

Horus: A WLAN-Based Indoor Location Determination System

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2003

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Motivation

- Ubiquitous computing is increasingly popular
- Requires
 - Context information: location, time, ...
 - Connectivity: 802.11b, Bluetooth, ...
- Location-aware applications
 - Location-sensitive billing
 - Tourist services
 - Asset tracking
 - E911
 - Security
 - ...



Location Determination Technologies

- GPS
- Cellular-based
- Ultrasonic-based: Active Bat
- Infrared-based: Active Badge
- Computer vision: Easy Living
- Physical proximity: Smart Floor
- Not suitable for indoor
 - Does not work
 - Require specialized hardware
 - Scalability



WLAN Location Determination

- Triangulate user location
 - Reference point
 - Quantity proportional to distance
- WLAN
 - Access points
 - Signal strength = $f(\text{distance})$
- Software based

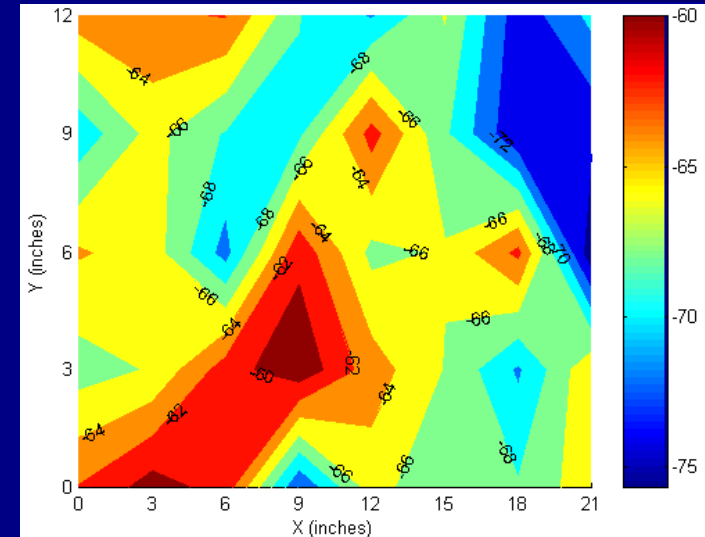
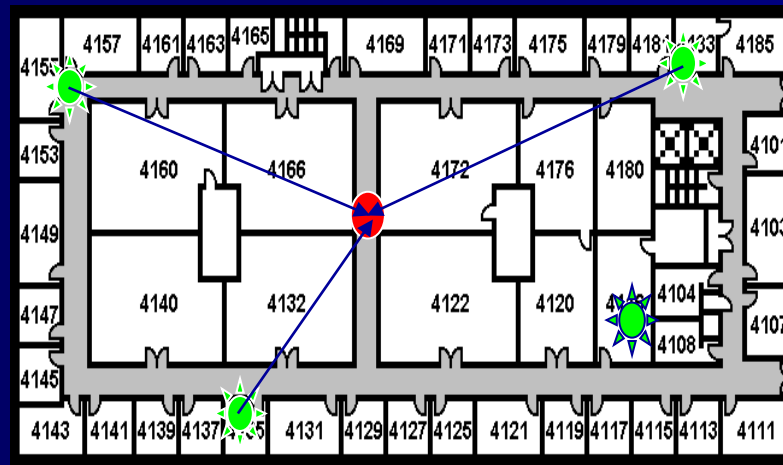


Roadmap

- ✓ Motivation
- ✓ Location determination technologies
- **Introduction**
 - Noisy wireless channel
 - Horus components
 - Performance evaluation
 - Conclusions and future work



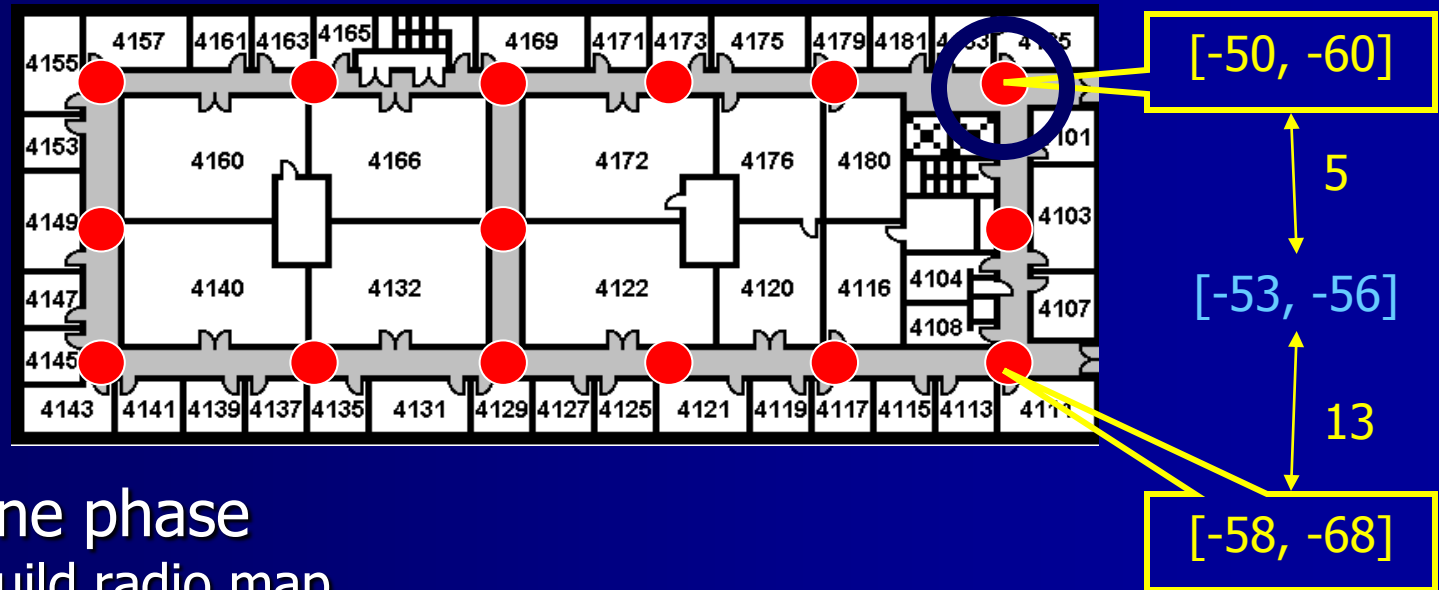
WLAN Location Determination (Cont'd)



- Signal strength = $f(\text{distance})$
- Does not follow free space loss
- Use lookup table $\equiv$ **Radio map**
- Radio Map: signal strength characteristics at selected locations



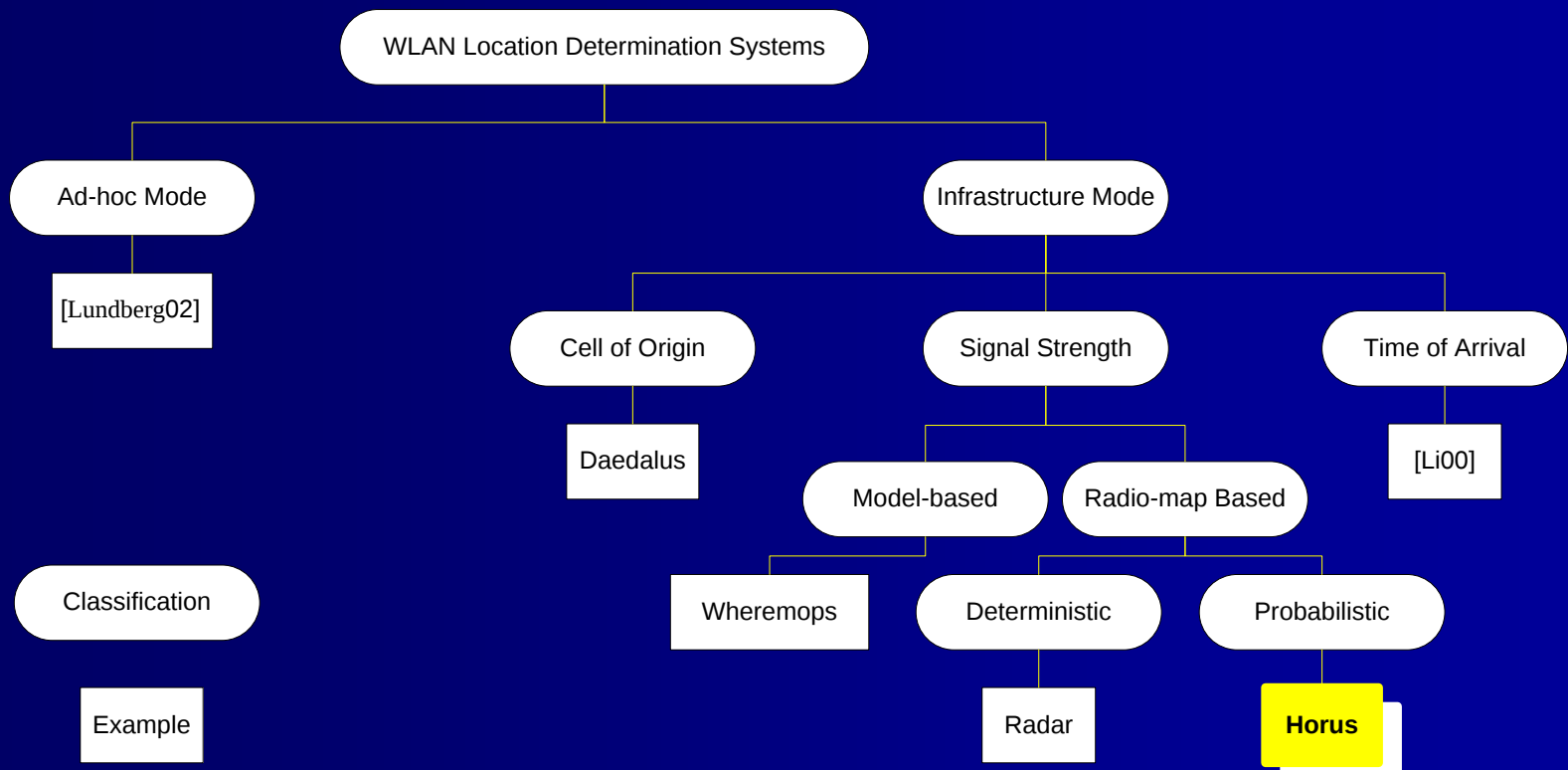
WLAN Location Determination (Cont'd)



- Offline phase
 - Build radio map
 - Radar system: average signal strength
- Online phase
 - Get user location
 - Nearest location in signal strength space (Euclidian distance)



WLAN Location Determination Taxonomy



Horus Goals

- High accuracy
 - Wider range of applications
- Energy efficiency
 - Energy constrained devices
- Scalability
 - Number of supported users
 - Coverage area



Contributions

- Taxonomy of WLAN location determination systems
- Modeling the signal strength distributions using parametric and non-parametric distributions
- Handling correlation between successive samples from the same access point
- Allowing continuous space estimation
- Clustering of radio map locations
- Handling small-scale variations
- Compare the performance of the *Horus* system with other systems



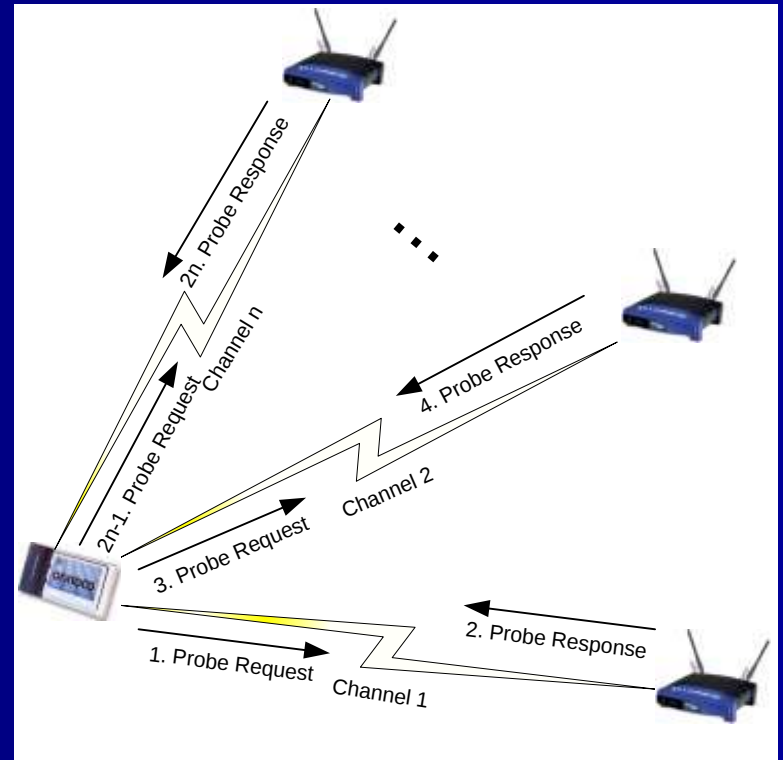
Roadio-map

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Sampling Process

- Active scanning
 - Send a probe request
 - Receive a probe response

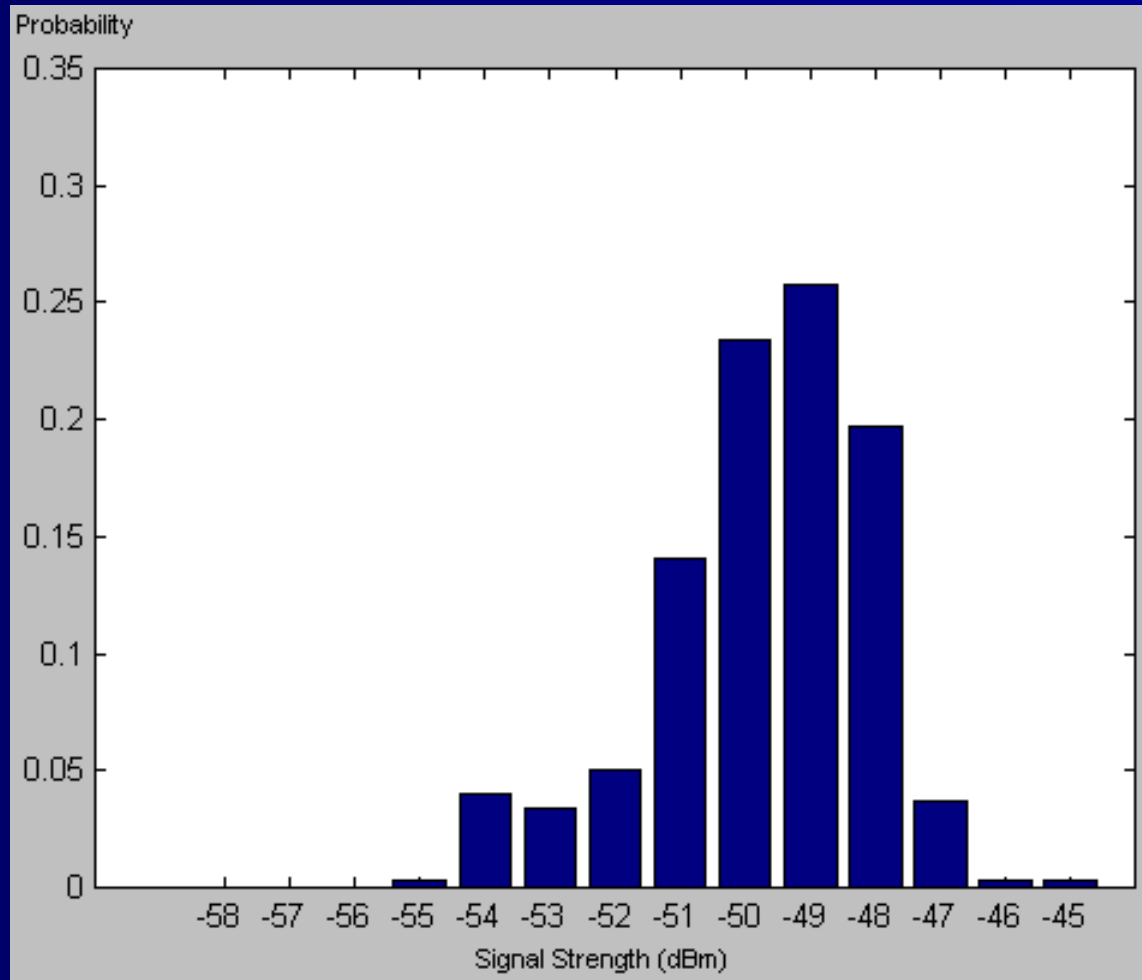


Signal Strength Characteristics

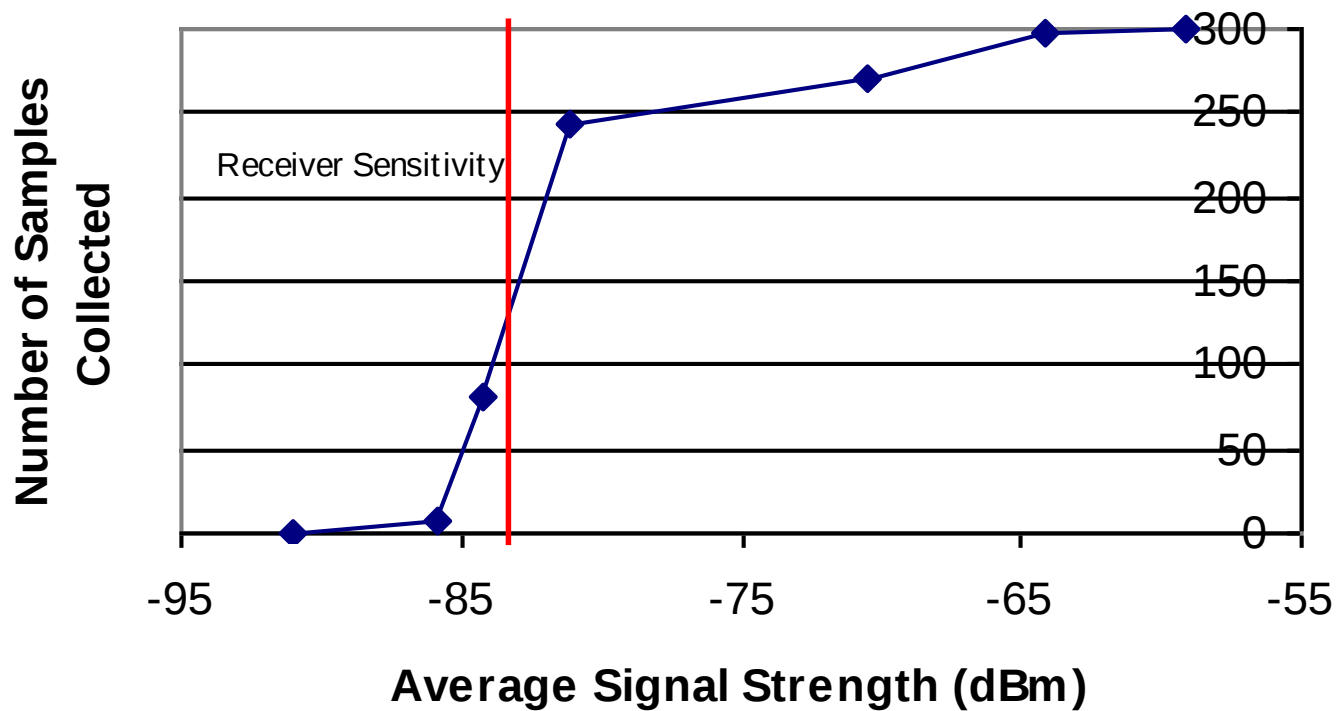
- Temporal variations
 - One access point
 - Multiple access points
- Spatial variations
 - Large scale
 - Small scale



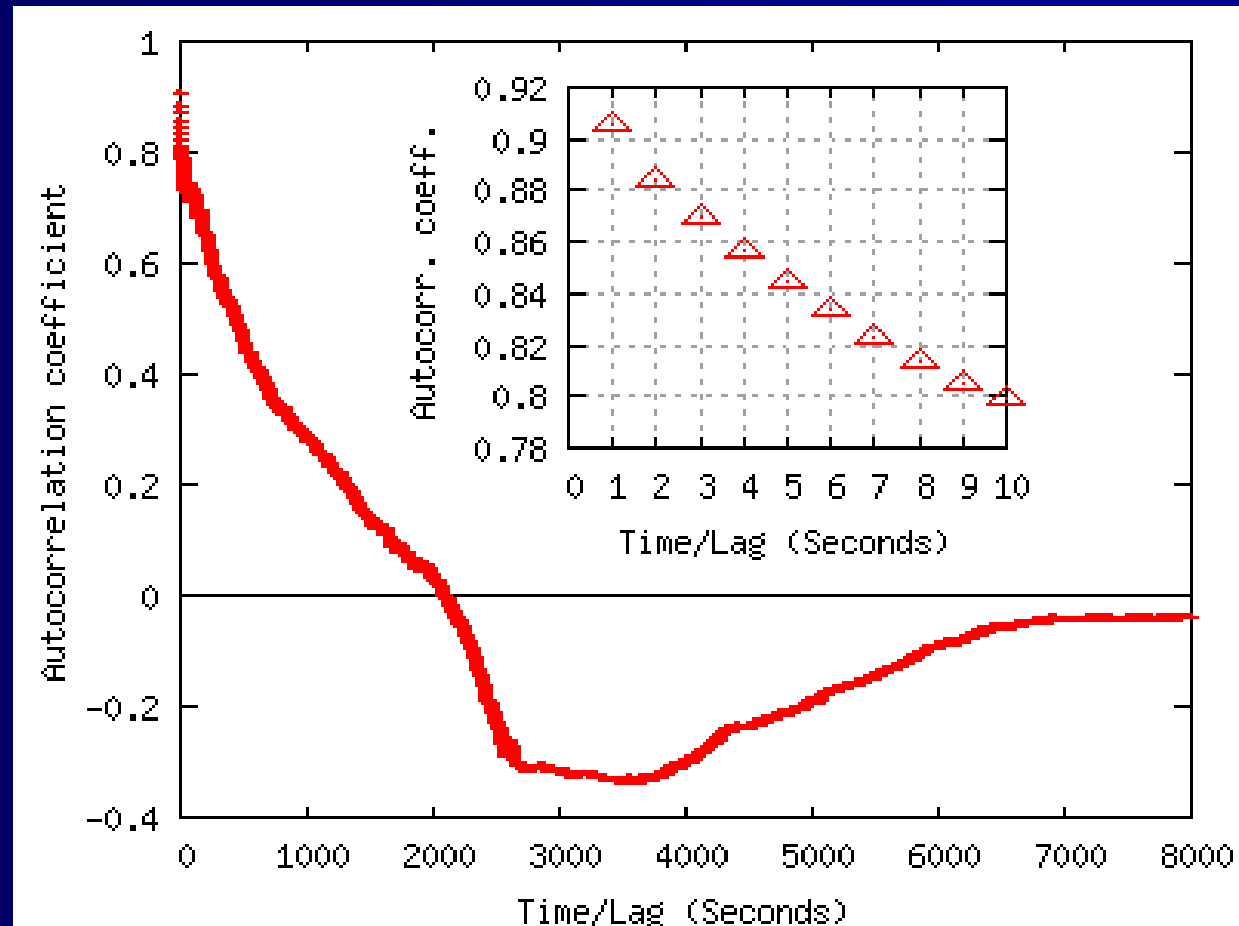
Temporal Variations



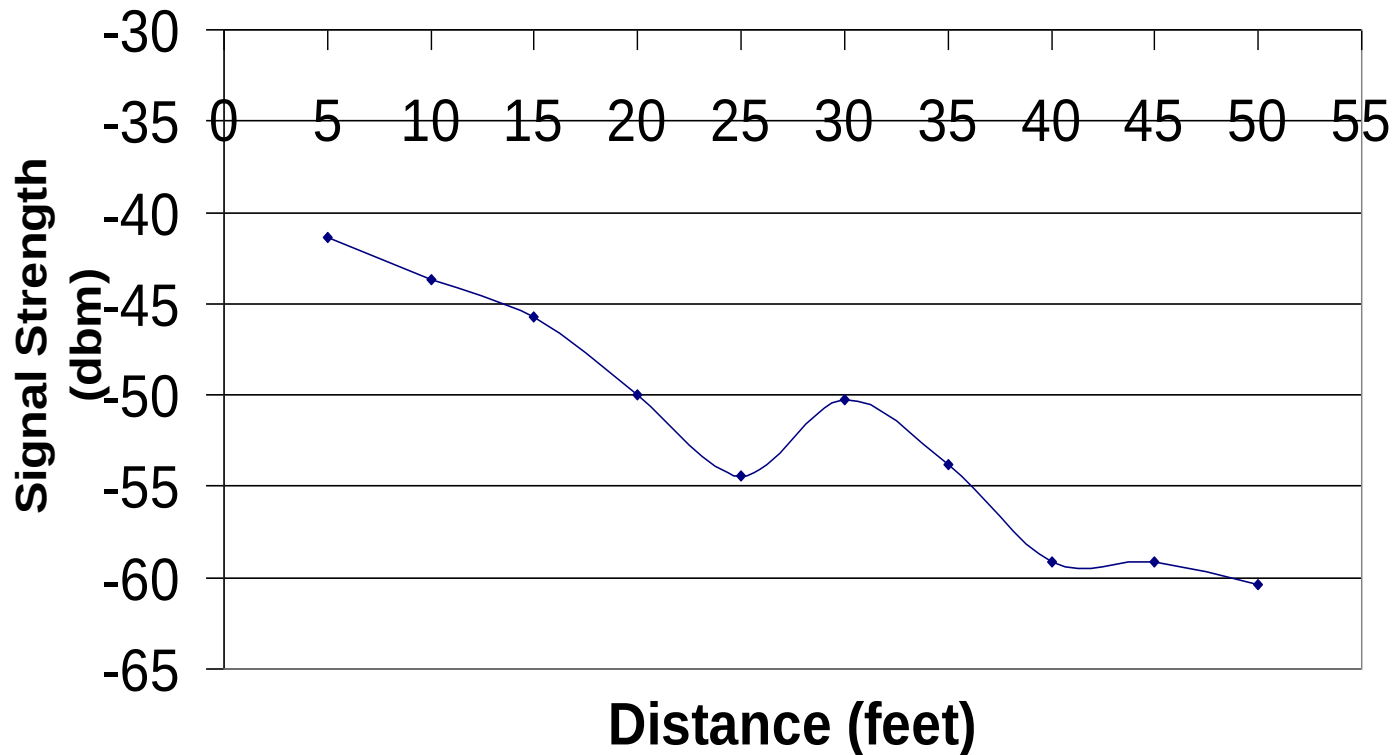
Temporal Variations



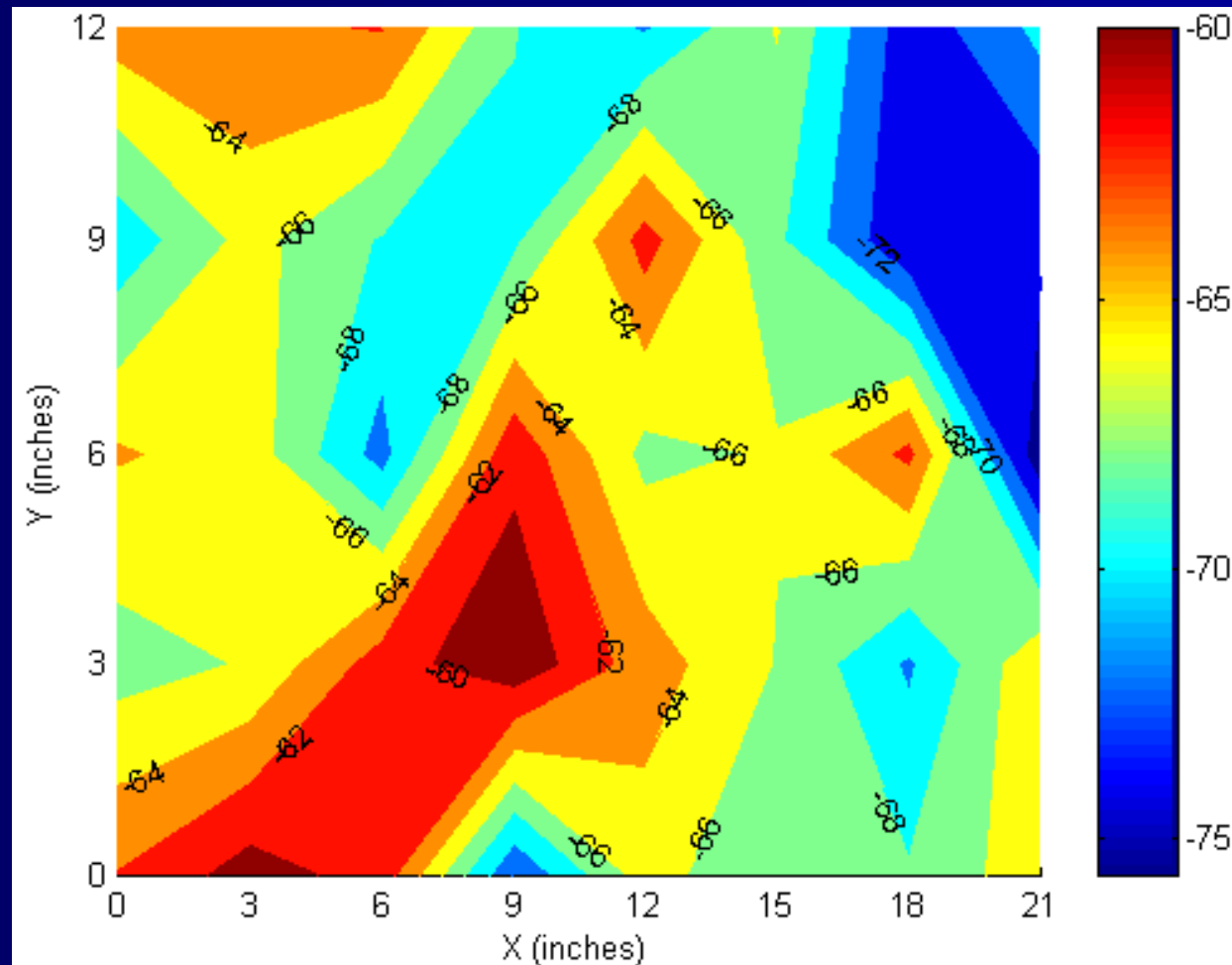
Temporal Variations: Correlation



Spatial Variations: Large-Scale



Spatial Variations: Small-Scale



Roadio-map

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Testbeds

■ A.V. William's

- 4th floor, AVW
- 224 feet by 85.1 feet
- UMD net (*Cisco* APs)
- 21 APs (6 on avg.)
- 172 locations
- 5 feet apart
- *Windows XP Prof.*

■ FLA

- 3rd floor, 8400 Baltimore Ave
- 39 feet by 118 feet
- LinkSys/*Cisco* APs
- 6 APs (4 on avg.)
- 110 locations
- 7 feet apart
- Linux (kernel 2.5.7)

Orinoco/Compaq cards



Horus Components

- Basic algorithm [Percom03]
- Correlation handler [InfoCom04]
- Continuous space estimator [Under]
- Locations clustering [Percom03]
- Small-scale compensator [WCNC03]

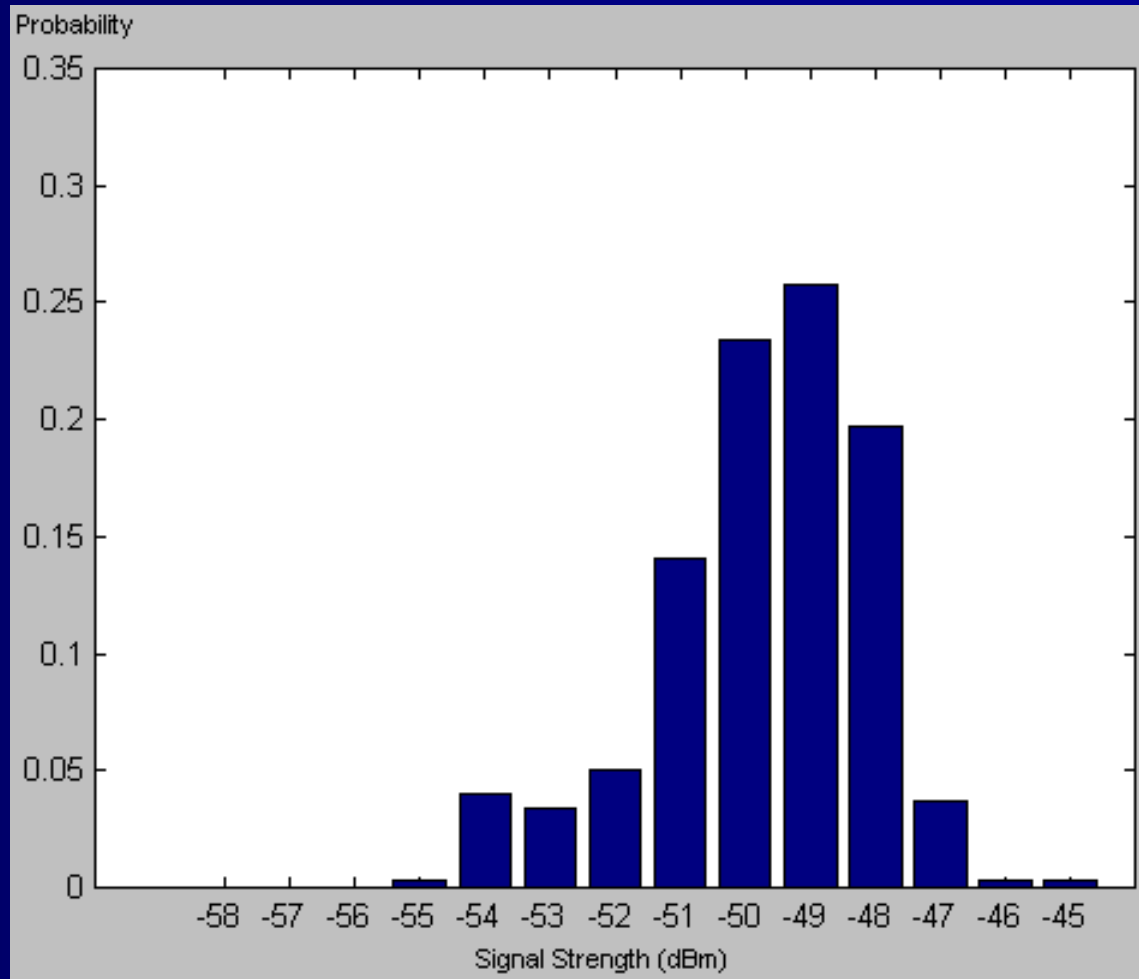


Basic Algorithm: Mathematical Formulation

- x : Position vector
- s : Signal strength vector
 - One entry for each access point
- $s(x)$ is a stochastic process
- $P[s(x), t]$: probability of receiving s at x at time t
- $s(x)$ is a stationary process
 - $P[s(x)]$ is the histogram of signal strength at x



Basic Algorithm: Mathematical Formulation



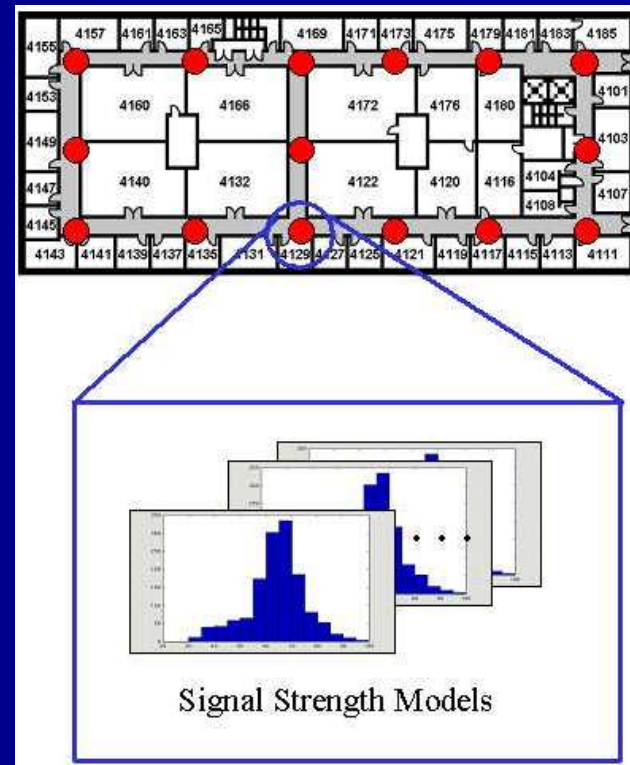
Basic Algorithm: Mathematical Formulation

- $\text{Argmax}_x[P(x/s)]$
- Using Bayesian inversion
 - $\text{Argmax}_x[P(s/x).P(x)/P(s)]$
 - $\text{Argmax}_x[P(s/x).P(x)]$
- $P(x)$: User history

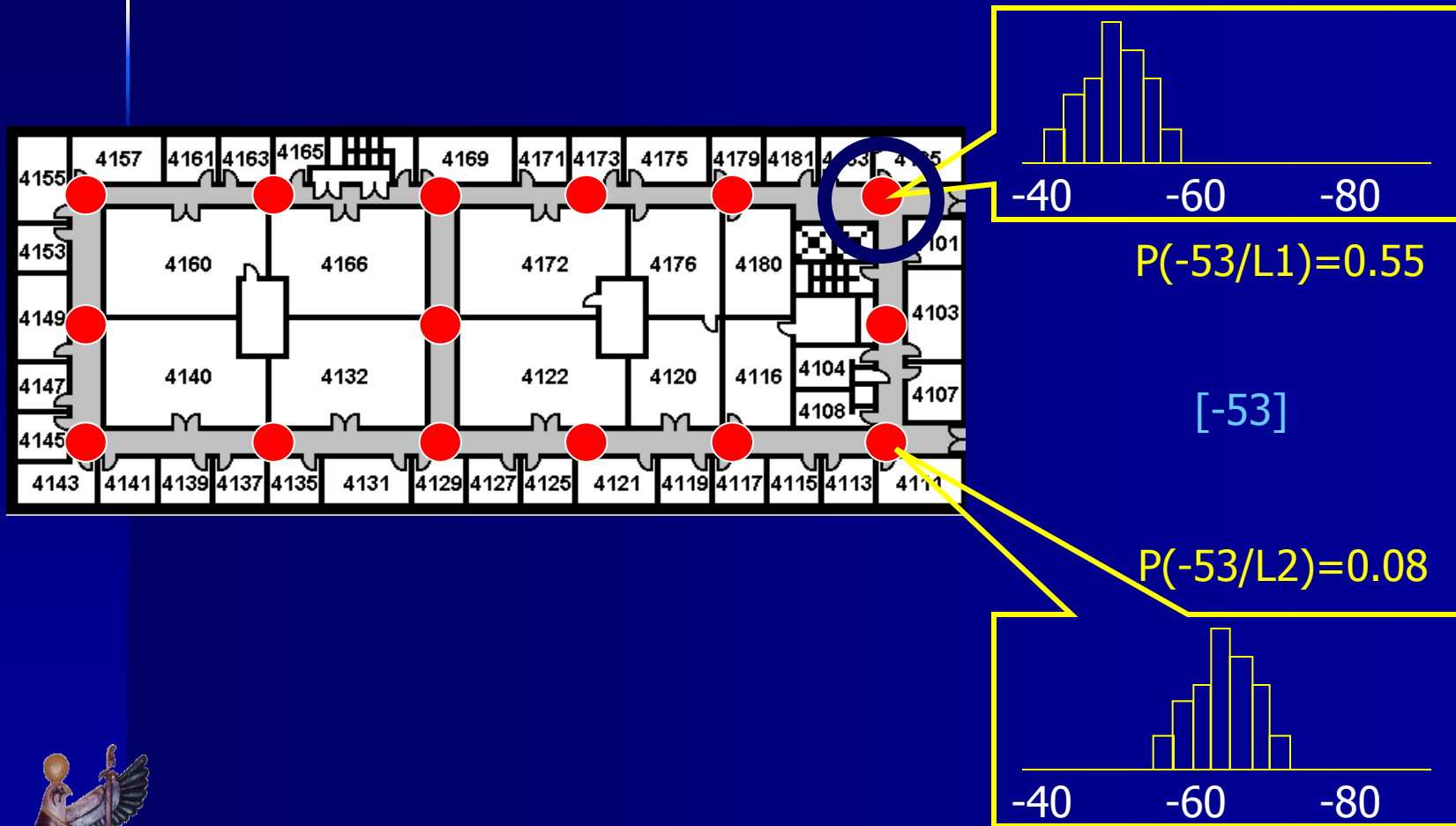


Basic Algorithm

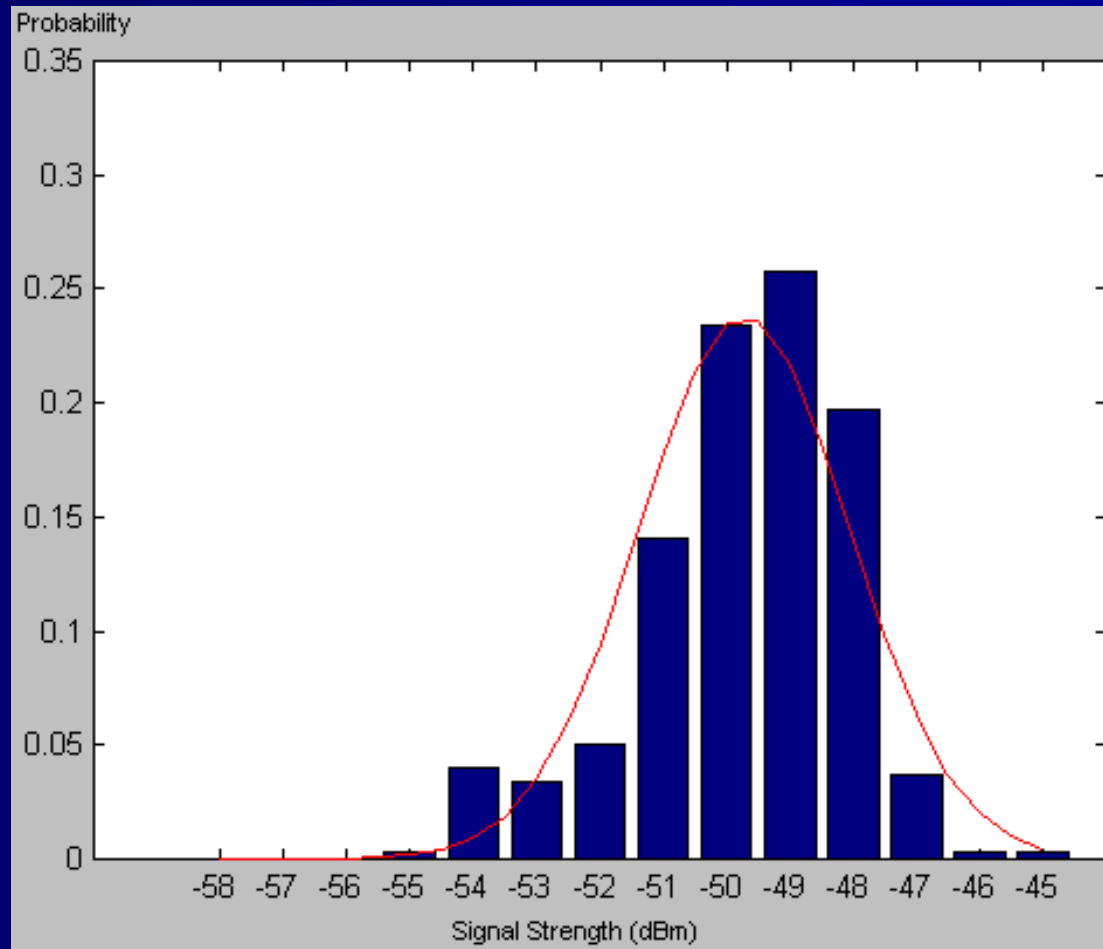
- Offline phase
 - Radio map: signal strength histograms
- Online phase
 - Bayesian based inference



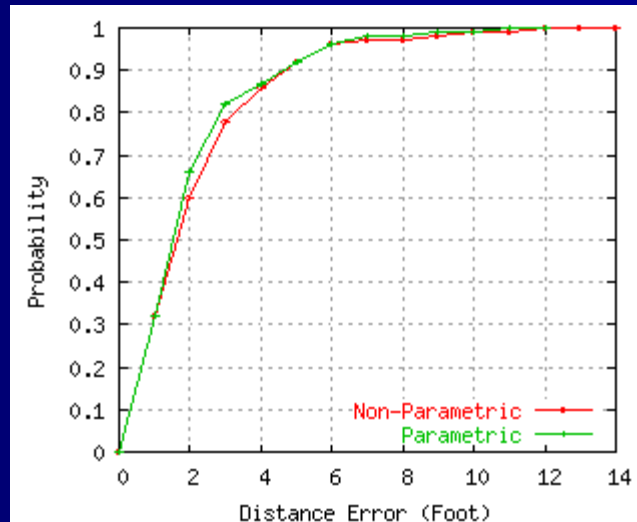
WLAN Location Determination (Cont'd)



Basic Algorithm: Signal Strength Distributions



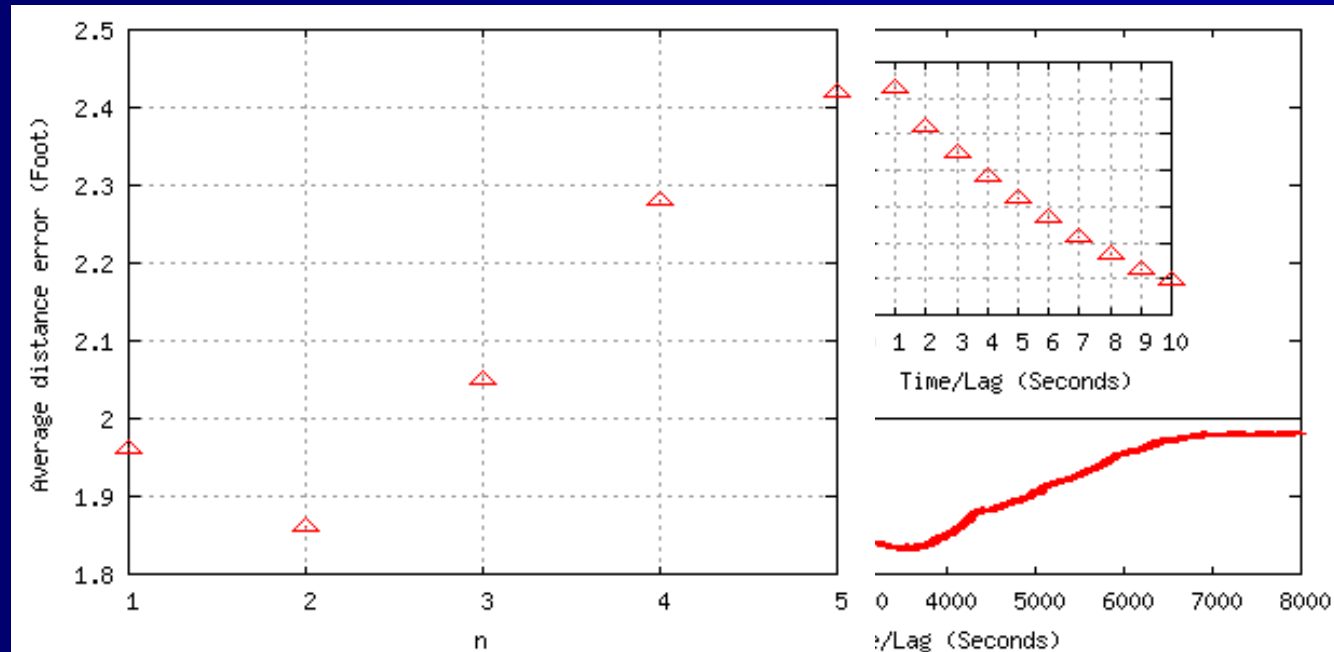
Basic Algorithm: Results



- Accuracy of 5 feet 90% of the time
- Slight advantage of parametric over non-parametric method
 - Smoothing of distribution shape



Correlation Handler



- Need to average multiple samples to increase accuracy
- Independence assumption is wrong



Correlation Handler: Autoregressive Model

- $s(t+1) = \alpha \cdot s(t) + (1 - \alpha) \cdot v(t)$
- α : correlation degree
- $E[v(t)] = E[s(t)]$
- $\text{Var}[v(t)] = (1 + \alpha) / (1 - \alpha) \text{Var}[s(t)]$

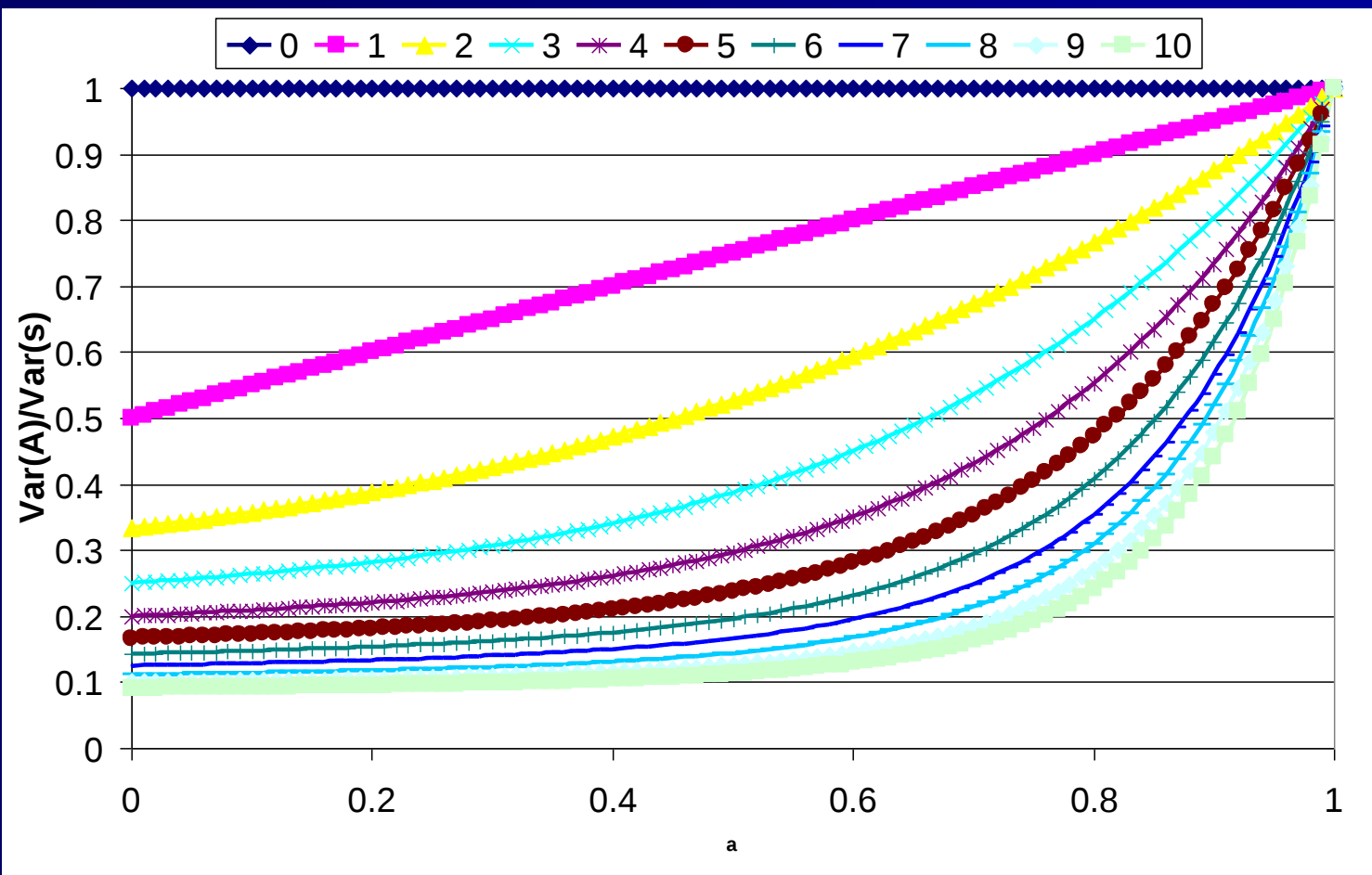


Correlation Handler: Averaging Process

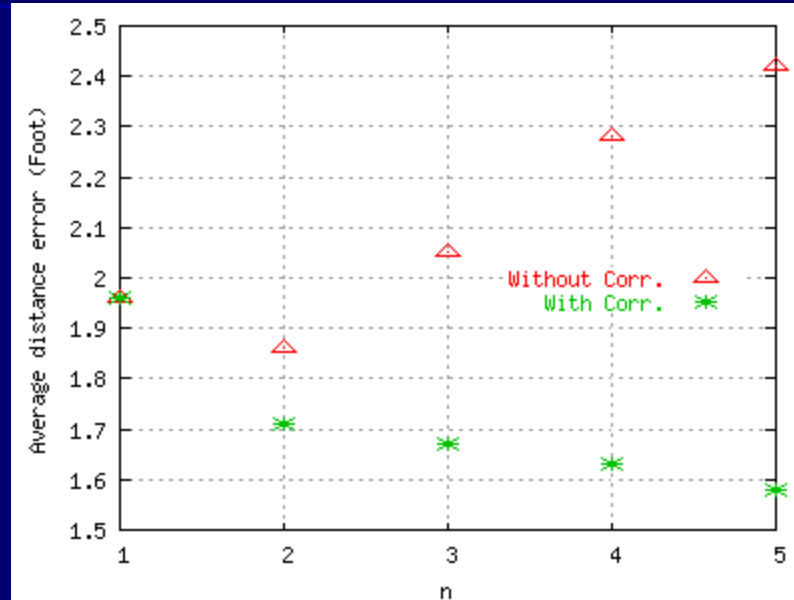
- $s(t+1) = \alpha \cdot s(t) + (1 - \alpha) \cdot v(t)$
- $s \sim N(0, m)$
- $v \sim N(0, r)$
- $A = 1/n (s_1 + s_2 + \dots + s_n)$
- $E[A(t)] = E[s(t)] = 0$
- $\text{Var}[A(t)] = m^2/n^2 \left\{ \left[\frac{(1 - \alpha^n)}{(1 - \alpha)} \right]^2 + \frac{n + 1 - \alpha^2 * (1 - \alpha^{2(n-1)})}{(1 - \alpha^2)} \right\}$



Correlation Handler: Averaging



Correlation Handler: Results



- Independence assumption: performance degrades as n increases
- Two factors affecting accuracy
 - Increasing n
 - Deviation from the actual distribution



Continuous Space Estimator

- Enhance the discrete radio map space estimator
- Two techniques
 - Center of mass of the top ranked locations

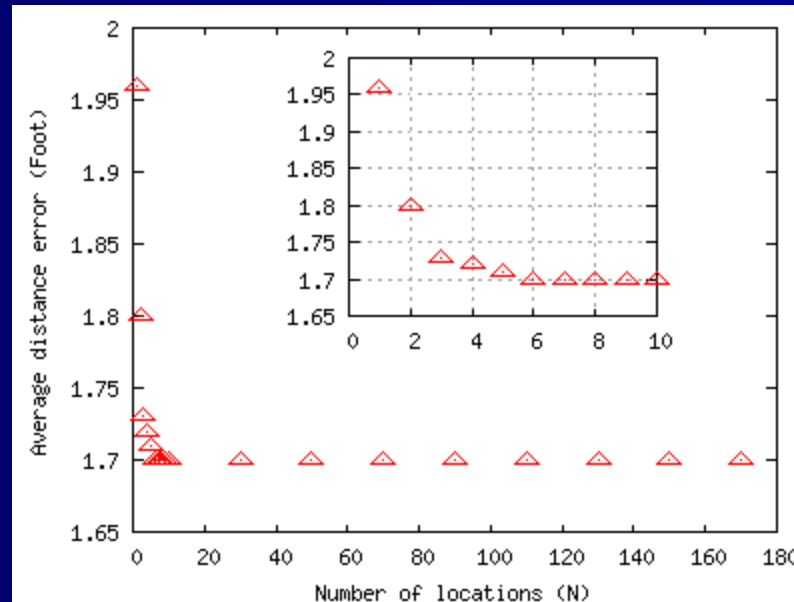
$$x \leftarrow \frac{\sum_{i=1}^{\min(N, \|\bar{X}\|)} p(i) * \bar{X}(i)}{\sum_{i=1}^{\min(N, \|\bar{X}\|)} p(i)}$$

- Time averaging window

$$\bar{x}_t = \frac{1}{\min(W, t)} \cdot \sum_{t-\min(W, t)+1}^t x_i$$



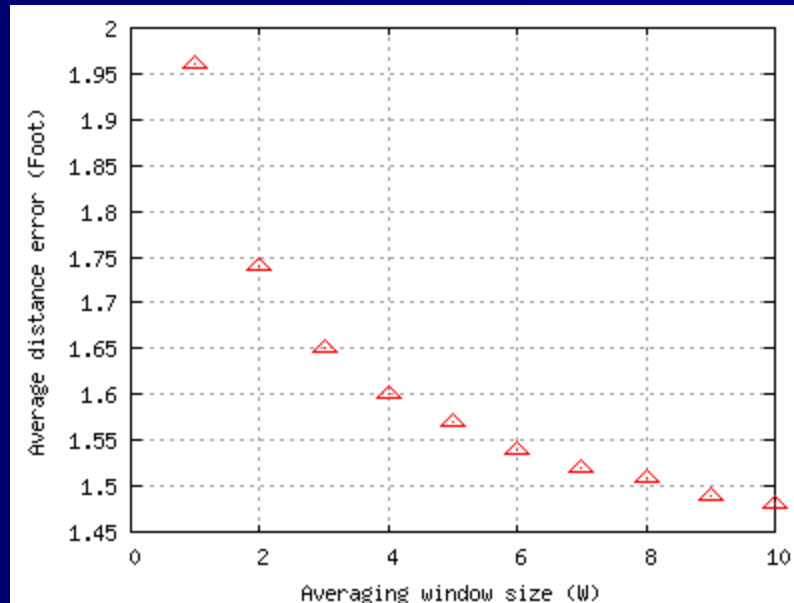
Center of Mass: Results



- $N = 1$ is the discrete-space estimator
- Accuracy enhanced by more than 13%



Time Averaging Window: Results



- $N = 1$ is the discrete-space estimator
- Accuracy enhanced by more than 24%

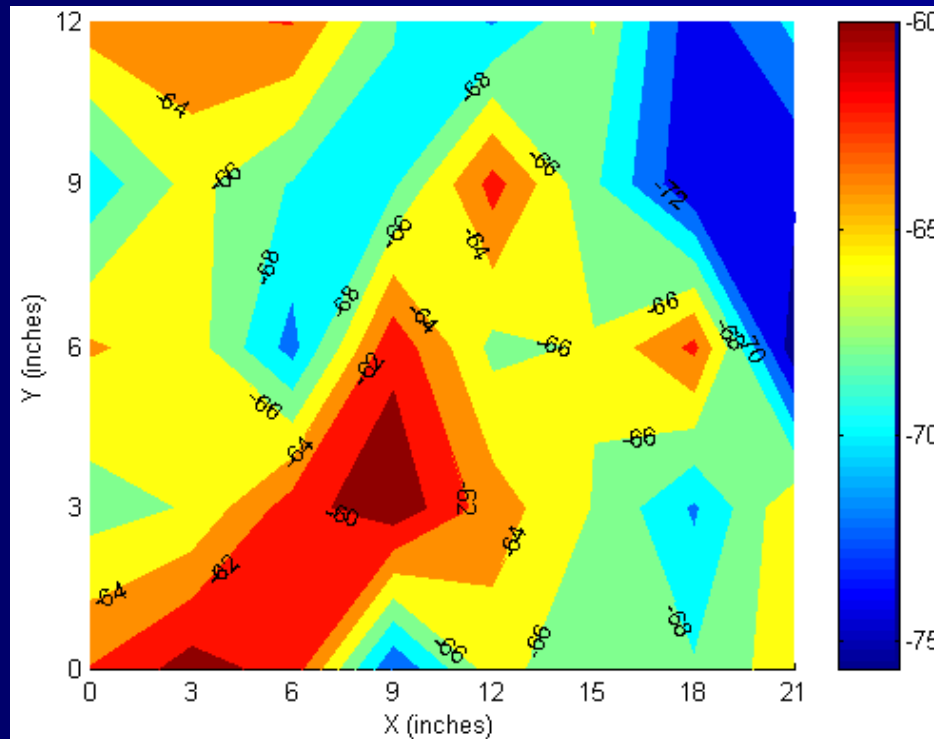


Horus Components

- Basic algorithm
- Correlation handler
- Continuous space estimator
- Small-scale compensator
- Locations clustering



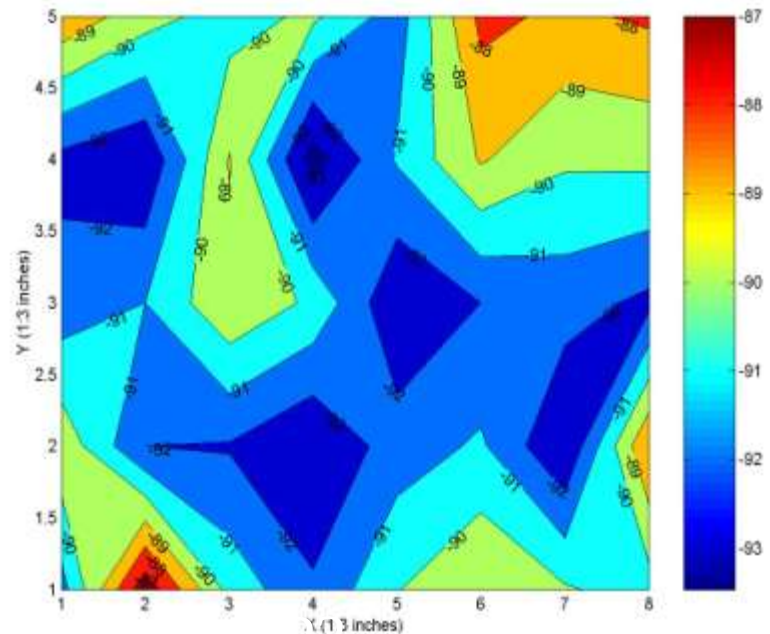
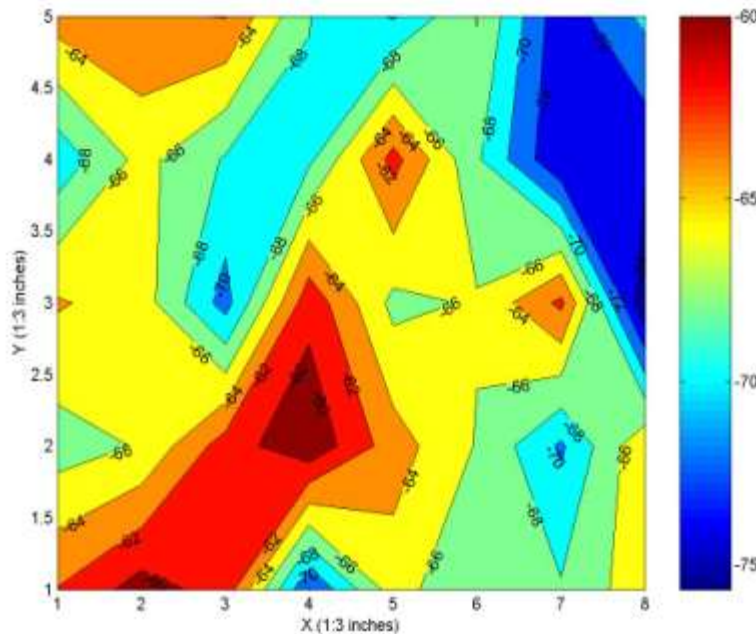
Small-scale Compensator



- Multi-path effect
- Hard to capture by radio map (size/time)



Small-scale Compensator: Small-scale Variations



- Variations up to 10 dBm in 3 inches
- Variations proportional to average signal strength

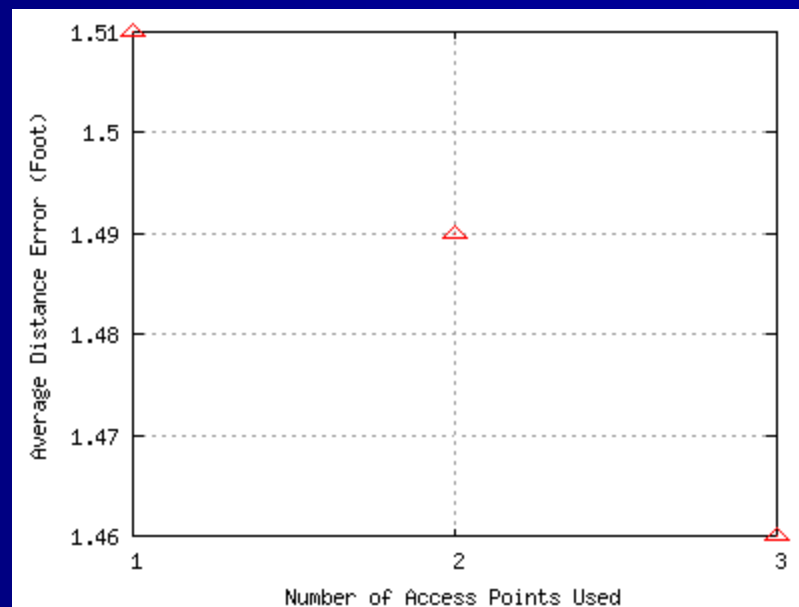
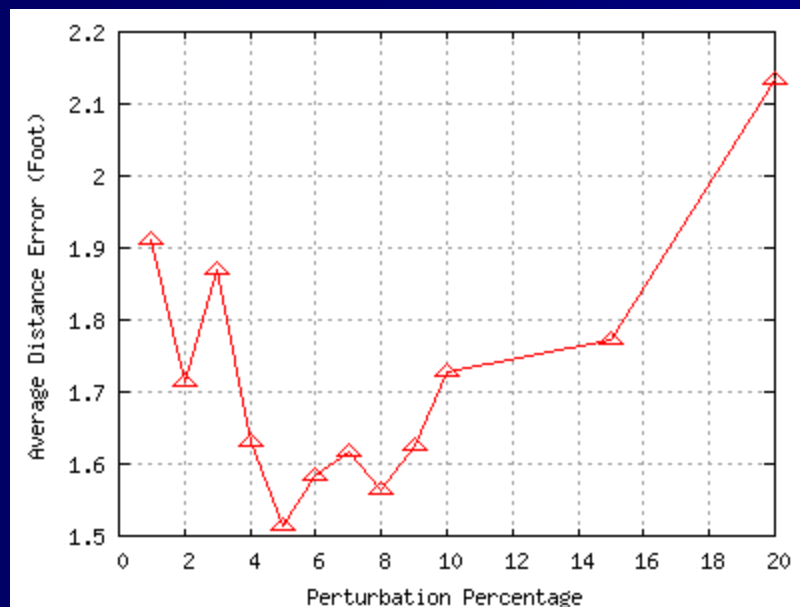


Small-scale Compensator: Perturbation Technique

- Detect small-scale variations
 - Using previous user location
- Perturb signal strength vector
 - $(s_1, s_2, \dots, s_n) \rightarrow (s_1 \pm d_1, s_2 \pm d_2, \dots, s_n \pm d_n)$
 - Typically, $n=3-4$
- d_i is chosen relative to the received signal strength



Small-scale Compensator: Results



- Perturbation technique is not sensitive to the number of APs perturbed
- Better by more than 25%



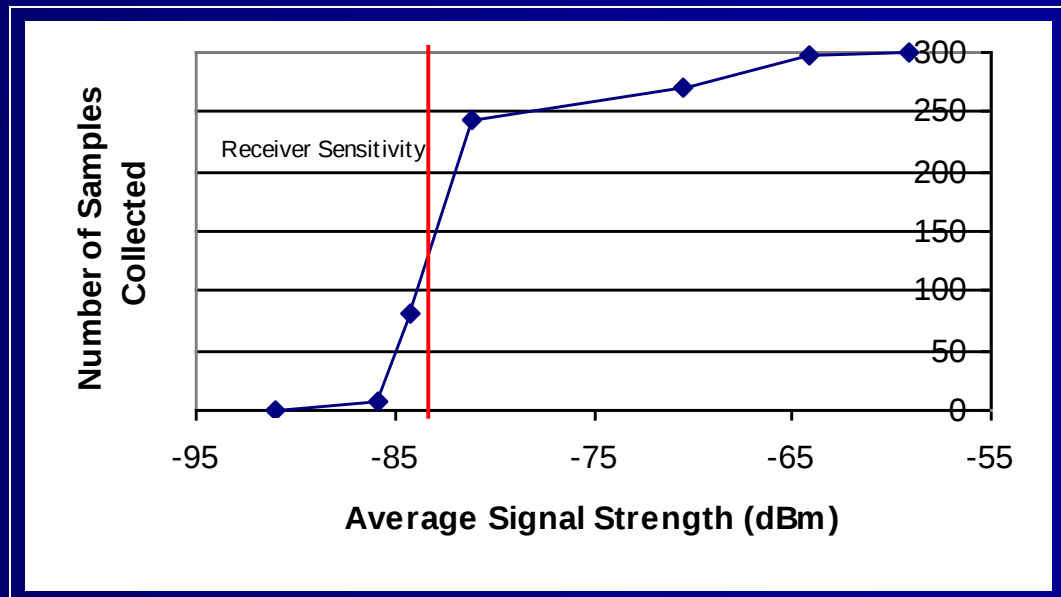
Horus Components

- Basic algorithm
- Correlation handler
- Continuous space estimator
- Small-scale compensator
- Locations clustering



Locations Clustering

- Reduce computational requirements
- Two techniques
 - Explicit
 - Implicit



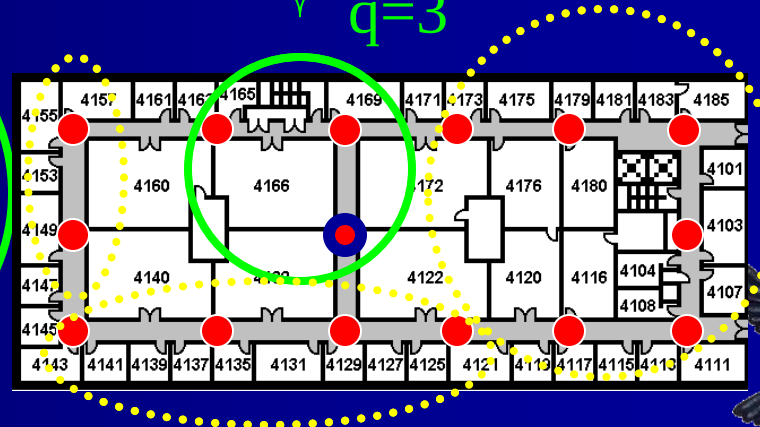
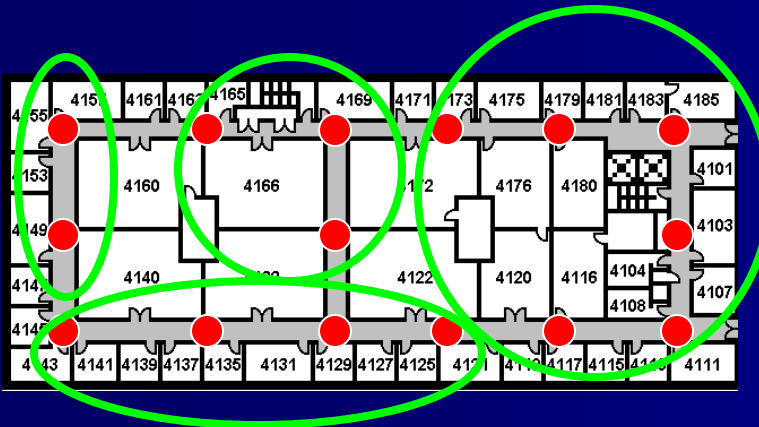
Locations Clustering: Explicit Clustering

- Use access points that cover each location
- Use the q strongest access points

$S = [-60, -45, -80, -86, -70]$

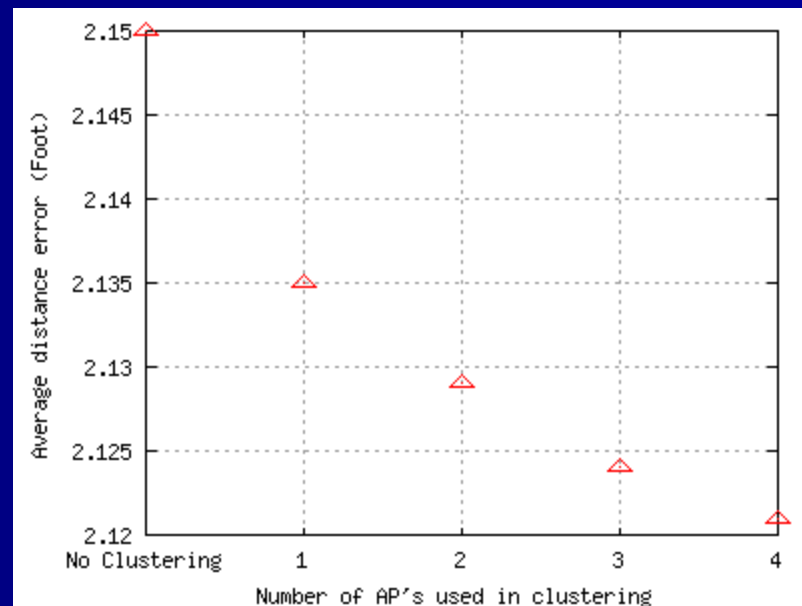
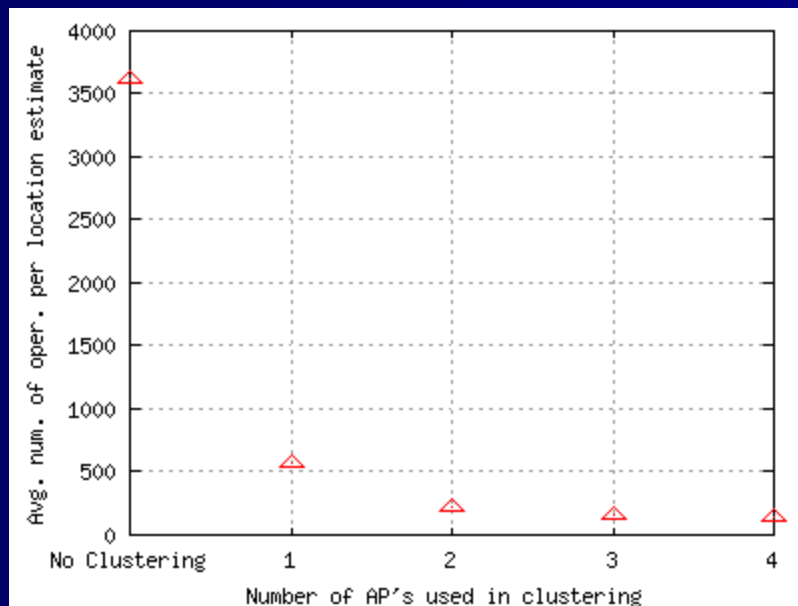
$S = [-45, -60, -70, -80, -86]$

$q=3$



Locations Clustering:

Results- Explicit Clustering



- An order of magnitude enhancement in avg. num. of oper. /location estimate
- As q increases, accuracy slightly increases

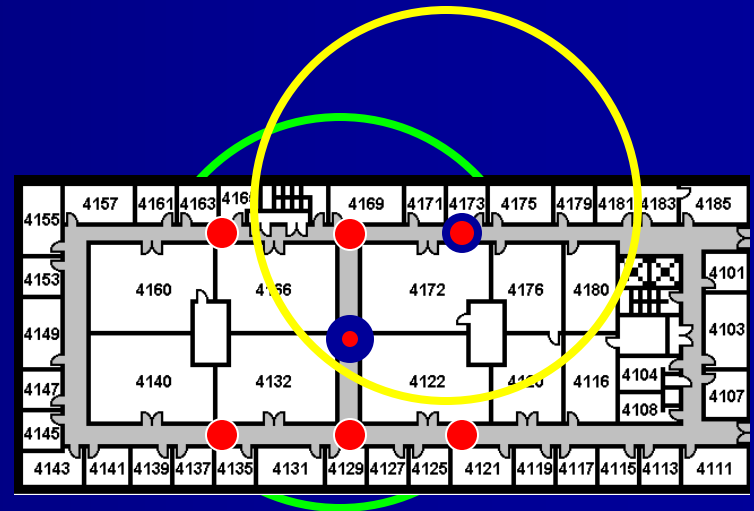


Locations Clustering: Implicit Clustering

- Use the access points incrementally
- Implicit multi-level clustering

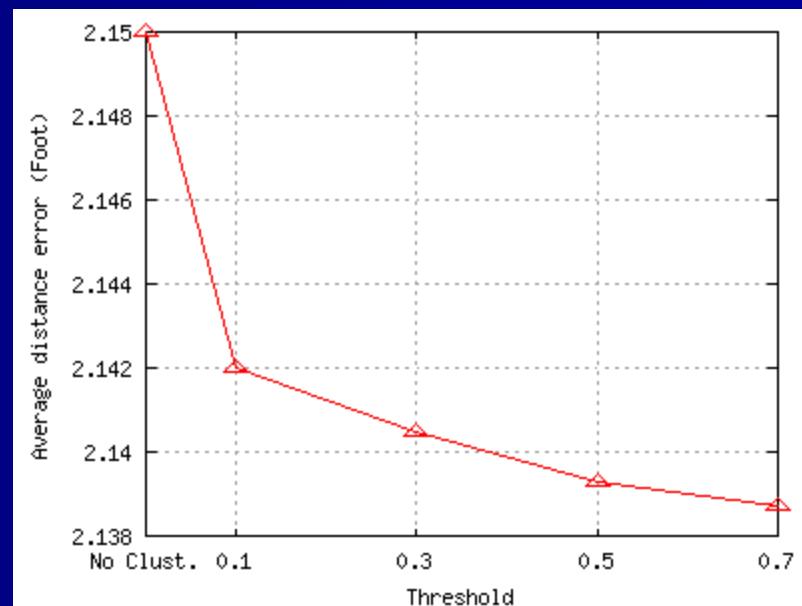
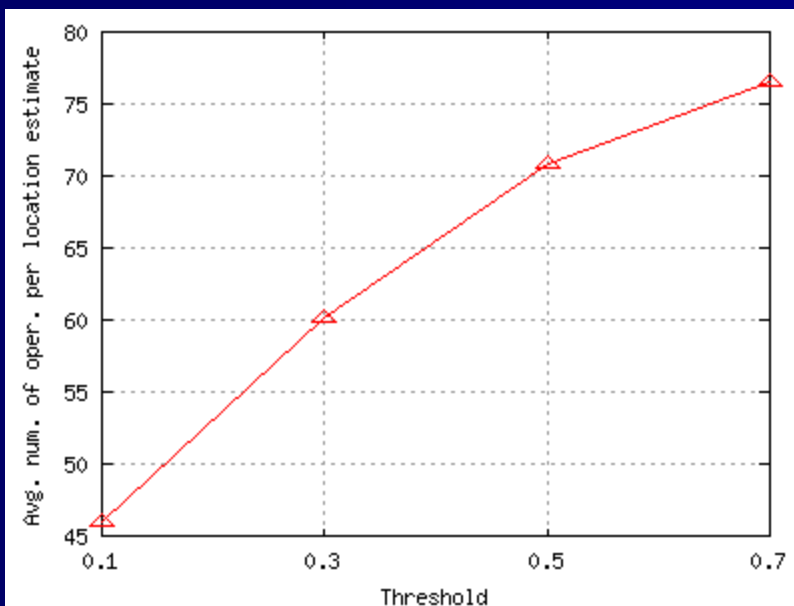
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$S = [-45, -60, -70, -80, -86]$



Locations Clustering:

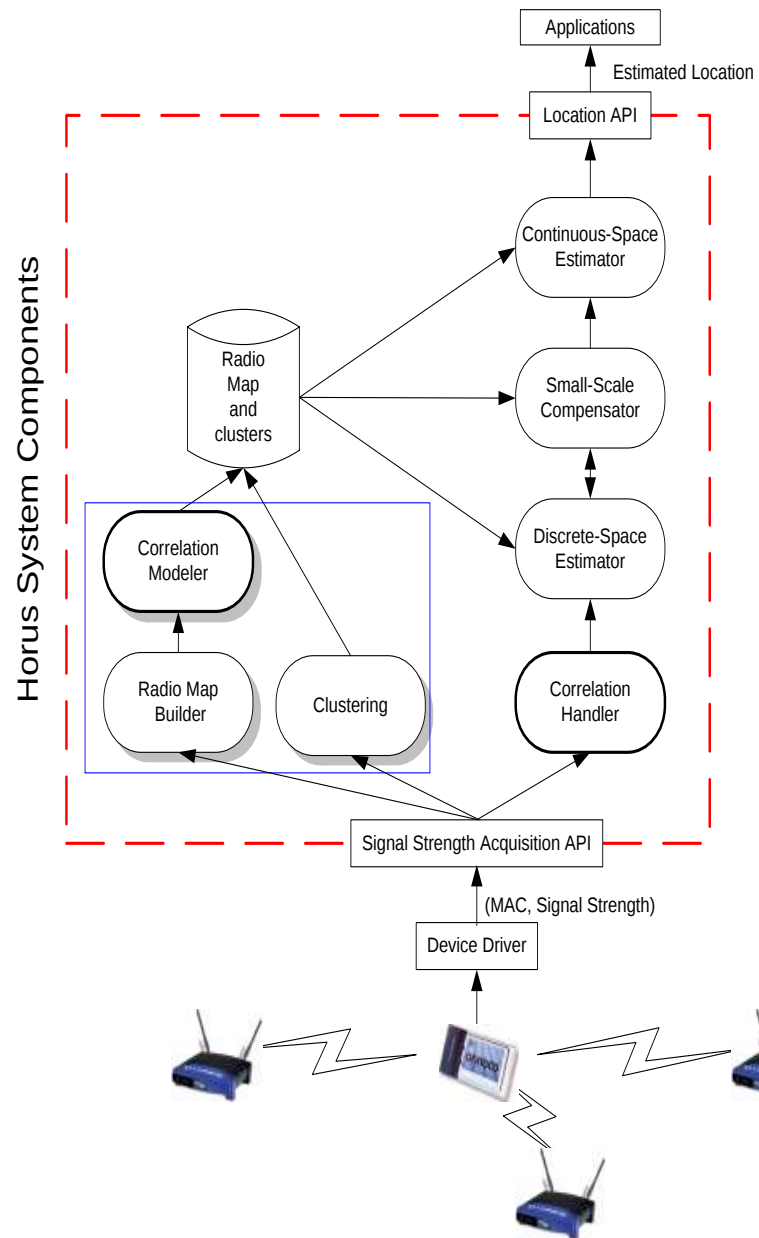
Results- Implicit Clustering



- Avg. num. of oper. /location estimate better than explicit clustering
- Accuracy increases with Threshold



Horus



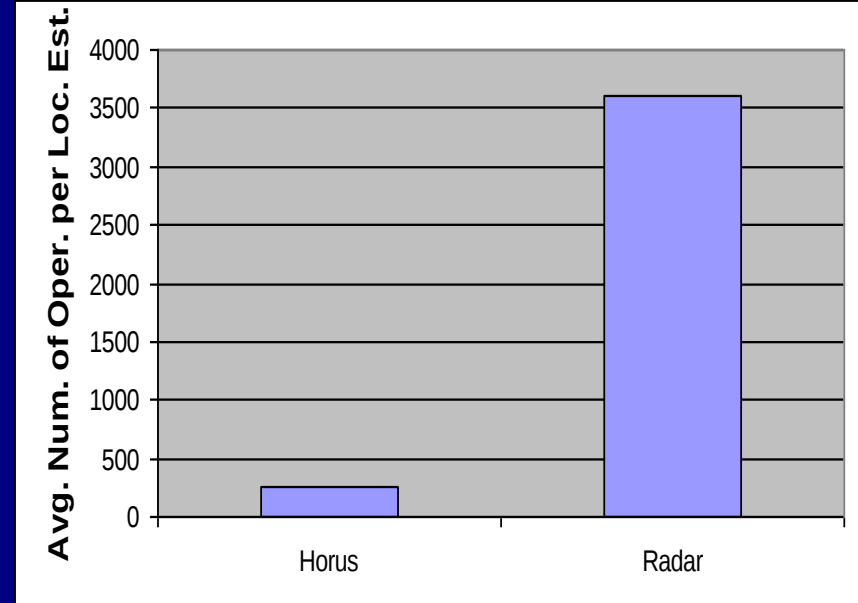
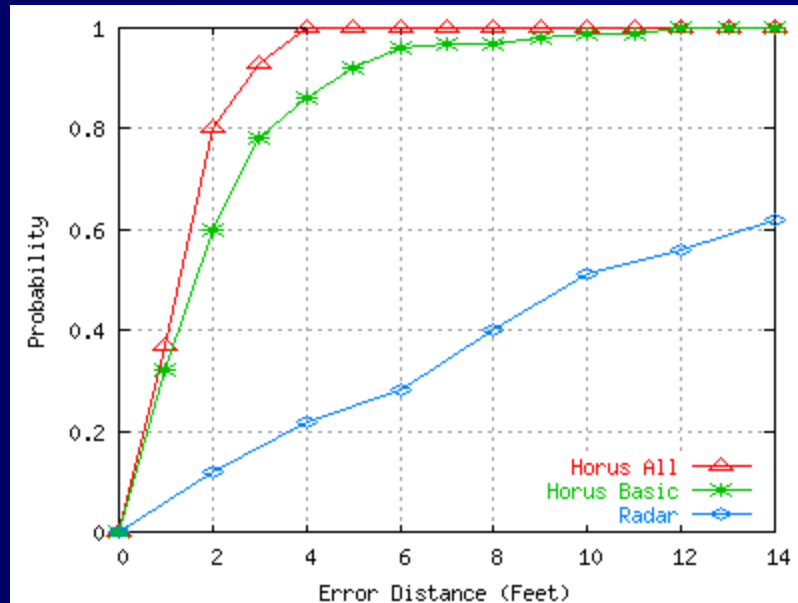
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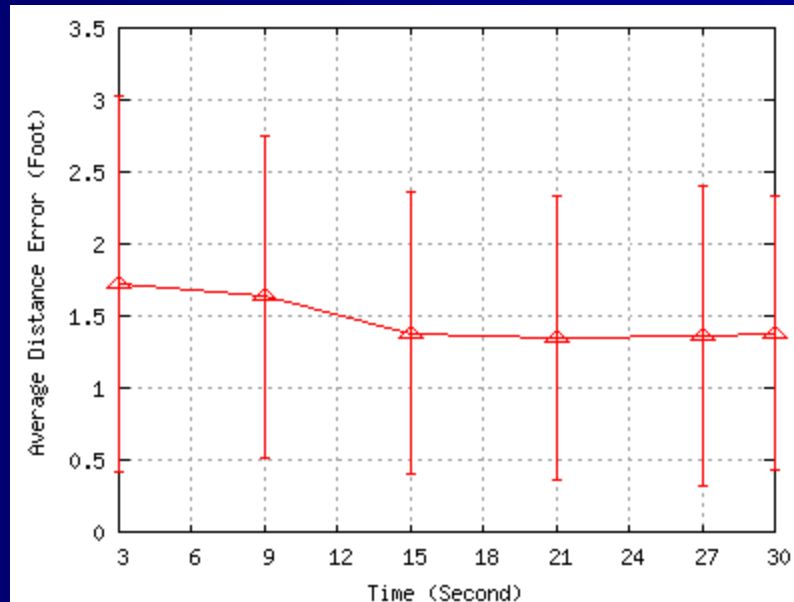
Horus-Radar Comparison



	Median	Avg	Stdev	Max
Horus (all components)	1.28	1.38	0.95	4
Horus (basic)	1.6	2.16	2.09	18.08
Radar	9.74	13.15	10.71	57.67



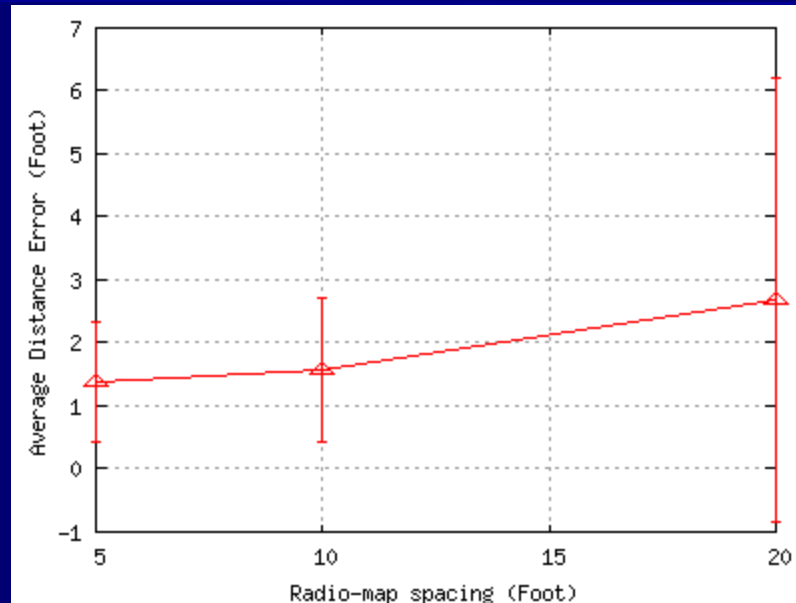
Training Time



- 15 seconds training time per location



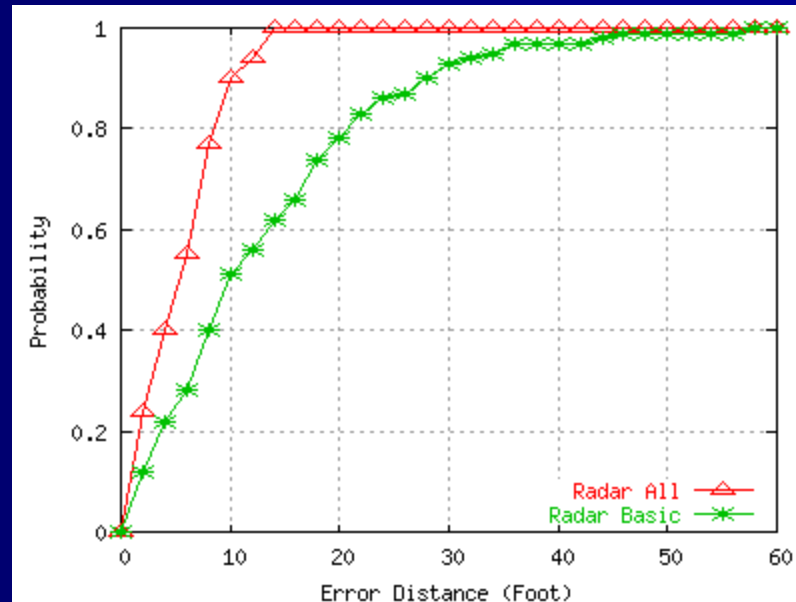
Radio map Spacing



- Average distance error increase by as much as 100% (20 feet)
- 14 feet gives good accuracy



Radar with Horus Techniques



- Average distance error enhanced by more than 58%
- Worst case error decreased by more than 76%



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Conclusions

- The *Horus* system achieves its goals
- High accuracy
 - Through a probabilistic location determination technique
 - Smoothing signal strength distributions by Gaussian approximation
 - Using a continuous-space estimator
 - Handling the high correlation between samples from the same access point
 - The perturbation technique to handle small-scale variations
- Low computational requirements
 - Through the use of clustering techniques



Conclusions (Cont'd)

- Scalability in terms of the coverage area
 - Through the use of clustering techniques
- Scalability in terms of the number of users
 - Through the distributed implementation
- Training time of 15 seconds per location is enough to construct the radio-map
- Radio map spacing of 14 feet
- Horus vs. Radar
 - More accurate by more than 11 feet, on the average
 - More than an order of magnitude savings in number of operations required per location estimate
- Horus vs. Ekahau



Conclusions (Cont'd)

- Modules can be applied to other WLAN location determination systems
 - Correlation handling, continuous-space estimator, clustering, and small-scale compensator
- Applied to Radar
 - Average distance error enhanced by more than 58%
 - Worst case error decreased by more than 76%
- Techniques presented thesis are applicable to other RF-technologies
 - 802.11a, 802.11g, HiperLAN, and BlueTooth, ...



Future Work

- Using the user history in location estimation and clustering
- Dynamically change the system parameters based on the environment
- Experimenting with other continuous distributions
- Optimal placement of access point to obtain the best accuracy
- Techniques to ensure user privacy



Future Work (Cont'd)

- Different clustering techniques
- Automating the radio-map generation process
- Changing the radio map based on the environment
- Effect of adding/removing access points
- Designing and developing applications and services
- Handling difference between different manufactures

