

A system for estimating crowd density based on Wi-Fi probe request frames

Department of Information Engineering and Computer Science

Supervisors Student

Fabrizio Granelli Samuel Bortolin

Daniele Miorandi

Academic year 2019/2020

About this Project



External internship at U-Hopper:

- → Big Data Analytics
- → Business Intelligence
 - → IoT solutions
- → Artificial Intelligence solutions



Problem Statement



Badly handled demand in company that provides services to physical customers can lead to overcrowding and inefficiency of the services

→ Inefficient and badly organized service leads to higher costs

→ It is important to detecting and monitoring crowds to reduce the risk of COVID-19 spreading during this global pandemic period

State of the Art



→ Analysis of different methods for estimating crowd density

Infrared sensors, LSE, treadle switch-based systems, Video methods, Audio methods, Wi-Fi, Bluetooth, BLE, LTE, Radar, RFID approaches

→ Many fields of application and several implementations

Why Wi-Fi Solution?



High diffusion of Wi-Fi devices

Low-cost implementation

Real-time data transmission

Customer privacy ensured (GDPR compliance)



Standard 802.11 → Management

frames → Probe request frames

Research Statement



Is it possible to continuously estimate the density of the crowd in a place of interest based on the Wi-Fi probe request frames?



Achievements

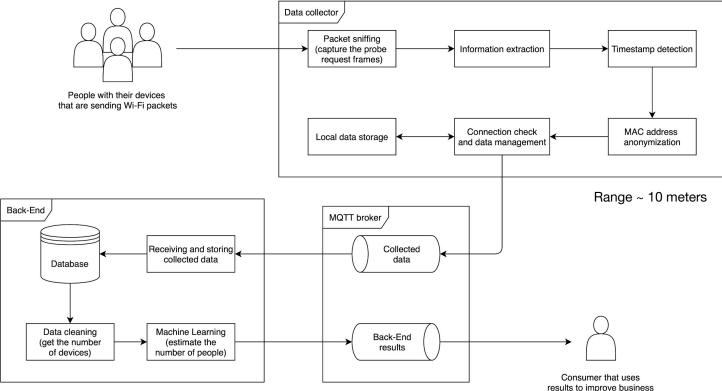


Designed and implemented a system for this problem that could work in several context

Tested the system in a Cafe and collected
4 weeks of data and manually-annotated
ground truth (the actual number of people)

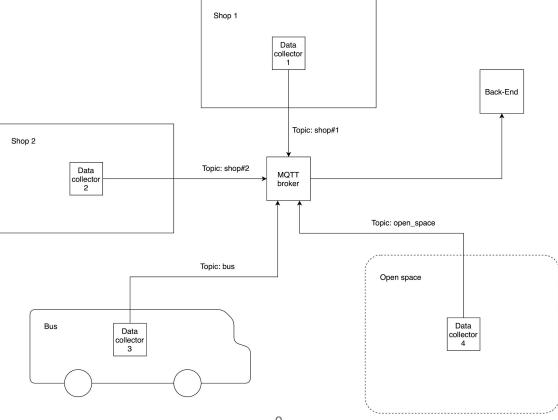
System Architecture





Scalable Architecture





Presence of Devices

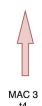


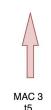
Probe request frames detections

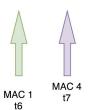












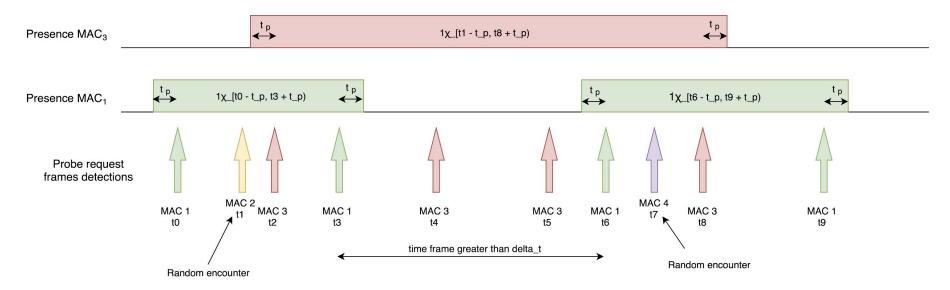


MAC 3

Samuel Bortolin

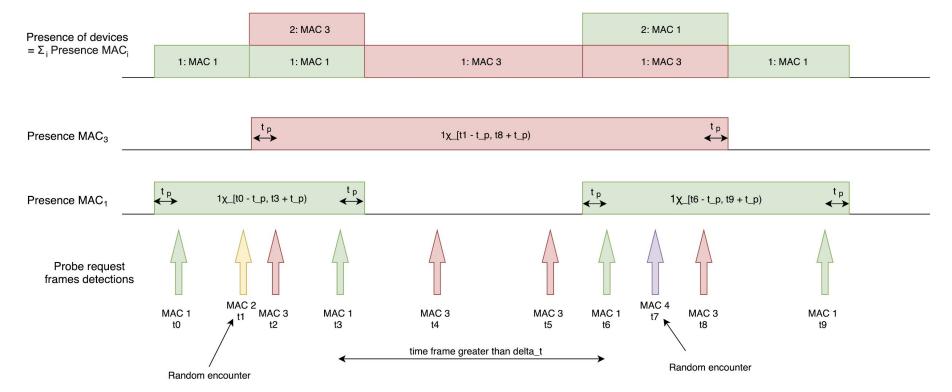
Presence of Devices





Presence of Devices





Feasibility Test at Home



Tests at home before validation \rightarrow 3 days of data collection

65928 probe request frames captured → 12 home devices detected

2 main range of RSSI -71 ÷ -91 dBm not in the kitchen, -35 ÷ -69 dBm in the kitchen

→ Feasibility of the method for detecting devices in the area

System Validation



→ Raspberry Pi in a Cafe where I annotate manually the ground truth

→ Eclipse Mosquitto Broker MQTT of U-Hopper on their cloud infrastructure

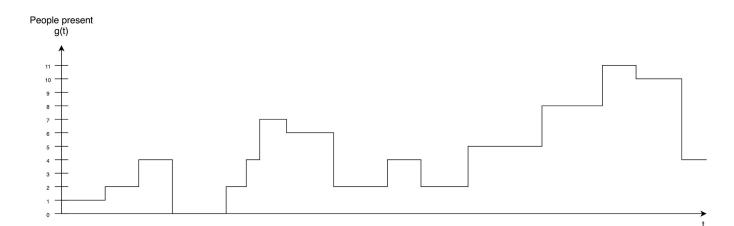
- → MQTT receiver and MongoDB on U-Hopper cloud infrastructure
- → Analyzer and Estimator on my pc to use on the data + annotated ground truth to test accuracy and reliability of the proposed system

Start Fit devices trend analysis Get trend and seasonality of the number of devices Split the dataset into train and test set Perform trend analysis Find the optimal polynomial approximation Save the optimal value for the trained time slot of the trained day in the local storage End

Ground Truth Annotation

15



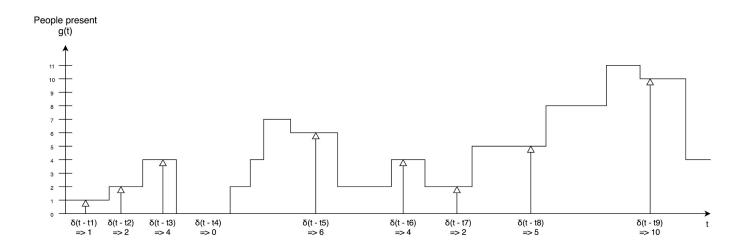


Samuel Bortolin

Start Fit devices trend analysis Get trend and seasonality of the number of devices Split the dataset into train and test set Perform trend analysis Find the optimal polynomial approximation Save the optimal value for the trained time slot of the trained day in the local storage End

Ground Truth Annotation

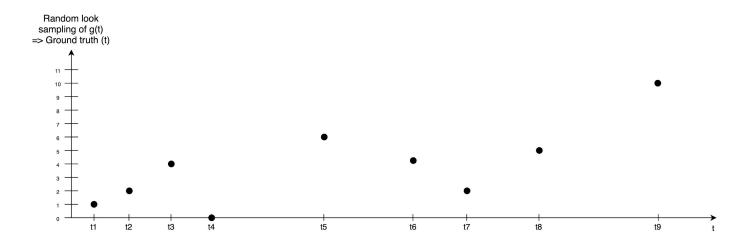




Start Fit devices trend analysis Get trend and seasonality of the number of devices Split the dataset into train and test set Perform trend analysis Find the optimal polynomial approximation Save the optimal value for the trained time slot of the trained day in the local storage End

Ground Truth Annotation





Test Results at Cafe



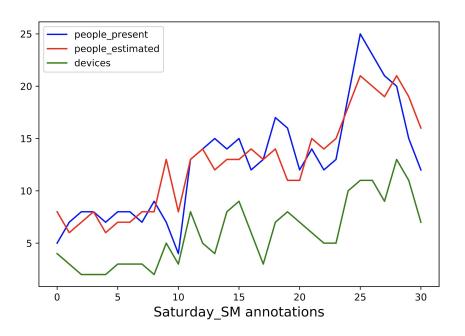
→ 4 weeks of data collection (24 days)

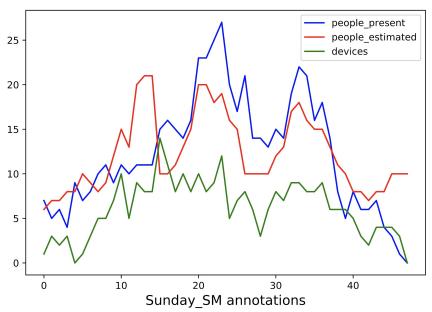
→ 1270 manual annotation of ground truth

- → 861979 probe request frames captured (~ 560 MB of data)
 - → 38771 unique MAC addresses detected

Test Results at Cafe

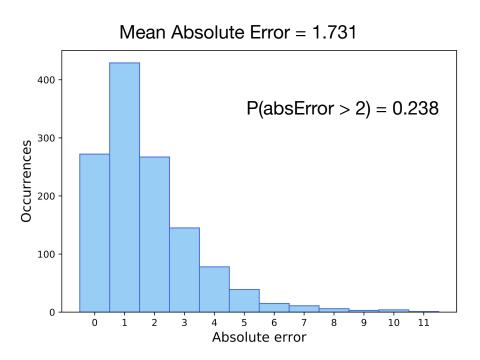


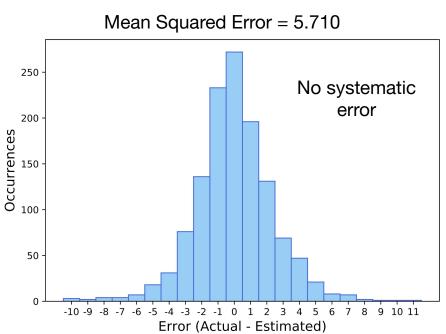




Error Distribution

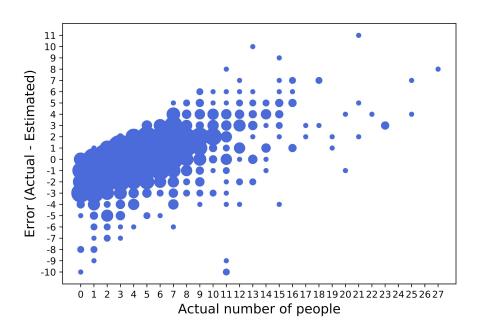


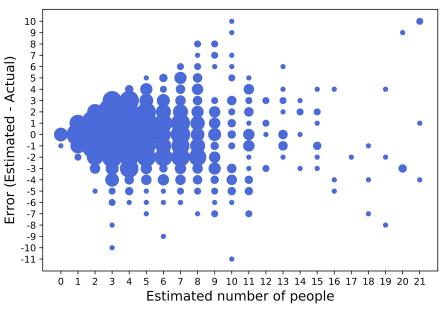




Scatter Plot of the Error







Summary



→ It is possible to continuously estimate the density of the crowd in a place of interest based on the Wi-Fi probe request frames

→ Designed and implemented a reliable system to do that

→ Tested the system in a Cafe and collected 4 weeks of data and manually-annotated ground truth with an overall MAE of 1.731

Future Works



→ Real-time execution

- → Test of the system in different contexts
 - → Extension to multiple data collectors

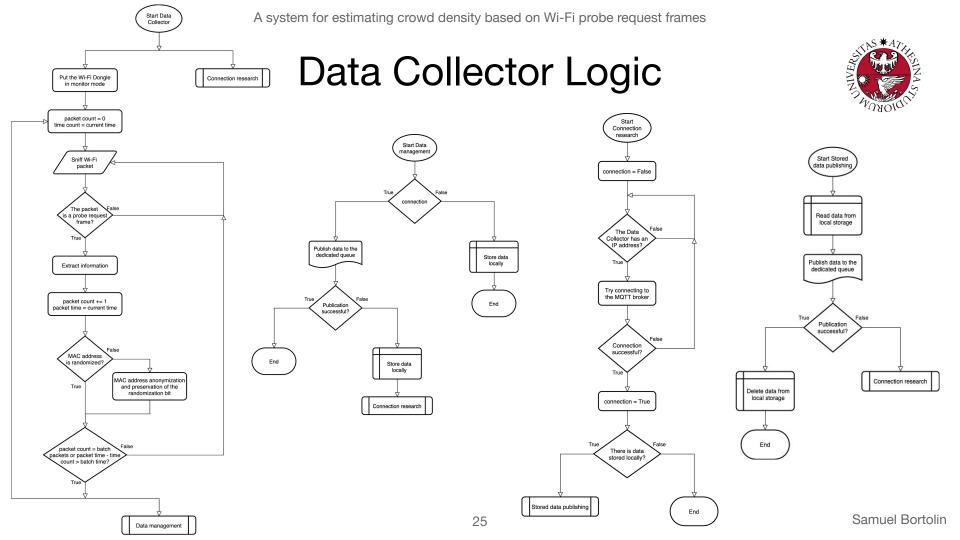
→ Improvement of the Machine Learning model



A system for estimating crowd density based on Wi-Fi probe request frames

Thank you for your attention

Samuel Bortolin



Start Back-End Subscribe to the collected data queue collected data Input manager, add corollary information Store data in the database The time slot is over? Get data from the database RSSI thresholding Remove random encounters Make blacklist Get the number of devices present for each timestamp Devices trend analysis

Back-End Logic

