Enhanced Location Estimation via Pattern Matching and Motion Modeling

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ISART 2004, Boulder

Kplus
Kompetenzzentren-Programm

Outline

- Motivation
- Signal Power Level localization method
- Motion model
- Trial and Results
- Conclusions

Measurement Sites (GPS)

Area Type	Inaccuracy	Availability
“Ring”	<~23meter	~85%
“Downtown”	<~50meter	~57%

Power Level bases Loc. Methods

Localize handset by „unique" RF pattern!

Extract feature vector:

Single position i:

$$\mathbf{f}^i = \left[\begin{array}{cccc} a_0^i & a_1^i & \ldots & a_6^i \end{array}\right]$$

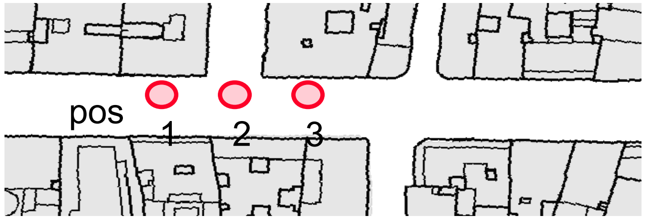
a … Cell ID
Index 0: … Serving cell
Index 1 to 6: … Neighboring cells

Compare with reference data:

Reference data: $\bar{\mathbf{d}}^1, \bar{\mathbf{d}}^2, \ldots, \bar{\mathbf{d}}^K$ with $\bar{\mathbf{d}}^i = [\bar{\mathbf{f}}_1^i, \ldots, \bar{\mathbf{f}}_M^i]$ M samples/pos

$\mathbf{f}^i$ → Location algorithm → Position estimate $\hat{i}$

Method continued



Assign model λ to every position:

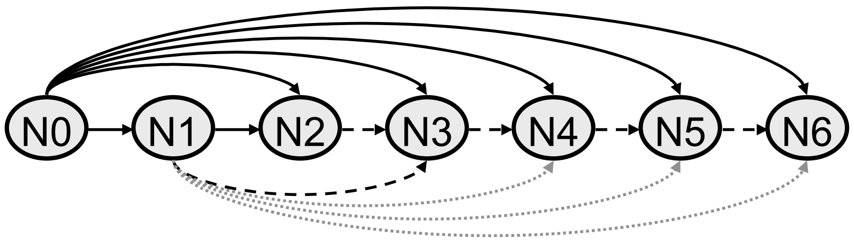
$$\lambda^i \Longleftrightarrow \text{pos } i$$

Model λ^i: represents all reference data d^i at position i

→ use non parametric models

→ Used prior information (of neighboring positions)

→ Use incomplete alphabet of nodes to decrease cardinality

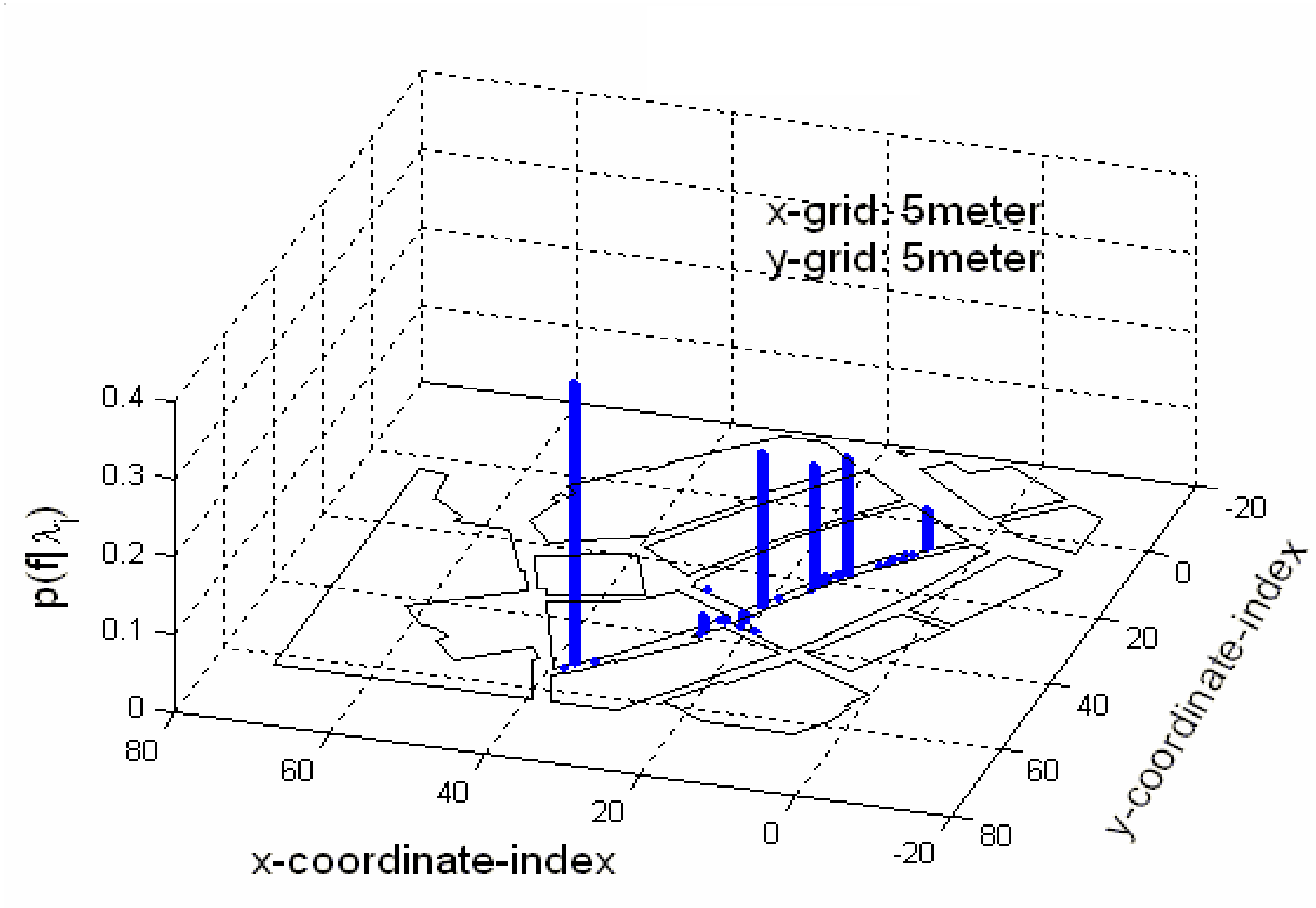


N0: Serving Cell

N1-N6: Neighboring cells

Likelihood of Position i

Localization: Find the model λ^i (and thus i) which emits most likely the measured fingerprint $\mathbf{f}^i$



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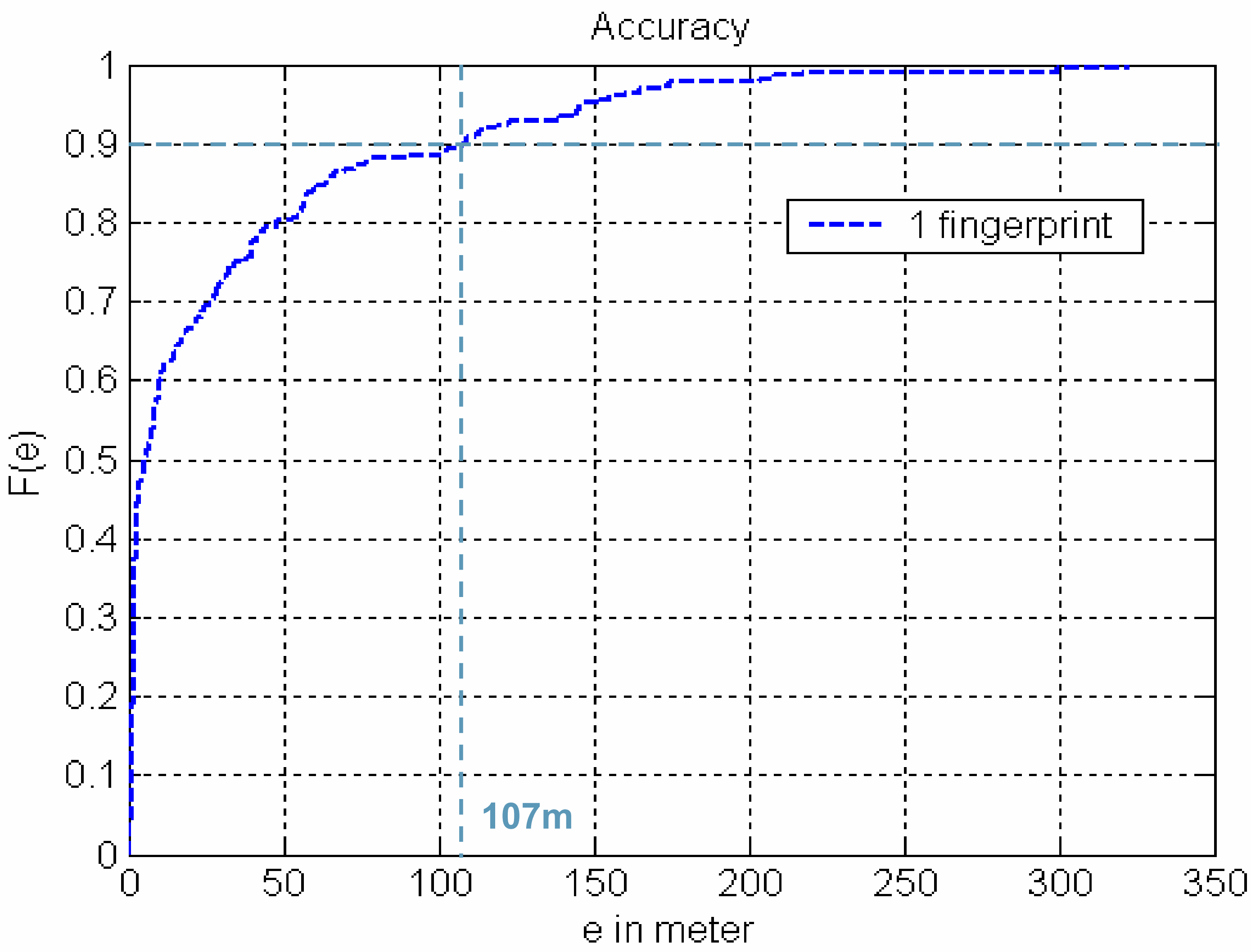
Accuracy in Urban Areas

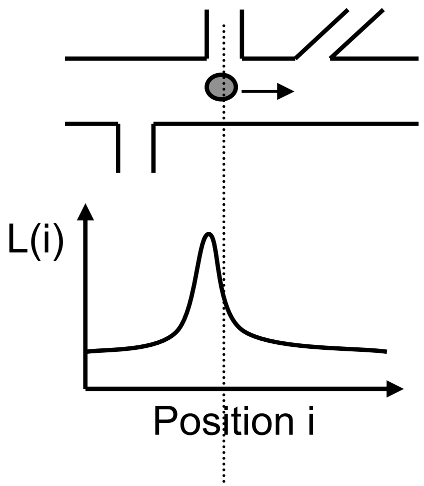


2800 test cases, uniformly distributed within target area

Utilization of Several Fingerprints

Step 1:

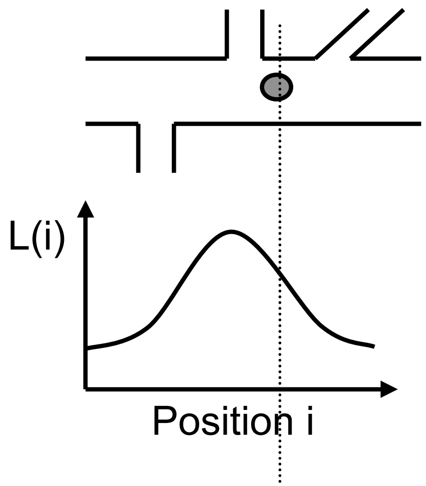
time t = k



First position estimate

Step 2:

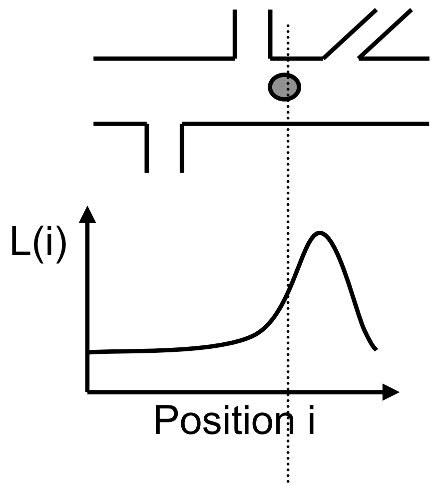
time t = k+1



Propagated position estimate

Step 3:

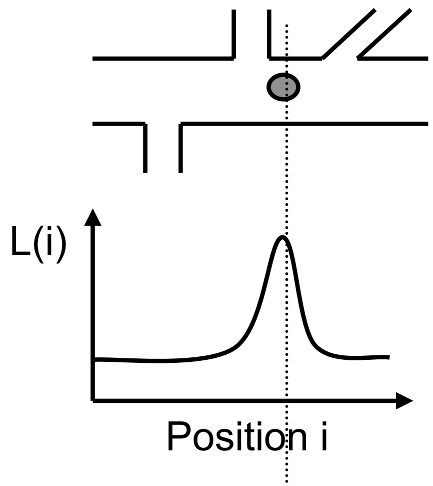
time t = k+1



Second position estimate

Step 4:

time t = k+1



Combined first and second position estimate

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Motion Model (based on [1])

- Constant velocity for the time under consideration
- Obstacles are viewed as perturbations upon the constant velocity
- Users try to reestablish their constant velocity

Use state space model:

$$\begin{aligned} \dot{x}(t) &= v(t) + u(t) \\ \dot{v}(t) &= -\alpha v(t) + w(t) \end{aligned}$$

$x(t)$ ... covered distance

$v(t)$ ... user's velocity variation

$u(t)$ ... user's desired velocity

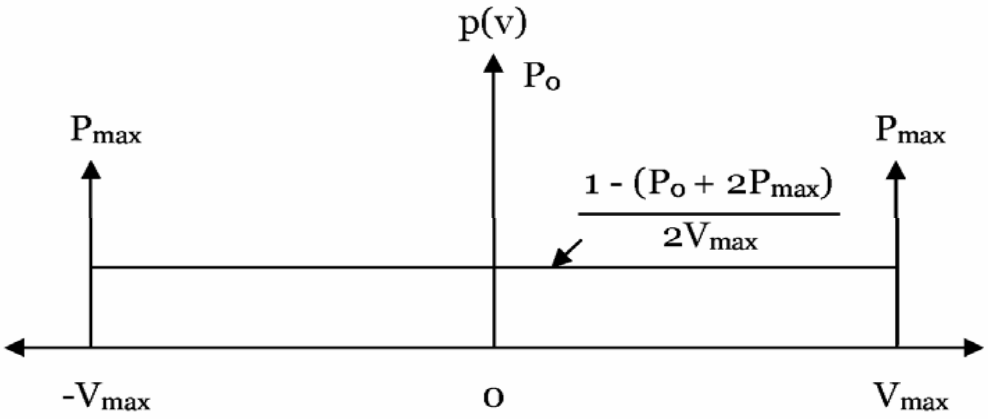
$w(t)$ ... white Gaussian noise

α ... reciprocal velocity time constant

[1] D. Helbing, "A mathematical model for the behavior of pedestrians, "*Behavioral Science*, vol. 36, pp. 298–310, 1991.

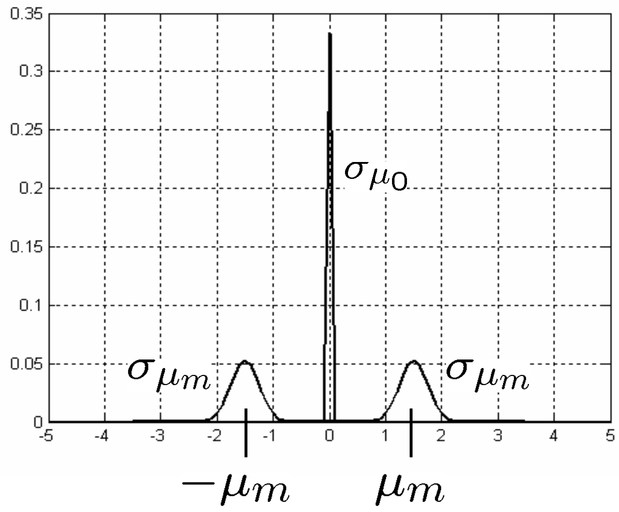
Density Assumptions

Assumed probability density of velocity variation v



V_{max}	...	max. speed increase
P_0	...	probability of no perturbation
P_{max}	...	probability of max. speed increase

Assumed probability density of velocity u



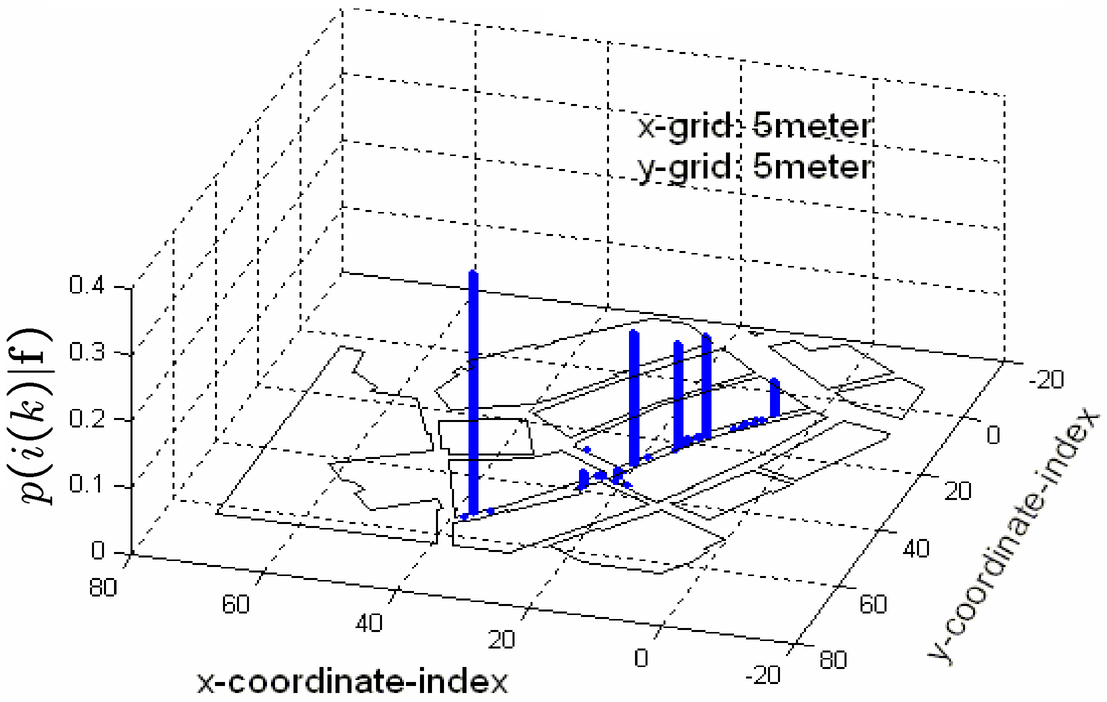
μ_m	...	mean velocity of moving user
$\sigma^2_{\mu_m}$	...	variance of moving user
$\sigma^2_{\mu_0}$	...	variance of motionless user

Propagation of Position Estimate

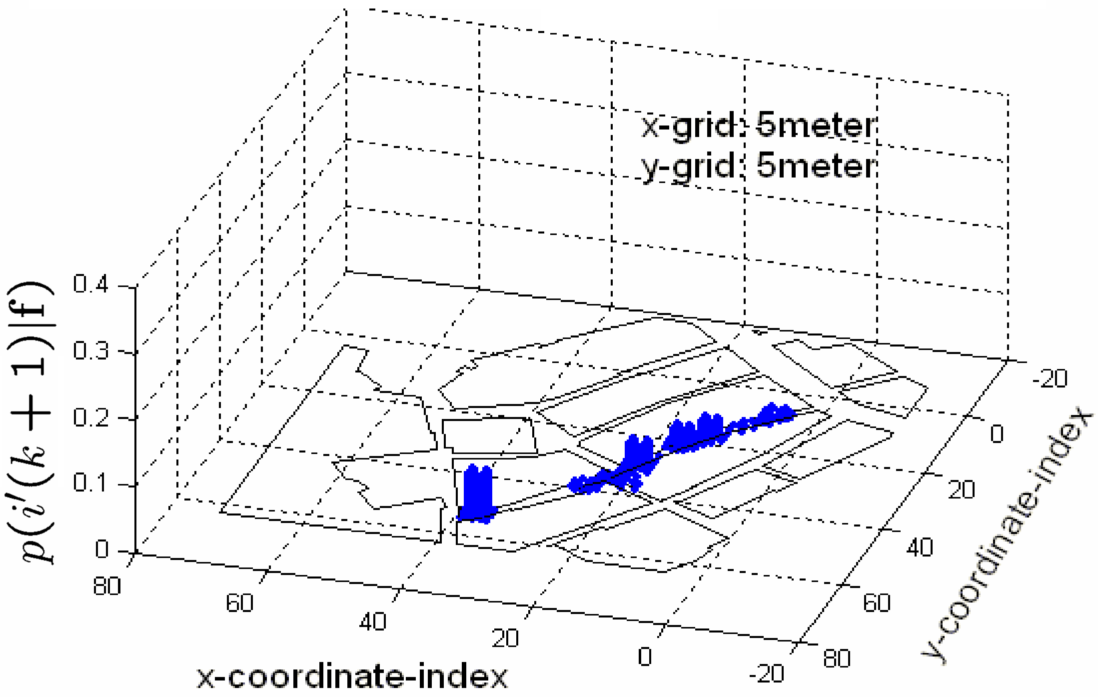
$\mathcal{L}(i) \propto p(i)$ ... if no prior belief about $p(\mathbf{f})$ and $p(i)$

$i'(k+1) = i(k) + x(k+1)$ ⟵ motion model's contribution; ⟵ position estimate at time k

Probability of being at position i at time k



Probability of being at position i' at time k+1



Combination of Position Estimates

$$\overbrace{i^*(k+1)}^{\text{new combined estimate}} = (I-K)\,\overbrace{i'(k+1)}^{\text{propagated estimate}} + K\,\overbrace{i(k+1)}^{\text{new estimate at time k+1}}$$

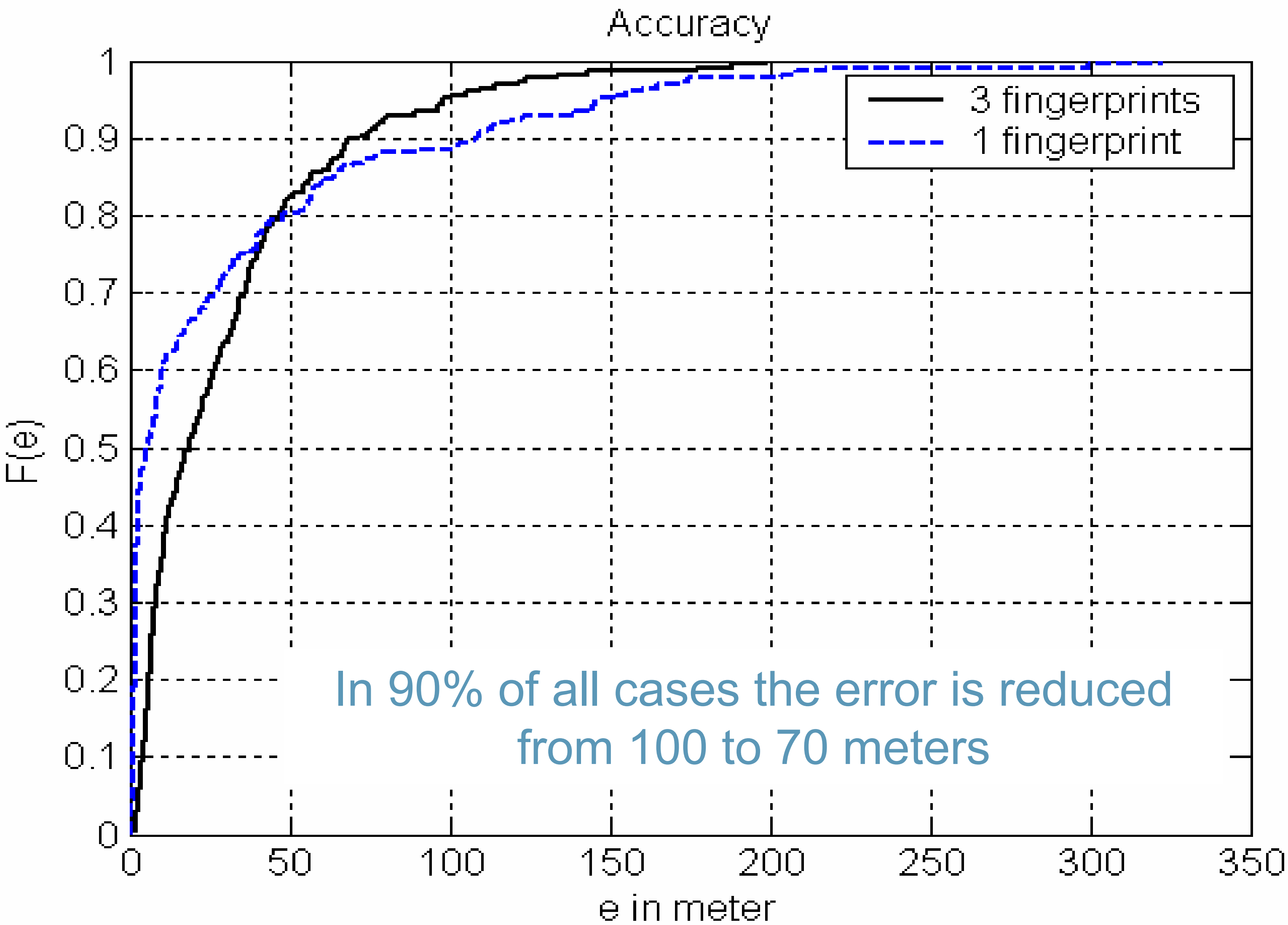
K ... Blending Factor

I ... Identity Matrix

- K is found by minimizing the variance of the estimator
- K can be computed in closed form

Test Area

Accuracy (3 Fingerprints)



2800 test cases, uniformly distributed within target area

Conclusions

Pro:
- → 100% availability (GPS less than 60% in downtown of Vienna)
- → Accuracy upper bounded by cell size.
- → Suitable for urban areas.

Con:
- → Reference data (still) necessary
- → Weak for indoor environments (but can be combined with other methods)

Thank you!