

# WiFi Based Indoor Positioning System

Under the guidance of “RIP NO GUDIANCE”

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# Global Positioning System (GPS)

- Prominent contributor in location tracking
- Works efficiently in open areas
- Used by many location tracking apps

Eg: Google Maps  
Apple Maps  
OpenStreetMap

# Indoor Positioning

- GPS needs an unobstructed line-of-sight to 4 or more satellites.
- Fails inside enclosed spaces
- Increasing need for Indoor Positioning inside Malls, Airports etc
- No common existing system for Indoor Positioning
  - Google's experimental Indoor Map

# How can it be done ?

- Radio Frequency Identification (RFID)
- Bluetooth
- WiFi
- Motion Sensors & Position Sensors

# Why choose WiFi ?

- Commonly available inside buildings
- All smart phones come with inbuilt WiFi adapters
- WiFi signal strength varies with distance
- Can be used as a distance indicator for positioning

# Design Overview

- Works in two phases.
  - Calibration phase
  - Positioning phase
- Calibration phase
  - Location Fingerprinting
- Positioning phase
  - Weighted K Nearest Neighbours Algorithm (WKNN)

# Location Fingerprinting

- Received Signal Strength (RSS) values from multiple routers act as a fingerprint for a location
- Different locations are most likely to have unique fingerprints
- Fingerprint of  $i^{\text{th}}$  location is denoted by  $r_i$

$$r_i = \{r_{i1}, r_{i2}, r_{i3}, \dots, r_{im}\}$$

$r_{ij}$  = RSS value of  $j^{\text{th}}$  router from  $i^{\text{th}}$  location

# Location Fingerprinting

- Several locations are chosen and RSS values from them are recorded and form a radio map.
- The recording at the  $i^{\text{th}}$  location is of the form

$(q_i, r_i)$

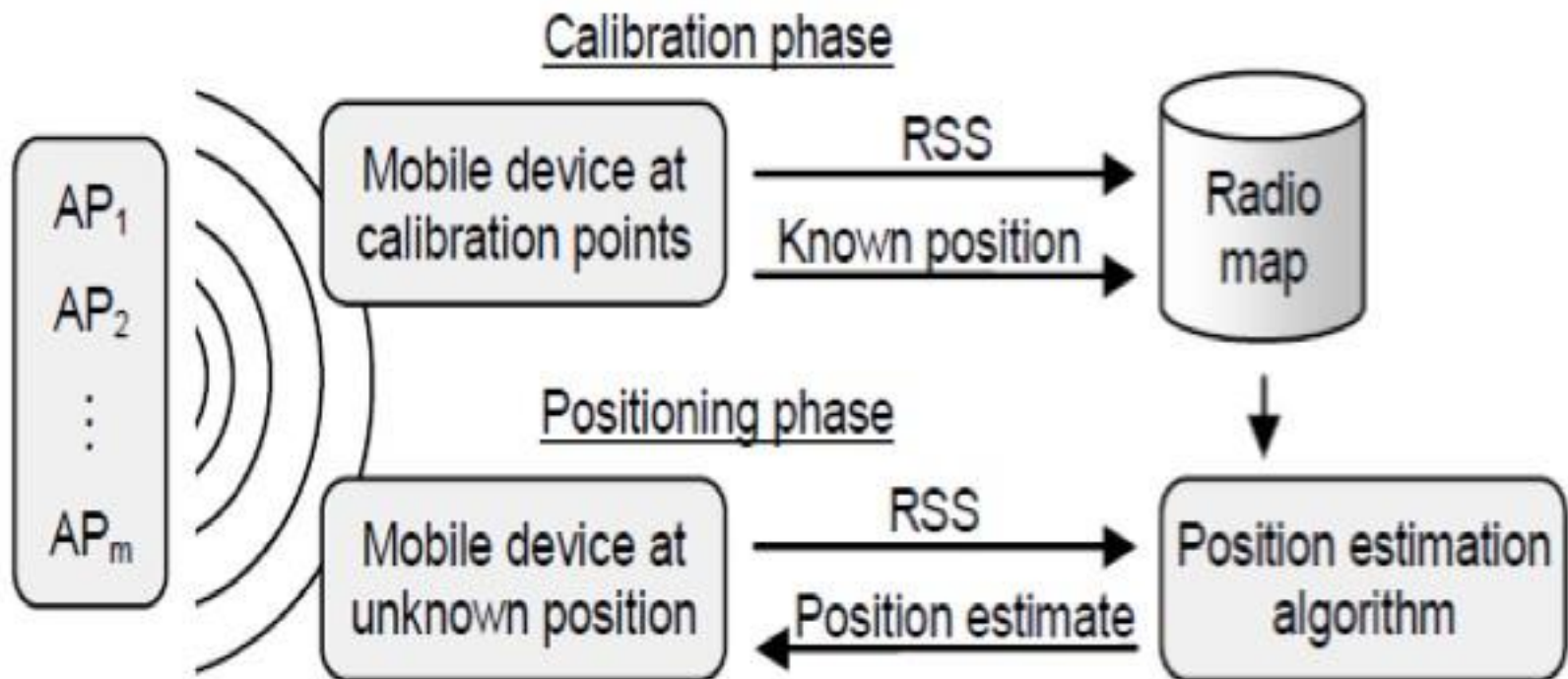
$q_i$  - the geometric coordinates  $(x_i, y_i)$

$r_i$  - the location fingerprint



# Location Fingerprinting

- A position estimator algorithm is used to find the coordinates of unknown location



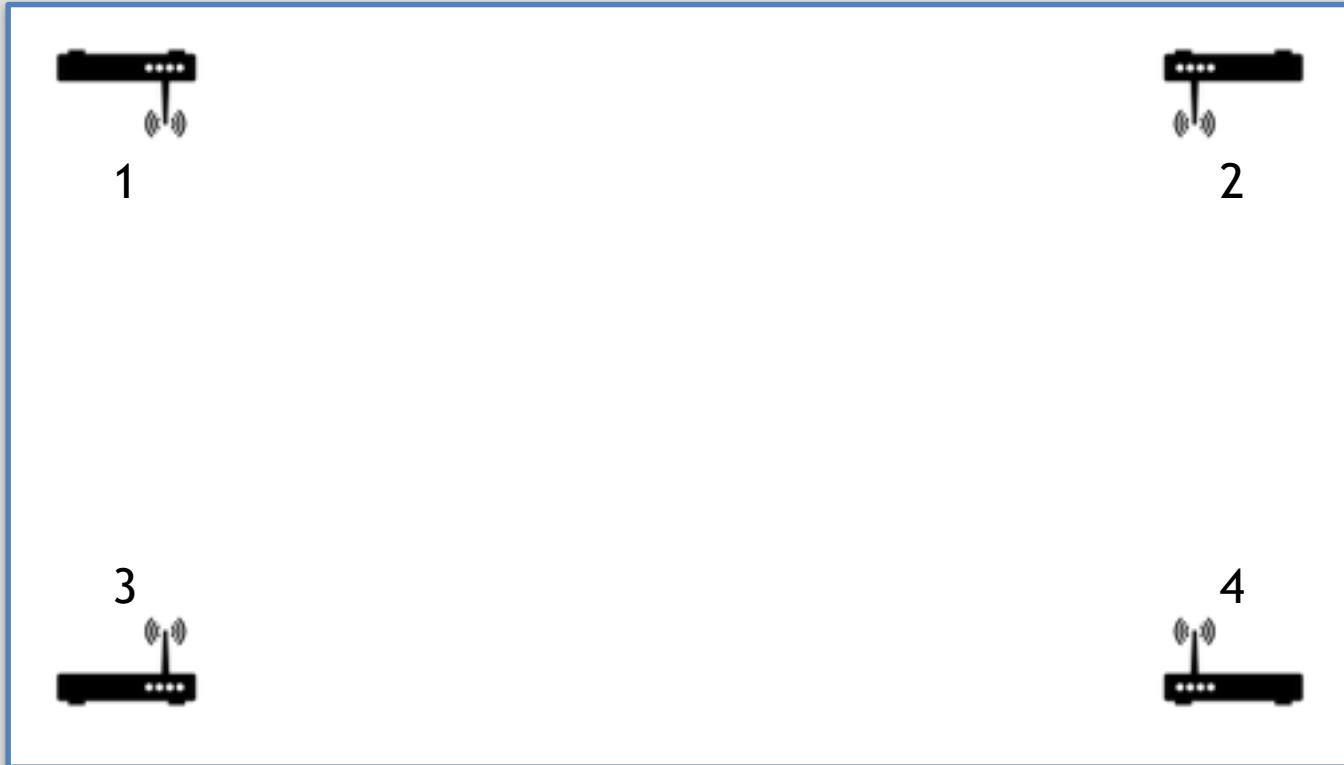
# Weighted k Nearest Neighbour

- The position estimator algorithm used is the Weighted k Nearest Neighbour (WkNN)
- Finds the k nearest chosen locations from unknown location based on Euclidean distance
- Calculates coordinates of unknown location as the weighted average of the nearest k points

# Weighted k Nearest Neighbour

- Weight is the inverse of the Euclidean distance
- $k$  can be considered as a tuning parameter in the algorithm
- When  $k=1$ , algorithm acts as a simple look up table

# How the algorithm works



Suppose there are 4 routers in the floor

# How the algorithm works



The signals emanating from routers leave a unique fingerprint at each location

# How the algorithm works



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# How the algorithm works



The signals emanating from routers leave a unique fingerprint at each location

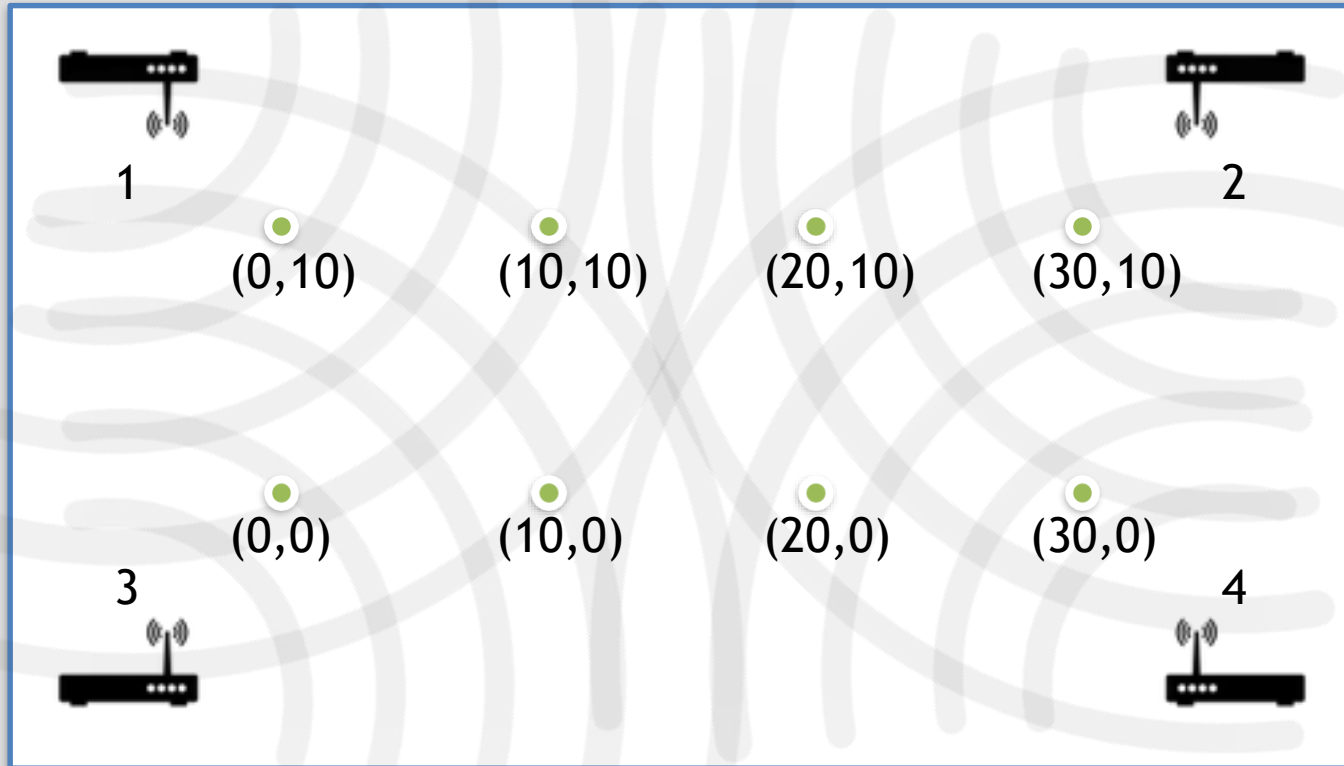
# How the algorithm works



The signals emanating from routers leave a unique fingerprint at each location

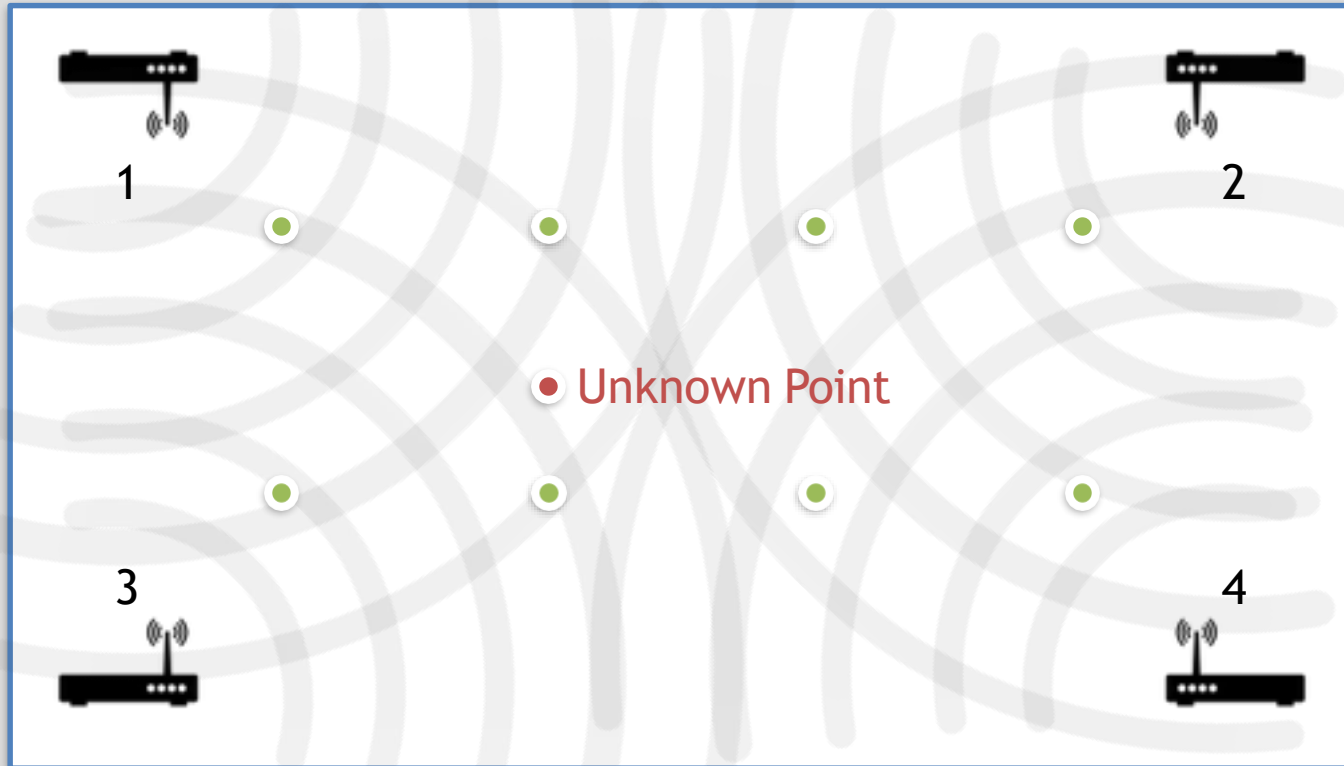


# How the algorithm works



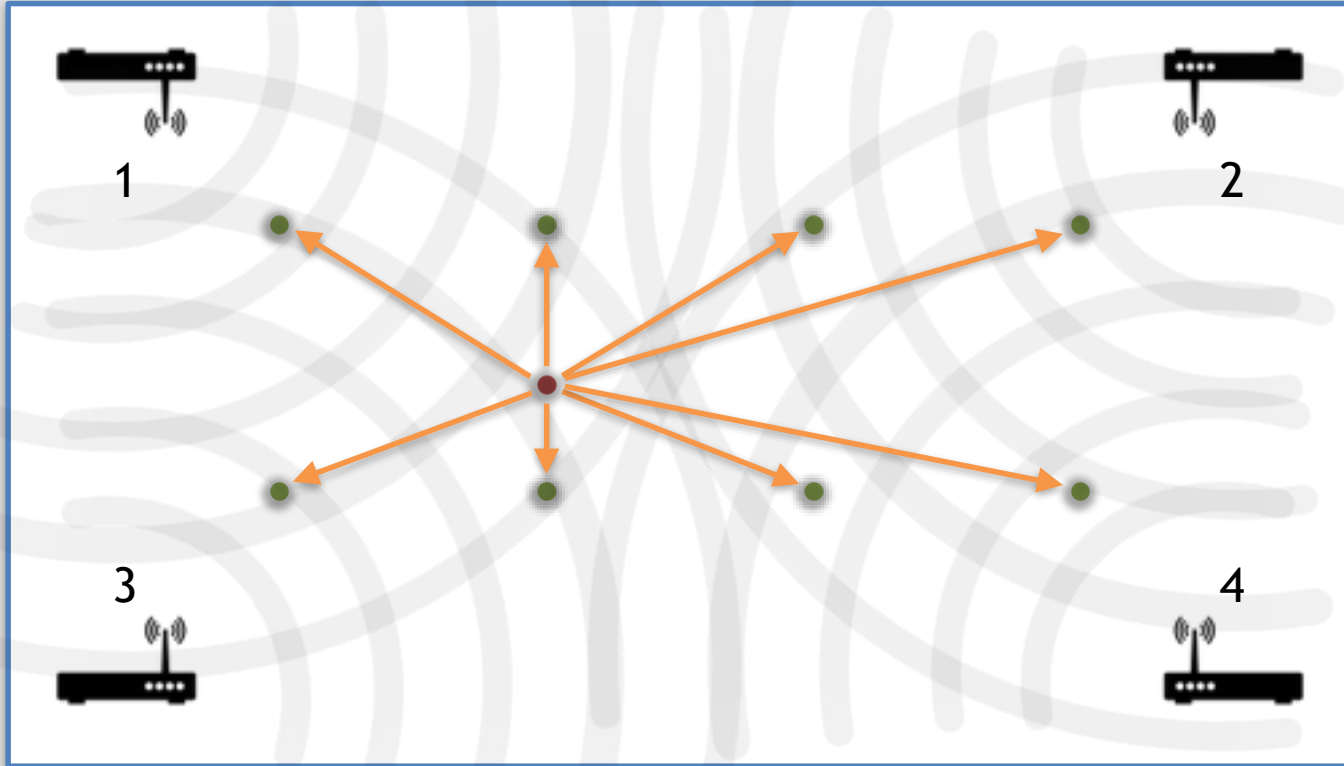
Choose 8 points on the floor and calibrate readings

# How the algorithm works



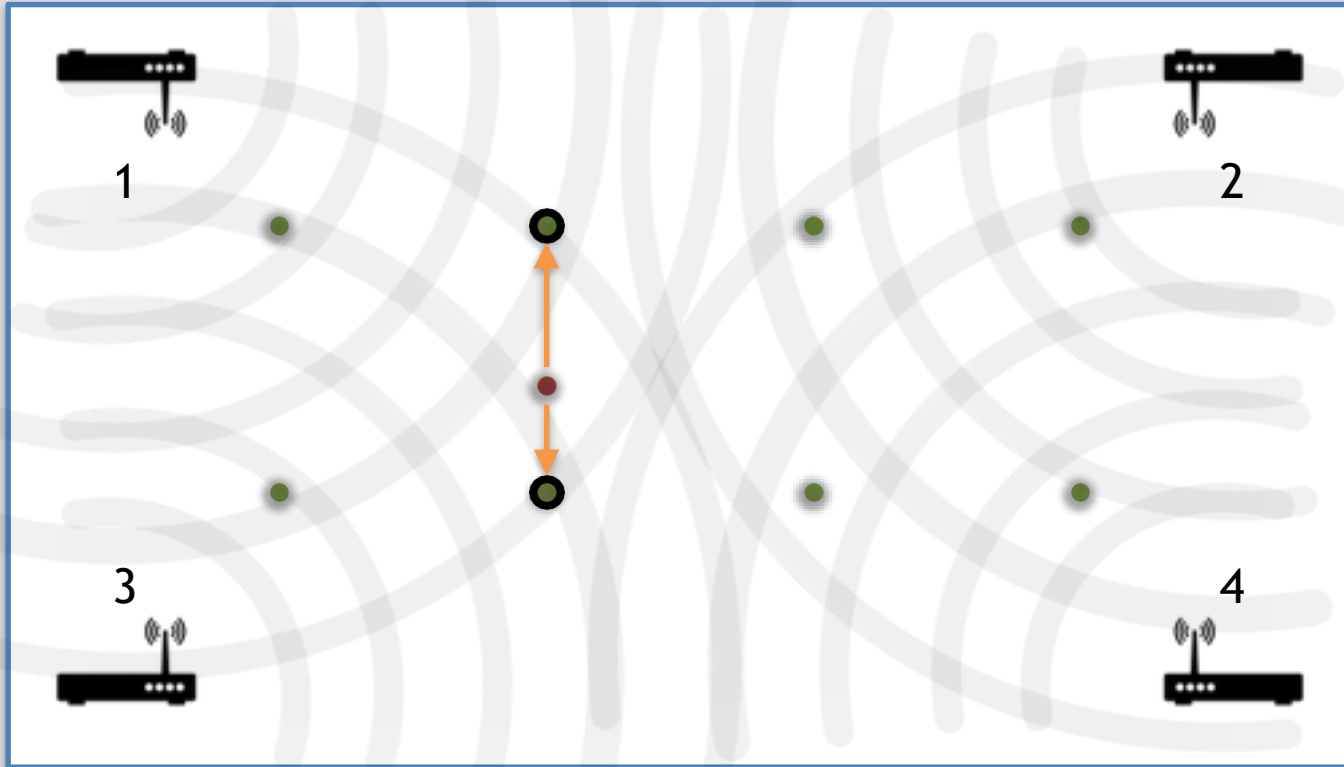
Calculate Euclidean distance from unknown point to all chosen points

# How the algorithm works



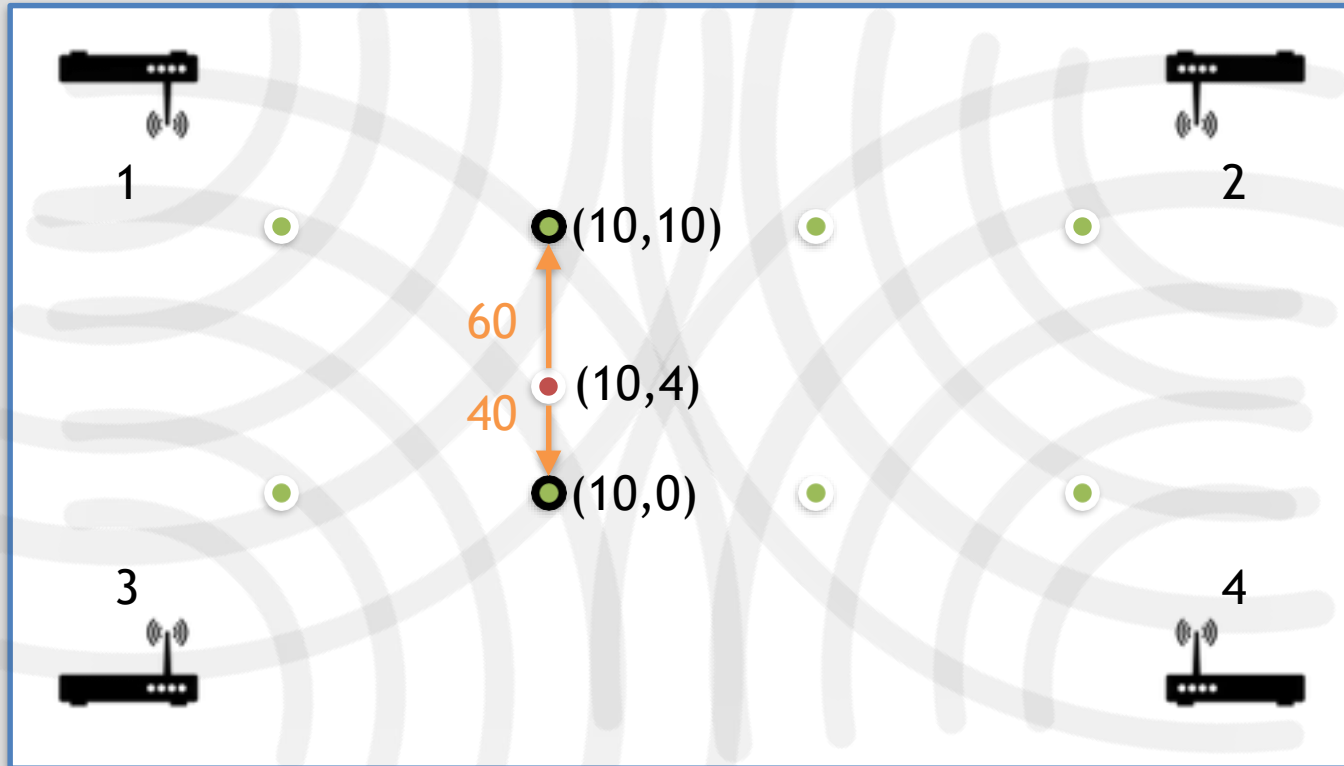
Euclidean distance between points  $x$  and  $y$  is calculated by  
$$\text{sqrt}((r_{x1}-r_{y1})^2+(r_{x2}-r_{y2})^2+(r_{x3}-r_{y3})^2+(r_{x4}-r_{y4})^2)$$

# How the algorithm works



Nearest  $k$  points are chosen. Here  $k = 2$ .

# How the algorithm works

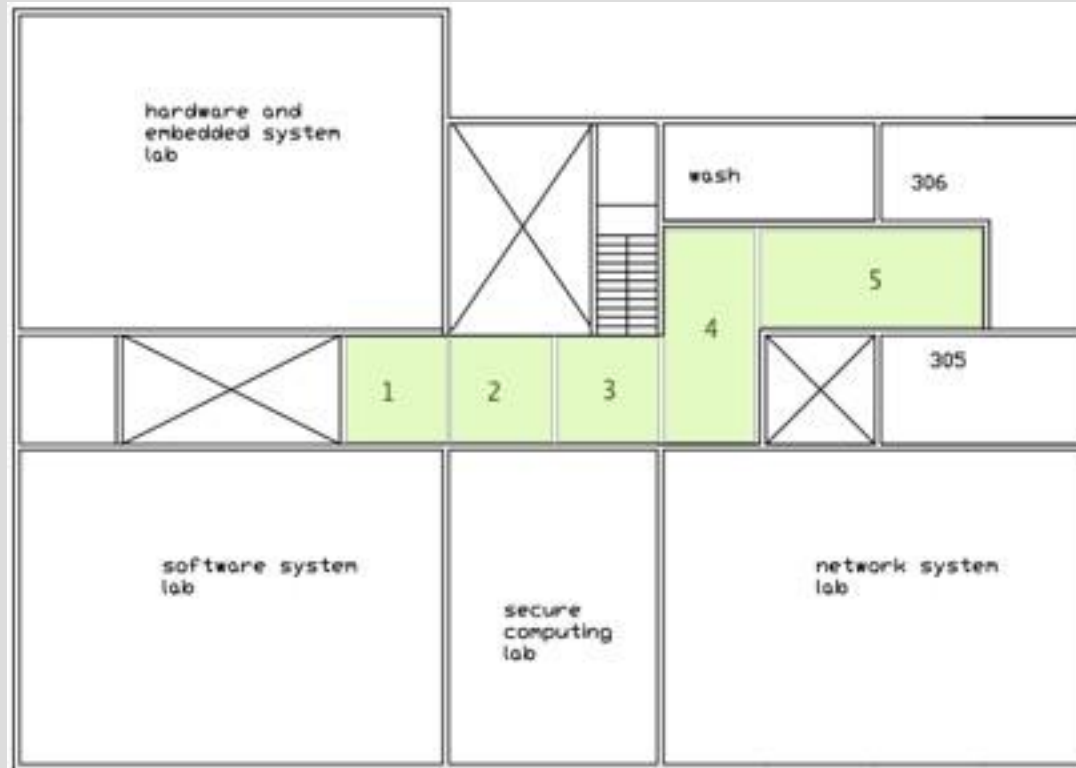


Coordinate of unknown location is the weighted average of the 2 nearest points

# Work to Done

- Prototype Android App which Performs calibration and positioning
- Mapping inside CSED Lab building
- WkNN algorithm with  $k=1$  used
- Device position logs could be tracked from website

# Work to done



- Application could distinguish the grids shown in the figure

# DEMO

AVAILABE SOON....

<https://github.com/git-preet/INAV>



# Practical Difficulties

- Fluctuating signal readings

*Considered the average of 30 readings*

- Readings might contain temporary APs

*Prepared a list of trusted APs to be used*

- User orientation can affect the readings

*Orientation specific readings should be recorded*

*(to be done)*

- Proper positioning of APs can reduce error

# Work we maybe can add

- WkNN should be modified to  $k > 1$  to increase accuracy
- Ideal value of  $k$  needs to be figured
- Readings taken should be normalised so as to work for all devices
- Calibrated readings should be made available for all devices

# Questions?

Thank You