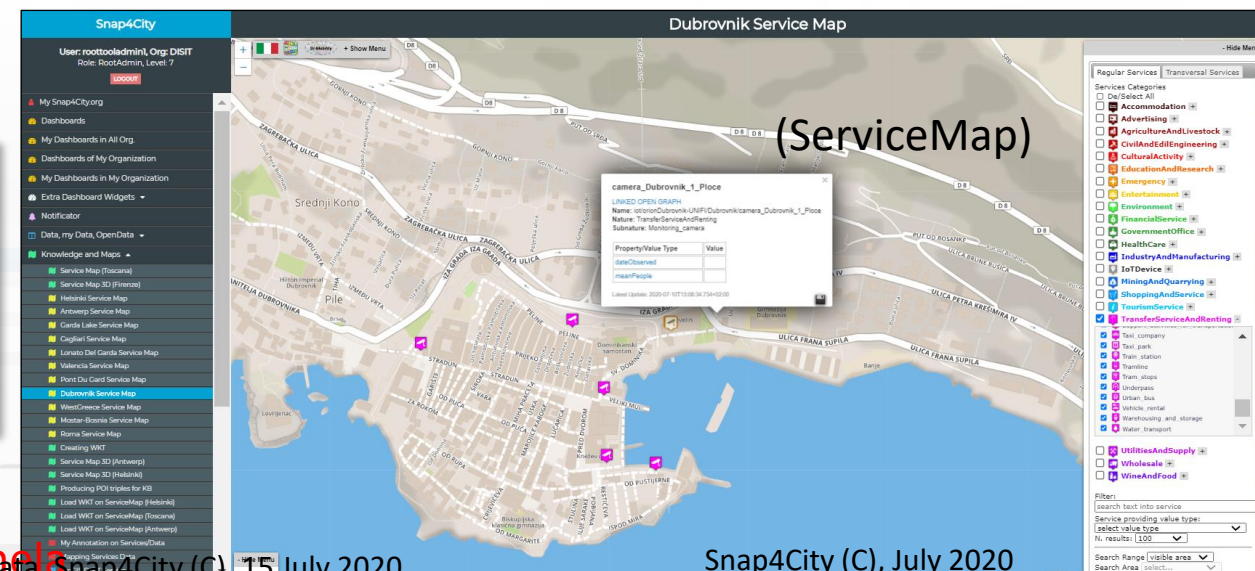
## 2) Static Flow to create IoTDevices



## 3) Add the license and Make Public the IoTDevices (according to the license)



## 3) Search for the Cameras on Map



## 5) Working on Dynamic Flow to save Average #people every 15 minutes for each IoTDevice



- 14:00-15:00
  - overview DISIT activity (herit data organization on snap4city)
  - demo of Twitter Vigilance
  - comparison with former tool of Almaviva
  - Integration of Twitter Vigilance with Dashboard and IOT App in Node-red
- 15:00-16:00
  - Acquired data from the cities
- 16:00-16:30
  - exploitation of data from IOT App and Dashboard
  - data analytic on accessible data vs COVID lockdown

	KB ready	Streets	Civic Numbers	IOT App ready	Big data store	MyKPI	Dashboards ready	Active loaded data
Dubrovnik	✓	✗	✗	possible	✓	✓	1 on TV	Twitter Vigilance, TV Cameras
Florence	✓	✓	✓	possible	✓	✓	1 on TV	Twitter Vigilance, WiFi people, traffic, POI, some Apps, parking, etc.
Pont Du Gard	✓	✓	✗	possible	✓	✓	1 on TV	Twitter Vigilance
Mostar	✓	✗	✗	possible	✓	✓	1 on TV	Twitter Vigilance
Valencia	✓	✓	✗	possible	✓	✓	1 on TV	Twitter Vigilance
WestGreece	✓	✗	✗	possible	✓	✓	1 on TV	Twitter Vigilance

## Presently

- Each Organization can access to its own data:
  - Firenze / Tuscany has a large set of data almost all accessible
    - KB, smart city API and MicroServices and thus IOT Apps
- Do multiple users and emails for accessing to multiple Organizations
- Ask for the UserName and Password for accessing to Twitter Vigilance Herit-Data Real Time

# Snap4City: Builder of Sentient Cities Solutions

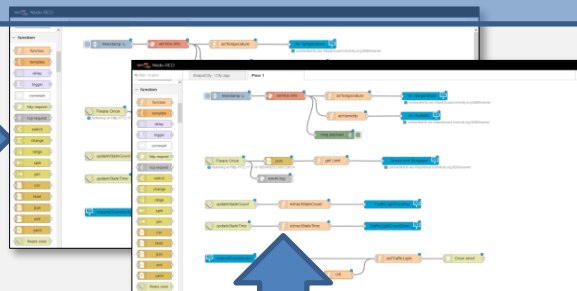
Dashboards with data driven IOT Applications enforcing intelligence

IOT and data World



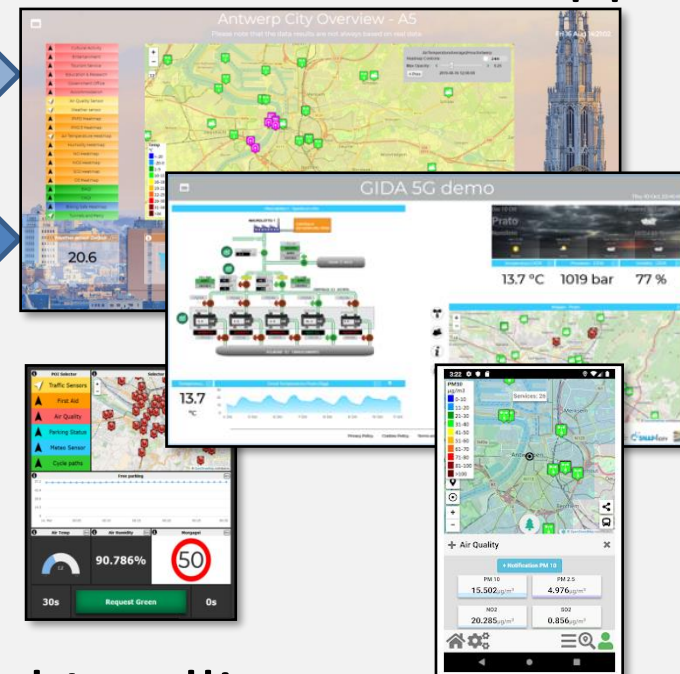
My IOT Devices

IOT Applications



Big Data Analytics, Artificial Intelligence

Dashboards and Apps



- 14:00-15:00
  - overview DISIT activity (herit data organization on snap4city)
  - demo of Twitter Vigilance
  - comparison with former tool of Almaviva
  - Integration of Twitter Vigilance with Dashboard and IOT App in Node-red
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  - data analytic on accessible data vs COVID lockdown

Irene & Paolo



# Data Analytics

- Daily data related to 85 research keys (features) associated to Covid lockdown periods in Italy, Spain, France, Greece and Croatia.
- Period from February 1<sup>st</sup> to May 30<sup>th</sup>, 2020.
  - marginally different Lockdown periods

## Goal

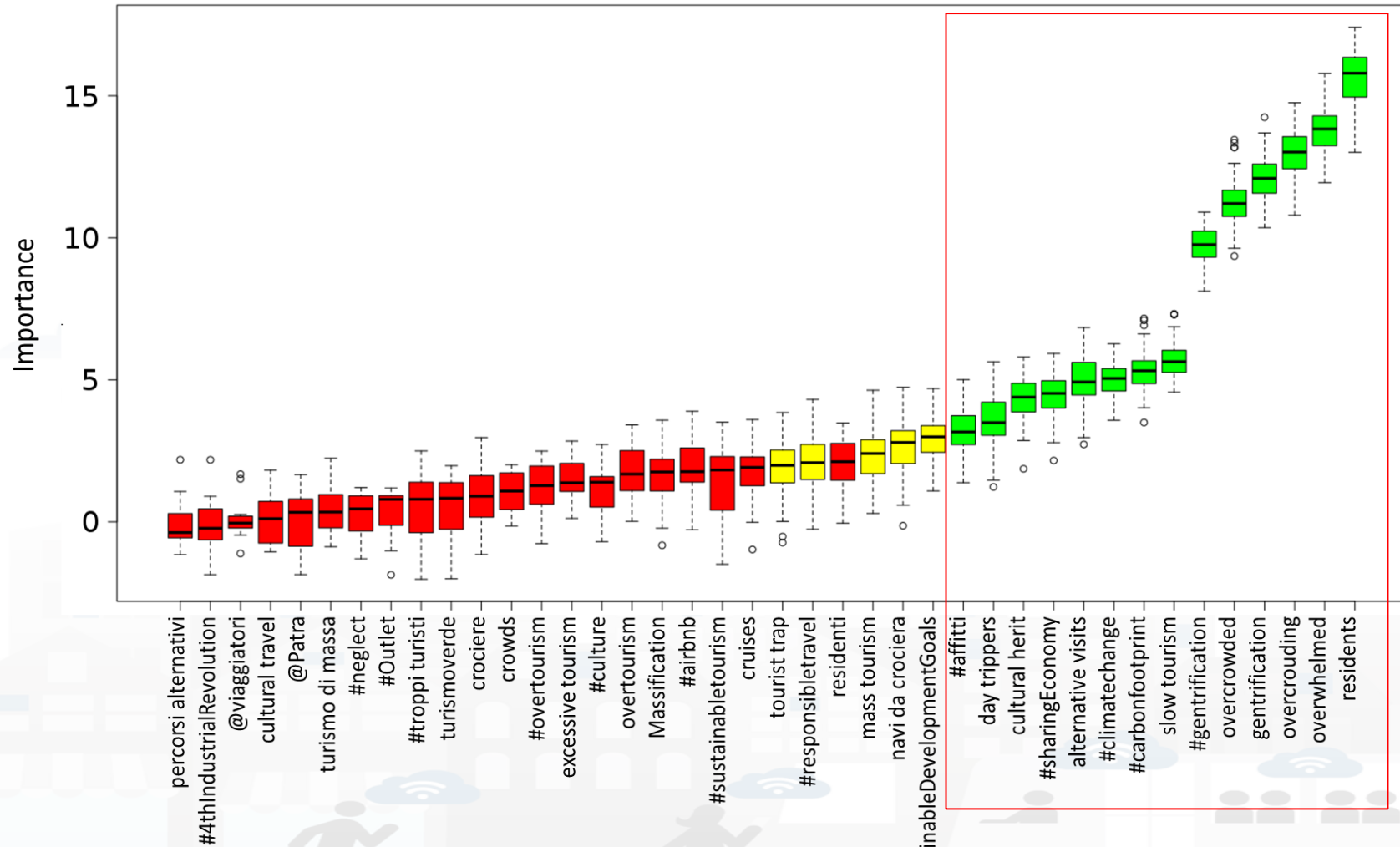
Study of the lockdown impact:

investigate about significant search keys related to the lockdown period.

# Research keys

- |                                |                          |                    |                         |                        |                     |
|--------------------------------|--------------------------|--------------------|-------------------------|------------------------|---------------------|
| • #4thIndustrialRevolution     | • #Πάτρα                 | • προστασία        | • overcrowded           | • #overtourism         | • residenti         |
| • #affitti                     | • #τουρισμός             | • συνωστισμός      | • papeleras             | • @viaggiatori         | • residents         |
| • #airbnb                      | • cultural herit         | • φθορά            | • petada                | • alternative visits   | • slow tourism      |
| • #carbonfootprint             | • cultural_travel        | • #Amunt           | • suciedad              | • crociere             | • tourisme de masse |
| • #climatechange               | • excessive tourism      | • #enfamilia       | • valencia              | • crowds               | • tourist trap      |
| • #gentrification              | • overtourism            | • #planesconniños  | • Dubrovnik City Walls  | • cruises              | • trop de touristes |
| • #Outlet                      | • patra                  | • #turismofamiliar | • Dubrovnik great       | • day trippers         | • troppi turisti    |
| • #overtourism                 | • west greece            | • #Valencia        | • Dubrovnik holiday     | • gentrification       | • turismo de masa   |
| • #responsibletravel           | • αρχαία ολυμπία         | • #VisitSpain      | • Dubrovnik Old City    | • mass tourism         | • turismo di massa  |
| • #sharingEconomy              | • δυτική Ελλάδα          | • congestión       | • Dubrovnik summer      | • navi da crociera     | • turistificación   |
| • #SustainableDevelopmentGoals | • κατάκολο               | • contaminación    | • Old City crowds       | • overcrowding         |                     |
| • #sustainabletourism          | • μνημεία                | • inseguridad      | • Old Town crowds       | • overtourism          |                     |
| • #turismoverde                | • παραμέληση             | • masificación     | • Stari grad            | • overwhelmed          |                     |
| • #culture                     | • πάτρα                  | • masificado       | • #gentrification       | • paquebots            |                     |
| • #neglect                     | • πολιτιστική κληρονομιά | • massification    | • #LasCasasNoSonHoteles | • percorsi alternativi |                     |

# Relevant Feature Selection - Italy



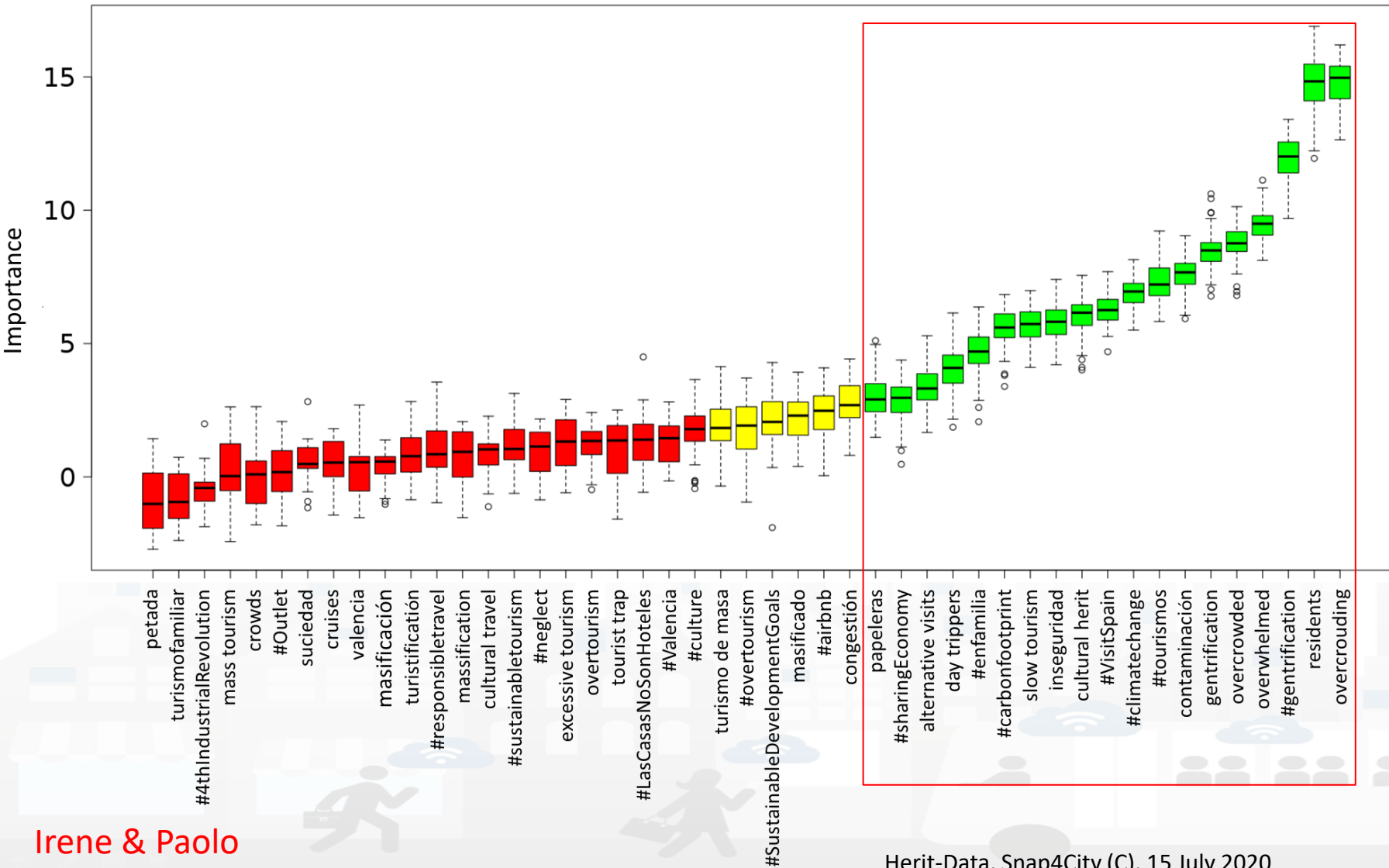
Random Forest Model  
Results exploiting the  
*confirmed* keys:

- Accuracy = 0.90
- Kappa index = 0.81

# Logistic Model Results - Italy

- The statistically significant research keys (with a 95% confidence level) are:
  - *Residents (+)*
  - *#gentrification (-)*
  - *Overwhelmed (+)*
  - *Overcrowded (+)*
  - *day trippers (-)*
- With the lockdown in Italy the **number of research keys**
  - *residents*, *overwhelmed* and *overcrowded* increased
  - *day trippers* and *#gentrification* decreased.
- The Pseudo R-squared of the model is **0.79**

# Relevant Feature Selection - Spain



Random Forest Model  
Results exploiting the  
*confirmed* keys:

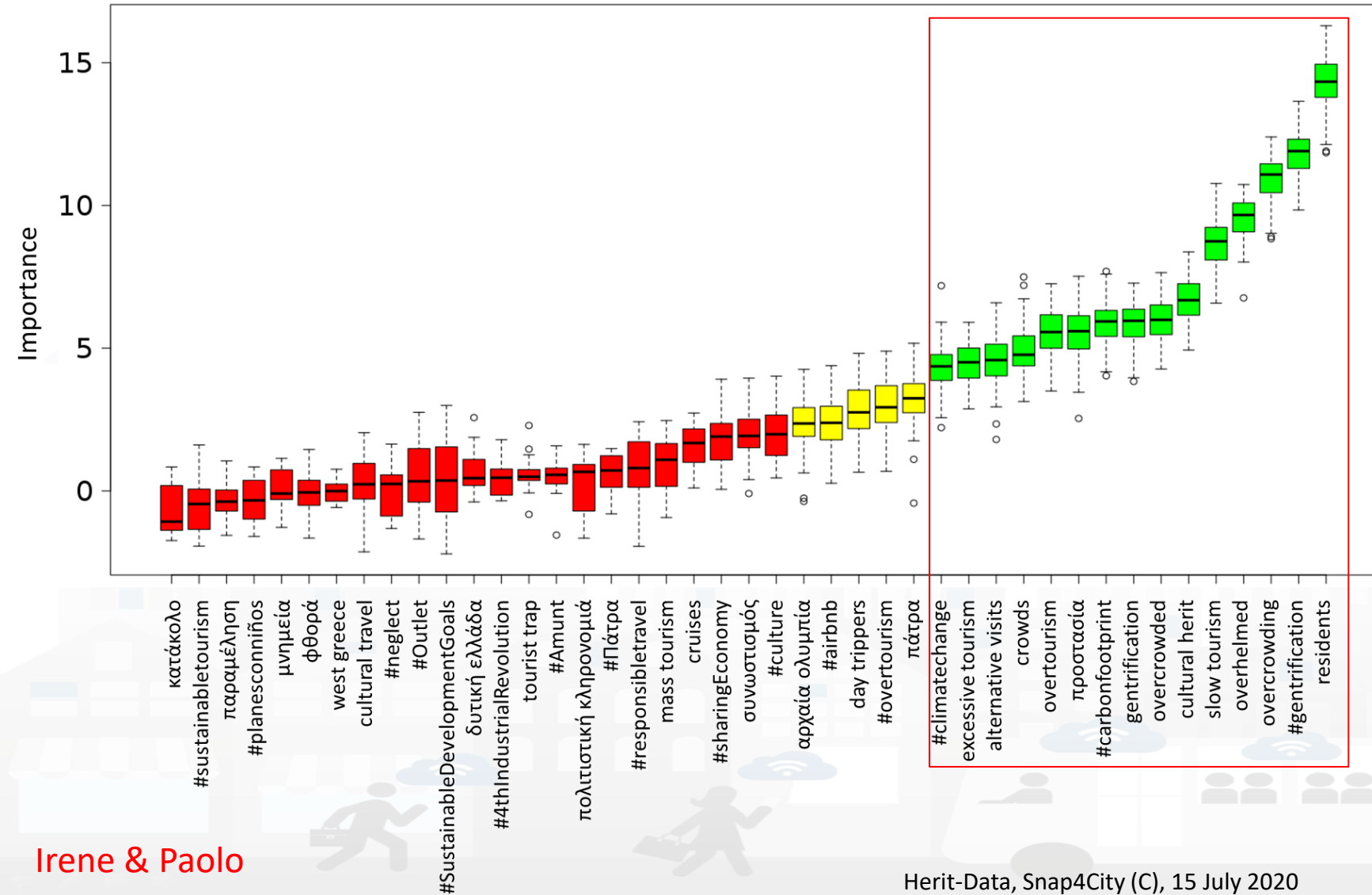
- Accuracy = 0.937
- Kappa index = 0.876



## Logistic Model Results - Spain

- The statistically significant research keys (with a 95% confidence level) are:
  - *#tourismos*
  - *#gentrification*
  - *#VisitSpain*
  - *#enfamilia*
  - *residents*
  - *day trippers*
  - *inseguridad*
  - *overcrowding*
  - *congestion*
- With the lockdown in Spain the **number of research keys**
  - *#enfamilia*, *residents* and *overcrowding* **increased**
  - *#tourismos*, *#VisitSpain*, *#gentrification*, *day trippers*, *inseguridad*, *congestion* **decreased**.
- The Pseudo R-squared of the model is **0.79**

# Relevant Feature Selection - Greece



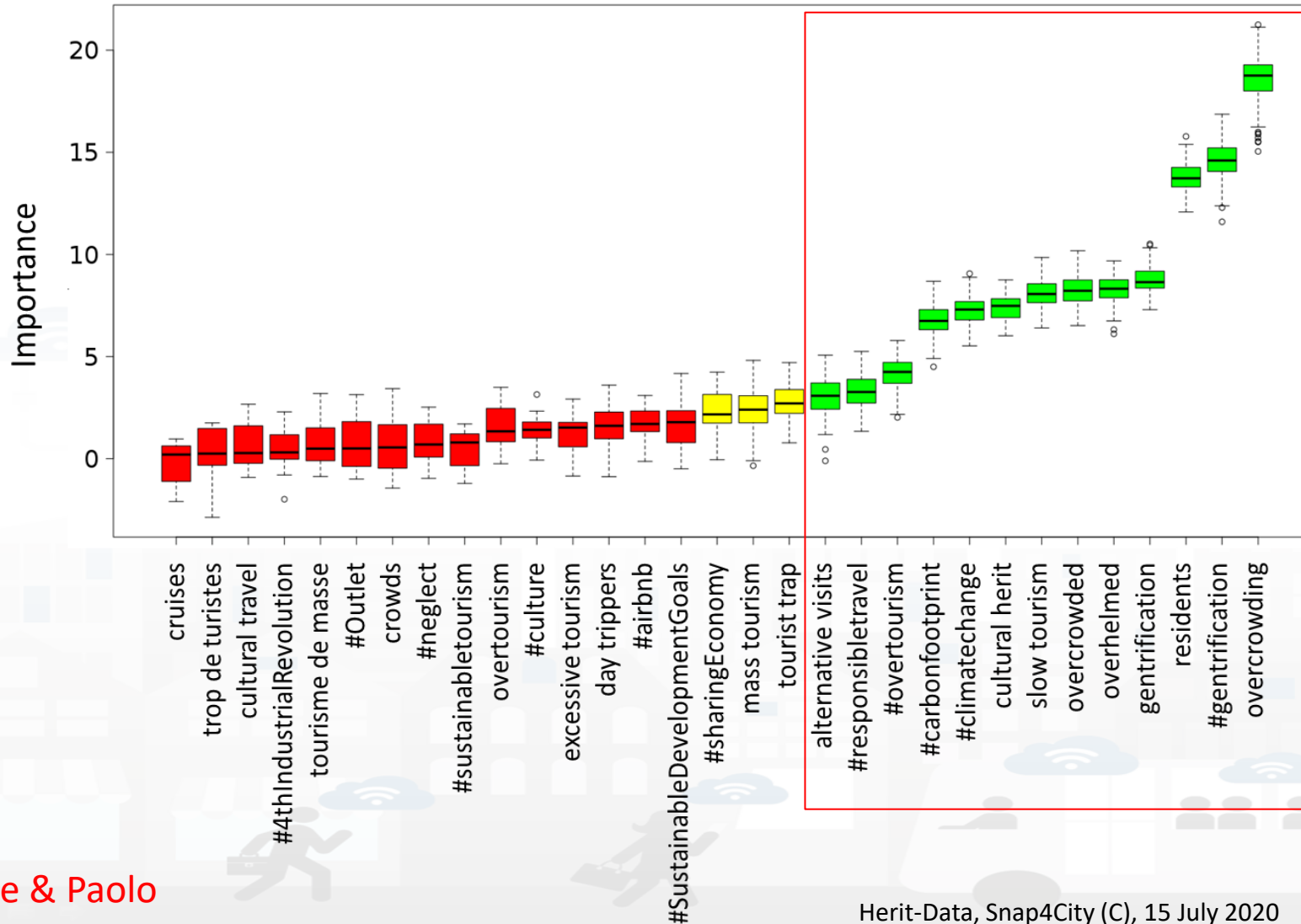
Random Forest Model  
Results exploiting the  
*confirmed* keys:

- Accuracy = 0.898
- Kappa index = 0.773

# Logistic Model Results - Greece

- The statistically significant research keys (with a 95% confidence level) are:
  - *residents*
  - *#gentrification*
  - *overwhelmed*
- With the lockdown in Greece the **number of research keys**
  - *residents* and *overwhelmed* increased
  - *#gentrification* decreased.
- The Pseudo R-squared of the model is **0.52**

# Relevant Feature Selection - France



Random Forest Model  
Results exploiting the  
*confirmed* keys:

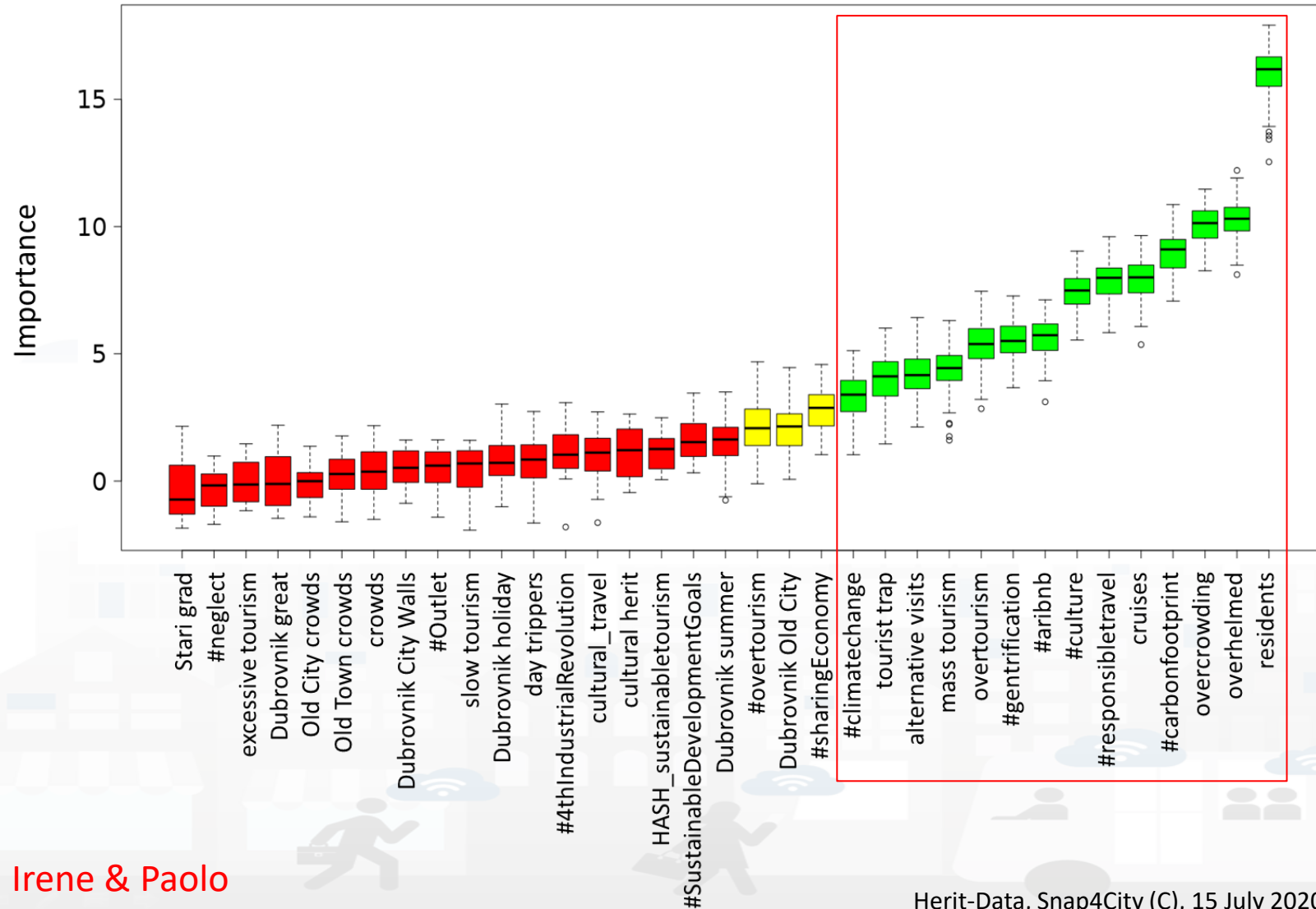
- Accuracy = 0.928
- Kappa index = 0.856

# Logistic Model Results - France

- The statistically significant research keys (with a 95% confidence level) are:
  - *#gentrification*
  - *#responsibletravel*
  - *overcrowding*
- With the lockdown in France the **number of research**
  - *overcrowding* increased
  - *#responsibletravel* and *#gentrification* decreased.
- The Pseudo R-squared of the model is **0.78**



# Relevant Feature Selection - Croatia



Random Forest Model  
Results exploiting the  
*confirmed* keys:

- Accuracy = 0.945
- Kappa index = 0.847

# Logistic Model Results - Croatia

- The statistically significant research keys (with a 95% confidence level) are:

- *#responsibletravel*
- *#gentrification*
- *#airbnb*
- *residents*
- *overtourism*
- *overcrowding*

1. With the lockdown in Croatia the **number of research keys**

1. *#airbnb*, *residents* and *overcrowding* **increased**
  2. *overtourism*, *#responsibletravel*, and *#gentrification* **decreased**.

2. The Pseudo R-squared of the model is **0.55**