

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

Department of Information Engineering and Computer Science

Supervisors Student

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About this Project



External internship at U-Hopper:

Big Data Analytics

Business Intelligence

Chatbot

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 bad organized service leads to higher costs

→ It is important to avoid generating 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

User privacy ensured



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

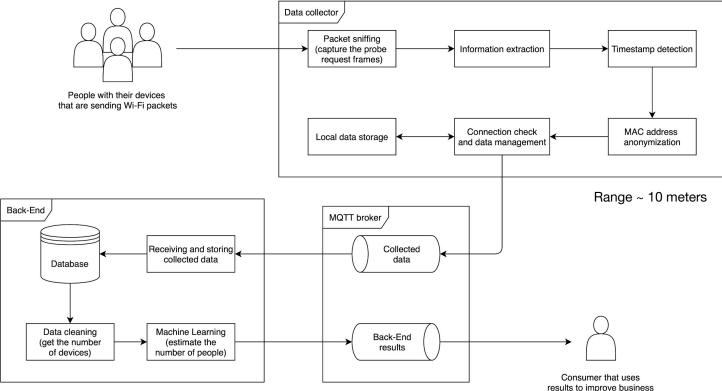


Designed and developed a system for this problem that could work in several context

Tested the system in a Cafe and collected 4 weeks of data and manually-annotated ground truth

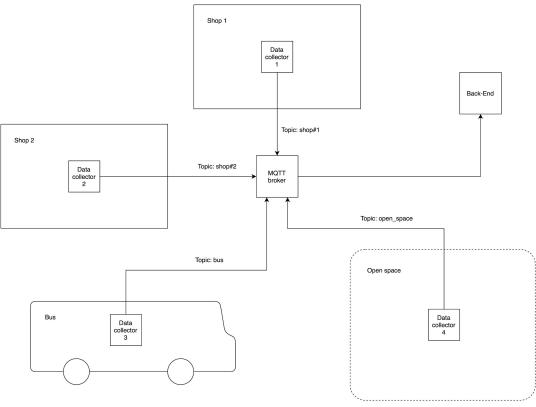
System Architecture

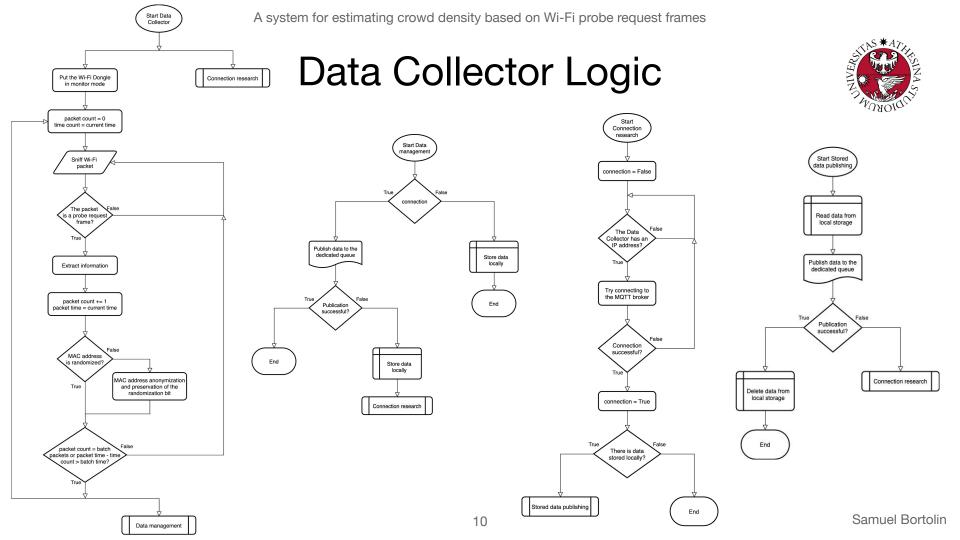




Scalable Architecture

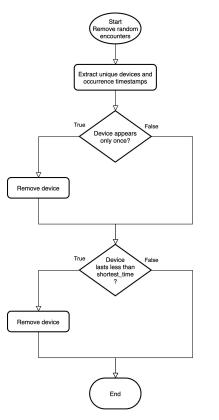


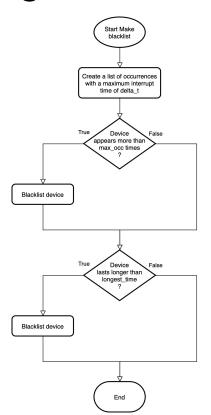




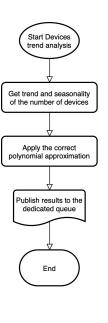
Start Back-End Subscribe to the collected data queue collected data Input manager, add corollary information Store data in the database The time slot is over? Get data from the database RSSI thresholding Remove random encounters Make blacklist Get the number of devices present for each timestamp Devices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

Back-End Logic







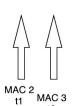


Presence of Devices



Probe request frames revelations

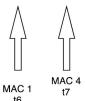










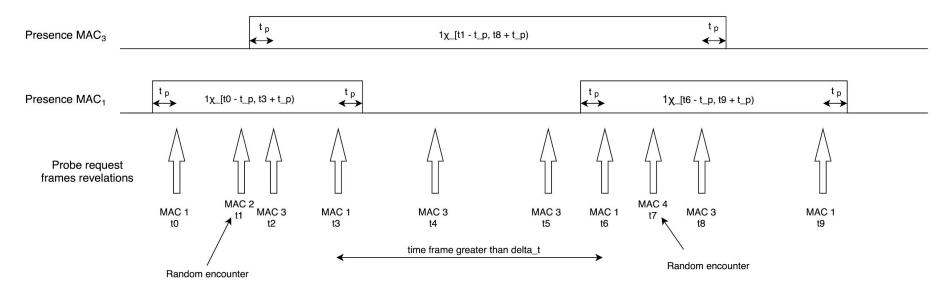


MAC 3



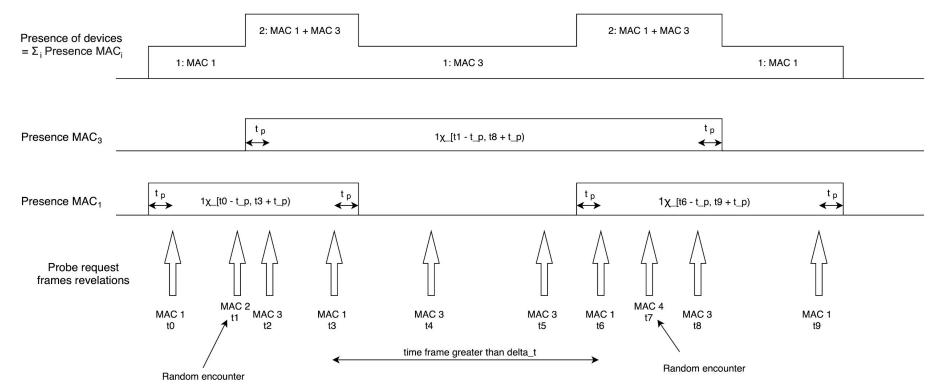
Presence of Devices





Presence of Devices





Feasibility Test at Home



Tests at home before validation

3 days of data collection

65928 probe request frames captured

12 home devices revealed

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

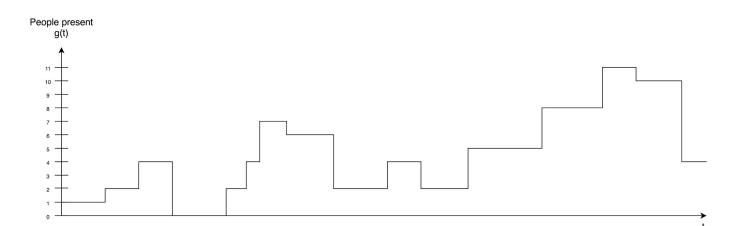
- → MQTT receiver and MongoDB on U-Hopper server
- → Analyzer and Estimator on my pc to use on the data + collected ground truth to test accuracy and reliability of the proposed system

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

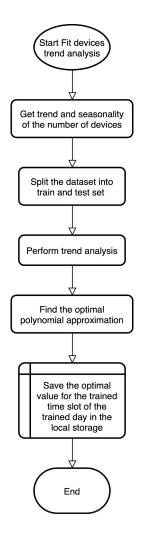
Ground Truth Collection

17





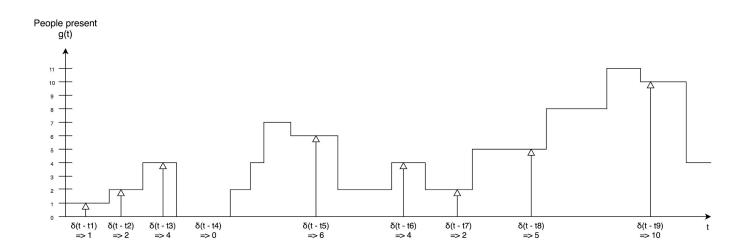
Samuel Bortolin

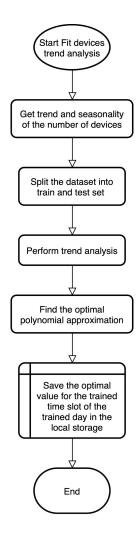


Ground Truth Collection

18

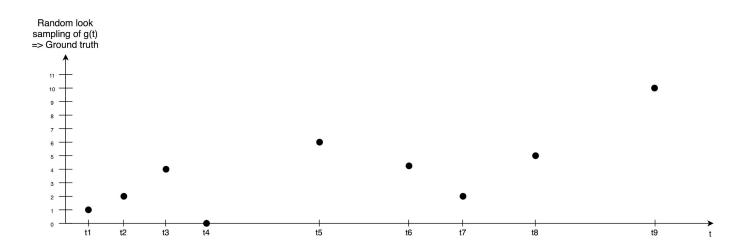






Ground Truth Collection





Test Results at Cafe

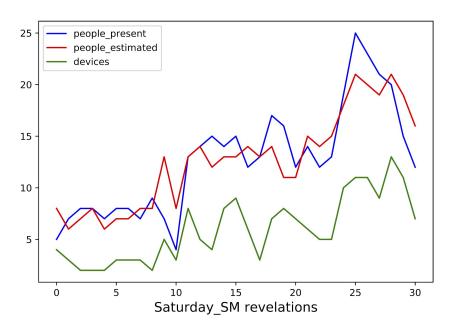


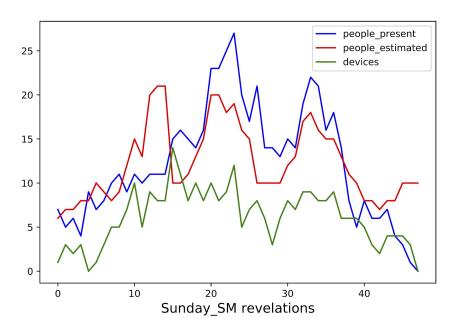
→ 4 weeks of data collection (24 days)

- → 1270 manual annotation of ground truth
- → 861979 probe request frames captured (~ 560MB)
 - → 38771 MAC addresses revealed

Test Results at Cafe



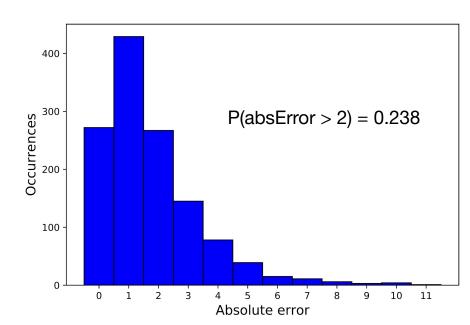




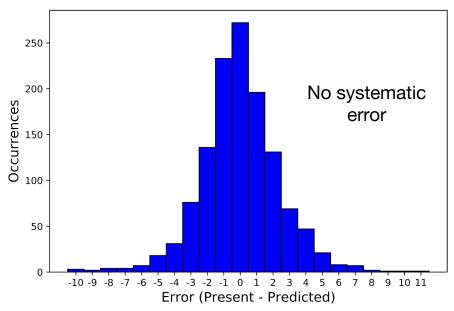
Error Distribution



Mean Absolute Error = 1.731



Mean Squared Error = 5.710



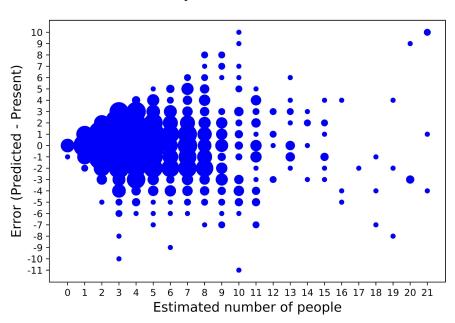
Scatter Plot of the Error



High number of people, lower predictions

11 10 Error (Present - Predicted) -3 6 7 8 9 101112131415161718192021222324252627 Actual number of people

No systematic error



Low number of people, higher predictions

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 developed 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



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Thank you for your attention

Samuel Bortolin