

# A MEASUREMENT STUDY FOR UNDERSTANDING WIRELESS FORENSIC MONITORING

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# Outline

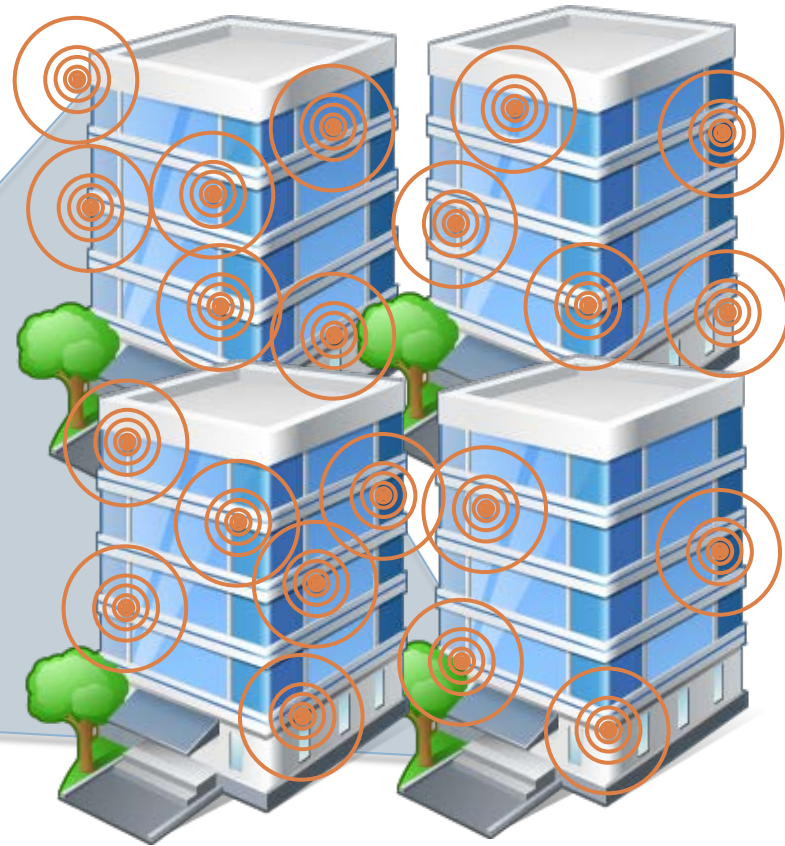
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- Background
- Experiment setup
- Observations
- Future work

# Background

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- Increasing number of WiFi and mobile devices
  - According to [wigo.net](http://wigo.net), 69.1M on 7/31/12, **73.3M** wifi reported on 9/20/12



# Background – Forensic Challenges

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- Lack of traceability through WiFi nets
  - ▣ (Open) WiFi – hacker comes and goes
  - ▣ Usually no logs kept at APs
- Uncertainty of **device identification**
  - ▣ Mac spoofing, etc.
- Uncertainty of **device location**
  - ▣ Densely populated WiFi in Metropolitan neighborhood – difficulty in localizing device both horizontally and vertically

# Background – Goal of this work

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- A *Measurement study* to
  - ▣ Explore Metropolitan WiFi nets characteristics
  - ▣ Explore representative Device Localization Approaches
    - K Nearest Neighbor (KNN): a fingerprinting based approach
    - Log-distance path loss model: a model based approach
    - A good survey: *Survey of Wireless Indoor Positions Techniques and Systems*, H. Liu, H. Darabi, P. Banerjee and J. Liu, IEEE Transactions on Systems, Man, and Cybernetics, Nov. 2007

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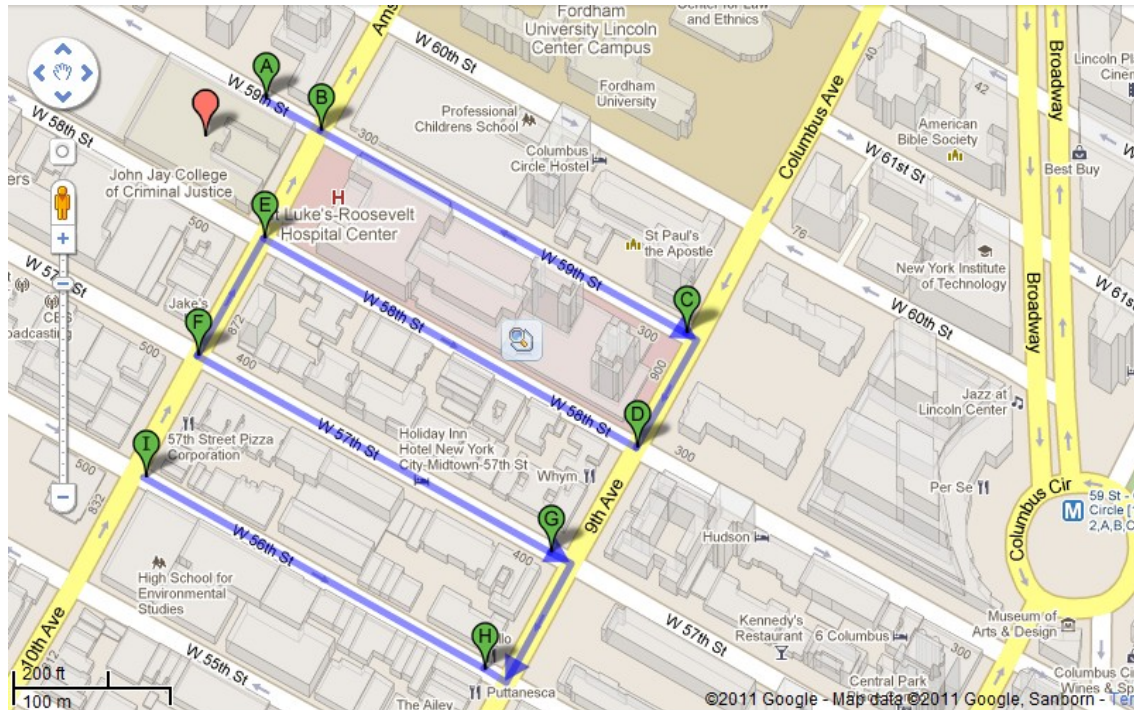
# Equipment Setting

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- Setting I
  - ▣ MacBook Pro laptop with Mac OS X10.7.4
  - ▣ integrated wireless card, BU353 GPS receiver
  - ▣ Kismet (Passive Scan)
- Setting II
  - ▣ MacBook Pro laptop, Backtrack in Virtual Machine running
  - ▣ Alfa wireless card, BU353 GPS receiver
  - ▣ Kismet
- Setting III
  - ▣ HTC Stream Tablet, Android,
  - ▣ integrated wireless card, integrated GPS receiver
  - ▣ [WiFum](#) (Active Scan)

# Testing Path and Duration

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- Three-block urban neighborhood of midtown west Manhattan
- Three data sets per test run
- Four test runs on Dec 26, 2011 [14:06-14:35 (29 mins, A-H), 14:36-15:05 (29 mins, H-A), 15:07-15:35 (28 mins, A-H) and 15:36-15:58 (22 mins, H-A)]



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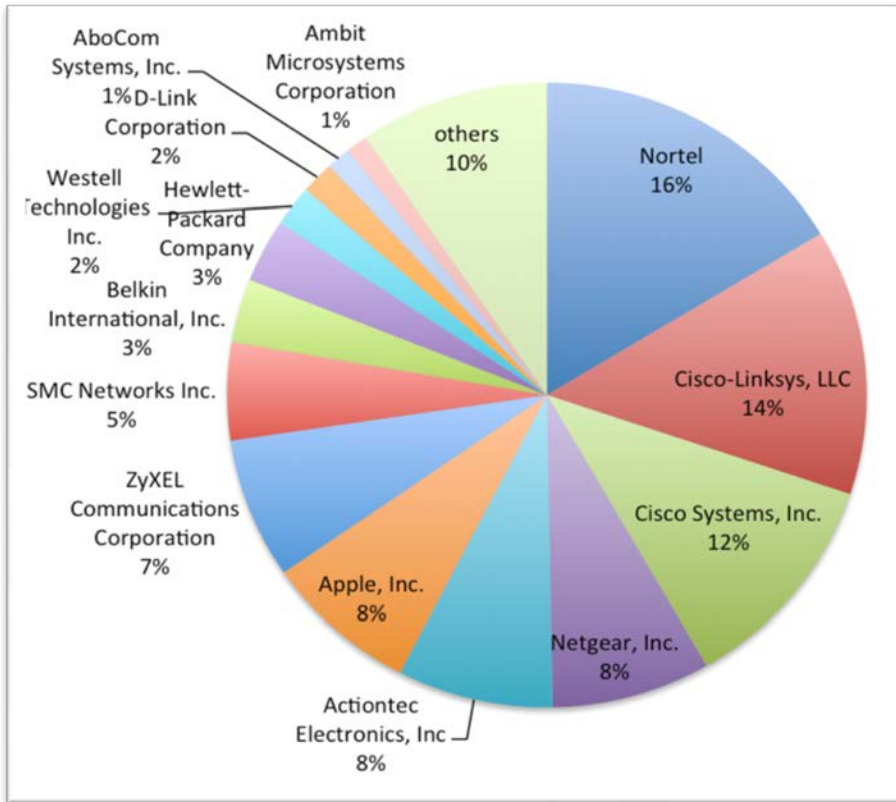
# Number of AP

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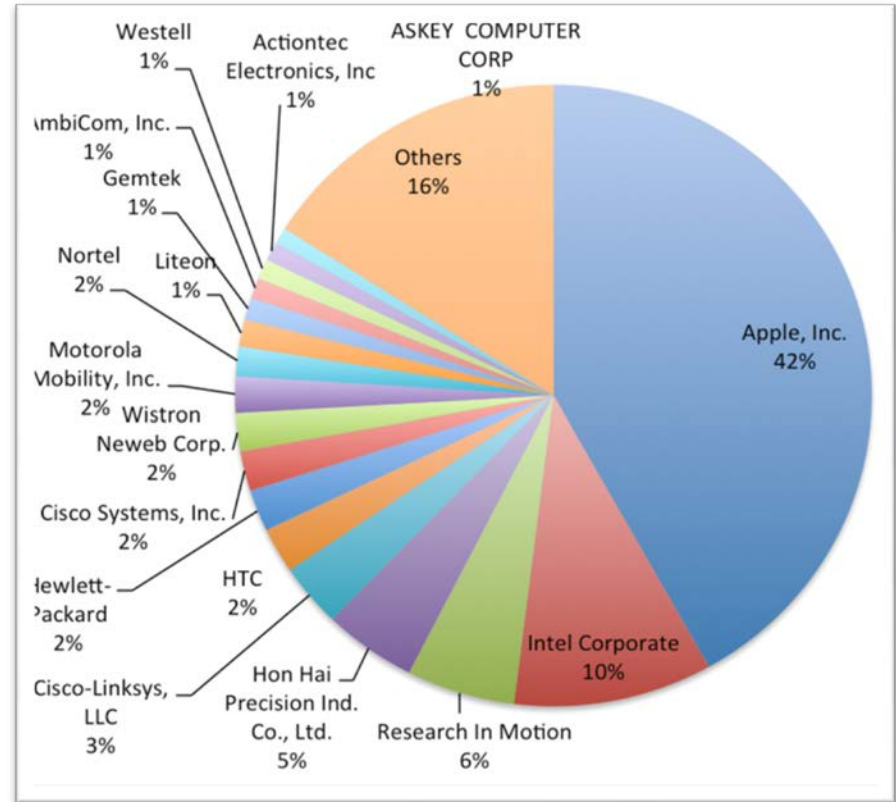
| Tool                                | Airport + Kismet | Alfa + Kismet | HTC + WiFum |
|-------------------------------------|------------------|---------------|-------------|
| Average # of AP/Position            | 15               | 11            | 24          |
| Average # of AP/Trip                | 1823             | 1829          | 1520        |
| Average percentage of Encrypted APs | 0.706            | 0.6835        | 0.8205      |

# Device Manufacturer

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(a) Access Points

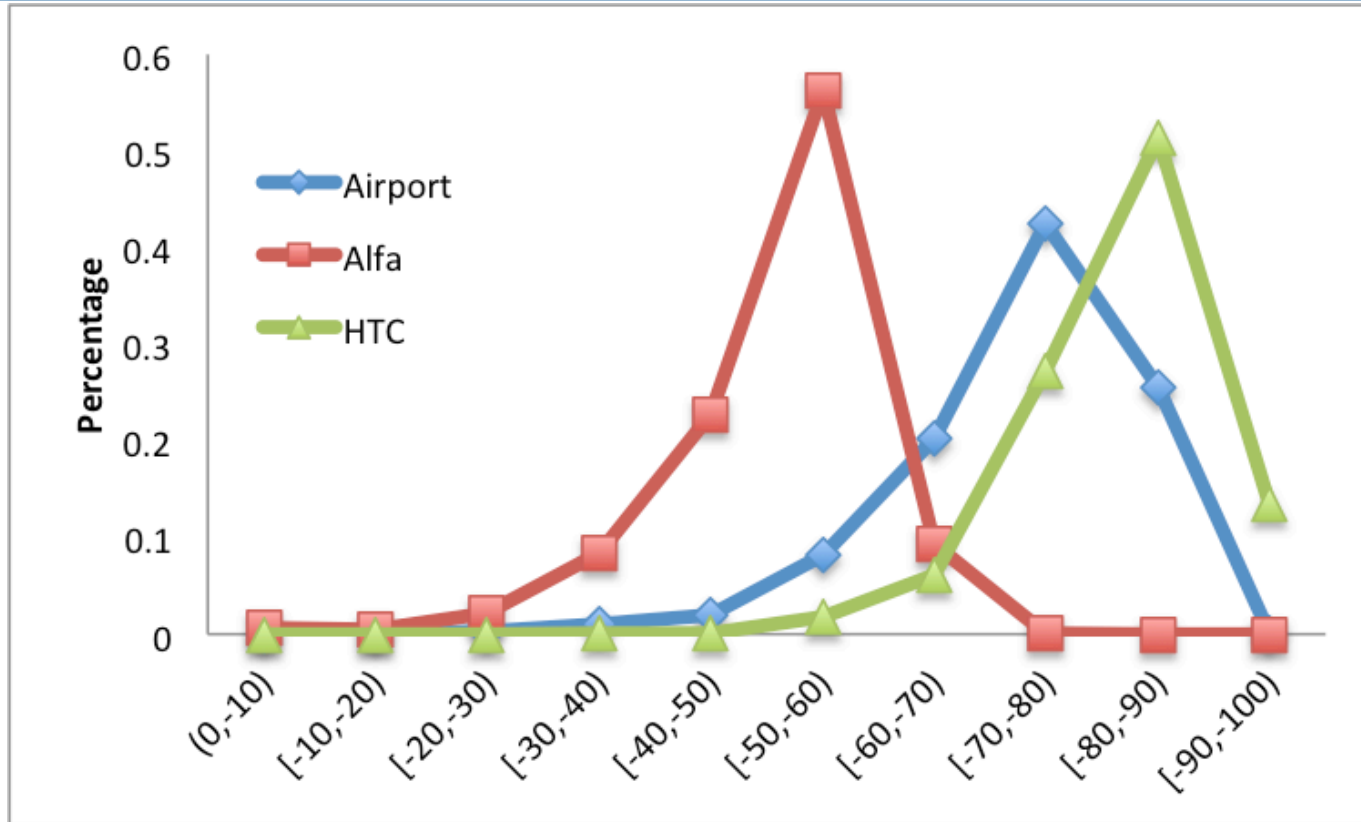


(b) Mobile Devices

**3807** Access Points and **6615** Mobile Devices from Kismet traces

# RSS Sensitivity

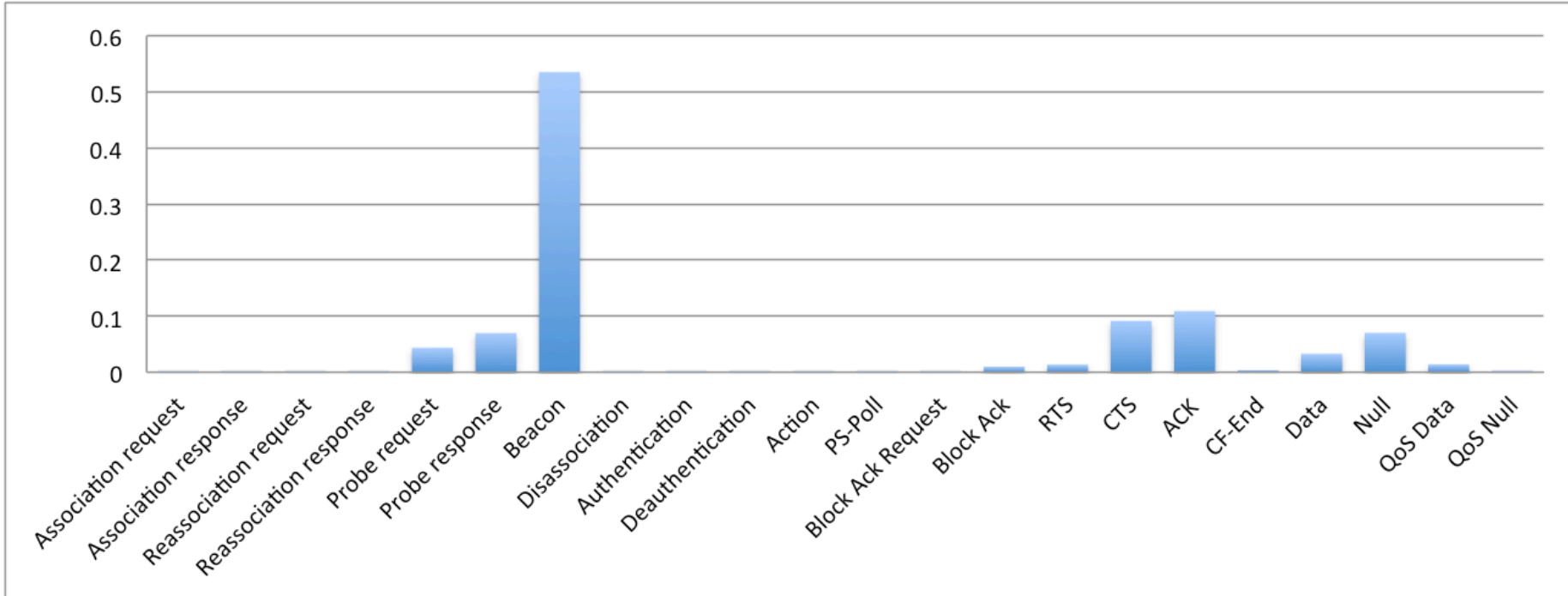
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- Average Received Signal Strength (RSS):  
External Alfa card  $>$  Cards of Mac computer & HTC tablet

# Packet Type

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- 362,305 packets from Kismet traces
- 50% beacons from APs, 20% CTS & ACK, 11% Probe req/rep
- only approx 3% data

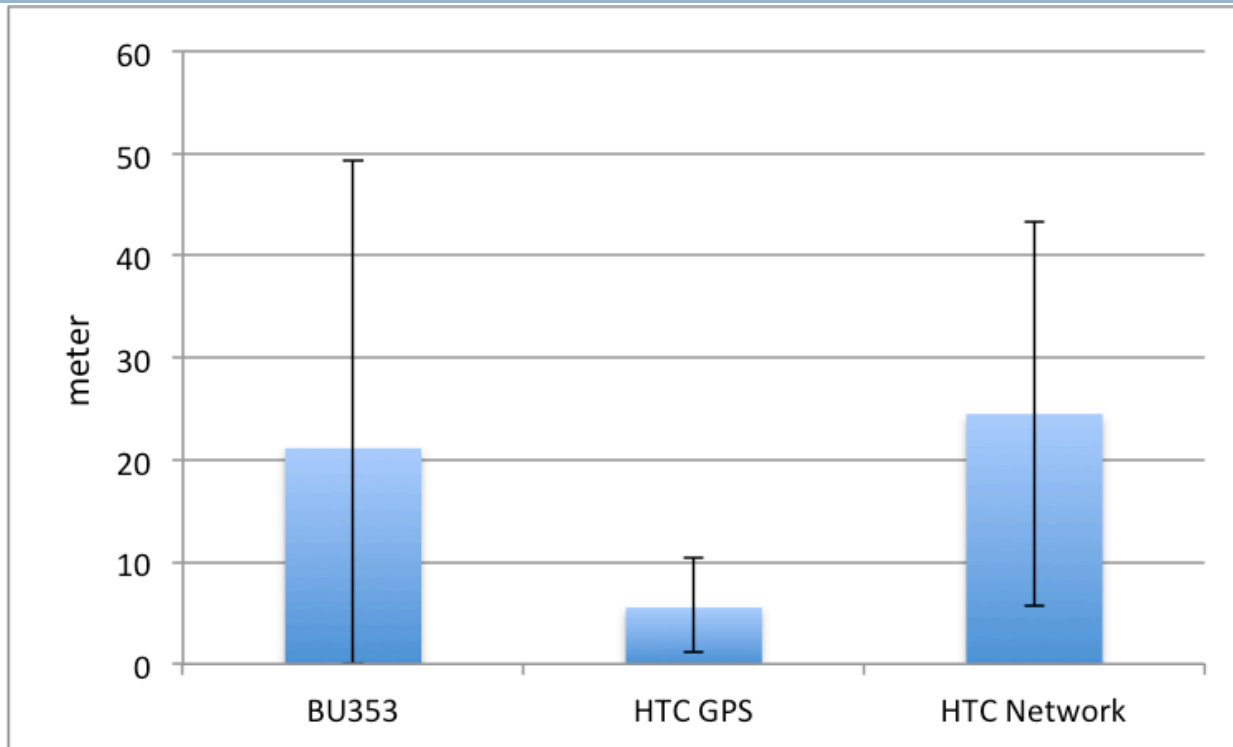
# Location Variance

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- Choose 10 locations along the path
- At each location gather trace continuously for 2 mins
- Record GPS coordinates from three different experiment settings
- Center Point = (avg latitude, avg longitude)
- Error Distance = |empirical location – center point|

# Average Error Distance

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- ❑ External GPS (BU353) provides worse consistency in location estimation
- ❑ Integrated GPS of HTC tablet provides relatively stable location info
- ❑ The HTC Network approach – combine cell tower & wireless APs – does not provide a significantly better position consistency as we thought

# K Nearest Neighbor (KNN) Approach

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- Hypothesis
  - ▣ *The set of APs and their associated signal strengths observed at a position represents a fingerprint that is unique to that position*
- Method:
  - ▣ Build an RF fingerprint database
  - ▣ Compare the observed RSS vectors to fingerprints in the database
    - Find K closest neighbors, estimate the new location via their positions
    - use Euclidean distance as the distance metric between unknown and known locations



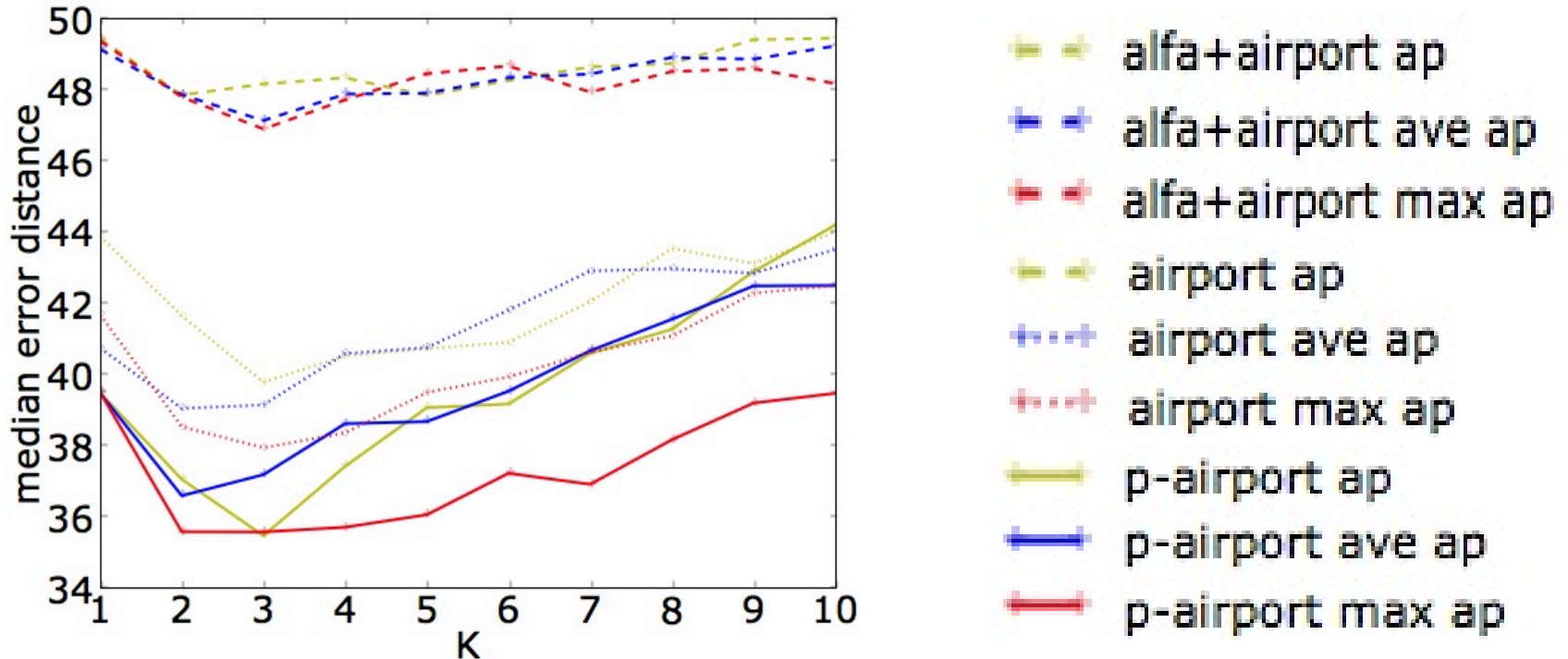
# Evaluating KNN Approach

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- Add 12 (kismet) datasets gathered in one week of April/11
  - ▣ Same setting with Setting I
- Data sets:
  - ▣ **p-airport**: the 12 traces from past (April/11)
  - ▣ **airport**: all 16 traces gathered through setting I
    - 12 traces from April/11, 4 traces from December/11
  - ▣ **alfa+airport**: 16 traces from above + 4 Dec. traced gathered through setting II
- AP selection mechanisms
  - ▣ **ap**: use all Aps
  - ▣ **avg ap**: select APs that show larger than Avg RSS at more than one location
  - ▣ **max ap**: only select APs that show Maximum RSS at more than one location

# KNN

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- Mostly, error distance smallest when  $K=3$
- Alfa USB wireless card provides significantly worse results than the integrated wireless card
- “Time” plays a role in localization accuracy => change of network parameter?

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# Future Work

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- More Measurement!
  - ▣ Establish “ground truth”
  - ▣ Static trace gathering: monitoring points close to static APs
  - ▣ More thorough investigation on localization algorithms
    - Compare and contrast the performance of existing ones, propose new one (?)
    - Combine traces from both static AP and moving monitoring points
    - Vertical localization

Thanks !  
Q&A