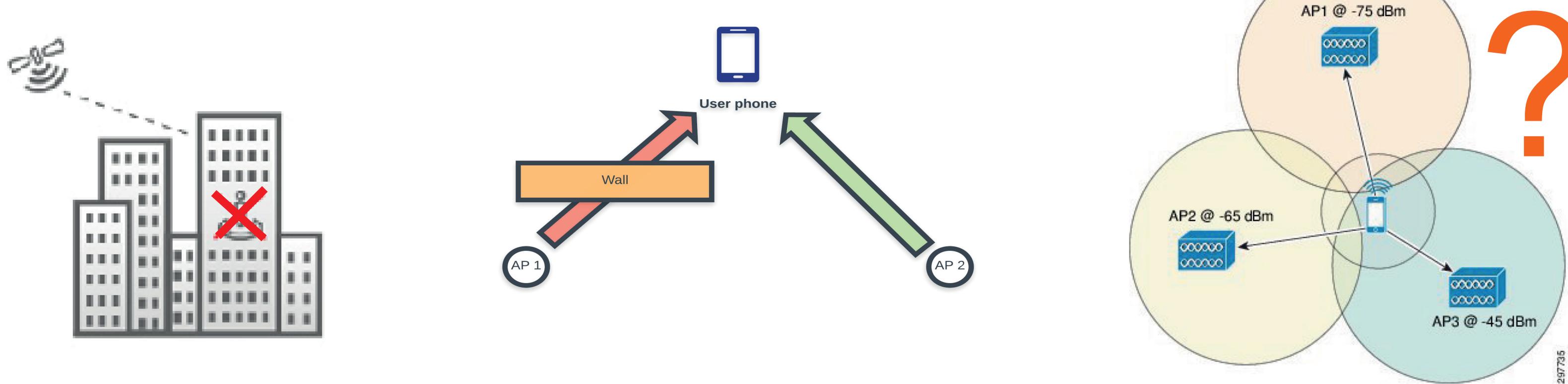


Introduction

- The majority of indoor location have Wi-fi coverage, users can connect to this wireless network. Therefore, it is possible to construct an indoor positioning system based on this infrastructure.

Problem

- The main goal is to perform a prediction of the indoor position of users.
- GPS signals cannot be used for indoor location due to attenuation caused by building walls and other materials.
- People turn to sensor-based systems to solve this problem such as WI-FI-based localization
- This technique performs poorly when used for determining a propagation model that describes the relationship between time to flight(TOF) and signal strength to predict the distance between the user and WI-FI routers.
- Static solution could not be applicable for long term usage (eg: mathematical model based on theoretical hypotheses), since those models need costly human recalibration, hence the need for a more dynamic solution.



Methods

1. Database creation and reconstruction:

- The database contains labelled fingerprints, each one includes a group of WI-FI signal measures with a reference to the point where it was measured.
- We regroup those fingerprints as clusters where every cluster refers to a specific region in the surveillance area (eg: classrooms, hallway, amphitheatres ...)

2. Model building and maintaining :

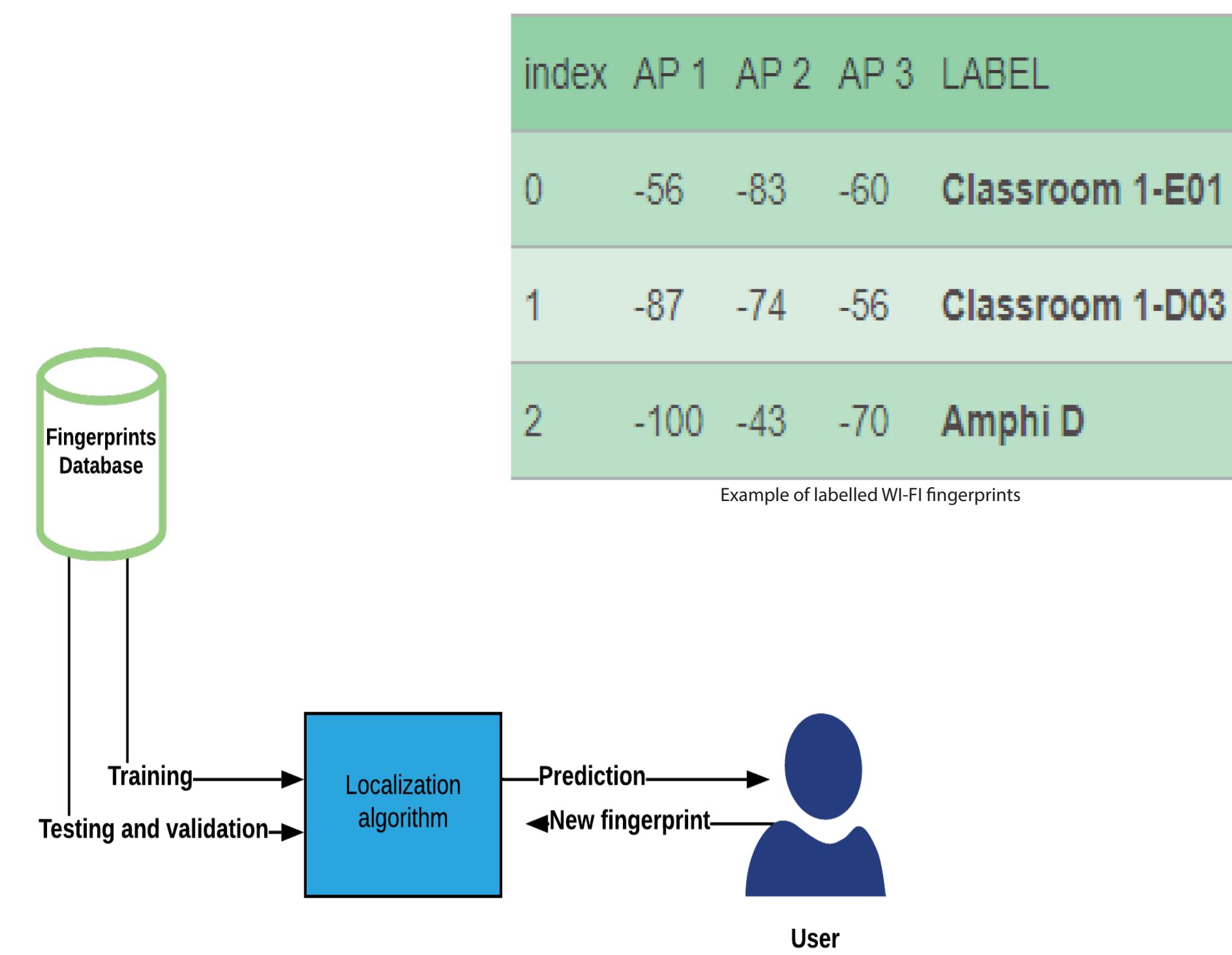
- Model training : We've chosen random forest where we try to map WI-FI fingerprints to those class labels.
- Model validation : recollect new fingerprints to test model accuracy, verify any possibility of overfitting and perform model parameters tuning for better results.

3. Deployment:

- The model is saved and then housed in a rest API that accepts both methods: post and request.
- The user scans for wifi signals, the fingerprint is then sent to the API, preprocessed to match the format of the input vector of our model. Finally, the API returns a prediction .

Solution

- We've chosen Wi-fi fingerprinting -based method to predict the user's position.
- It consists of mapping a space of physical measurement(signal strength from every router in the surveillance area(Db m)) to a space of positions.
- The mapping operation utilizes the similarity in WI-FI signal strength within the region to assign every wifi fingerprint to the appropriate cluster.
- Instead of traditional methods ,this problem will be handled as a machine learning problem to reduce human intervention and guarantee dynamicity



Experimentation

Surveillance area specifications: • First floor of the classrooms hall bloc E (8 classrooms)
• Second floor of the classroom hall bloc E (3 classrooms)

Number of wifi routers detected: 182 routers

Number of fingerprints collected for training: 1966 samples

Data collection method :

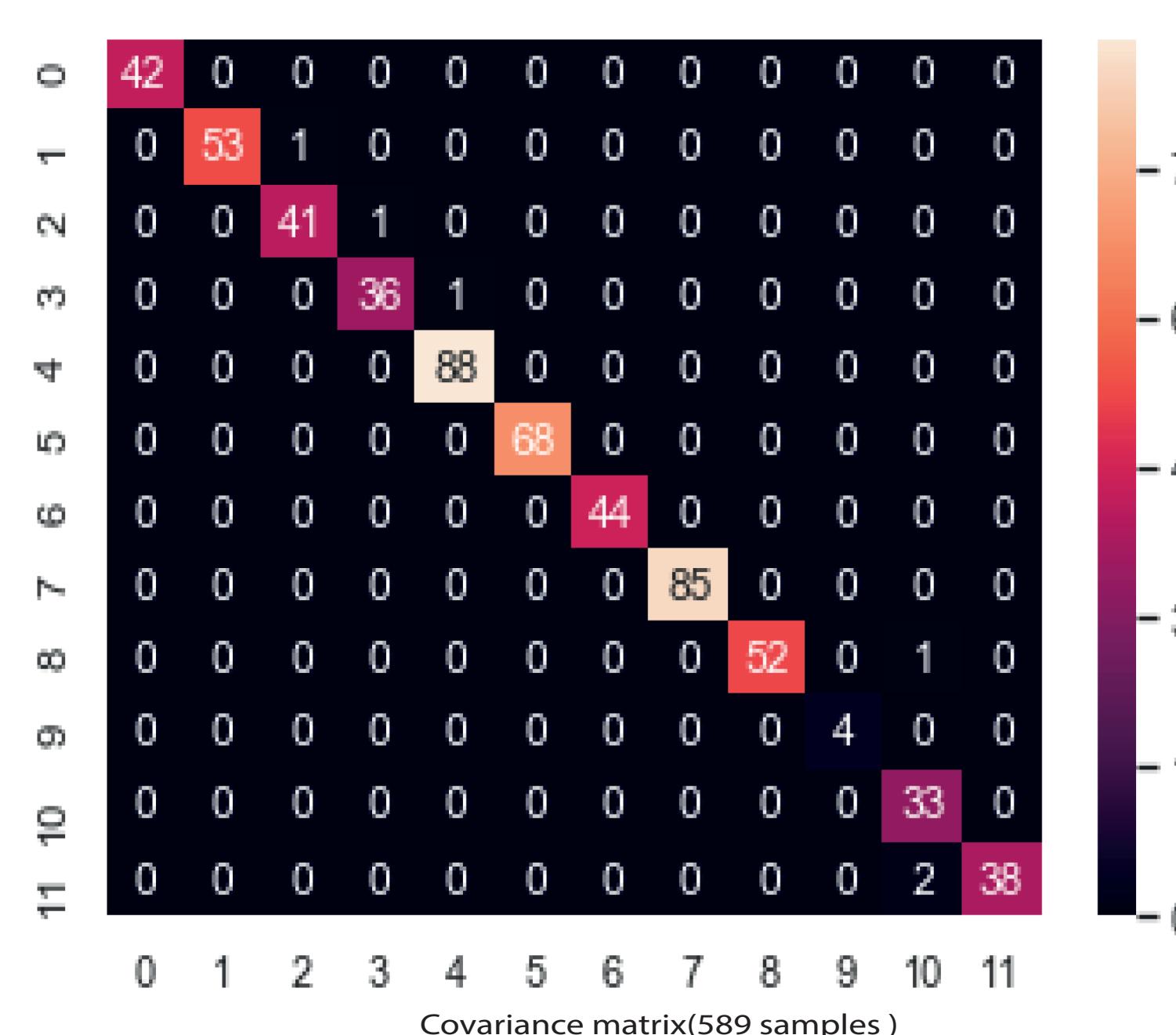
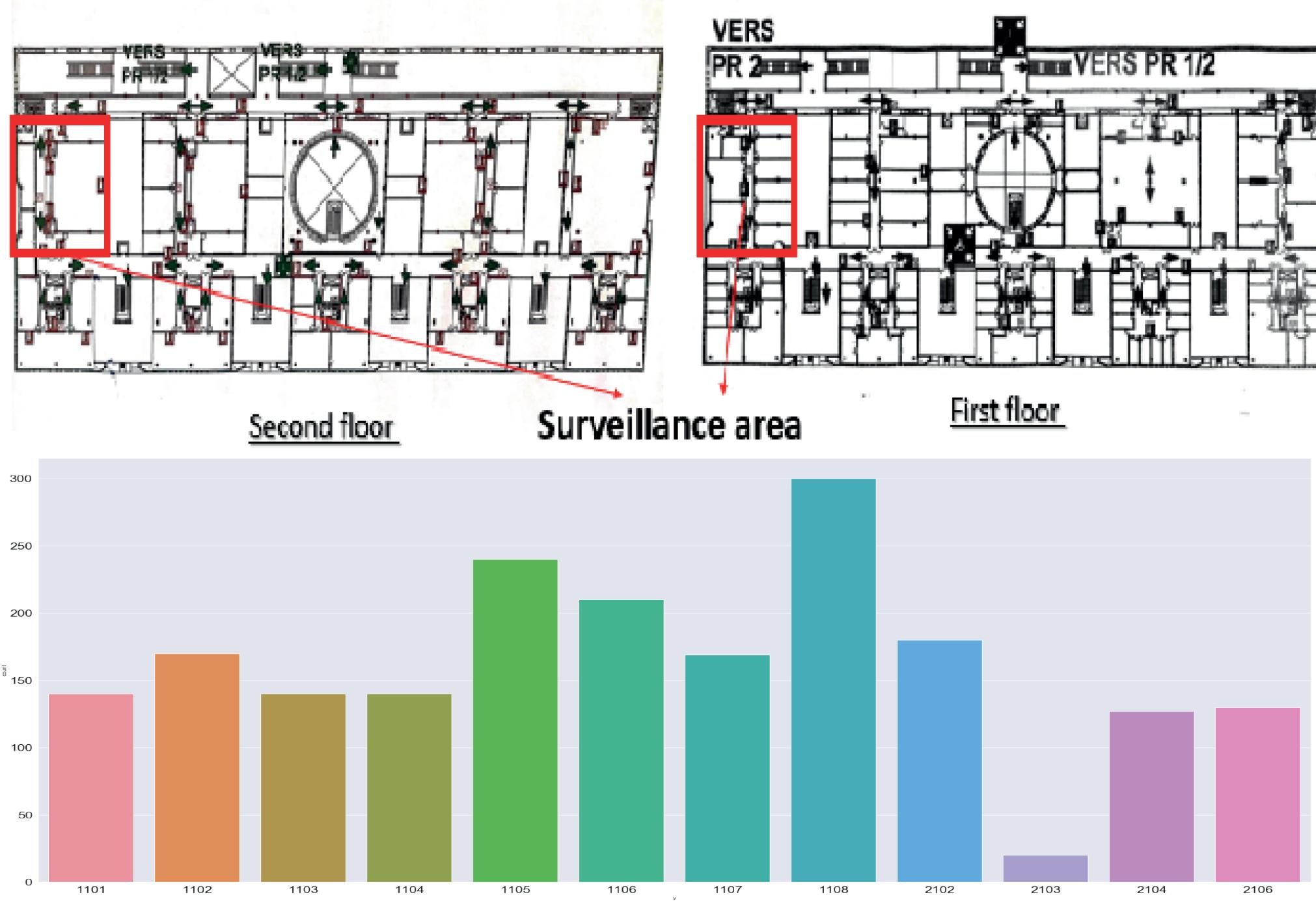
- We used a basic Android app to scan for WI-FI signals (mac address of the router and relative RSSI value).
- The data then preprocessed to create a normalized dataset and fed to the model after splitting between test and train data-set.

• model used : Random forest from Scikit-learn implementation

• Parameters: 100 estimator, max depth:50 ,criterion: entropy

Model accuracy :

- The model performed well with a + 95 % accuracy.
- We tried an unseen dataset to verify bias or overfitting, and it returned an identical level of accuracy.



Conclusion

- Accuracy must be verified periodically using a new testing dataset to see if we need to tune model parameters.
- This experiment phase(includes 11 classrooms) showed good accuracy that encouraged us to enlarge the surveillance area so we can verify the scalability of this solution.
- We are aiming to move to a more challenging problem: predicting the exact position of the user (spatial coordinates of the user) ,thus the need for a regression model.

Contact

- Anas Ait Aomar
- anasaitaomar1999@gmail.com
- 00212 605140465