

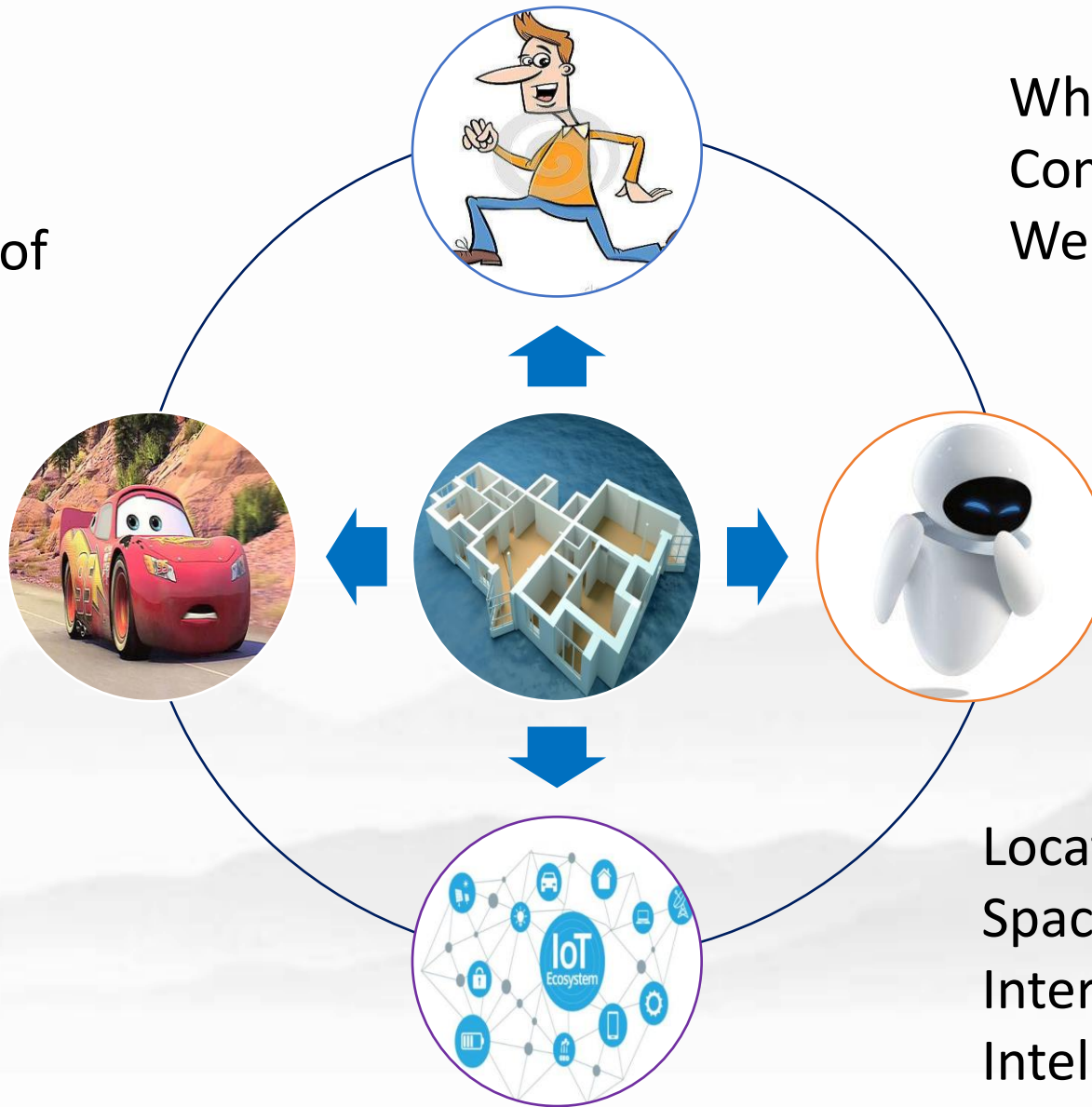
Smartphone Positioning and 3D Mapping Indoors

Ruizhi Chen

Wuhan University
Oct. 4, 2018, Delft

Adding a Smart LIFE to 3D

People spend 80% of their time indoors



When People Communicates to a Robot, We Need Locations

Locating Actors Living in 3D Space will Facilitate Smart Interactions and Enable Intelligent Applications

Mixed Reality Games in 3D Spaces



Autonomous Driving for Underground Parking



Precisely Locate the Vehicles in a 3D Space

Precision Marketing



Interaction between human and goods

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Precise Smartphone Positioning Based on Built-in Sensors and RF Radios

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Indoor Mapping

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Indoor Mapping

Your Phone Knows Where You Are



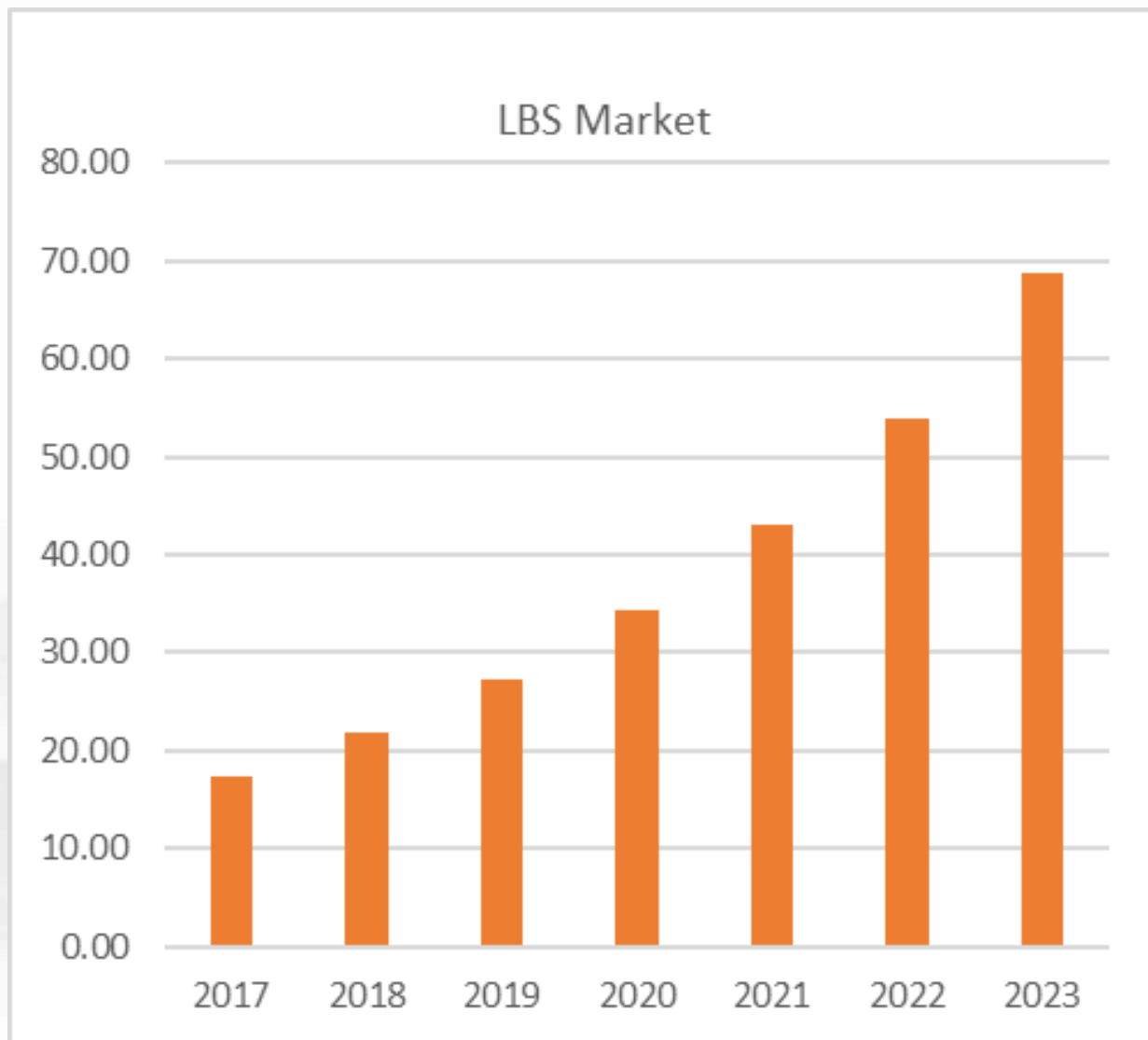
“Where am I?”

The Market Size of LBS and RTLS

The Location-Based Services (LBS) and Real-Time Location Systems (RTLS) market size was valued at **USD 17.38 billion in 2017** and is projected to reach **USD 68.85 billion by 2023**, at a Compound Annual Growth Rate (CAGR) of **25.4%** during the forecast period. The base year considered for the study is 2017 and the forecast period is from 2018 to 2023.

MarketsandMarkets™

<https://www.marketsandmarkets.com/Market-Reports/location-based-service-market-96994431.html>



Challenges for Indoor Positioning



Complex topology

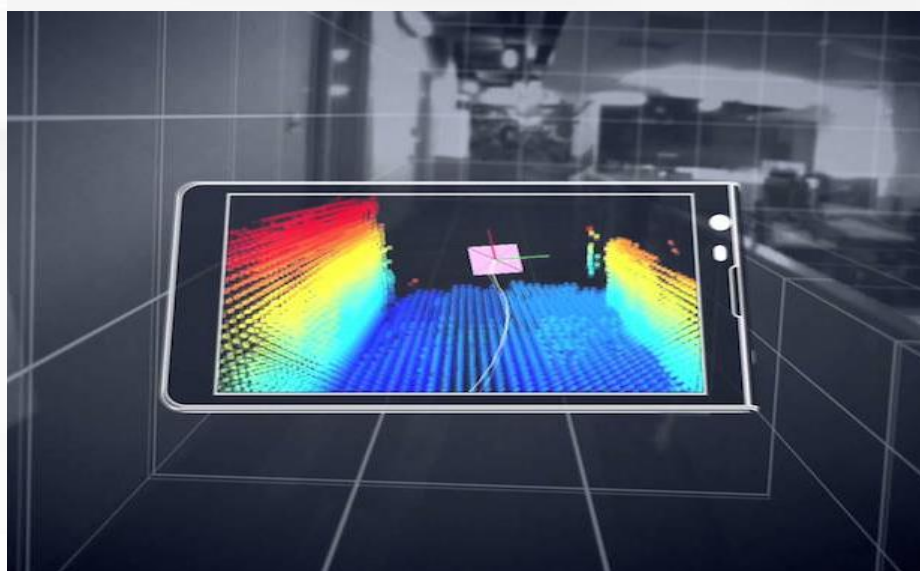


Complex radio environment



Complex human motion patterns

Visual Positioning Service – A Google Core Technology



Google

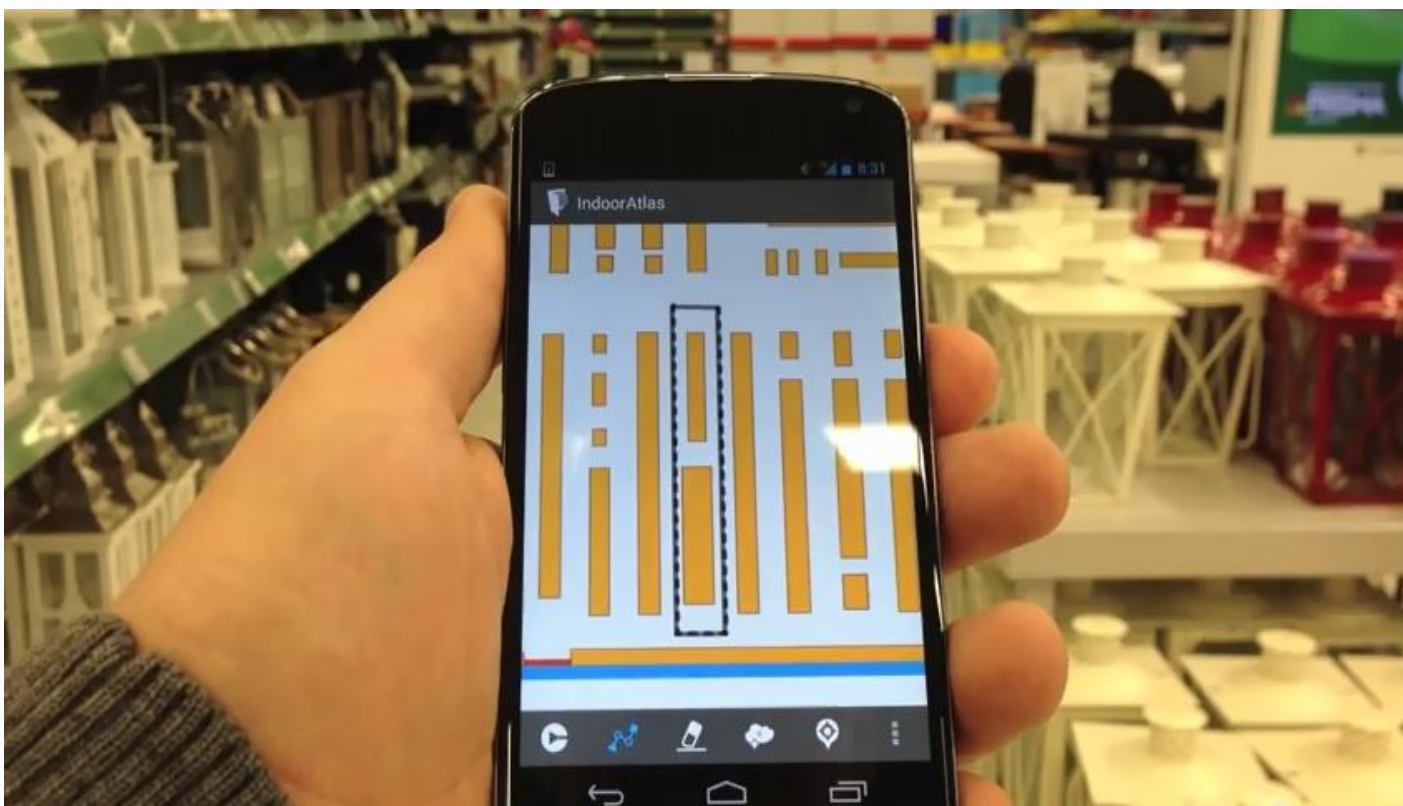
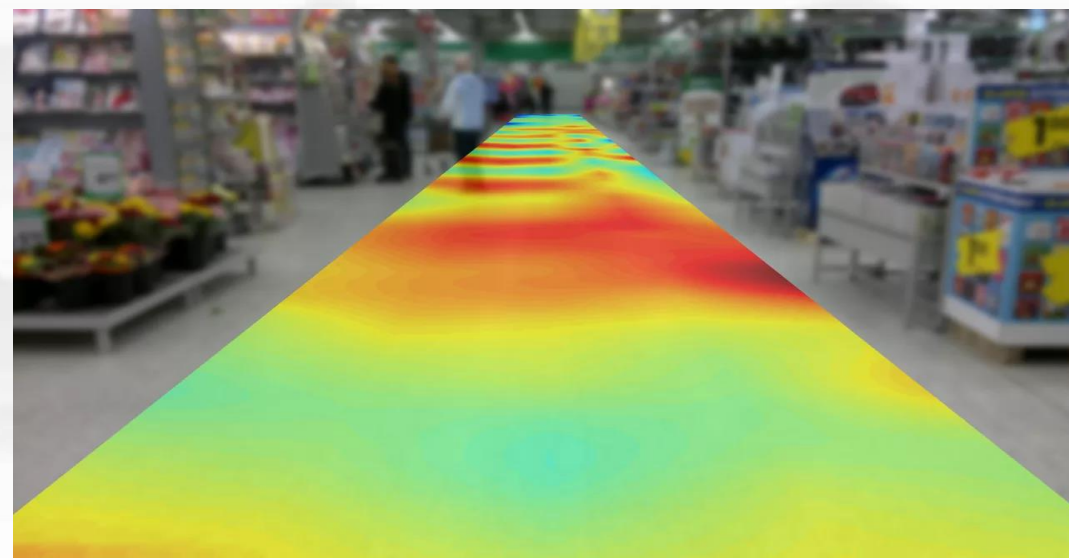
iBeacon – An Apple Technology



iBeacon



Baidu: Magnetic Fingerprinting



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Current Smartphone Positioning Technologies

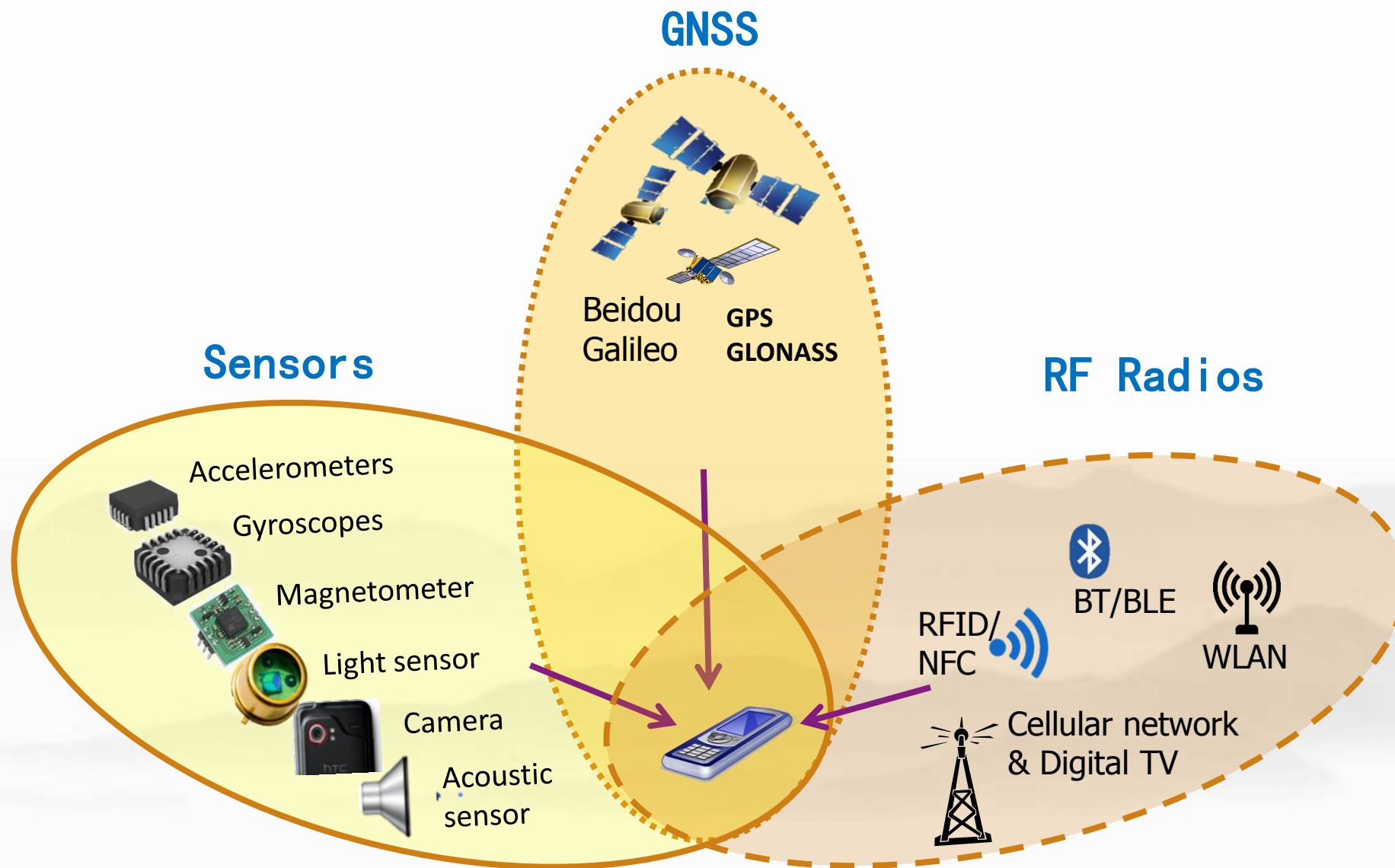
3

Precise Smartphone Positioning Based on Built-in Sensors and RF Radios

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Indoor Mapping

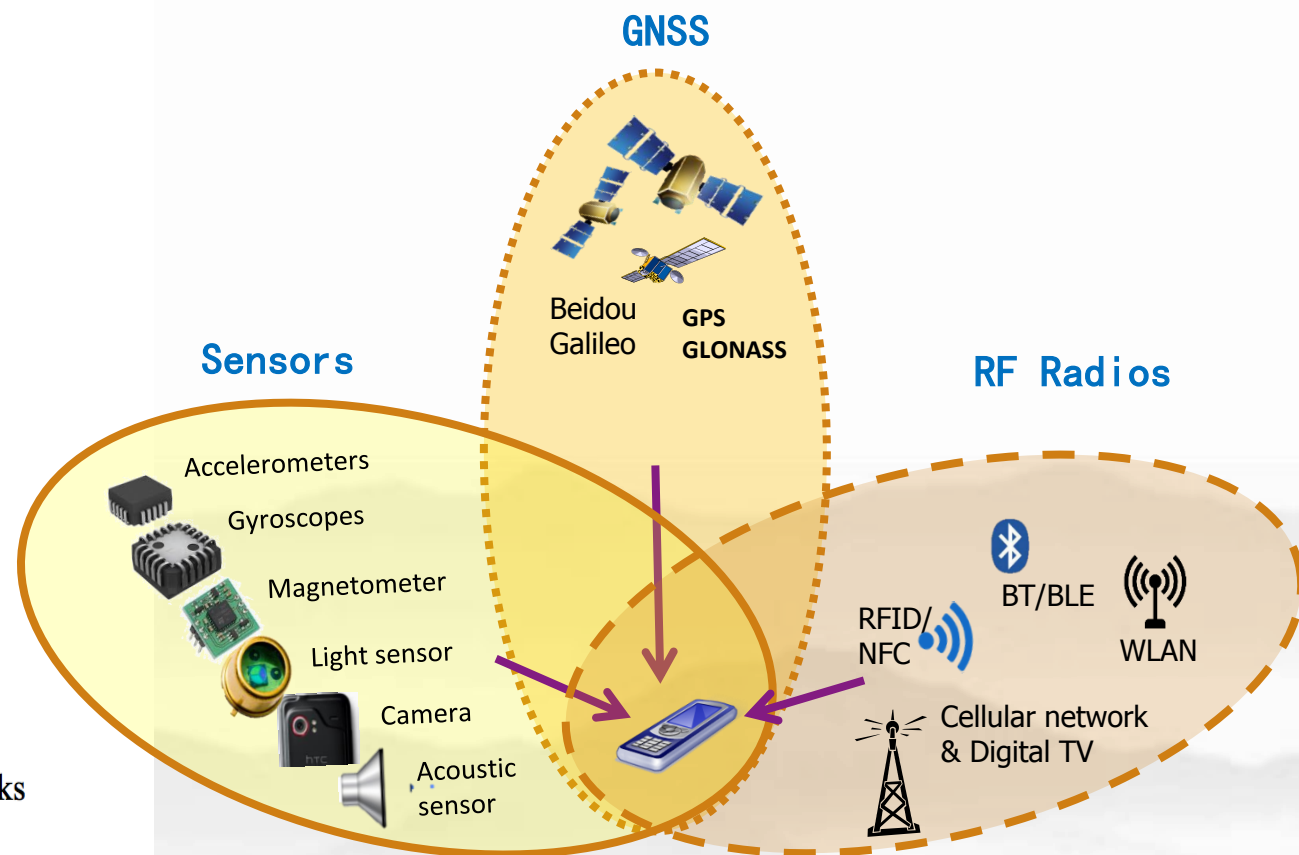
Positioning Sensors and RF Radios in Smartphones



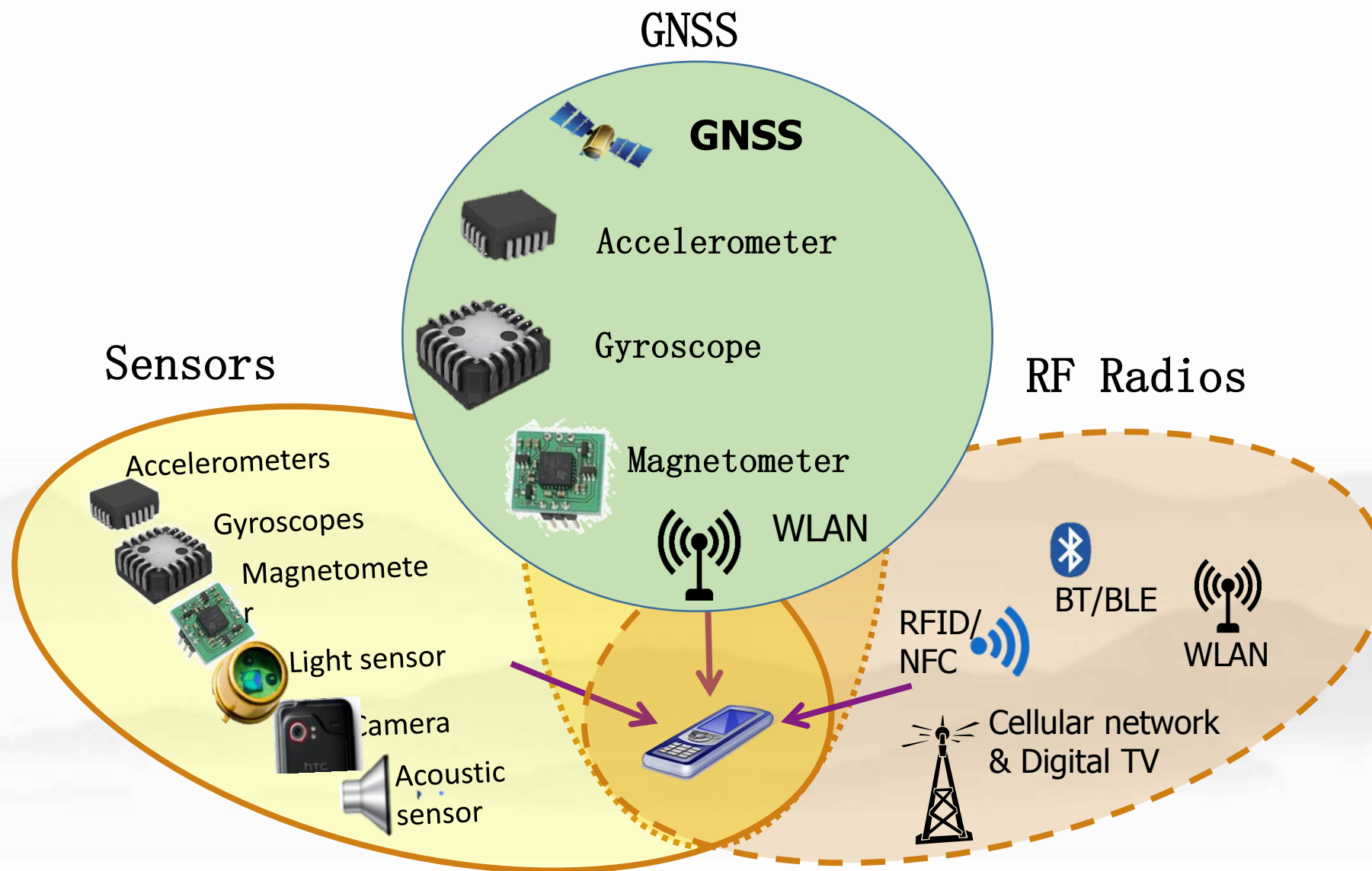
Typical Observables

List of Mobile Positioning Observables

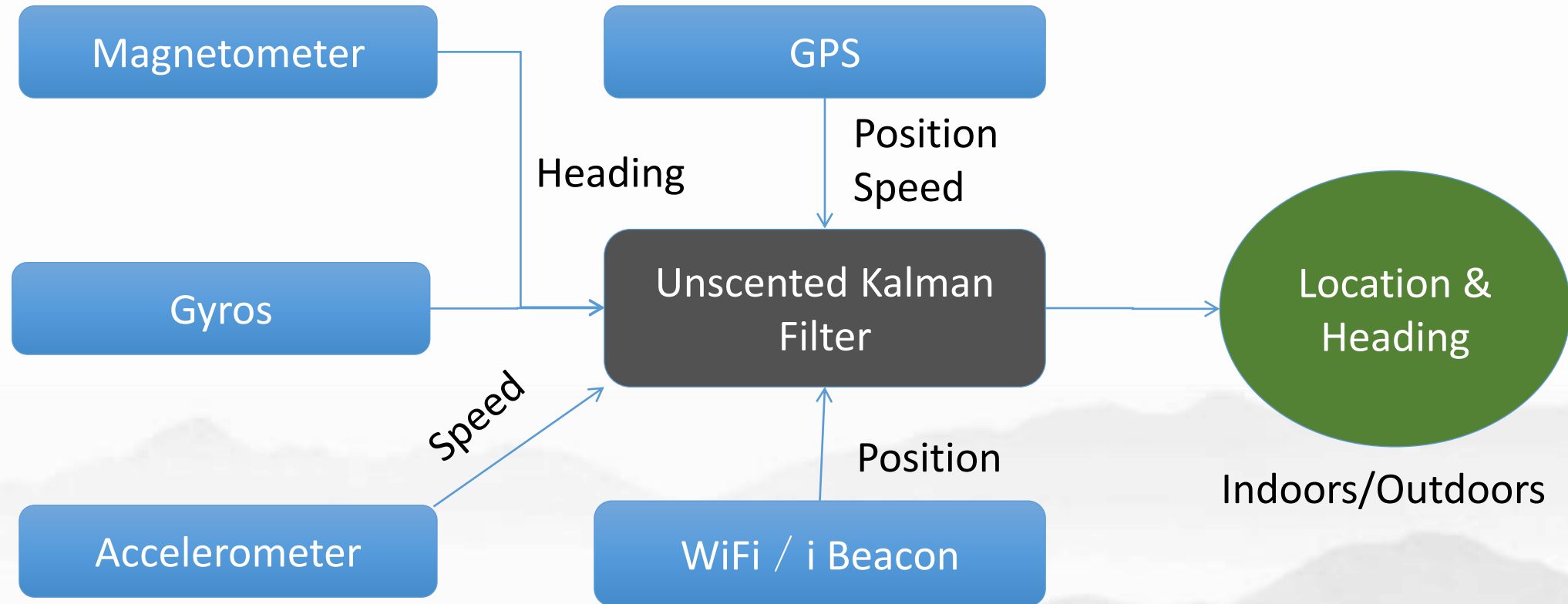
Observables	Sensors or Networks
Range	GNSS receiver, cellular networks
Ranging difference	GNSS receiver, cellular networks
Traveled distance	Accelerometer, camera
Speed	GNSS receiver, accelerometer, camera
Acceleration	Accelerometer
Angles/azimuth	Digital compass, cellular network
Angle rates	Gyroscope
Signal strength	WLAN, Bluetooth, RFID, cellular network
Cell-ID	MAC address, base stations in cellular networks
Image/image features	Camera



Positioning With WiFi, Sensors and GNSS

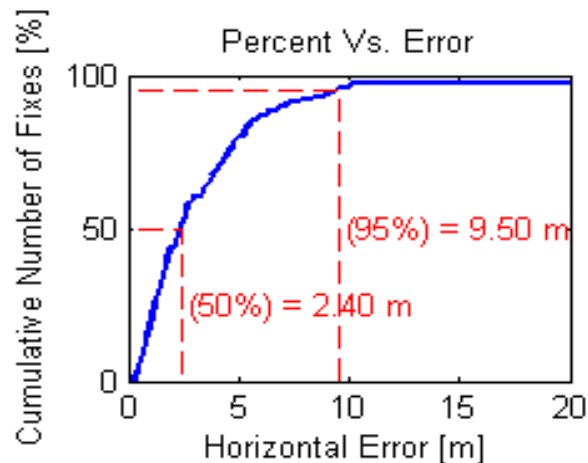
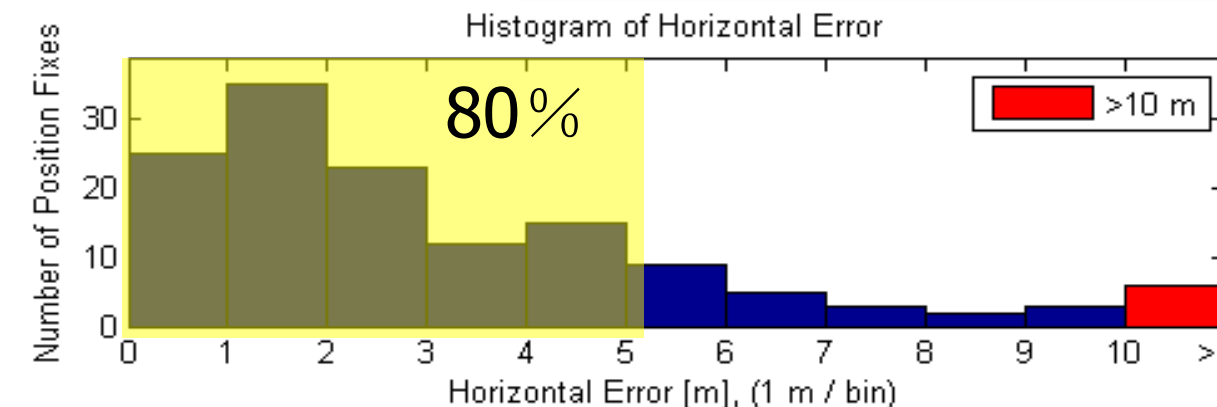
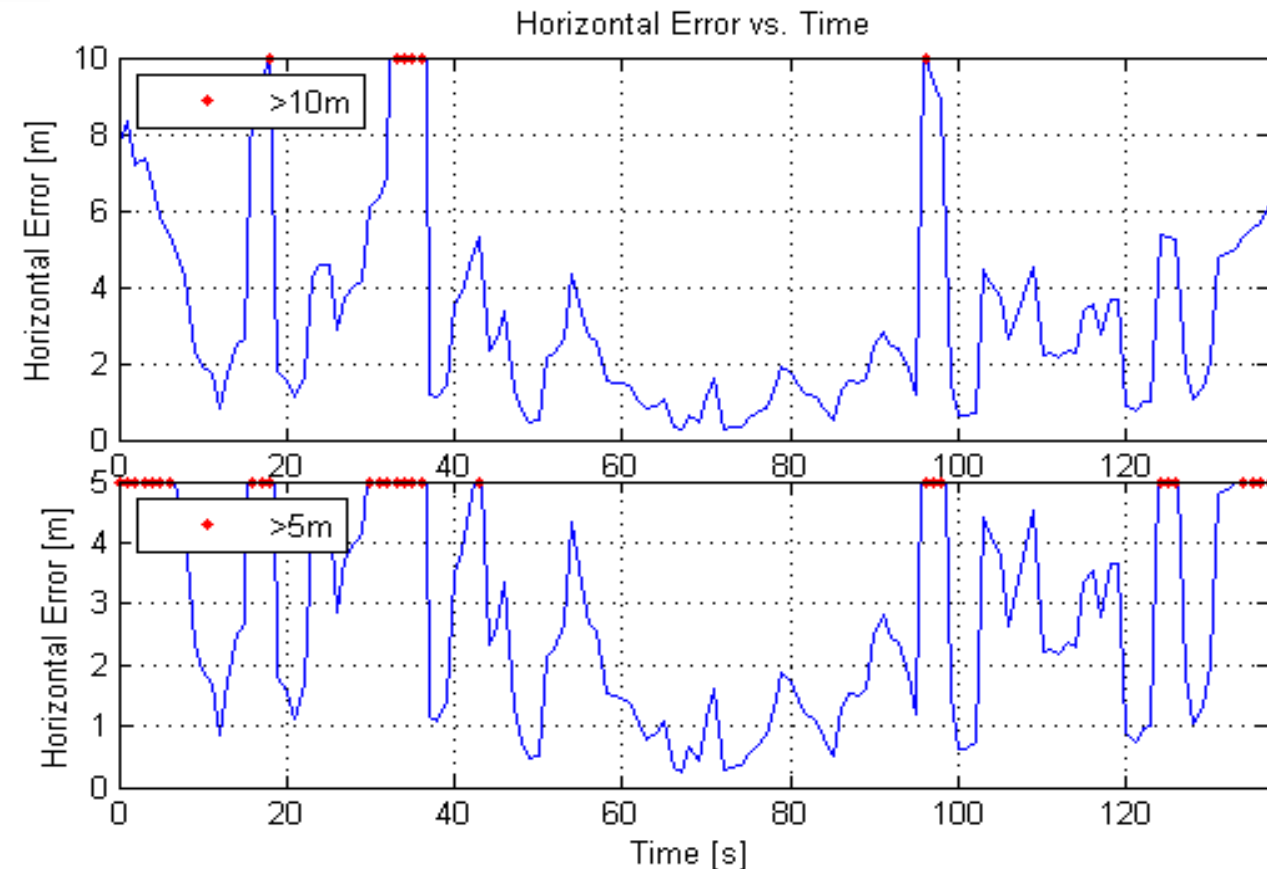


Fusing Sensor and RF Measurements



Chen, R., Chu, T., Liu, K., Liu, J., & Chen, Y. (2015). Inferring Human Activity in Mobile Devices by Computing Multiple Contexts. *Sensors* , 15(9), 21219–21238. <http://doi.org/10.3390/s150921219>

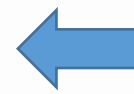
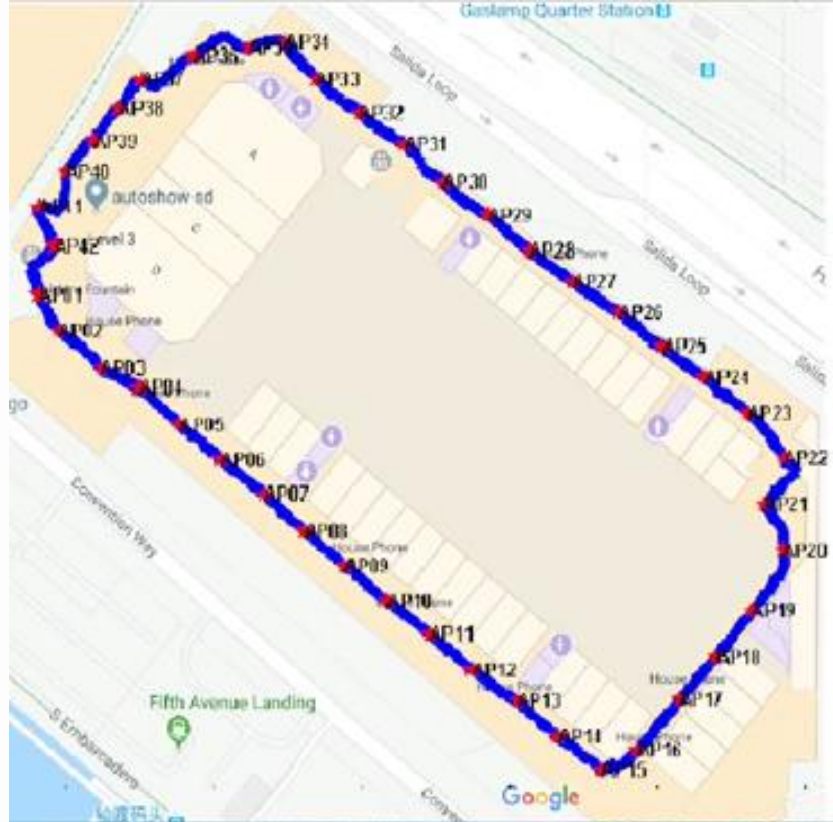
Positioning Accuracy



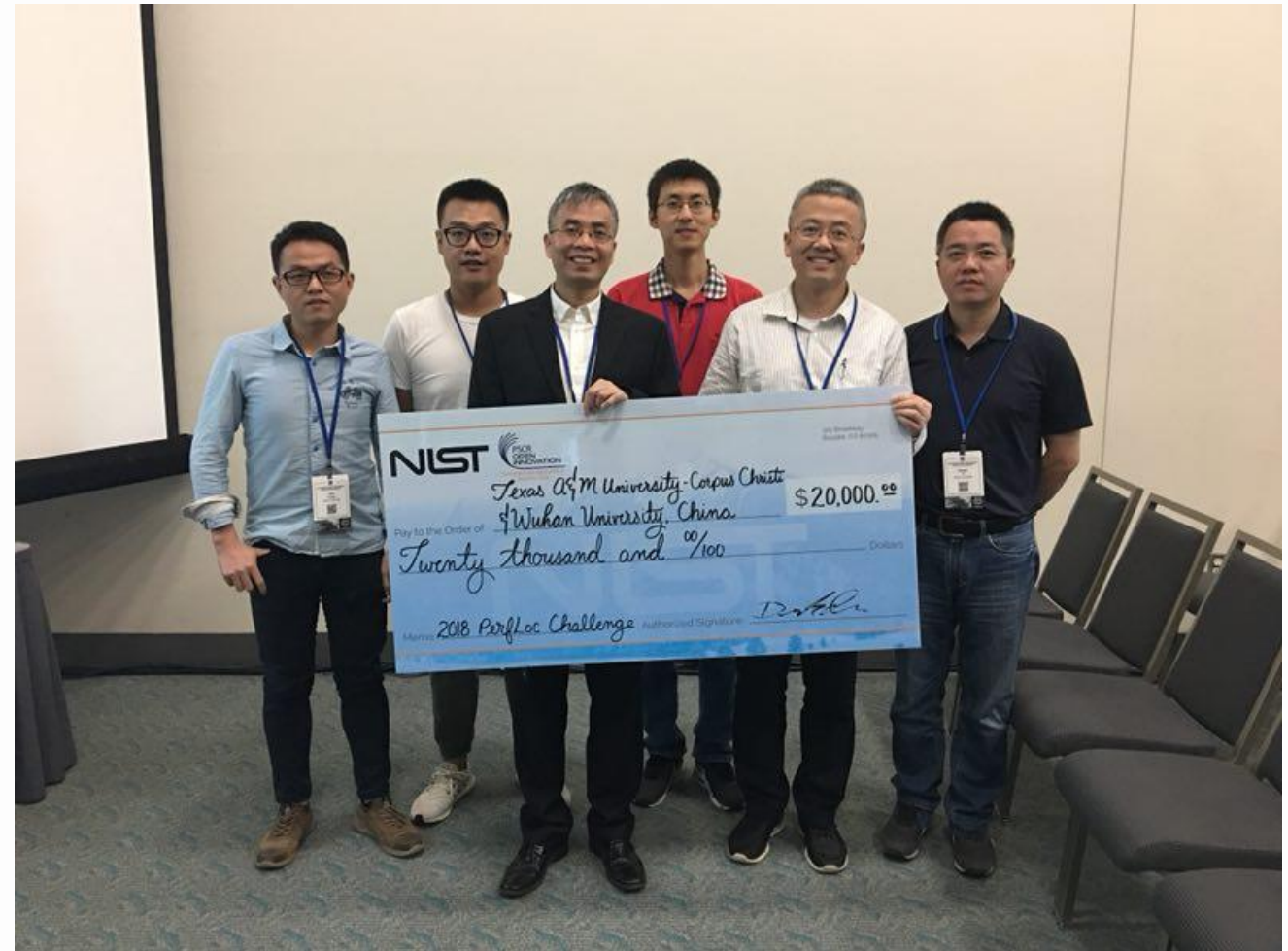
Testing Date: June 21, 2015
First Fix: 01:57:32 (UTC)
Last Fix: 01:59:49 (UTC)
Availability: 100 %
Error < 5 m: 79.71 %
Error < 10 m: 95.65 %
Maximum Error: 21.95 m
Mean Error: 3.52 m
Standard Deviation: 3.83 m

Ground Truth: GPS/INS





Real-Time : 2-5 meters under typical indoor environment



PerfLoc: NIST indoor Positioning Competition³⁰

Post-Processing Accuracy



A photograph of an award ceremony stage for the IPIN 2018 Track 3 competition. A large projection screen displays the results table. On the left, a man in a blue shirt is gesturing towards the screen. In the center, four men are standing and clapping. On the right, a man is speaking at a podium. The podium has a blue banner that reads "AWARD CEREMONY" and "WELCOME TO IPIN".

IPIN 2018
NINTH INTERNATIONAL CONFERENCE ON
INDOOR POSITIONING
AND INDOOR NAVIGATION
SEPTEMBER 24-27, 2018, BANGKOK, THAILAND

Track 3

Error	Team
1.1m	Five-WHU
1.5m	EGEC
2.5m	TENCENT

1st place medal icon
2nd place medal icon
3rd place medal icon

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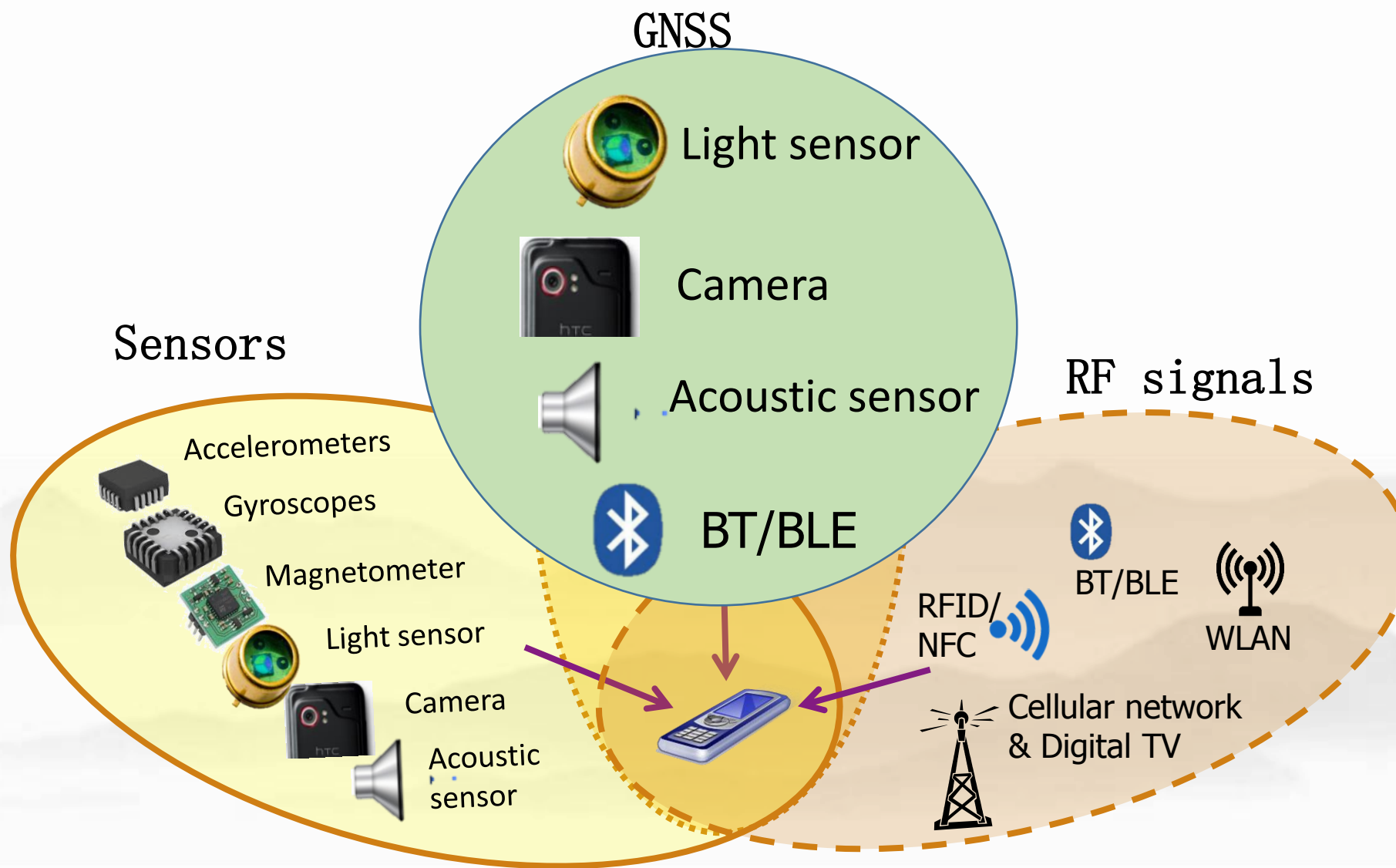
3

Precise Smartphone Positioning Based on Built-in Sensors and RF Radios

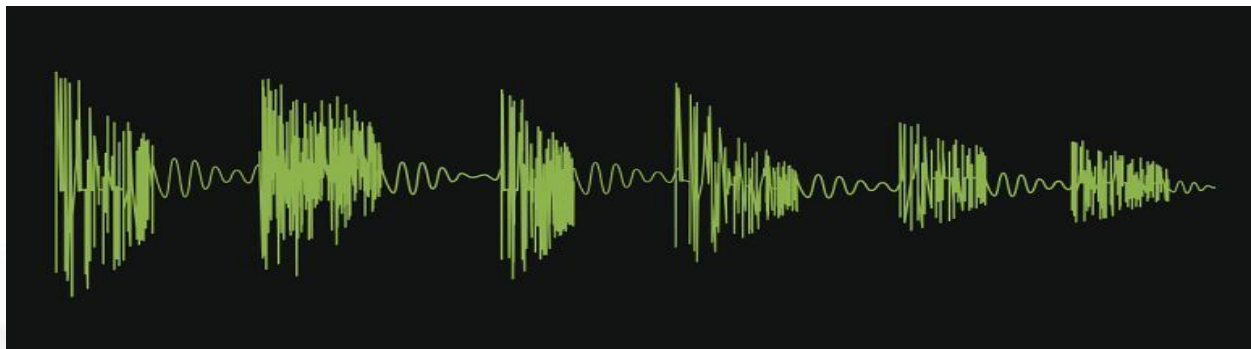
4

Indoor Mapping

Positioning with New Sensors

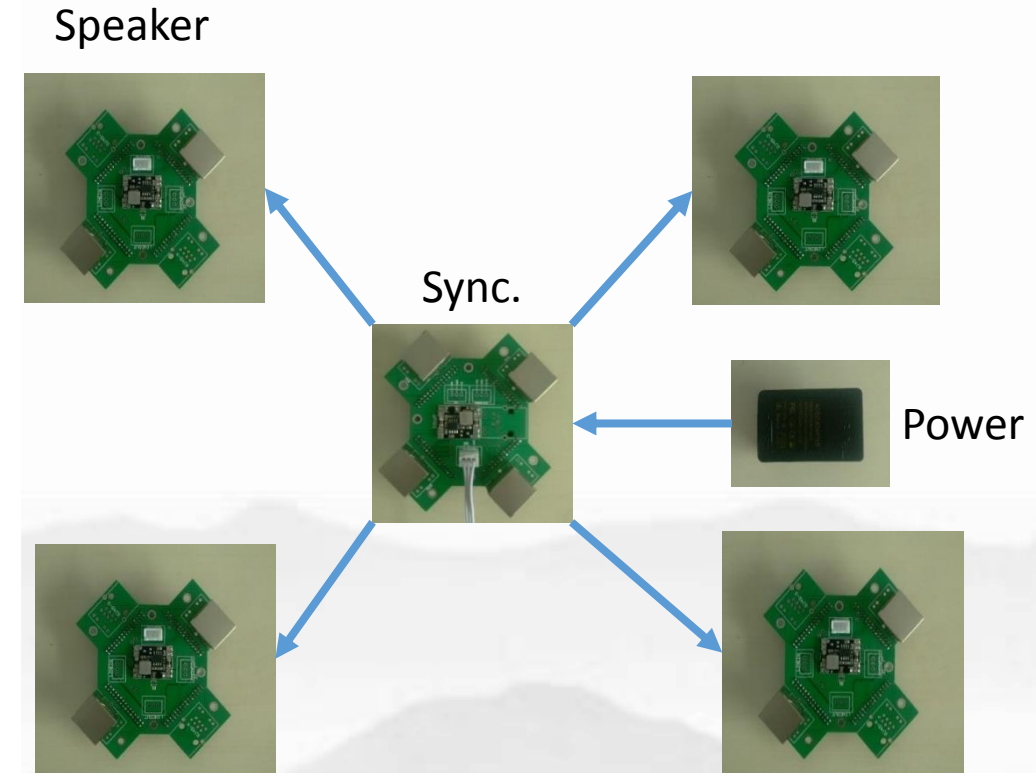


Positioning Based on Acoustic Signal

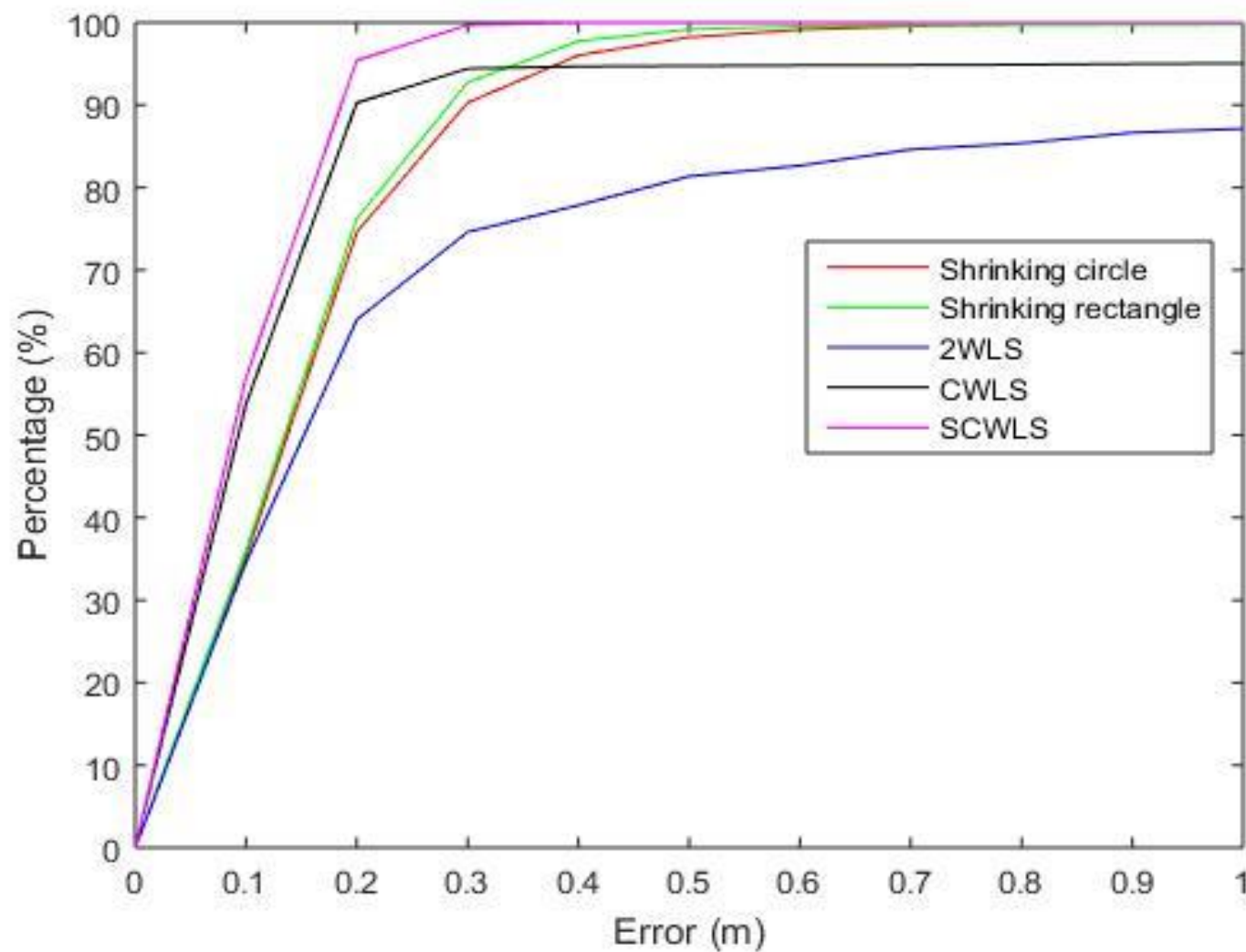


Acoustic Ranging Positioning

- **Using the Mic and Speakers of the Smartphone**
- **Working spectrum ranges from 16-21KHz
Not hearable by human, not interfered by human voices**
- **The speed of sound is slow compared to RF signals, therefore, the clock synchronization requirement is not high.**
- **Measure TOA**
- **Positioning accuracy: decimeters**
- **Effective Range: 5-20m**



Positioning Accuracy



0.2 m at 95%

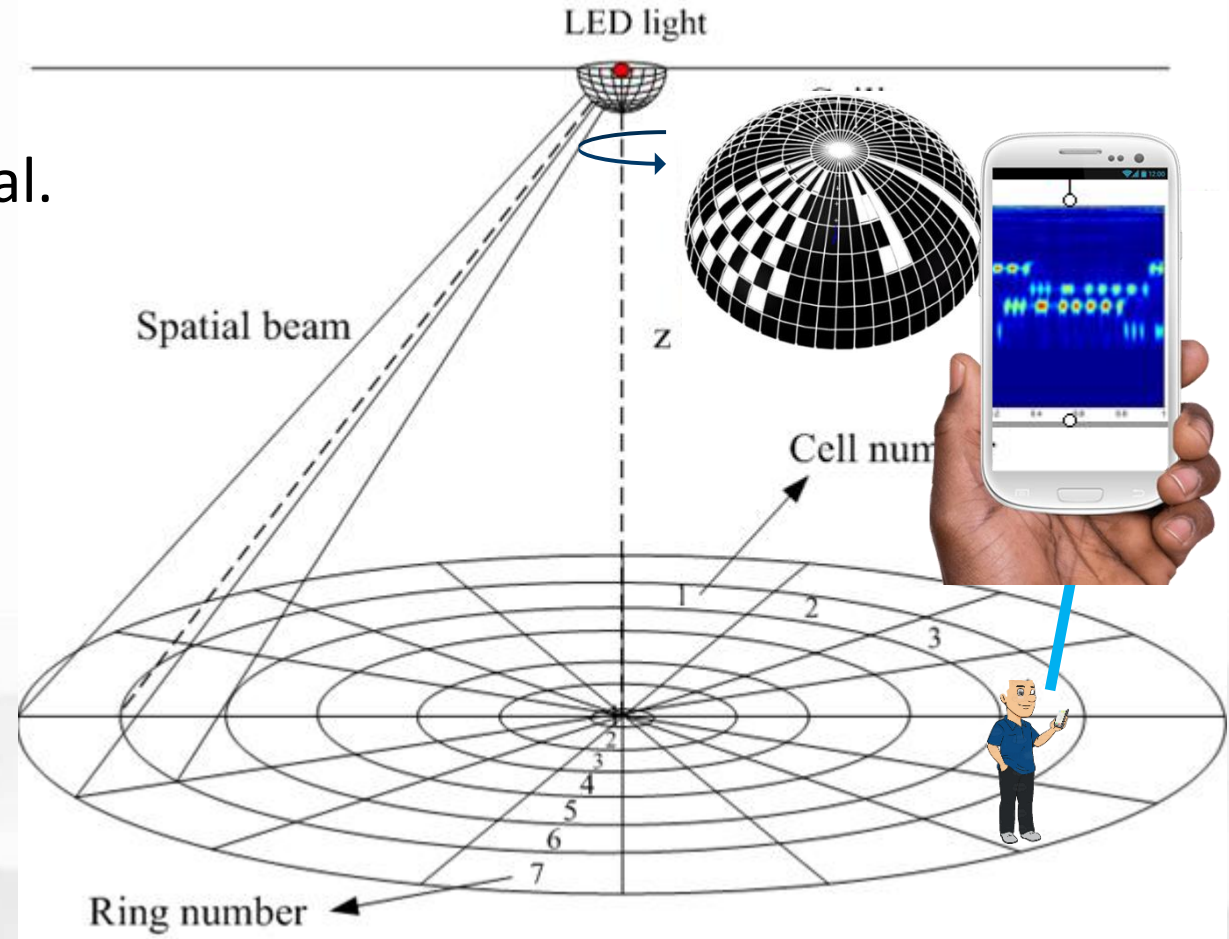
Positioning Based on Light Signal



Positioning Using Light

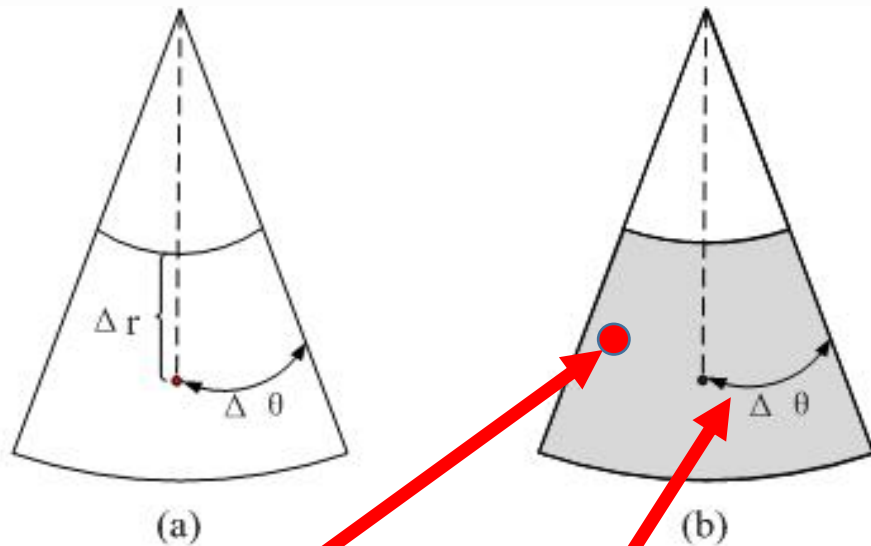


- An light shade is divided into 8 rings, each ring has 48 grids, there are 384 sectors in total.
- Each sectorial grid can be opened (0) or closed (1), by rotating the shade, the light sensor of the smartphone can receive different light patterns in different sectors.
- A sector is identified by the light patterns.
- No hardware change is needed from smartphones
- Positioning accuracy is 5-10cm.
- Single Station Positioning for Small Indoor Space

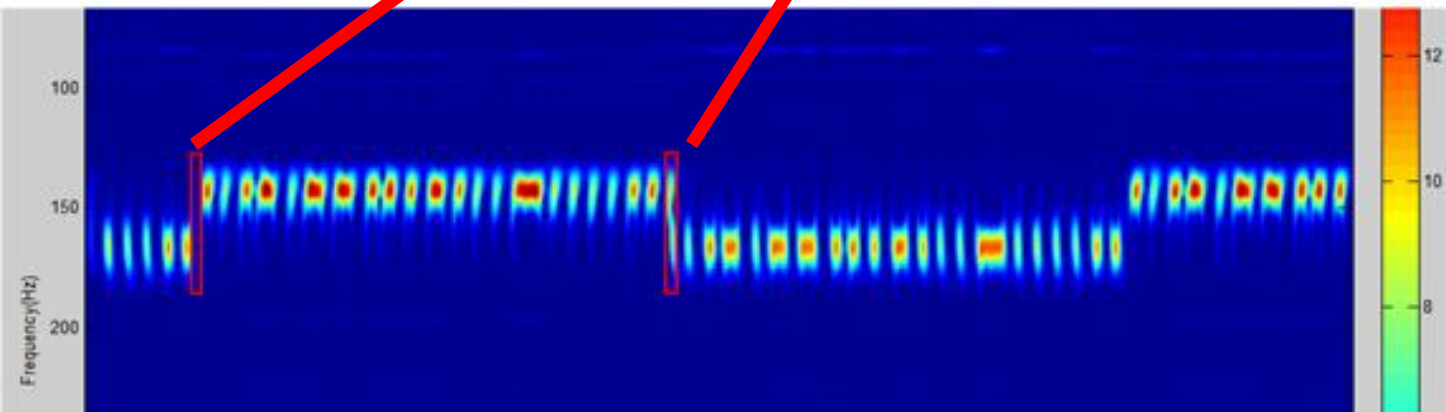


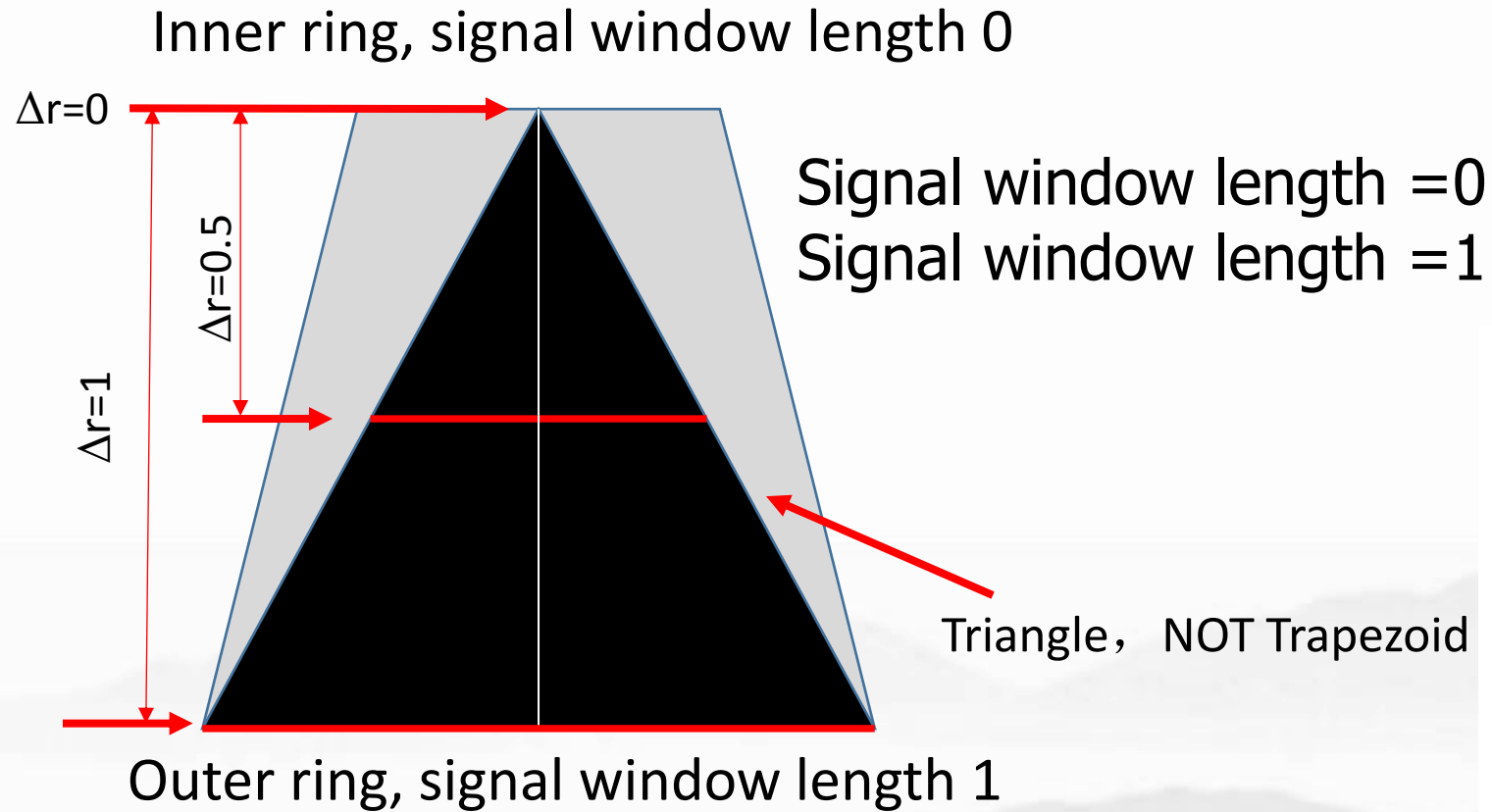
Light source: 850 nm Infrared

Correction in Longitude Direction

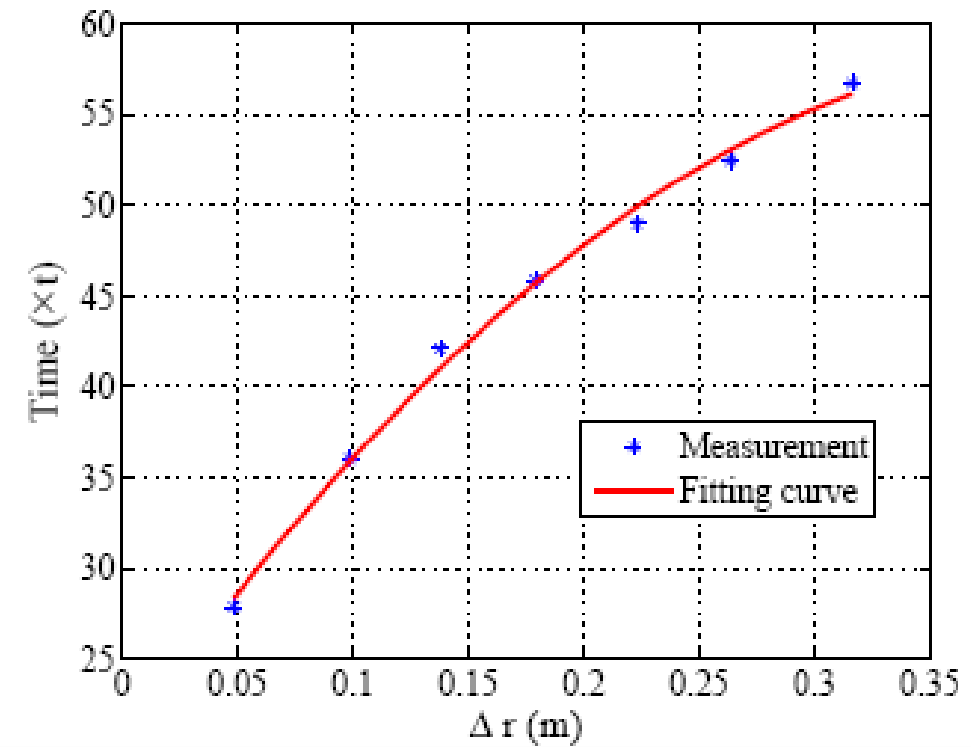
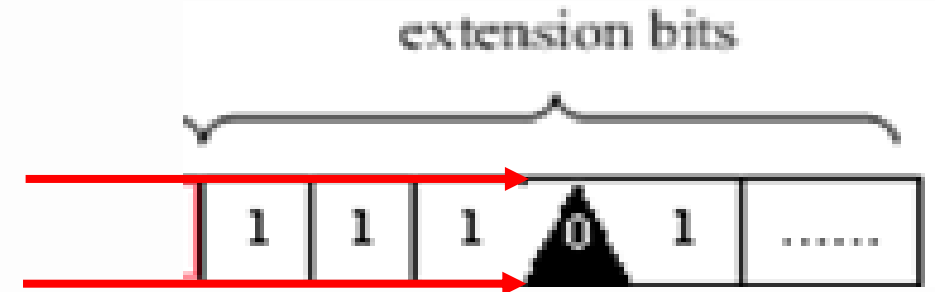


- (ring number, cell number) is too coarse grained (e.g., 0.5 m)
- Need to obtain two offsets: $\Delta \Theta$ and Δr
- A fraction of cell period (signal window length) is used to estimate $\Delta \Theta$





Signal window length varies between 0 and 1 for different Δr



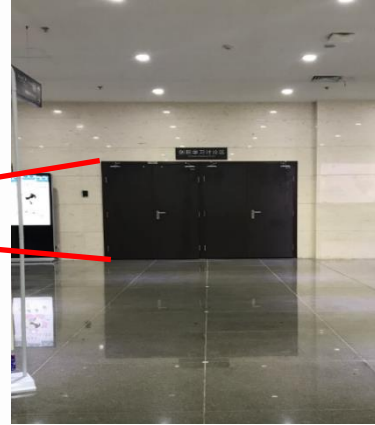


Visual Positioning with Point-Line-2D-3D Objects



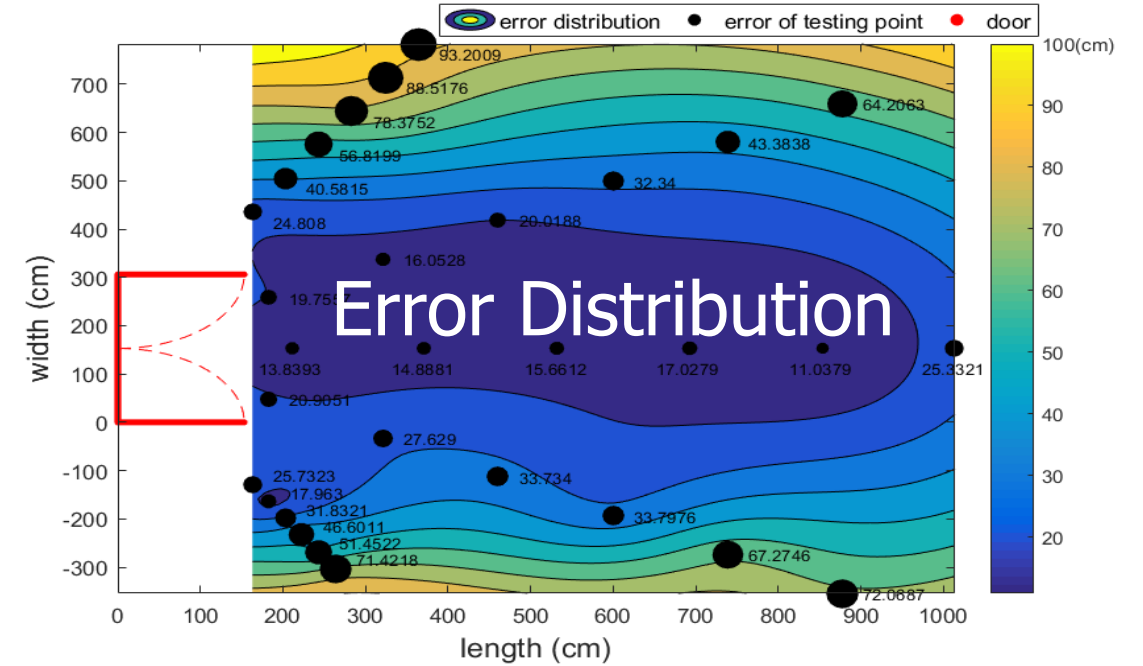
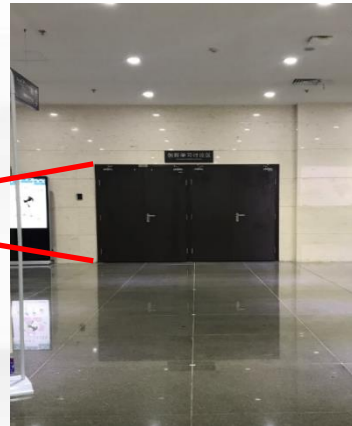
Human Eyes vs Smartphone Camera

5 Types of smartphones



3 Test Fields

10 Students



Positioning Error

Human Brain

Phone Camera

0.73m

0.31m

Indoor Visual Positioning aided by CNN-based Image Retrieval

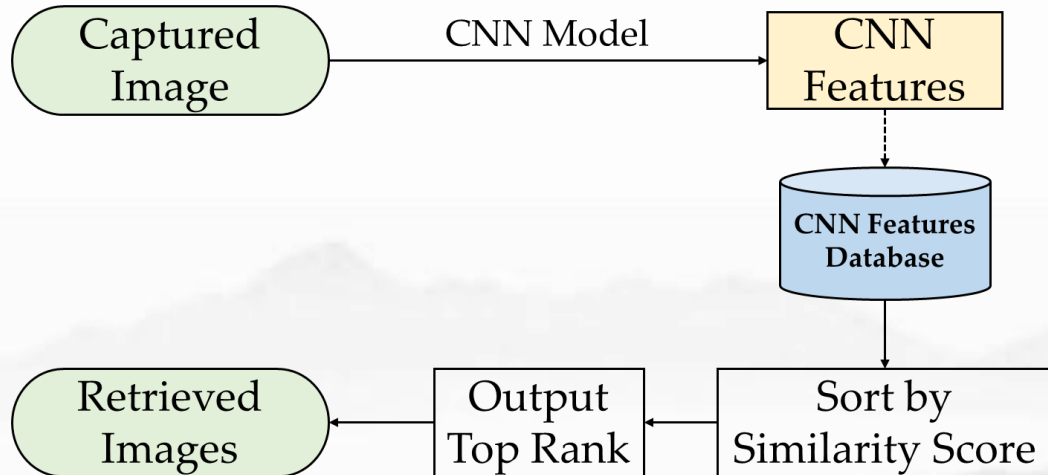
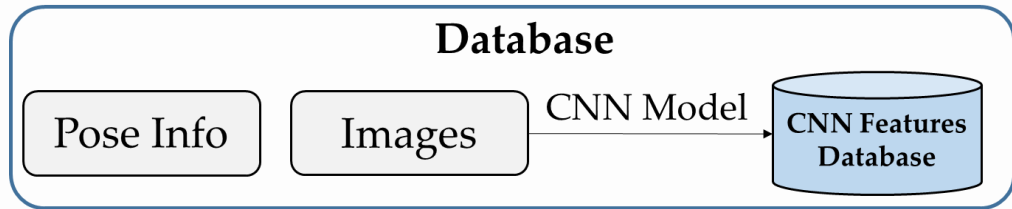
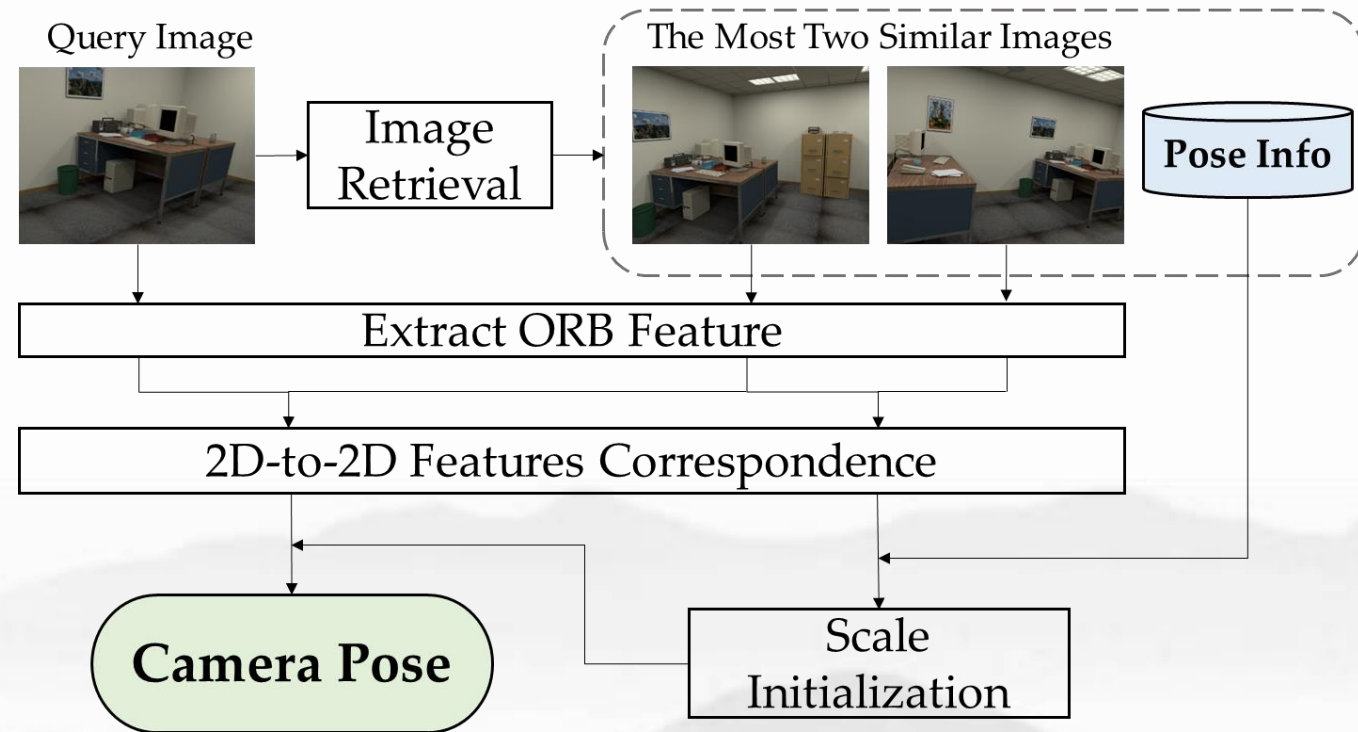


Image Retrieval

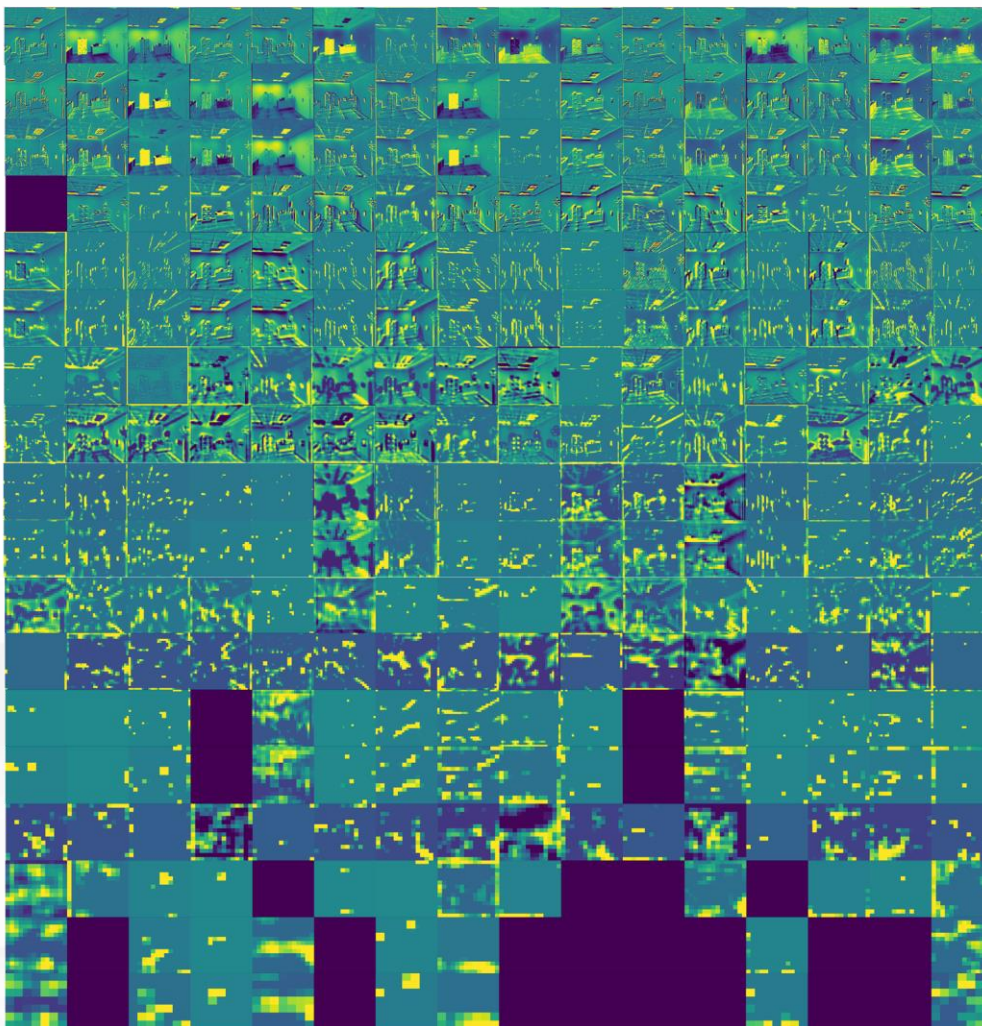


Pose Estimation

CNN-Features

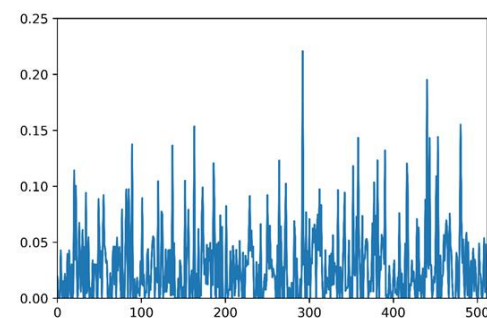


Block1_conv1 224×224×64
Block1_conv2 224×224×64
Block1_pool 112×112×128
Block2_conv1 112×112×128
Block2_conv2 112×112×128
Block2_pool 56×56×128
Block3_conv1 56×56×256
Block3_conv2 56×56×256
Block3_conv3 56×56×256
Block3_pool 28×28×256
Block4_conv1 28×28×512
Block4_conv2 28×28×512
Block4_conv3 28×28×512
Block4_pool 14×14×512
Block5_conv1 14×14×512
Block5_conv2 14×14×512
Block5_conv3 14×14×512
Block5_pool 7×7×512

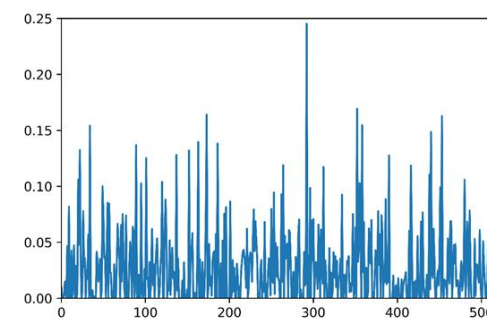


Convolution layers visualization

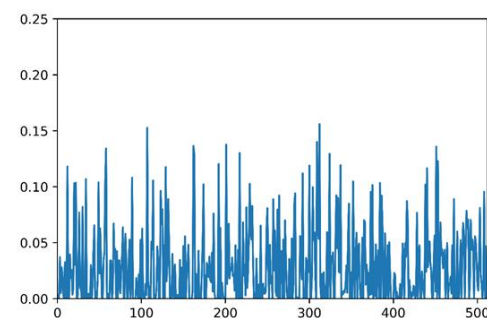
- Employ CNN model to extract features
- Rank images from database by feature similarity



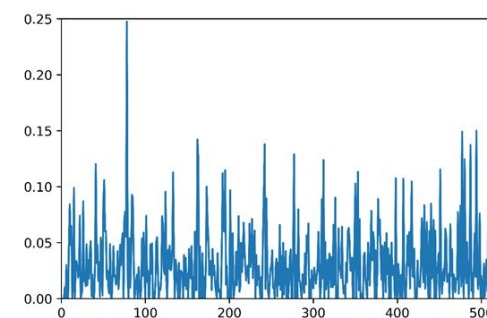
(a)



(b)



(c)



(d)

Image feature vectors visualization

Positioning Errors

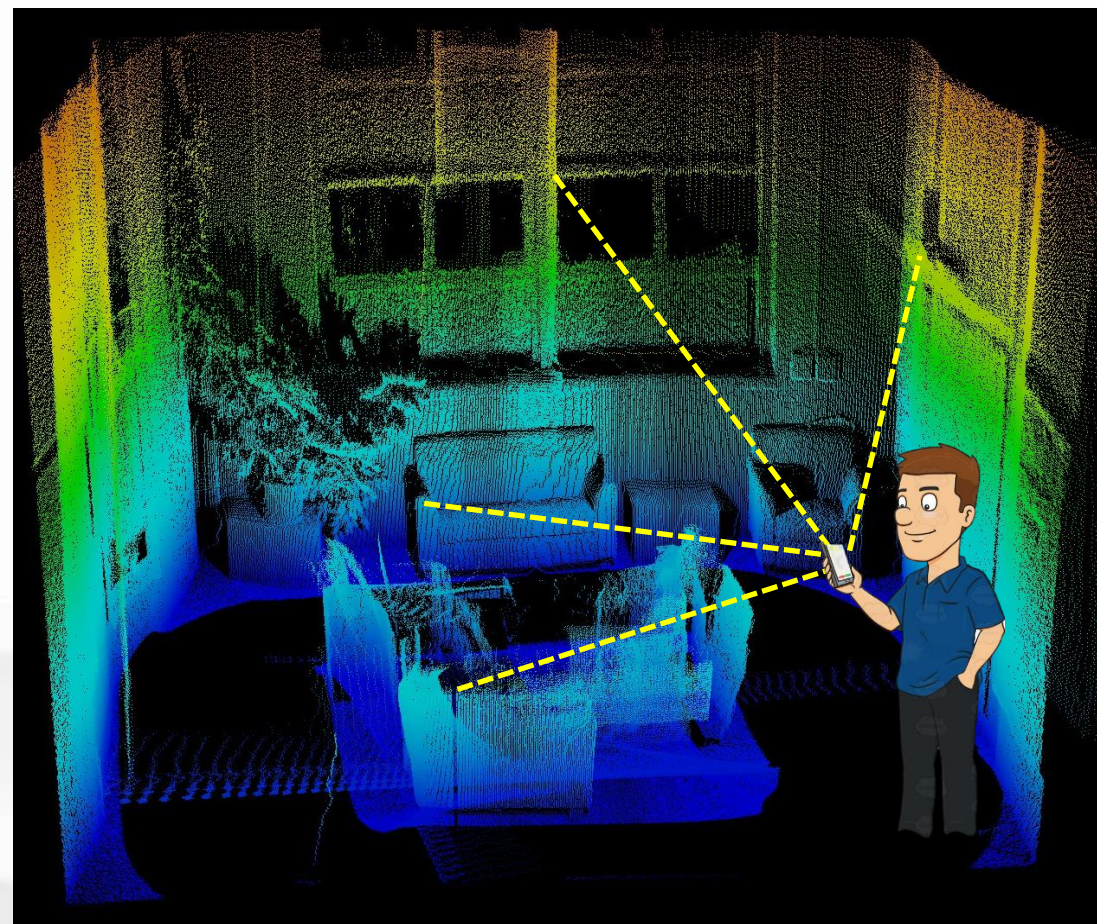
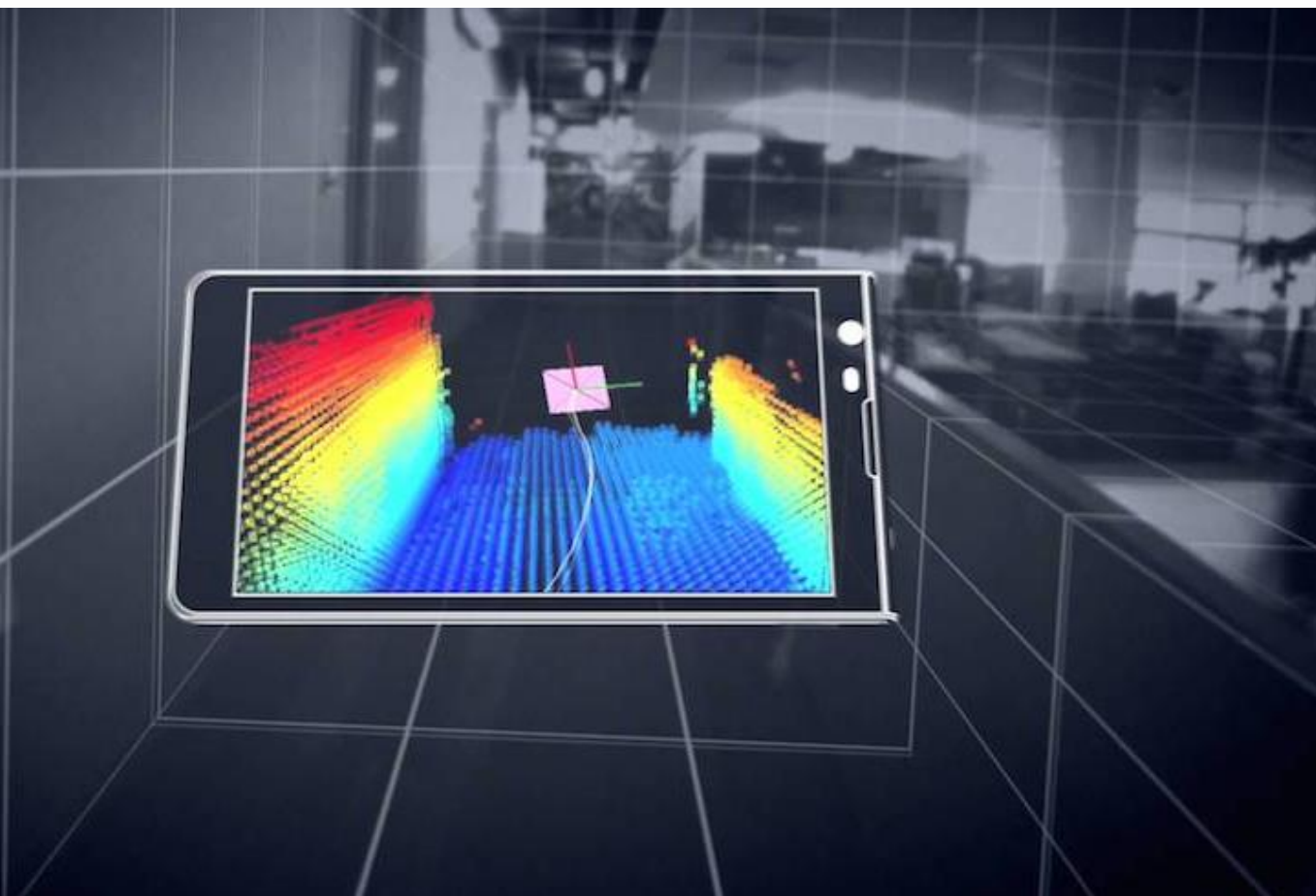


Comparison based on the ICL-NUIM dataset

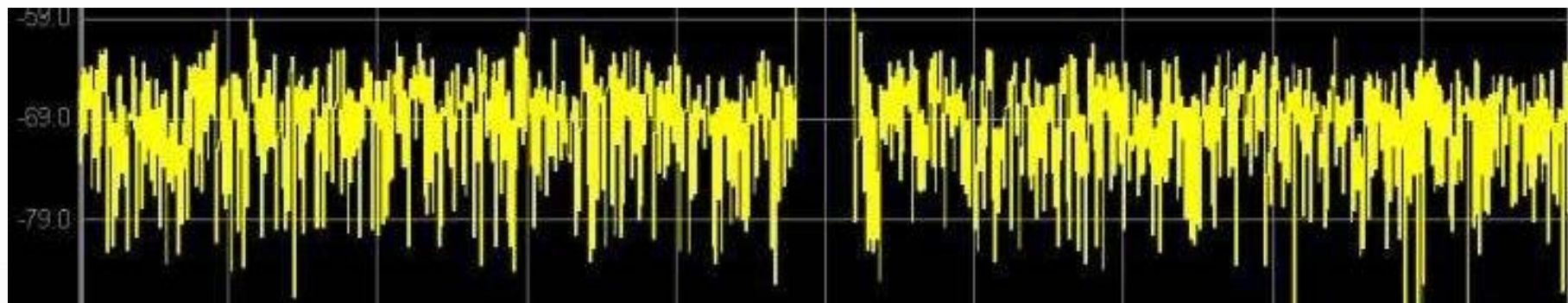
Method	Living Room 1495 Samples	Office Room 1533 Samples
PoseNet	0.60m, 3.64°	0.46m, 2.97°
4D PoseNet	0.58m, 3.40°	0.44m, 2.81°
CNN+LSTM	0.54m, 3.21°	0.41m, 2.66°
ours	0.36m, 4.36°	0.31m, 2.47°

- Better position accuracy, Comparable orientation accuracy;
- Much fewer images in database construction period (Training images vs. Reference images);
- 3D-Modeling Free;
- Training Free;
- A set of images with high-precision pose is the key.

Visual Positioning With Depth Camera

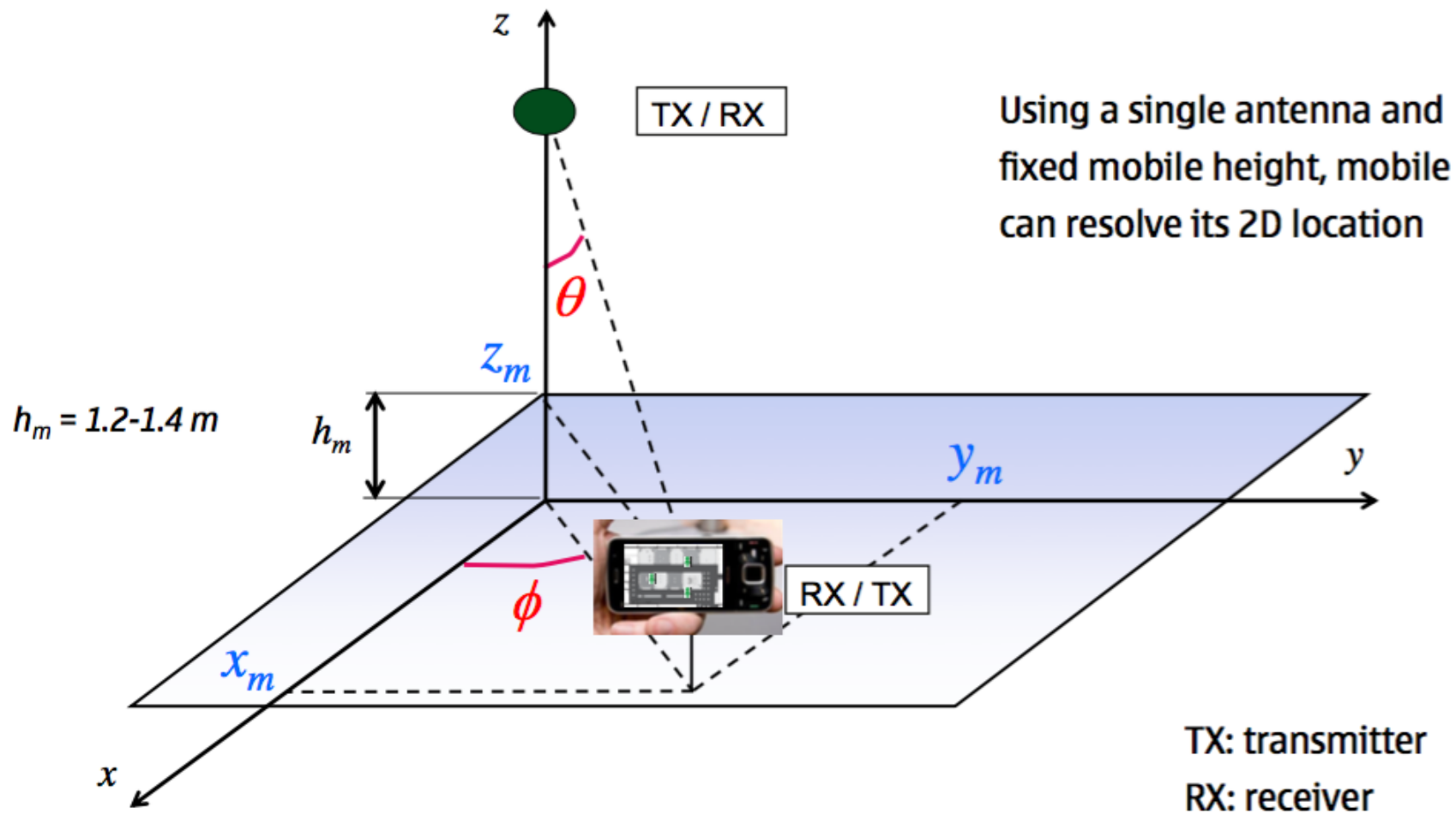


Positioning Based on RF Signal



Nokia BLE Antenna Array

Localization Principle with a Single Positioning Beacon



Positioning With an BT Antenna Array



+



- A pseudolite-based approach
- Broadcast BS positions in WGS-84
- TTFF (Time To First Fixed) 0.1 Sec.
- Low-cost, easy for installation
- Positioning update rate 1-10Hz
- Positioning accuracy: 1-2m

Wi-Fi Round Time Trip Ranging

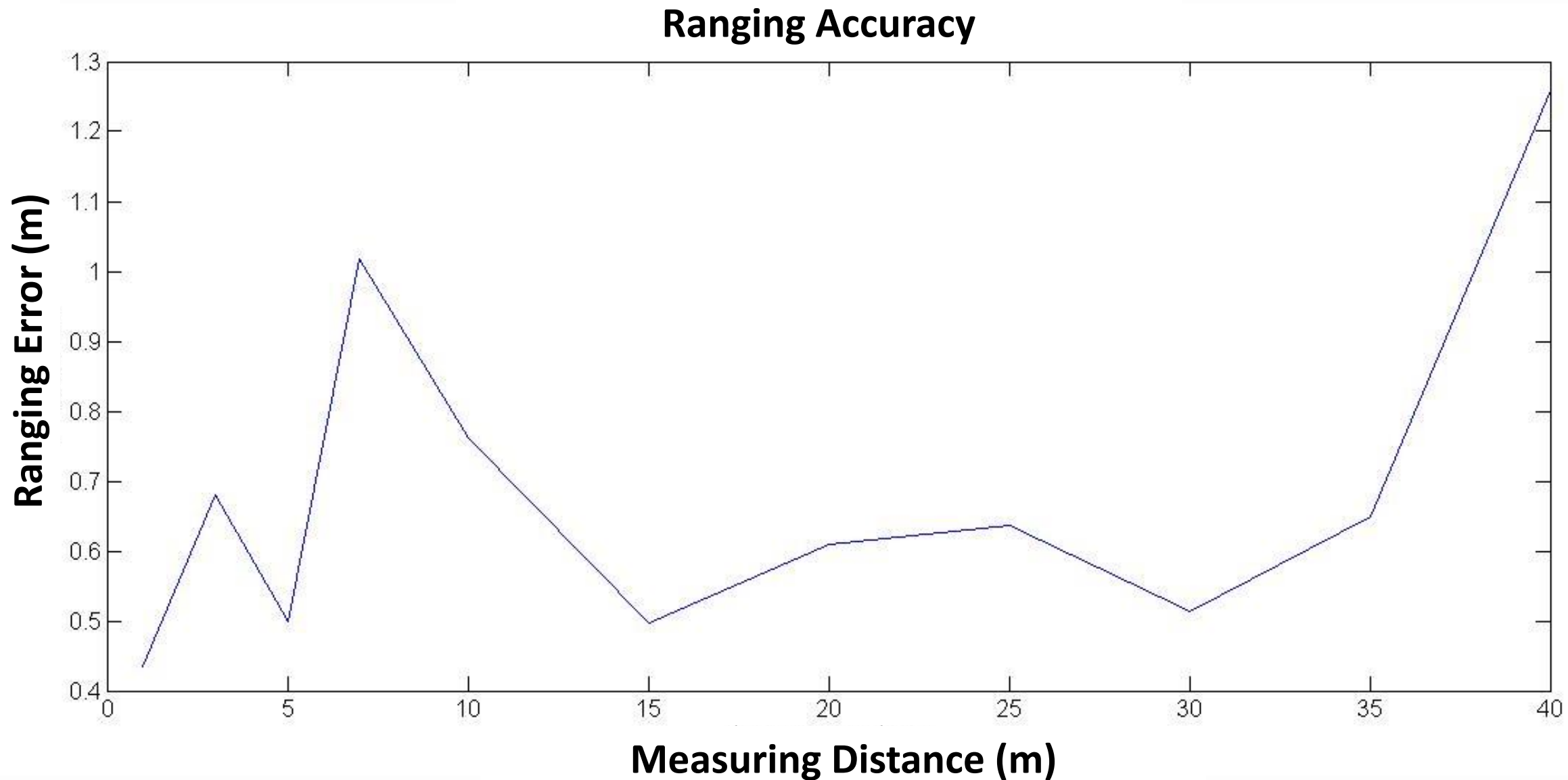


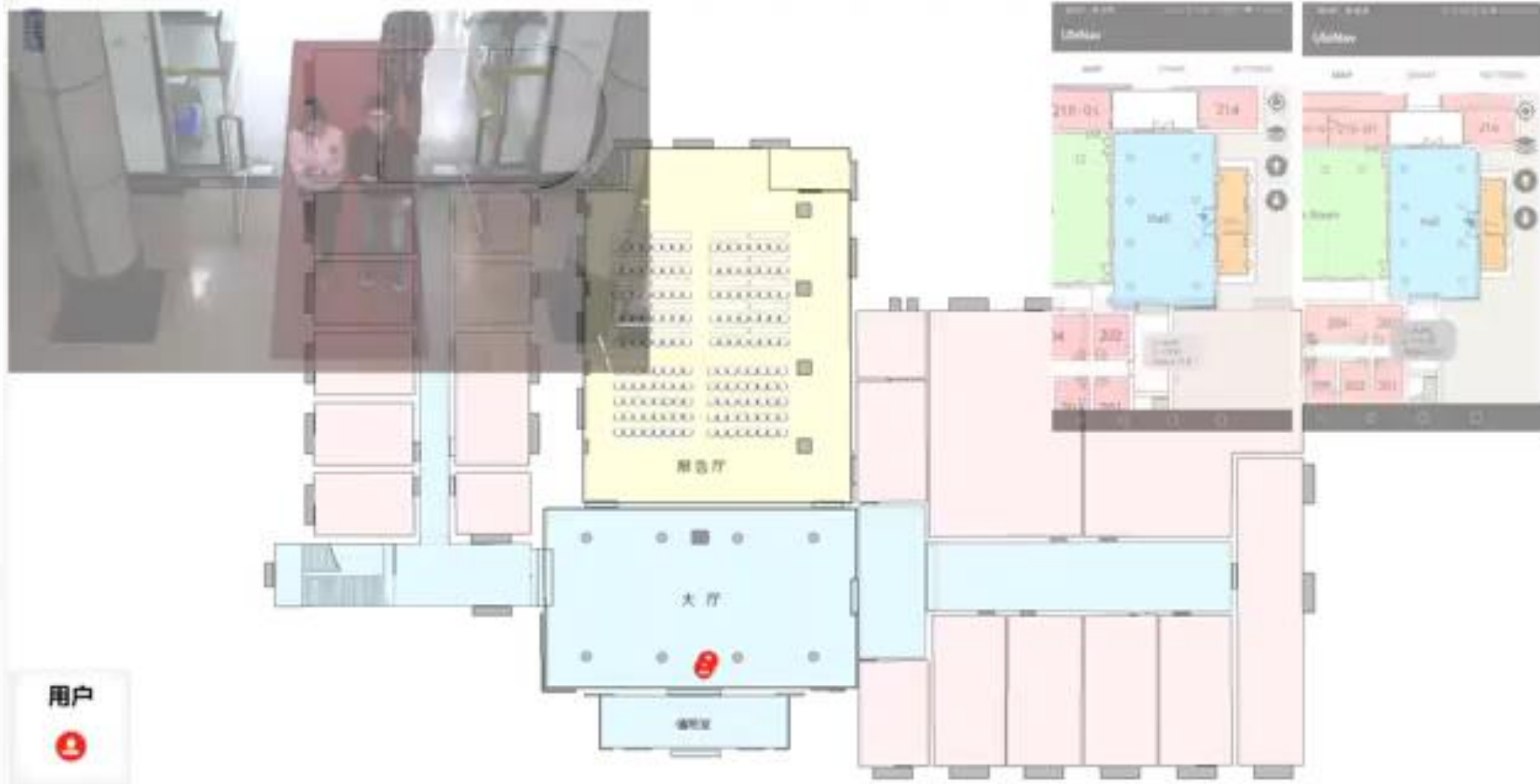
Wi-Fi AP



Based on 802.11mc

Wi-Fi –RTT Ranging Accuracy





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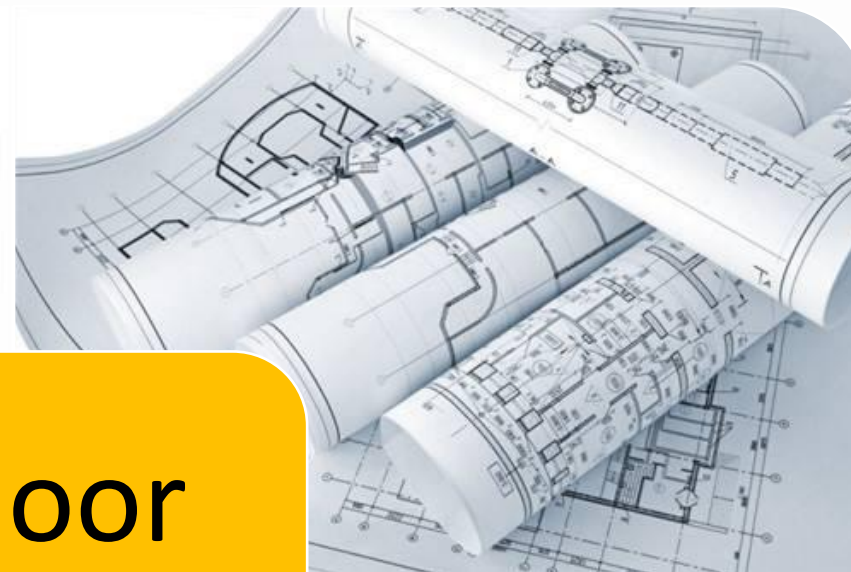
Precise Smartphone Positioning Based on Built-in Sensors and RF Radios

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Indoor Mapping

Indoor Mapping Approaches

Scanned
FloorPlan



CAD/BIM

Crowded
Sources

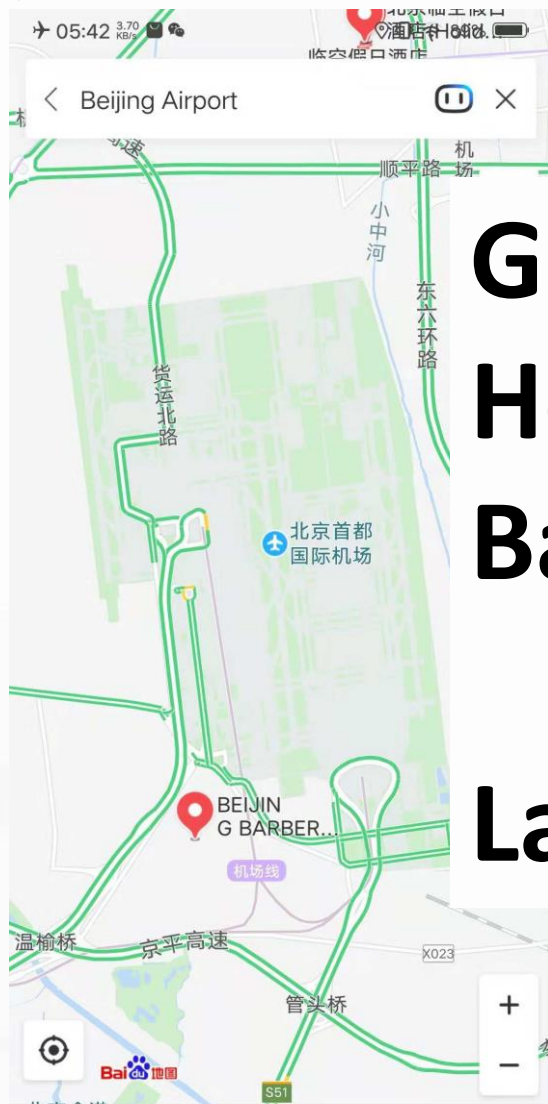


Indoor Mapping

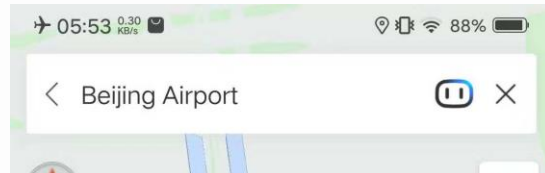


SLAM

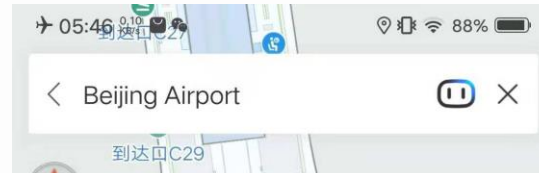
An Example of Indoor Map



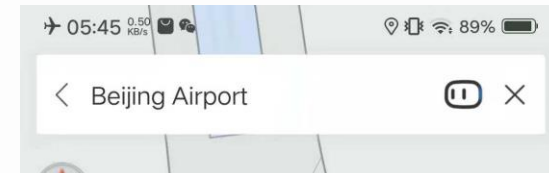
共找到 "Beijing Airport" 相关3个结果



共找到 "Beijing Airport" 相关3个结果



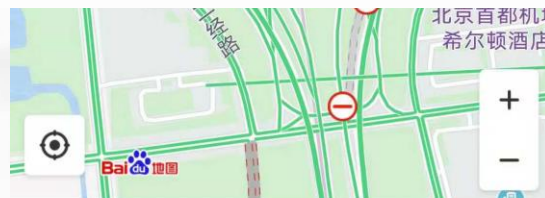
共找到 "Beijing Airport" 相关3个结果



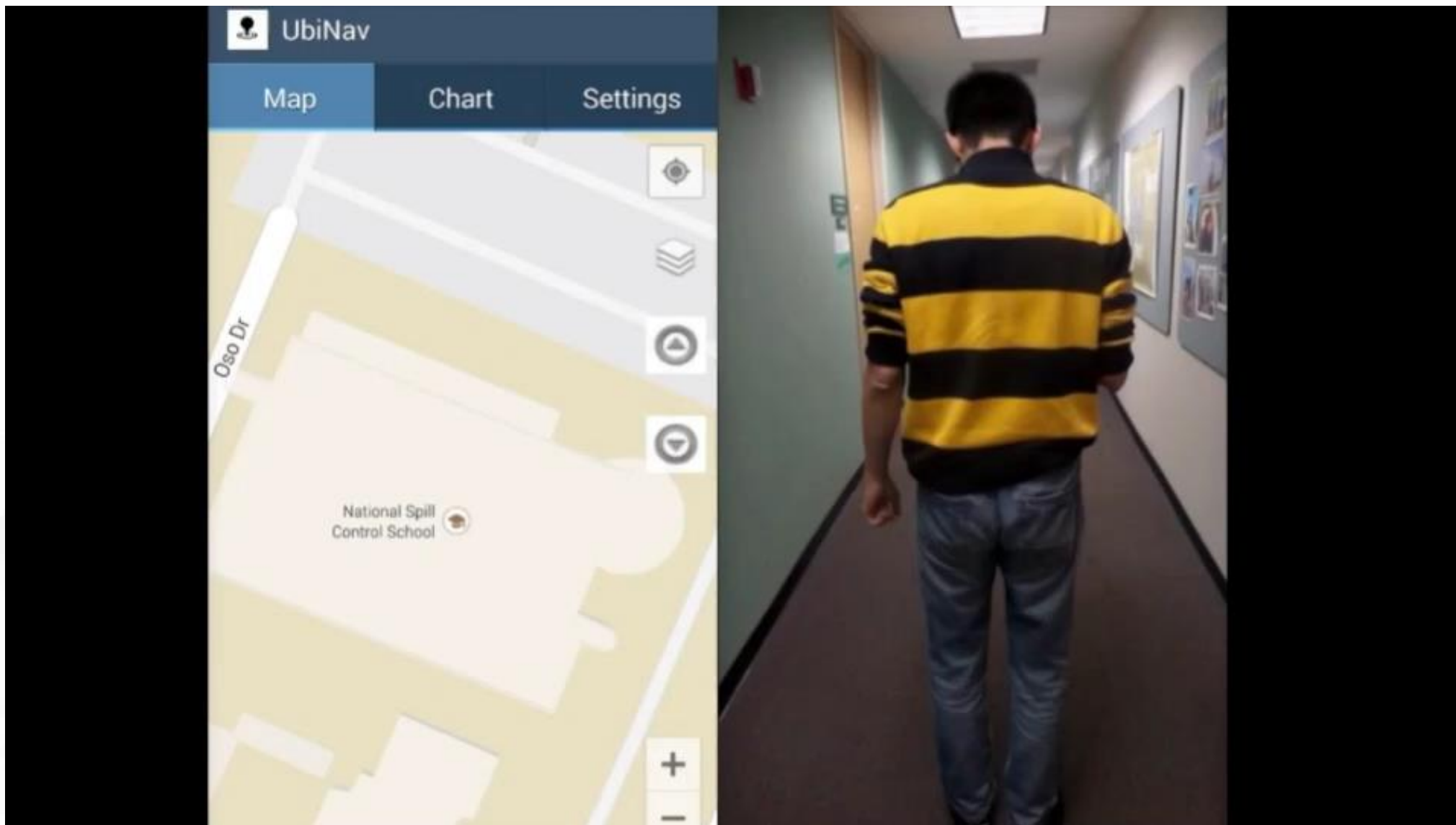
共找到 "Beijing Airport" 相关3个结果

Google: 10000+
Here Map: 50000+
Baidu: 2000+ in China

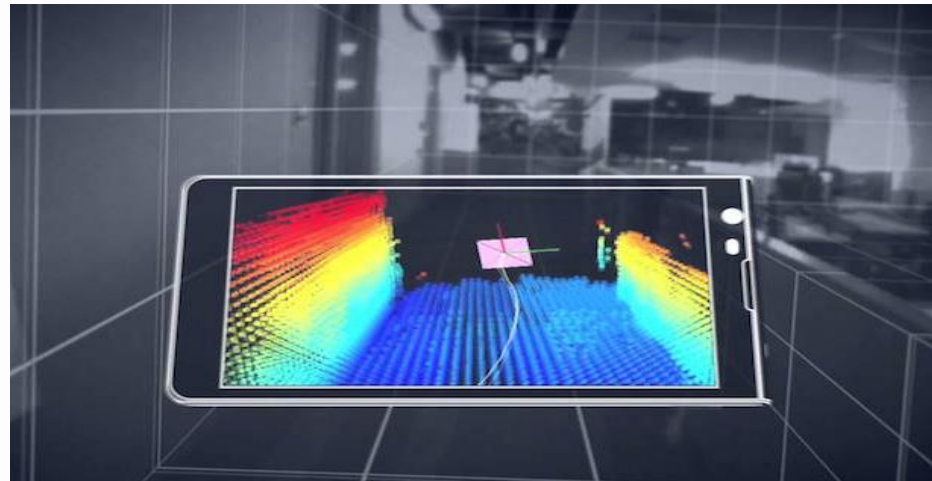
Lack of international standard



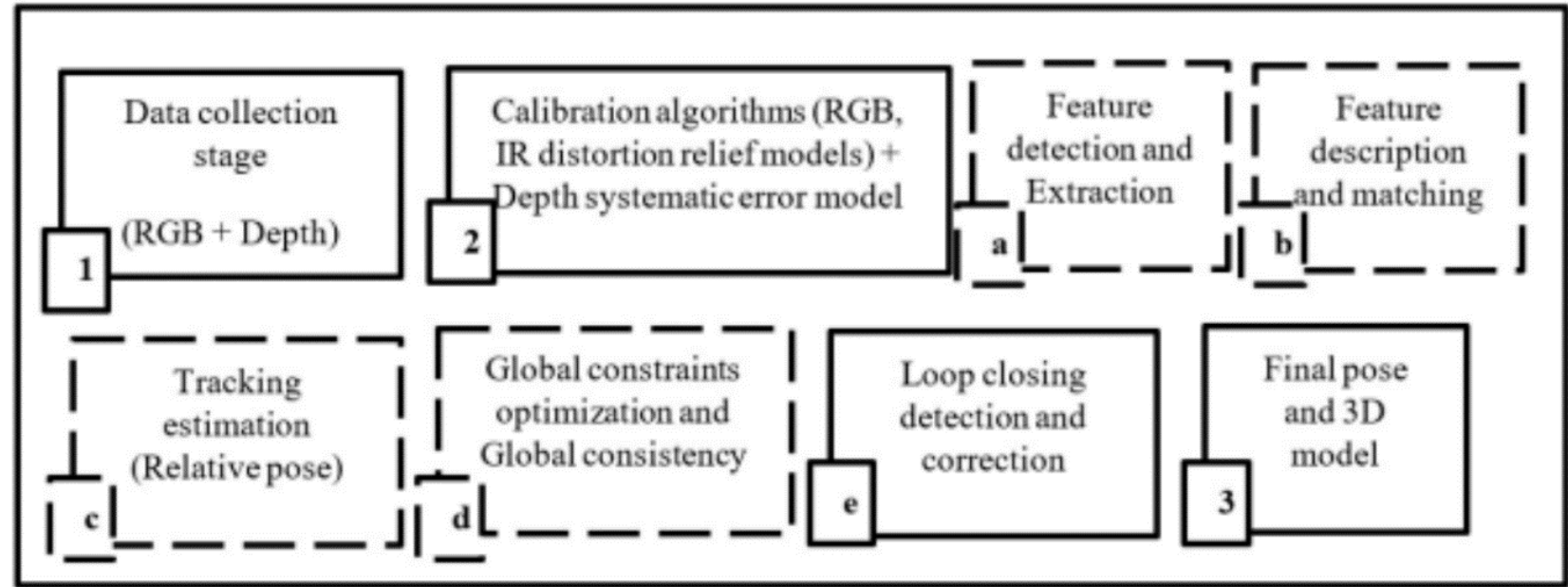
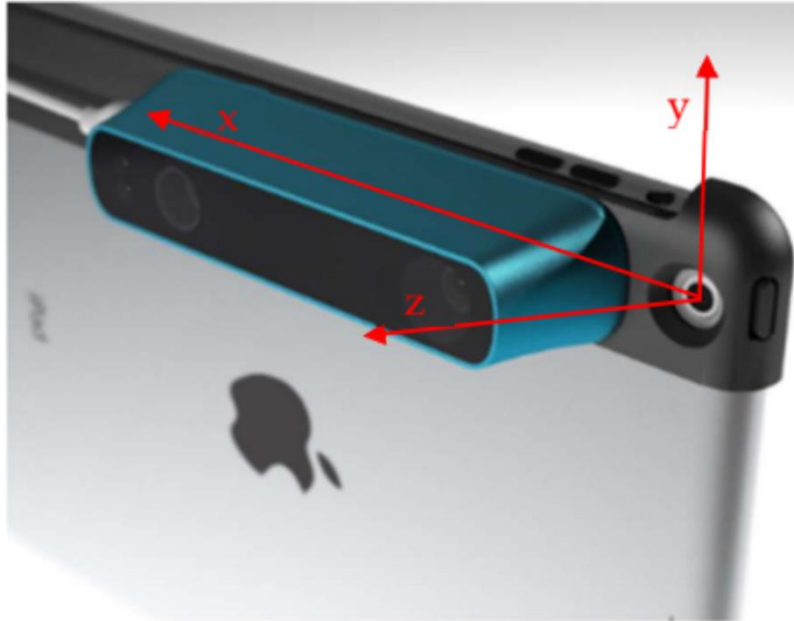
Indoor Mapping Demo



3D Modelling Based on Depth Camera



A Mobile SLAM Solution – FC RGB-D SLAM



PhD Thesis: **PRECISE RECONSTRUCTION OF INDOOR ENVIRONMENTS USING RGB-DEPTH SENSORS**

By

WALID ABDALLAH ABOUMANDOUR DARWISH

Supervisor: Wu Chen

Department of Land Surveying and Geo-Informatics

Hong Kong Polytechnic University

Hong Kong Central Metro Station



自強弘毅
求是拓新



Conclusions

- There are lots of positioning technologies for indoor, however, there is no such an indoor positioning technology that works like GNSS for outdoor.
- Using the built-in sensors and RF radios, smartphone positioning can achieve an accuracy of about 2-5meters in real time and about 1 meter by post processing.
- High precise indoor positioning technologies are capable to deliver centimeter level accuracy, but effective coverage of a single base station is limited. The new Wi-Fi ranging technology will resolve this problem partly.
- Integration of multiple positioning sources is probably the best option for complex indoor environments.
- Mobile devices with depth camera are capable of deliver 3D indoor models.

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求是拓新



Thank You!