

#### Multimodal Indoor Localization using Heterogeneous Technologies

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## Outline

#### Introduction

## Heterogeneous Localization Technologies

#### Combination Approach

- Experiment Results
- Conclusion

## **Location based services**

- Information customization based on user location
- Navigation guide
- Location-based advertising
- Security surveillance, alert, notification, warning,...

## **Indoor localization**

- GPS generally works only outdoor → search for indoor localization schemes
- Many approaches proposed for indoor localization: cellular networks, infrared, ultrasonic, computer vision, RFID...
  - All suffer either from the limited accuracy, range, lacking of the infrastructure, or high deployment price

Combination of multiple technologies to overcome the limitation of individual ones

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# **GPS, GALILEO**

- Principle:
  - TOA  $\rightarrow$  distance to satellites
  - Least square solution
- Accuracy: 30m
- Advantage: global

#### Problems:

- Obstruction  $\rightarrow$  only outdoor
- Multipath propagation
- Signals weakened through atmosphere, walls, trees



# RFID

#### Main approaches:

- Fixed readers, mobile tags
- Fixed tags, mobile readers

## Accuracy: 1m

#### Problems:

- Proximity localization
- Scalability



# WiFi signals

#### Two main approaches:

- Geometrical calculation: angulation, lateration,...
- Fingerprinting
- Accuracy: 5m
- Problems:
  - Complex propagation characteristics (low stability)

73 dB)

Pre-deployment efforts required

 $(x_1, y_1)$ 

Y

**У**1

y<sub>k</sub>

Х

**X**1

 $(\mathbf{x}_k, \mathbf{y}_k)$ 

**AP**₁

ap<sub>11</sub>

ap<sub>k1</sub>

AP<sub>n</sub>

ap<sub>1n</sub>

ap<sub>kn</sub>



Self localization mechanism

#### Problems

- Additional orientation sensor required
- Calibration needed
- Inapplicable to robots

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## System architecture



# **Aggregation approach**

#### Probability based

For each point (x,y,z), calculate aggregation probability ρ<sub>Σ</sub>



Maximizing

$$\rho_{\Sigma}(x, y, z) = \Omega_{i=1..n} \left( \rho_i(x, y, z) e^{-\lambda_i t}, R_i \right)$$

- $\Omega$ : probability aggregation function
- n: number of technologies
- $\rho_i$ : probability of technology *i* (sum, product,...)
- *R<sub>i</sub>*: reliability constant of technology *i*
- $\lambda_i$ : time decay constant of technology *i*

## **Parameter estimation**

- Parameters: λ<sub>i</sub>, R<sub>i</sub>
- Using genetic algorithms



#### Cost function: RMS of localization error

$$\Phi = \left(\frac{1}{N}\sum_{i=1}^{N}(\hat{x}_{i}-x_{i})^{2}+(\hat{y}_{i}-y_{i})^{2}+(\hat{z}_{i}-z_{i})^{2}\right)^{1/2}$$



- $(x_0, y_0, z_0)$ : returned location by GPS
- $\sigma$ : function of accuracy by 3-sigma rule



## WiFi

Gaussian probability

$$\rho = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(r-r_0)^2}{2\sigma^2}}$$

- r<sub>0</sub>: nominal distance from empirical propagation model
- $\sigma$ : function of  $r_0$



Distance (m)

## Pedometer

Gaussian probability



- $(x_0, y_0, z_0)$ : nominal user location
- σ: function of (step-length x step-count)
- d: Euclidean distance function

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## **Historical & map information**

Gaussian probability

$$\rho_i(x, y, z) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{d^2(x, y, z, x_0, y_0, z_0)}{2\sigma^2}} <$$

- $(x_0, y_0, z_0)$ : previous user location
- $\sigma$ : function of user speed by 3-sigma rule
- *d*: distance function with environment map awareness
  - ⋆ Shortest-path based
  - ⋆ Impossible location avoidance



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## Test scenario: user 1



## **Test scenario: user 2**



## Results

- WiFi only:
  - ◆ <u>video</u>
- WiFi + RFID + step count:
  - ♦ video
- WiFi + RFID + step count + historical & environment info:
  - ♦ video





## Conclusion

- Probability based Multimodal localization approach
- System parameters tuned by using genetic algorithms with collected training data
- Highly extensible with heterogeneous technologies
- Significantly high accuracy of user localization is achieved

#### Perspectives

- Integration of other technologies: camera (fixed or mobile)
- Calculation speed

# Thank you for your attention!



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