COMP9336/4336 Mobile Data Networking www.cse.unsw.edu.au/~cs9336 or ~cs4336

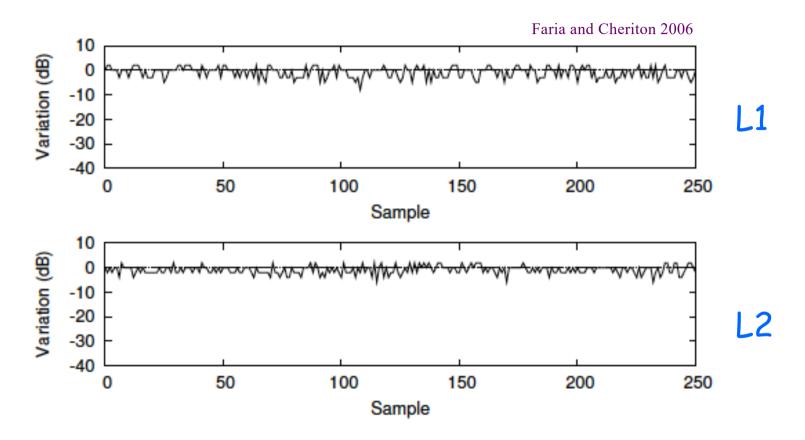
WiFi Fingerprinting

Adapted from Faria and Cheriton 2006

Signal fingerprint based positioning

- Received signal is extremely location-specific
 - dependence on terrains and obstacles
- Multipath structure is unique to every location
 - considered a fingerprint or signature of the location
- Create fingerprint database for locations of interest
- Received signal is matched against database
 - to identify location of the transmitted signal

RSSI Oscillation



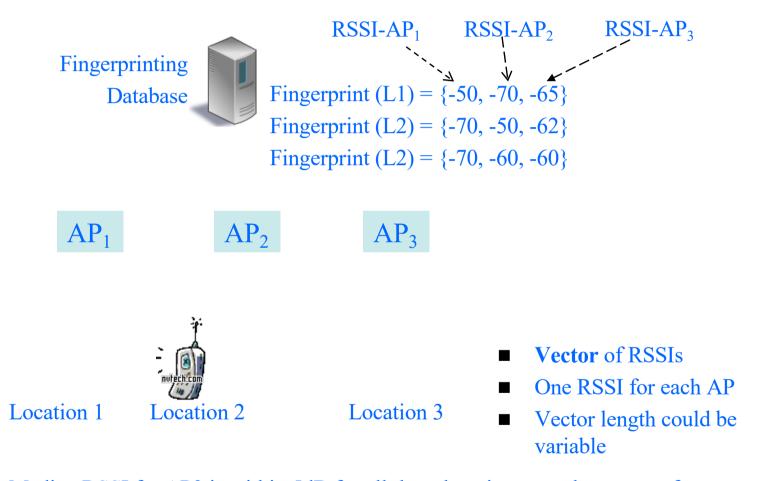
In most cases, for a given location, RSSI remains within a few dBm of the median value (median shown as '0')

L1 could be differentiated from L2 using a **single** WiFi AP if the **RSSI medians** were 10dB apart in this case

Why a single WiFi AP is not adequate?

- In the previous example, L1 and L2 could not be always separated if the median RSSIs were less than say 5dB
- A single WiFi AP therefore cannot provide high-resolution localization with good accuracy
- What if the mobile device can hear from multiple WiFi APs?

Basic WiFi Fingerprinting Example



Median RSSI for AP3 is within 5dB for all three locations, yet the vector of three APs provide unique WiFi fingerprint for these locations!

A basic algorithm for identifying locations with WiFi fingerprint

- 1. Mobile obtains a real-time fingerprint
- 2. Compare the real-time fingerprint with each signature in the database (RSSI differences in vector elements)
- 3. Attach a score to each comparison (number of elements differed less than Δ dBm)
- 4. Maximum match = signature with max score

Example

- 2 signatures in the database for two different locations
 - $S1 = \{-50, -70, -45\}$ and $S2 = \{-40, -70, -35\}$
- Real-time fingerprint of a mobile = {-44,-66,-34}
- Assuming a Δ =5dBM (needs to be finetuned for real environments)
 - Score for S1 = 1, and
 - Score for S2 = 3
- Maximum match is with location 2 (s2)
- The client positioning is predicted as 'location 2'