

# Deterministic indoor detection from dispersions of GPS satellites on the celestial sphere

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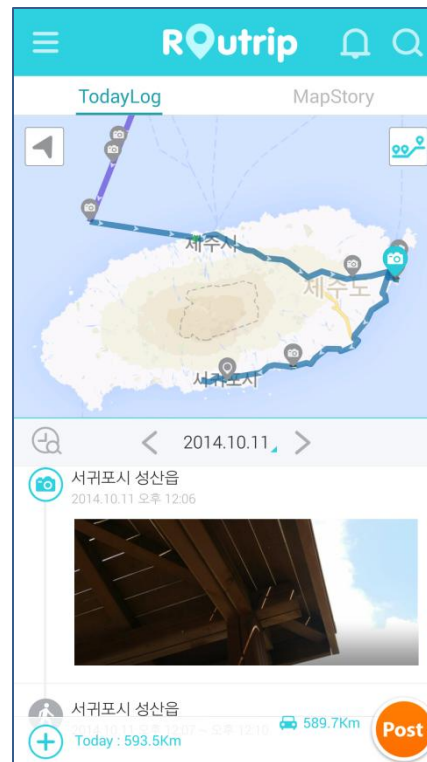
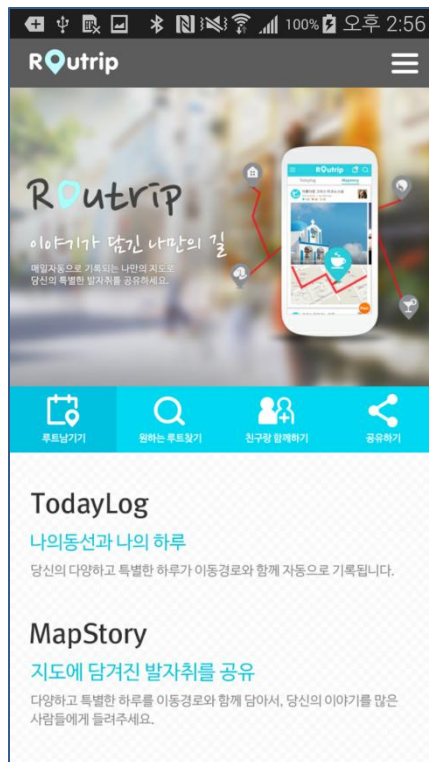
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# Background

- We are making a location-based service, called Routrip.
  - TodayLog : Record my whole day movements with my activities.
  - MapStory : Compose the routes to a story and share with friends.

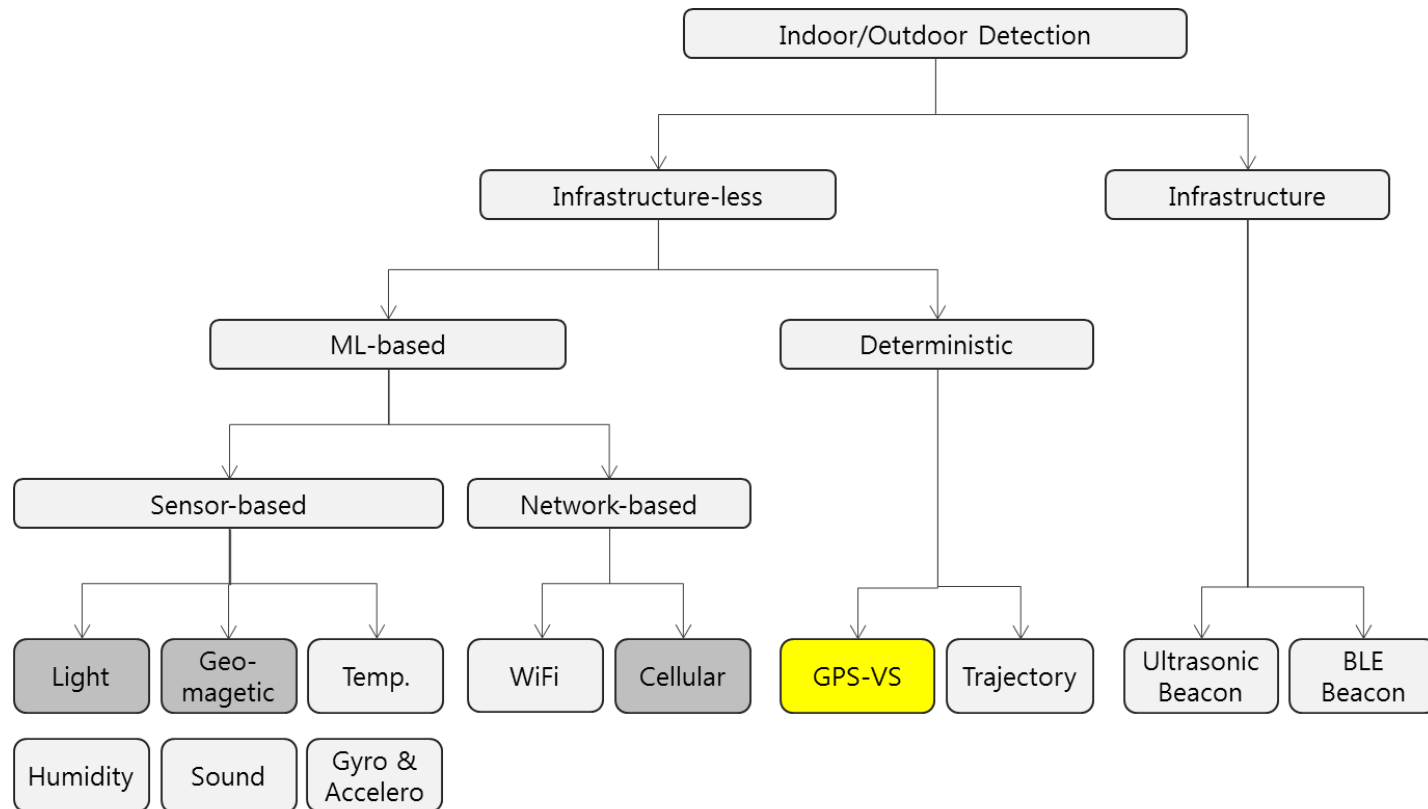


# Background

- One of the biggest hurdle is the power consumption while tracking the location of users continuously.
  - Coarse sampling → loss in movement points
  - Fine sampling → battery drain
- After many trial and errors, Routrip consumes
  - 1.6mA in the stationary state
  - 5.6mA in the walking/running state
  - 18.3mA in the vehicle/transport state
- As most users spend much time in the stationary state, we want to reduce the power consumption while stationary much less.

# Indoor Detection

- If the mobile device could detect that it is inside buildings, the service can use longer intervals to check its location.



# ML-based vs. Deterministic

- In machine-learning based approaches,
  1. Collect the sample data (training set)
  2. Generate the classifier model using ML algorithms
  3. Decide the status from real-time test data
- As the classifier model is generated by training set data, we need sample data as many as possible for high accuracy, but even with many data, the model may be over-fitted to the specific environments.
- So, we want to make a method to detect indoor or outdoor without any prior knowledge on the environments.

# GPS & NMEA

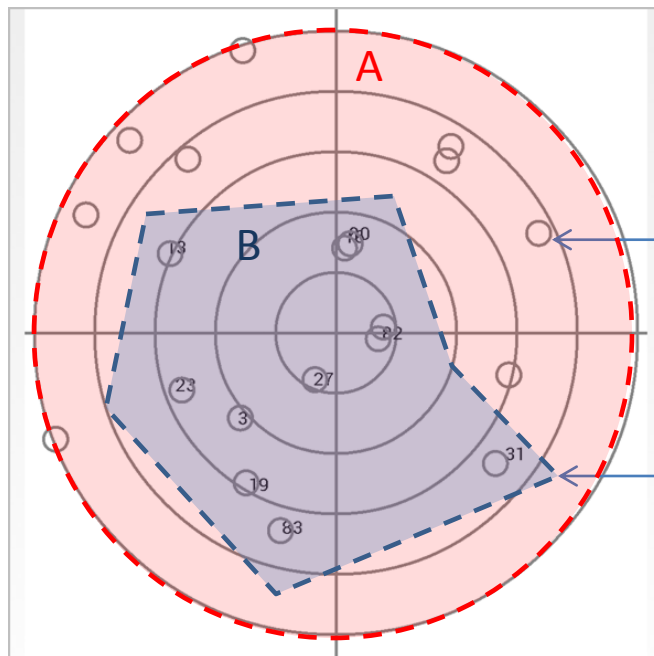
- The fixed location from GPS chipset is delivered to the host system using the format of NMEA0183 standard.
  - GPGGA: Global Positioning System Fix Data
  - GPGSV: GPS Satellites in View

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$GPGGA,092750.000,5321.6802,N,00630.3372,W,1,8,1.03,61.7,M,55.2,M,,*76
$GPGSA,A,3,10,07,05,02,29,04,08,13,,,,,1.72,1.03,1.38*0A
$GPGSV,3,1,11,10,63,137,17,07,61,098,15,05,59,290,20,08,54,157,30*70
$GPGSV,3,2,11,02,39,223,19,13,28,070,17,26,23,252,,04,14,186,14*79
$GPGSV,3,3,11,29,09,301,24,16,09,020,,36,,,*76
```

- Usually, we concern only fix data for the current location.
- But, there are more information for all visible satellites including their azimuth and elevation.

# Plotting on the Celestial Sphere

- First, we calculated the position of all visible GPS satellites using information from NORAD and NASA.
- And, plotted both calculated and received GPS satellites information on the celestial sphere.



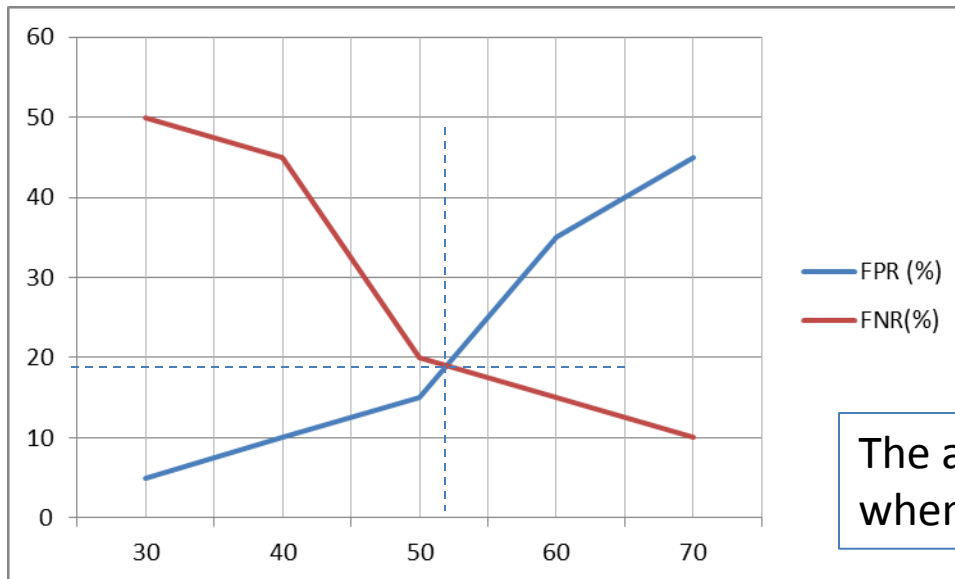
A GPS satellite should be there, but no or weak signal received.

DOOS (Degree-of-open-skies)  
 $= (\text{Area of B}) / (\text{Area of A})$

# Determining DOOS threshold

- We measured false positive and false negative ratio according to the threshold value of the DOOS metric.

|                | Ground Truth | System Result |
|----------------|--------------|---------------|
| False positive | Outdoor      | Indoor        |
| False negative | Indoor       | Outdoor       |

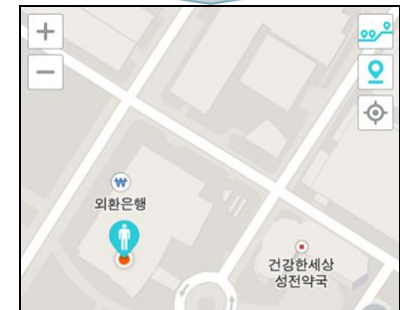
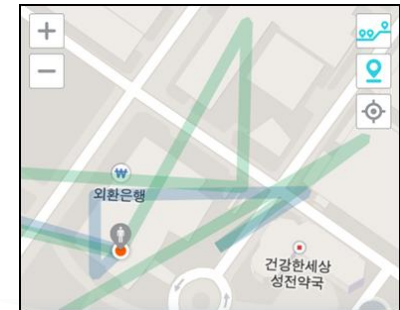
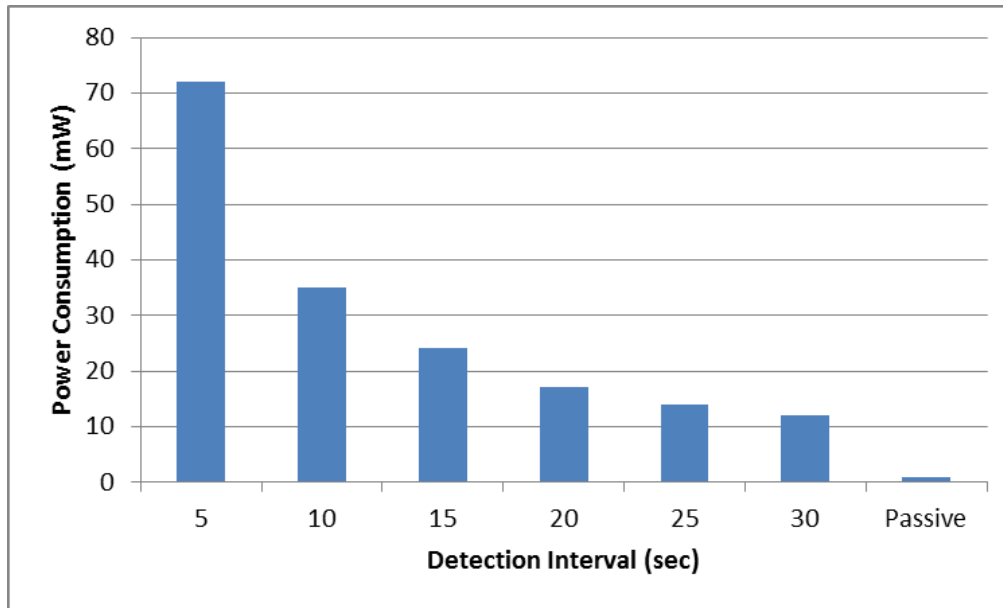


The accuracy shows optimal when the DOOS threshold is about 53.



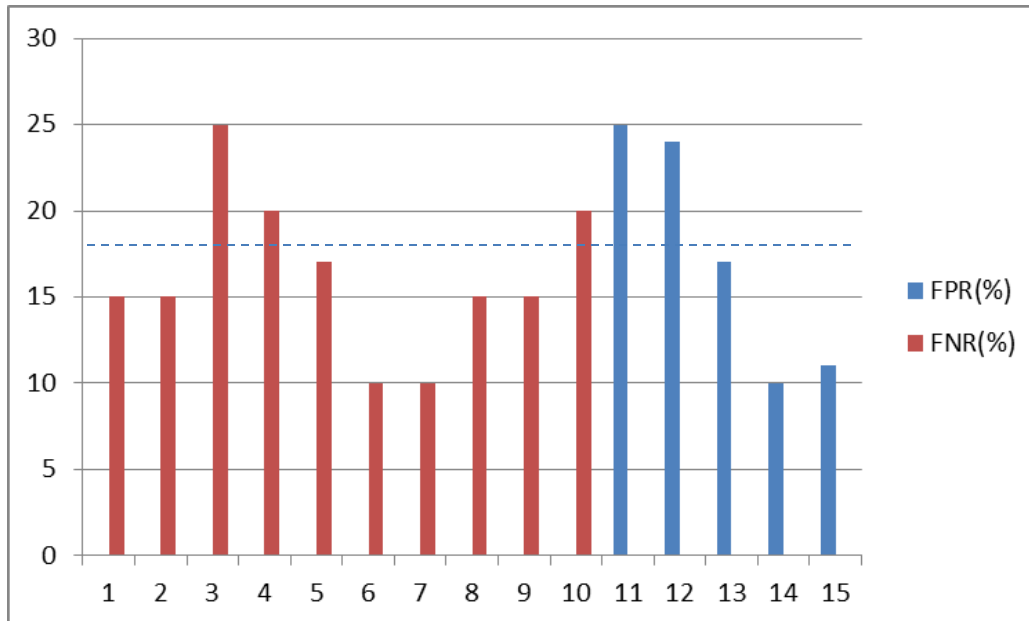
# Power Consumption

- Periodic use of GPS might consume much power.
- But the detection mechanism can utilize the GPS location requests from other location-based services passively.



# Availability

- To evaluate the availability, we selected 15 identical locations and performed the detection 10 times over two days.



The false ratio showed about 17%

# Conclusion

- Detecting indoor/outdoor status of the smart phone is valuable context to most location-based services.
- We tried to find a solution with
  - Low-power consumption
  - High availability
  - High accuracy
- The accuracy cannot meet the QA requirement currently, but we will try to improve the method itself and to hybridize with sensor-based approaches.