

vCityMap: Crowdsensing Towards Visible Cities

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Visualization of City

- Sound
- Roads with respect to maintenance
- View (Google Maps, Google Earth)
- ...
- ...

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600m

600m

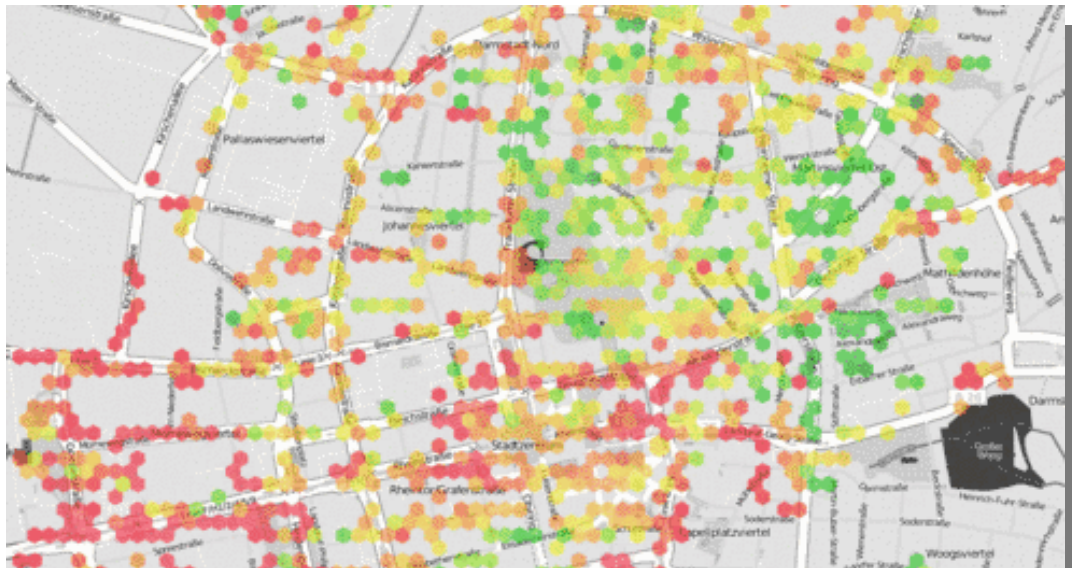
Crowdsensing

- or Participatory Sensing for data with non-real-time property

Sensor Data Collection in Wide-Area by People

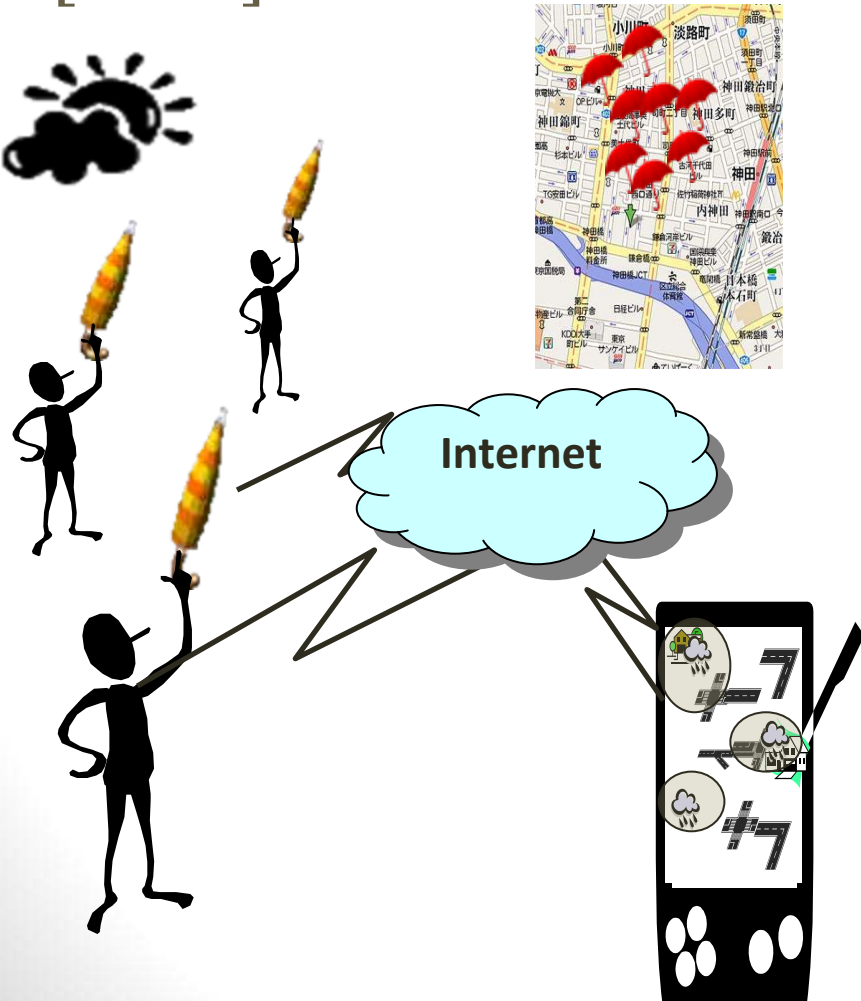
(Example □ Noisemap - Discussing Scalability in Participatory Sensing

□ Meurisch, 2013 □

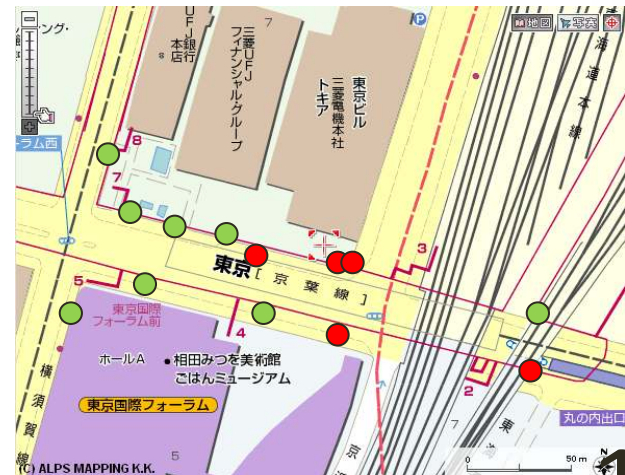
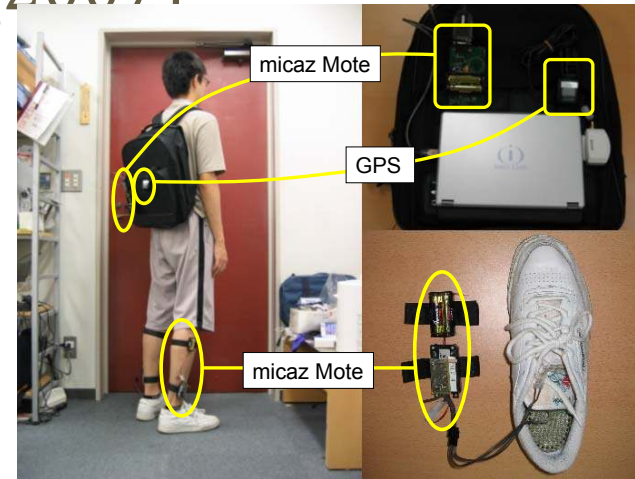


Our Previous Gadgets

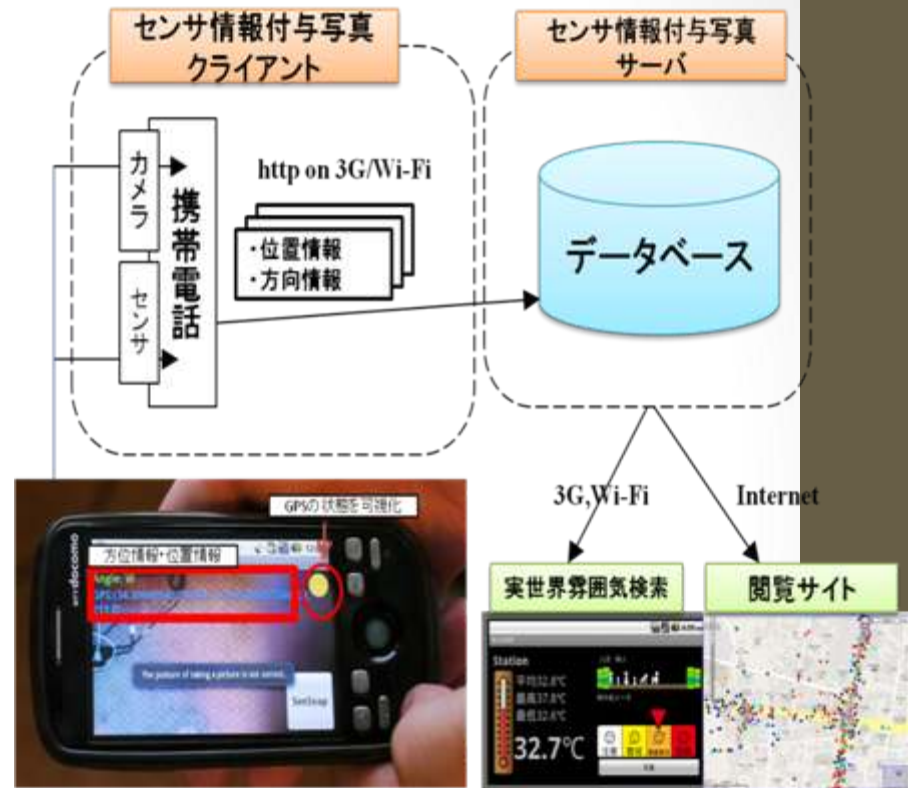
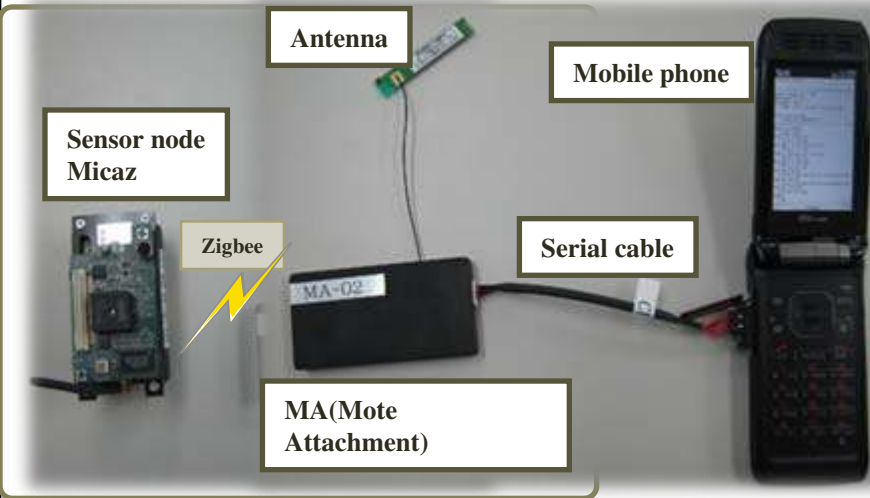
Umbrella Human Probes [2005]



Shoes Human Probes [2007]



Additional Dedicated Hardware ?



Dedicated Hardware

Software

Sensors are embedded into Phones.

Visualization of City

- Sound
- Roads with respect to maintenance
- View (Google Maps, Google Earth)
- ...
- ...

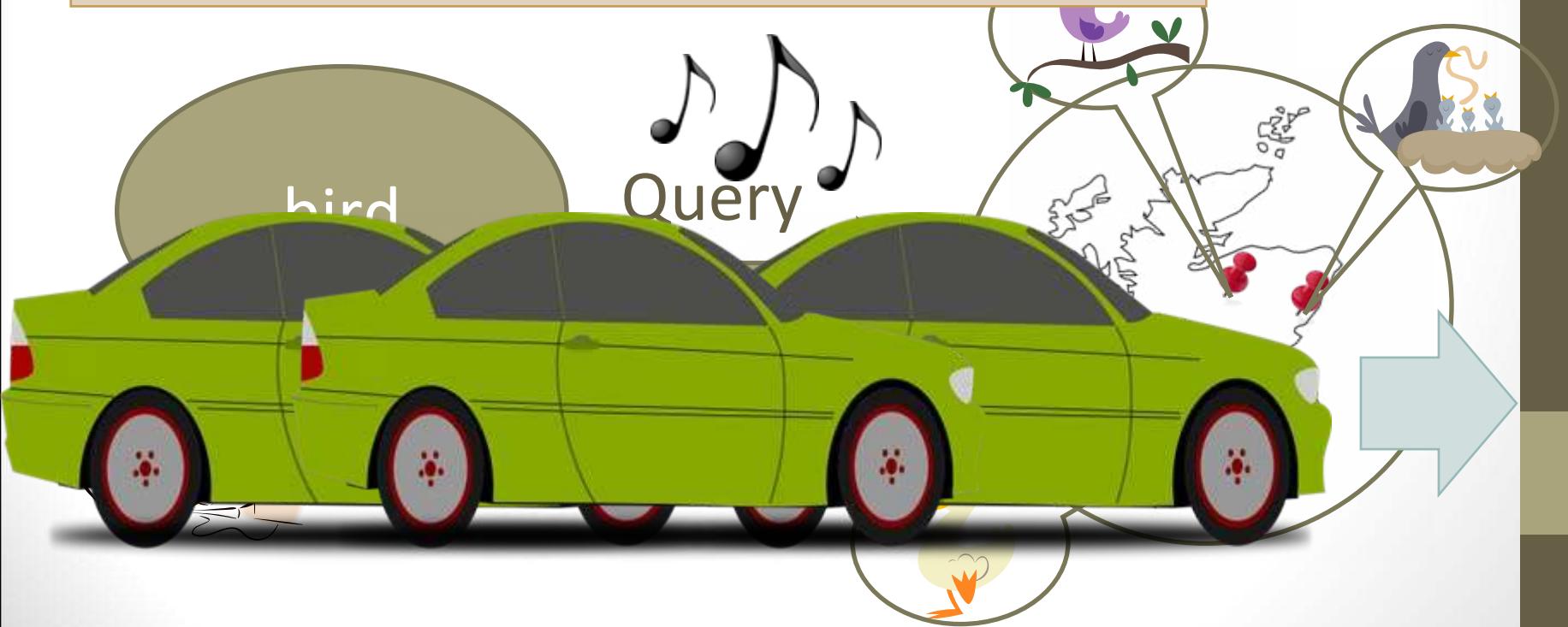
Why Sound ?

- Sound represents the atmosphere of Place
- Less concern about privacy
- Analyzable

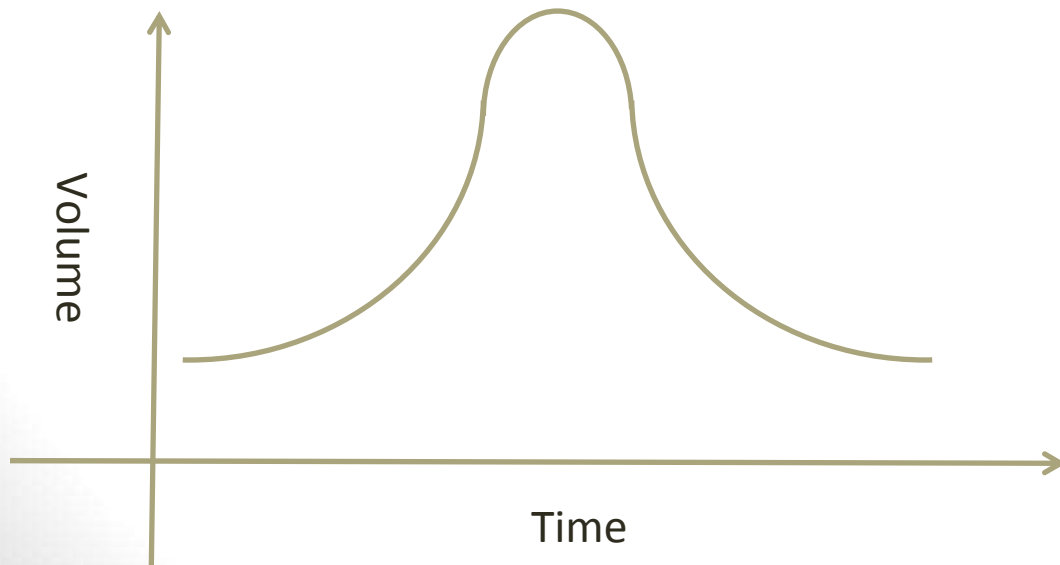
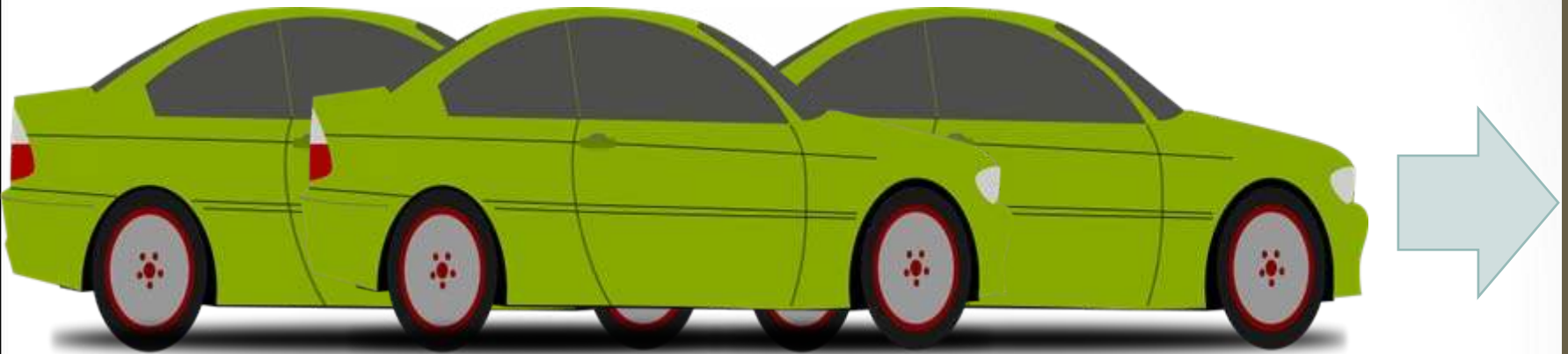
First Goal

- Classification of sound for queries
 - what kind of sound
 - Which area has a particular kind of sound ?

Estimating rough number of passing cars

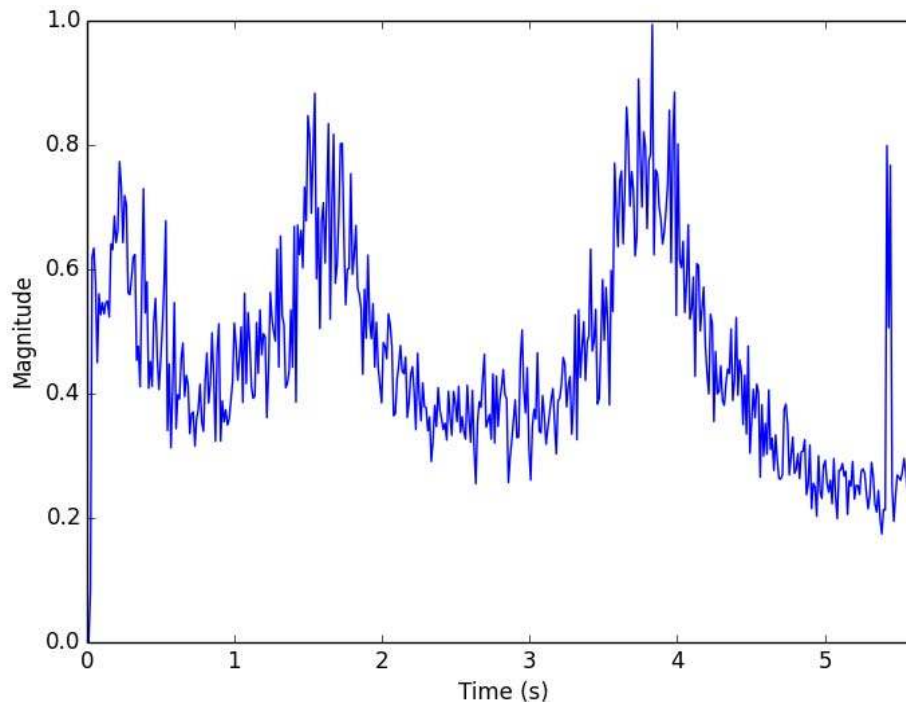


Classification Method



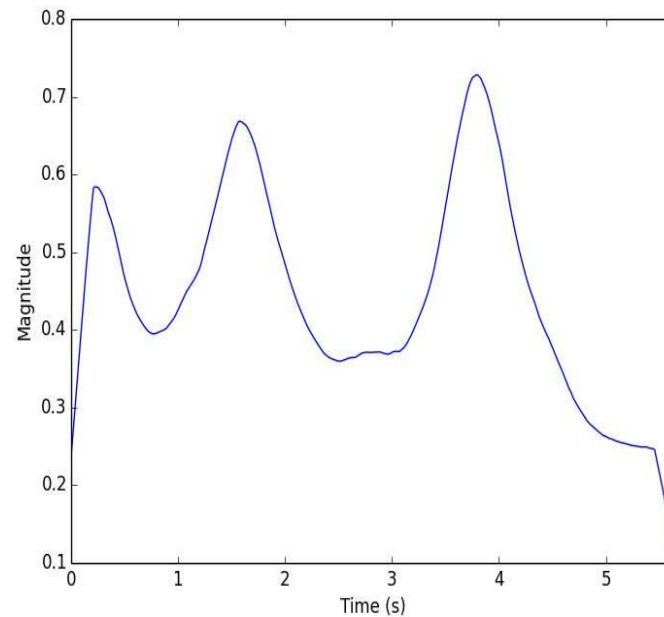
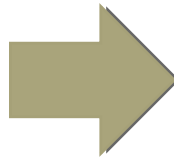
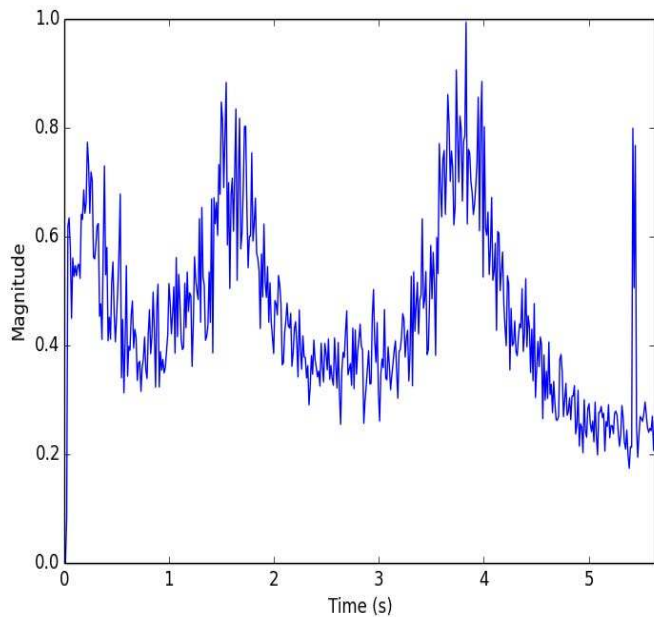
Preprocessing

- Applying short-time Fourier Transform to the originally acquired signal
- Calculate the mean of spectral power of 2-3 kHz



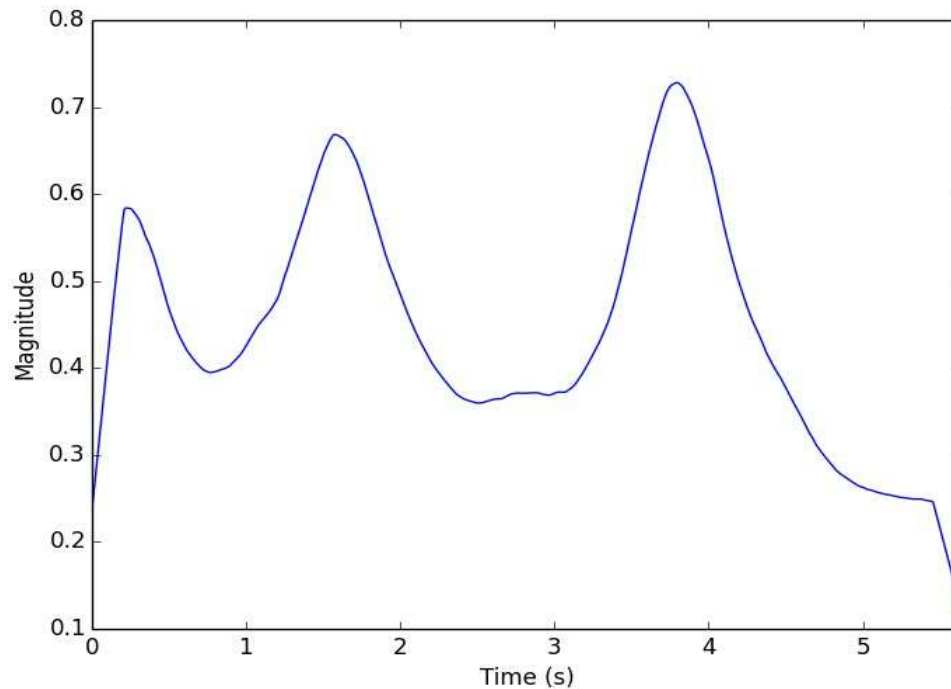
Filtering

- Applying moving average and median filter to subband mean time series
- Filtering impulse noise by median filter
- Smoothing signal transition by moving average filter



Determination

- Filtered Signal $\square S_i$ ($i = 1, 2, \dots, N$)
- Differential Signal $\square \Delta S_j = S_j - S_{j-1}$ ($j = 1, 2, \dots, N-1$)



Preliminary Experiment

Format	Wave 16-bit LPCM
Sampling rate	44.1 kHz
Recording time	5 □ 10 sec
Channels	Mono

- 100 sound files were collected with and without passing-by cars.
- Car sound
 - A recorder is on a sidewalk
 - 2-lane road
- Non-car sound
 - Inside a train
 - On campuses
 - On streets where pedestrians were walking



Experimental Result

	Car	Non-Car
Correct	45	48
Error	5	2
Accuracy	90%	96%

Correct	Error	Mean Error per file	Accuracy
38	12	1.08	76%

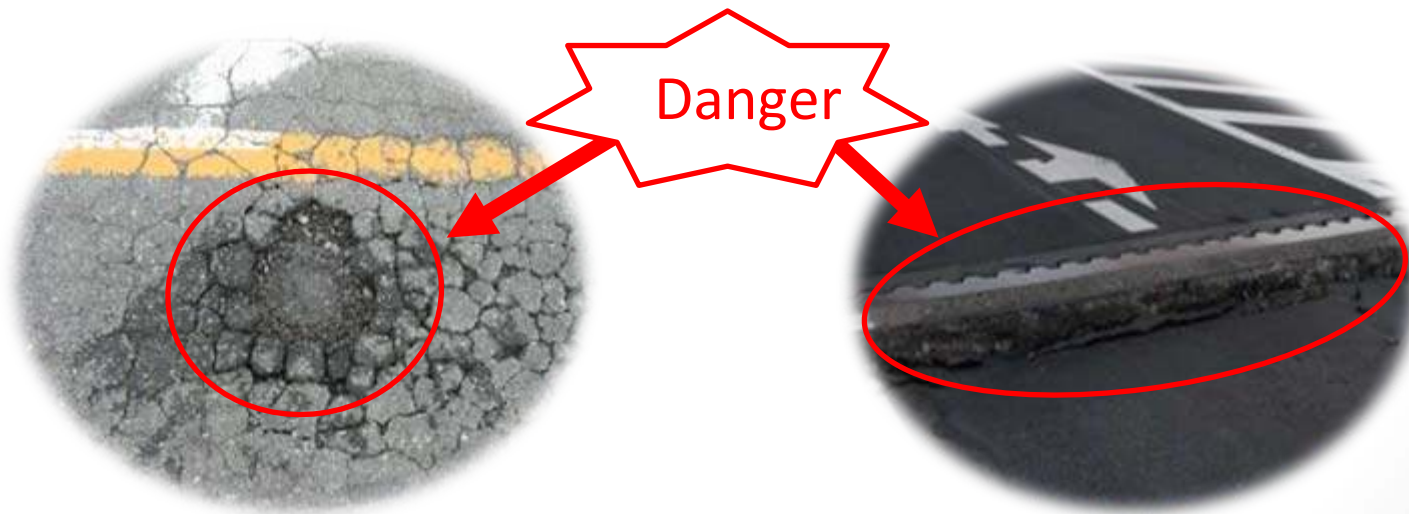
Visualization of City

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Road Surface

- Long-Term Maintenance of Social Infrastructure
 - Roads, Cycling Roads

Needs Early Detection of Anomaly.



Our Approach

- YKOB □ Your Kinetic Observation Bike □
 - ✓ Investigation by Participatory Bike Riders
 - ✓ Android Phones

Collection &
Transmission

Analysis

Visualization



Accelerometer
+
GPS Signal



YKOB

Collection of Data

Extraction of Road-
Surface Signal

Detection of Anomaly

Classification of
Anomaly

YKOB

Collection of Data

- Smartphone in a pocket of bike rider

Extraction of Road-Surface Signal

Detection of Anomaly

Classification of Anomaly

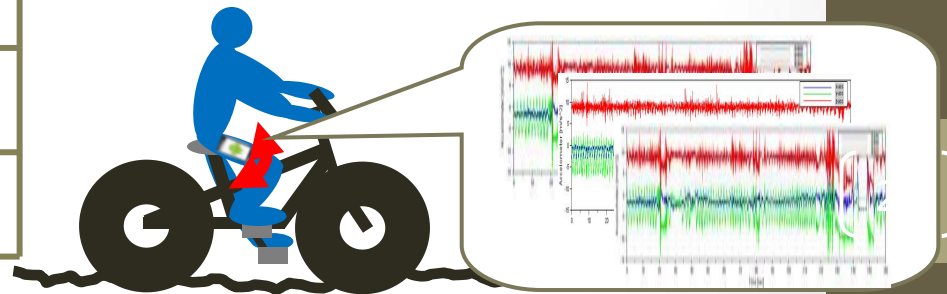
Collection of Data

- Installation
 - Pocket of pants
- Format

Time	Latitude	Longitude	X_{acc}	Y_{acc}	Z_{acc}
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- Conditions

Sampling Frequency	50 Hz
Assumed Speed	12~14 km/h
Type of Bike	City Bike



YKOB

Collection of Data

Extraction of Road-Surface Signal

▶ Collected signal contains signals other than source from road surface

□ Extract only road-surface signal

Detection of Anomaly

Classification of Anomaly

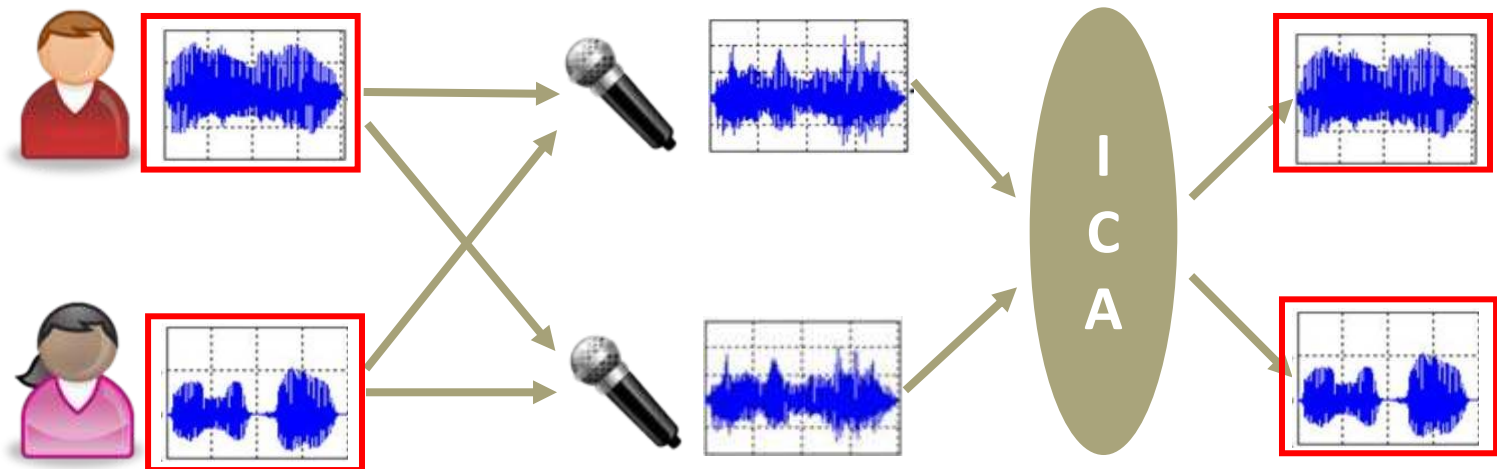
Extraction of Road-Surface Signal (□/□)

- ▶ Mixture of Pedaling and Road-Surface Signals

Separation based on ICA

- Independent Component Analysis (ICA)


Separate Independent Signals from Observed Signals

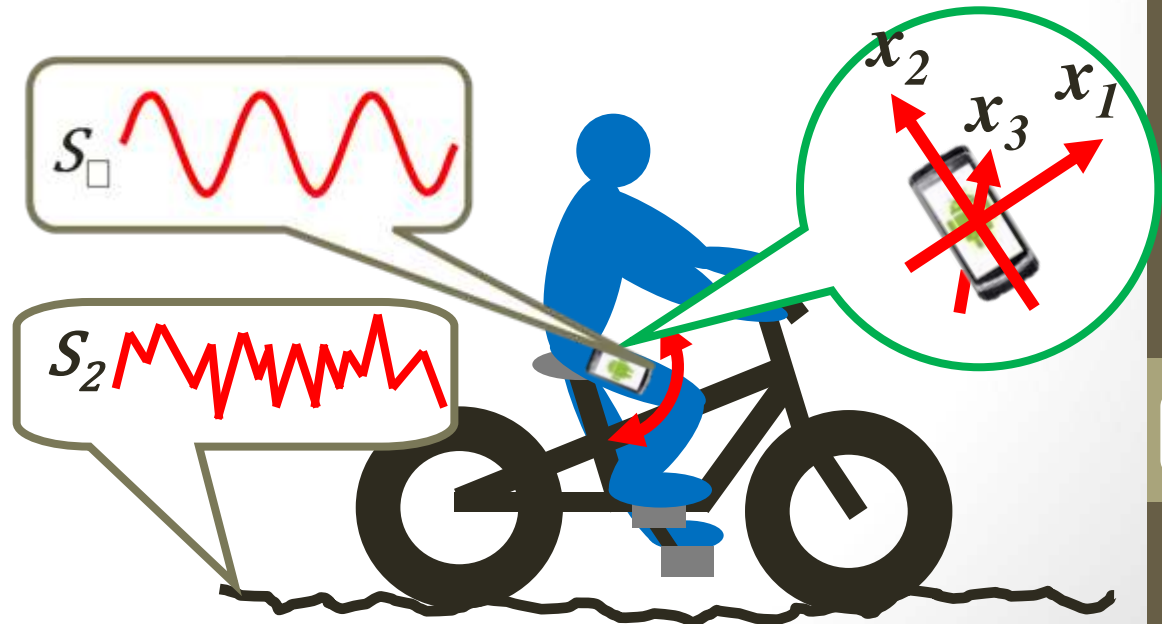


Extraction of Road-Surface Signal(□/□)

▶ Applying ICT to the raw signal

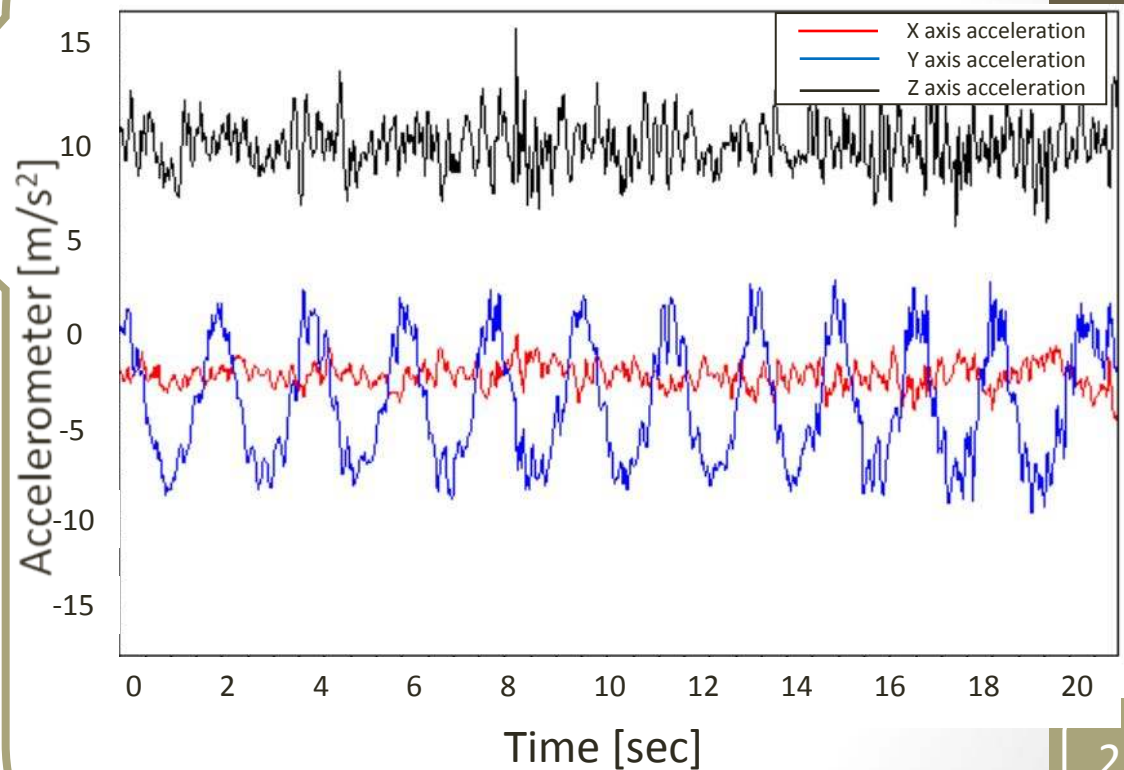
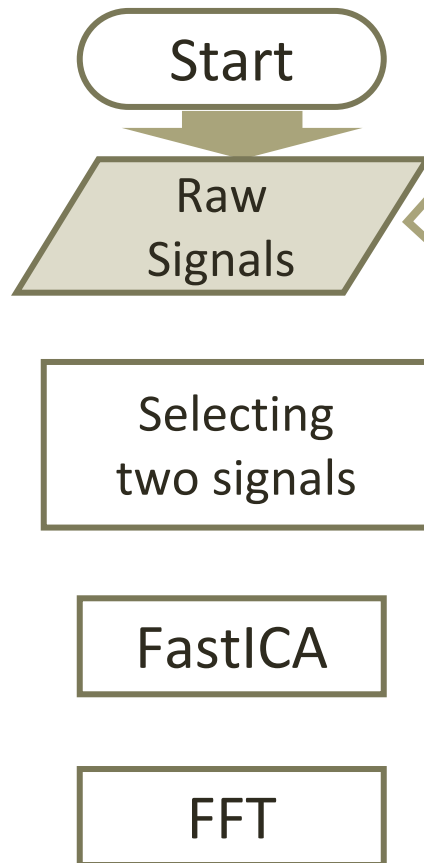
- Independent Signals $s = \begin{bmatrix} s_1 \\ s_2 \end{bmatrix}$ (s_1 : Peddaling, s_2 : Road Surface)
- Observed Signals $x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$ (x_1 : x acceleration, x_2 : y accel., x_3 : z accel.)
- Mixing Matrix A

$$x = As$$

$$\hat{s} = \hat{A}^{-1}x$$



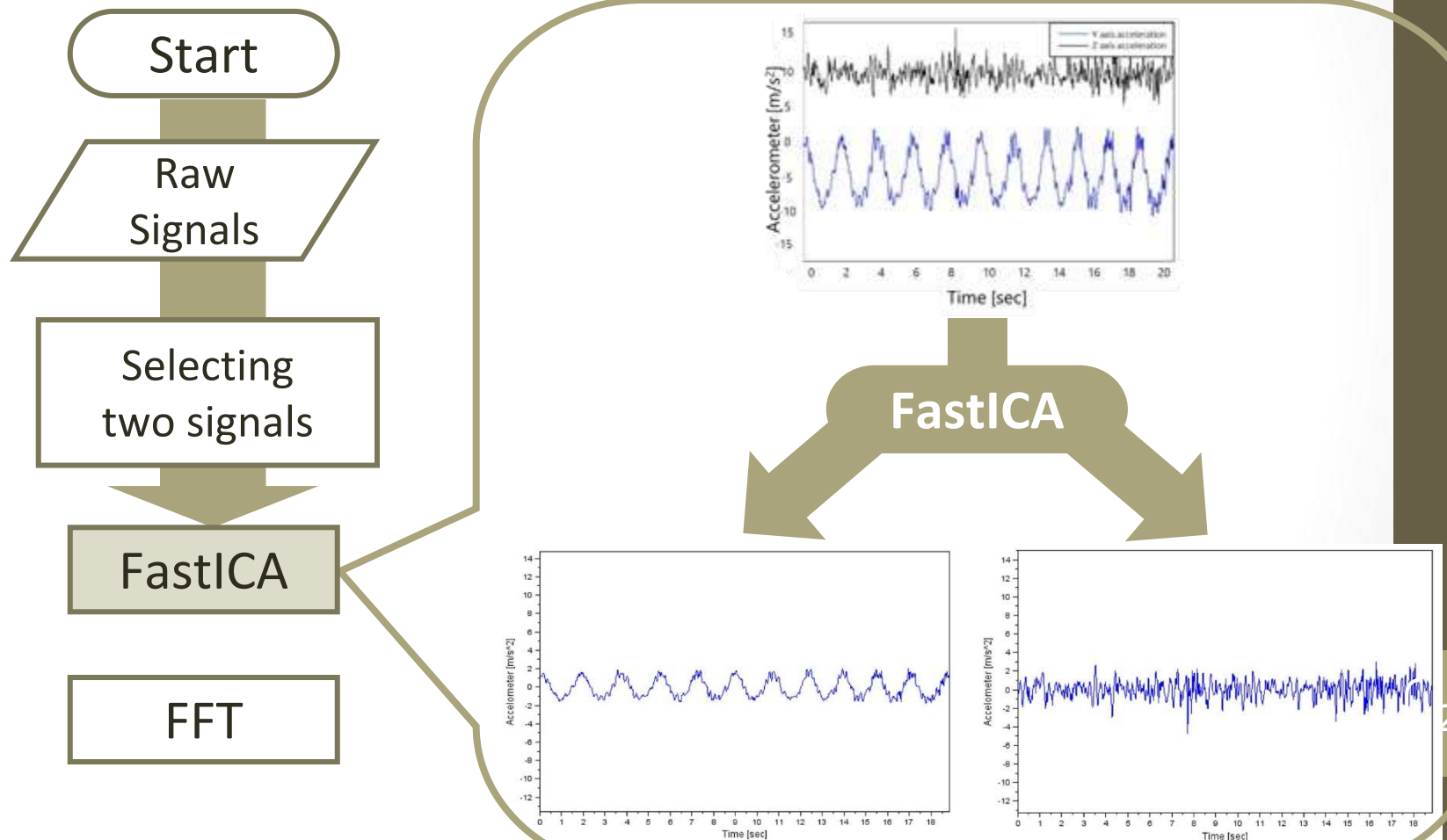
Extraction of Road-Surface Signal(3/5)

- Obtaining Acceleration Signal



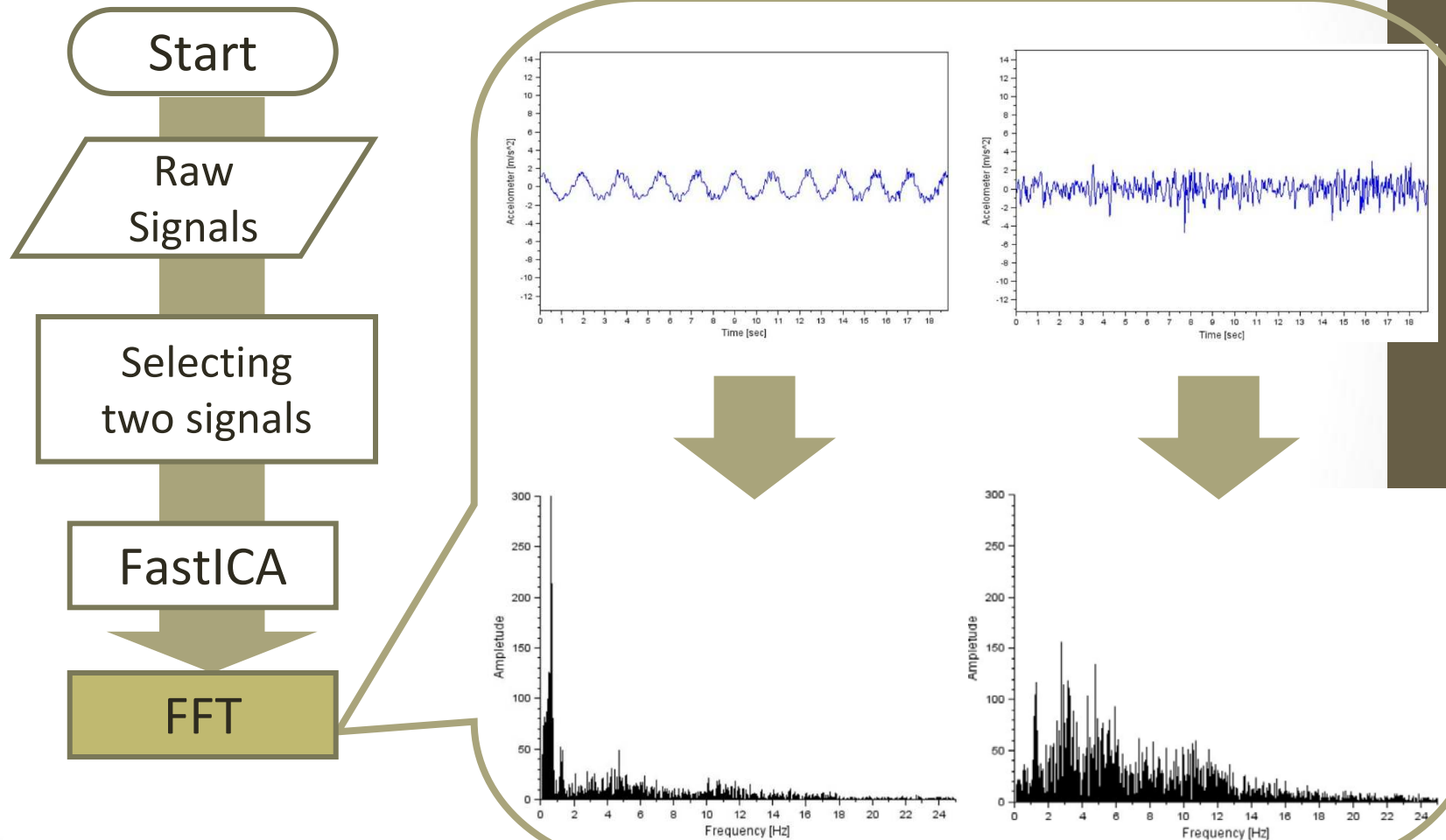
Extraction of Road-Surface Signal (4/5)

- Separation of signals using FastICA



Extraction of Road-Surface Signal (5/5)

- Extraction of Periodicity using FFT



YKOB

Collection of Data

Extraction of Road-
Surface Signal

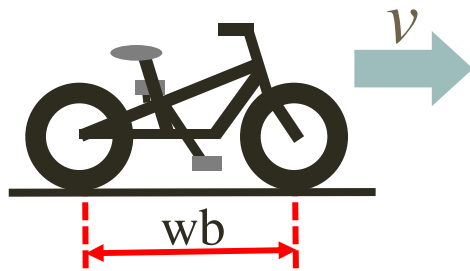
Detection of Anomaly

▶ Algorithm of Detecting Anomaly Location

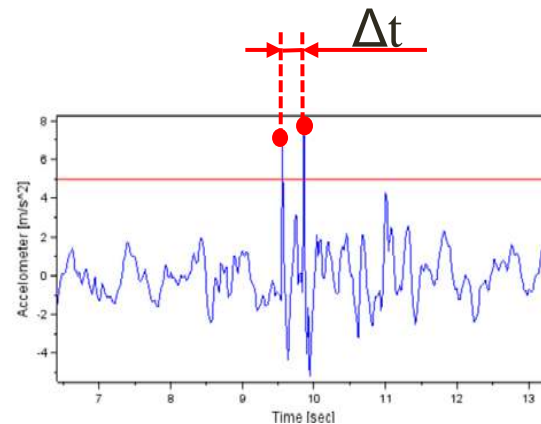
Classification of
Anomaly

Detection of Anomaly

- Use two peak values corresponding to the front and the rear wheel



$$wb = 1102 \text{ [mm]}$$



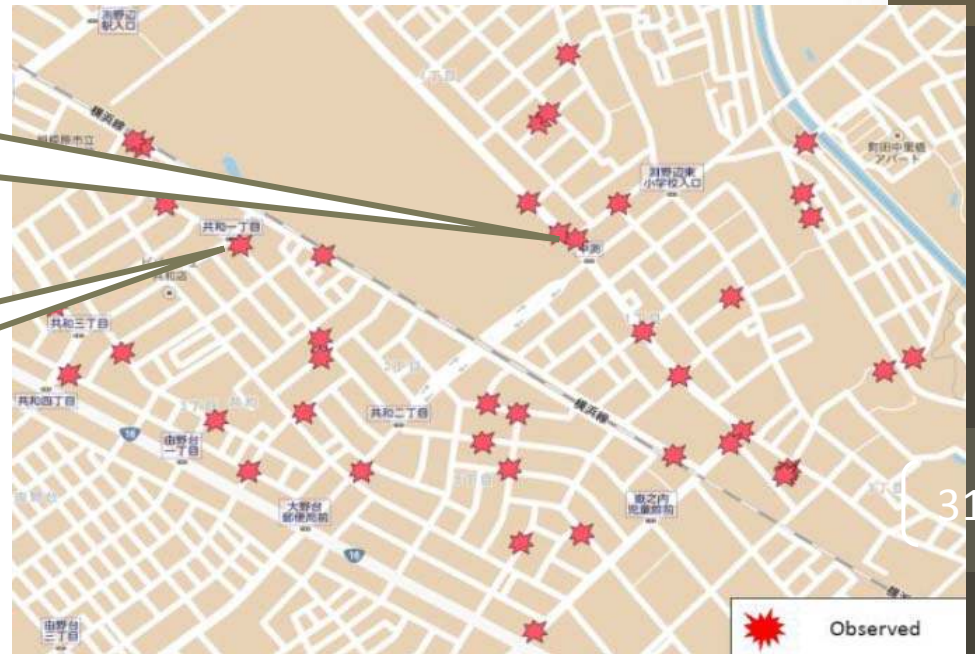
$$\left\{ \begin{array}{l} \Delta t = 0.30 \text{ [s]} \\ v = 13.4 \text{ [km/h]} \end{array} \right.$$

$$d = \Delta t * v = 1102 \text{ [mm]}$$

- Threshold

$$\Delta t < 0.4 \text{ s (based on experiments)}$$

Detection of Anomaly □ □ / 3 □



Detection of Anomaly □ 3/3 □

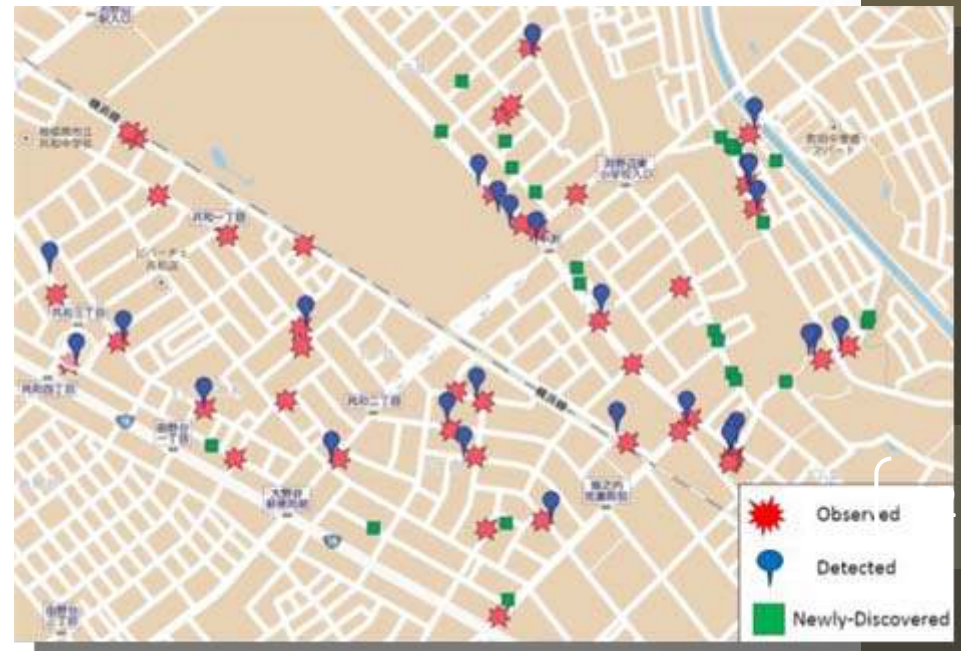
- Results

N_o : # of observed anomaly (42 locations)

N_d : # of correctly detected anomaly (28 locations)

N_n : # of incorrectly detected anomaly (26 locations)

	Definition	Ratio
Correct	$\frac{N_d}{N_o}$	0.67
Incorrect	$\frac{N_n}{N_d + N_n}$	0.48



vCityMap Sagamihara

- OpenStreetMap
- Collector
 - Android Application Downloadable from the Web
- Visualizer
 - JavaScript
- Participants
 - Began in July 2014
 - 9 participants
 - Facebook Community

vCityMap

[HOME](#)[MAP](#)[APPLICATION](#)[LINKS](#)

お問い合わせ

ご不明な点がございましたら、下記メールアドレスまでお問い合わせください。

vcity-info@rcl-aoyama.jp

はじめに

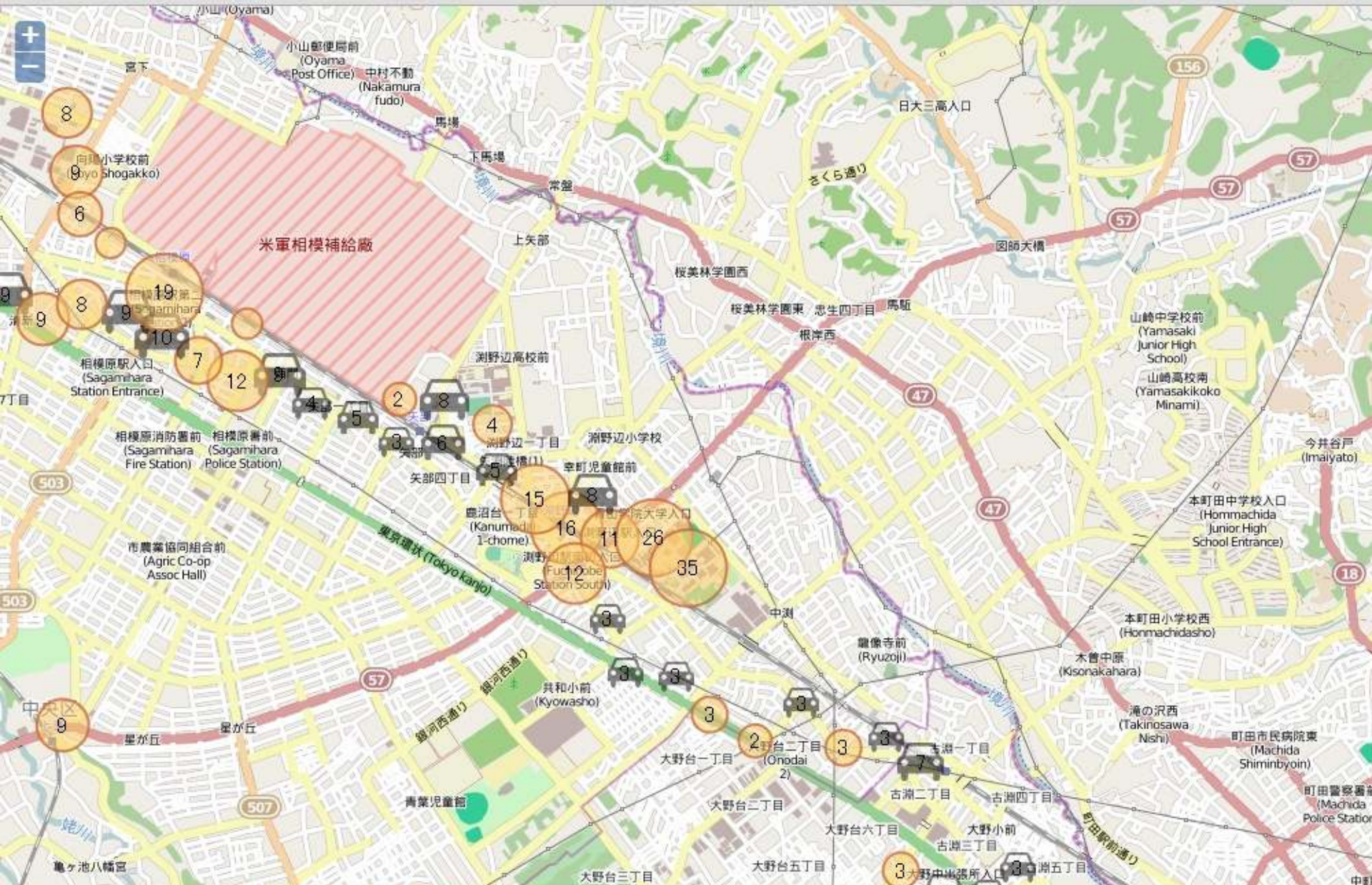
vCityMap-Sagamihara は、相模原市およびその周辺の環境の変化を地図とむすびつけて記録に残していこうという試みです。具体的には、スマートフォンを保持している方から、場所・時刻付きのデータをアップロードしていただき、サーバで管理します。あらゆる種類の状態取得に挑戦しますが、2014年はまず、

- (1) 環境音
 - (2) 自転車走行に伴う路面の状態
- の2つに取り組む予定です。

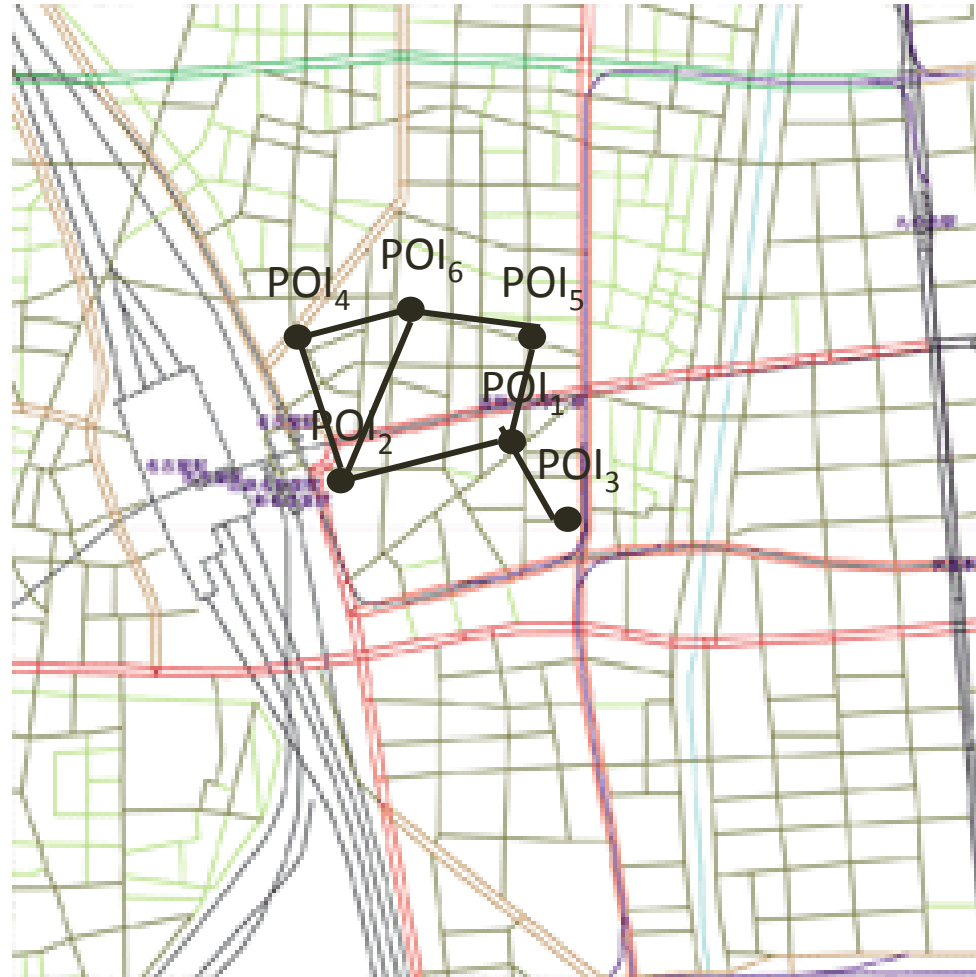
情報収集にご協力お願いいたします。

MAP





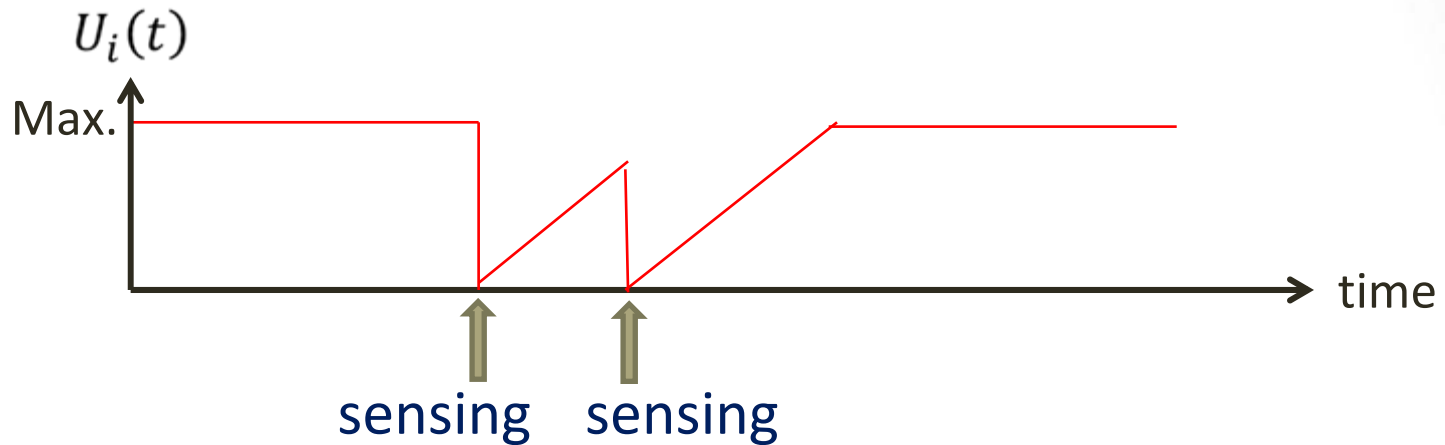
Point of Interest (POI)



Consumers determine POIs

Utility Function

Sensing data at a POI i are independent of other POIs



$$U(t) = \max\left(U_{min}, \min\left(U_{max}, a(t - t_{prev})\right)\right)$$

$U_i(t)$ = Utility at POI i at time t

U_{min} = Minimum utility

U_{max} = Maximum utility

a = Constant

t = Current time

t_{prev} = Latest sensing time at POI i

Summary

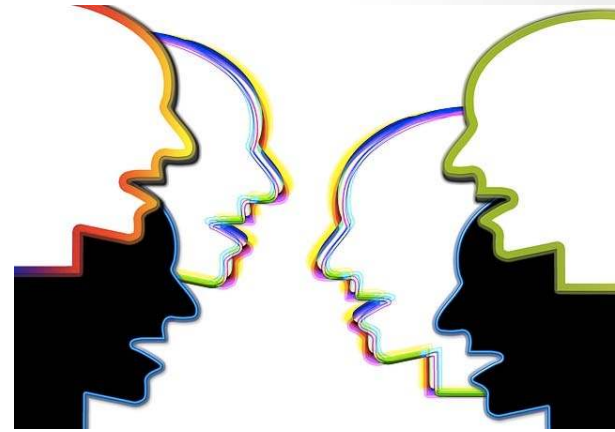
- We have created a crowdsensing kit.
- Starting from
 - Sound Maps
 - Road-Surface Status Maps
- Future Plans
 - Community in Android-phone users
 - Decision of POIs
 - Machine Learning for classification

Sensor-Data Efficiency Index

- $\mu = C / (U + C)$
 - C: Consumed Data
 - U: Unconsumed Data
- Low μ means energy inefficiency
 - “Hey, don’t produce that many data!”
- Not Consumed by One Person

Discussion

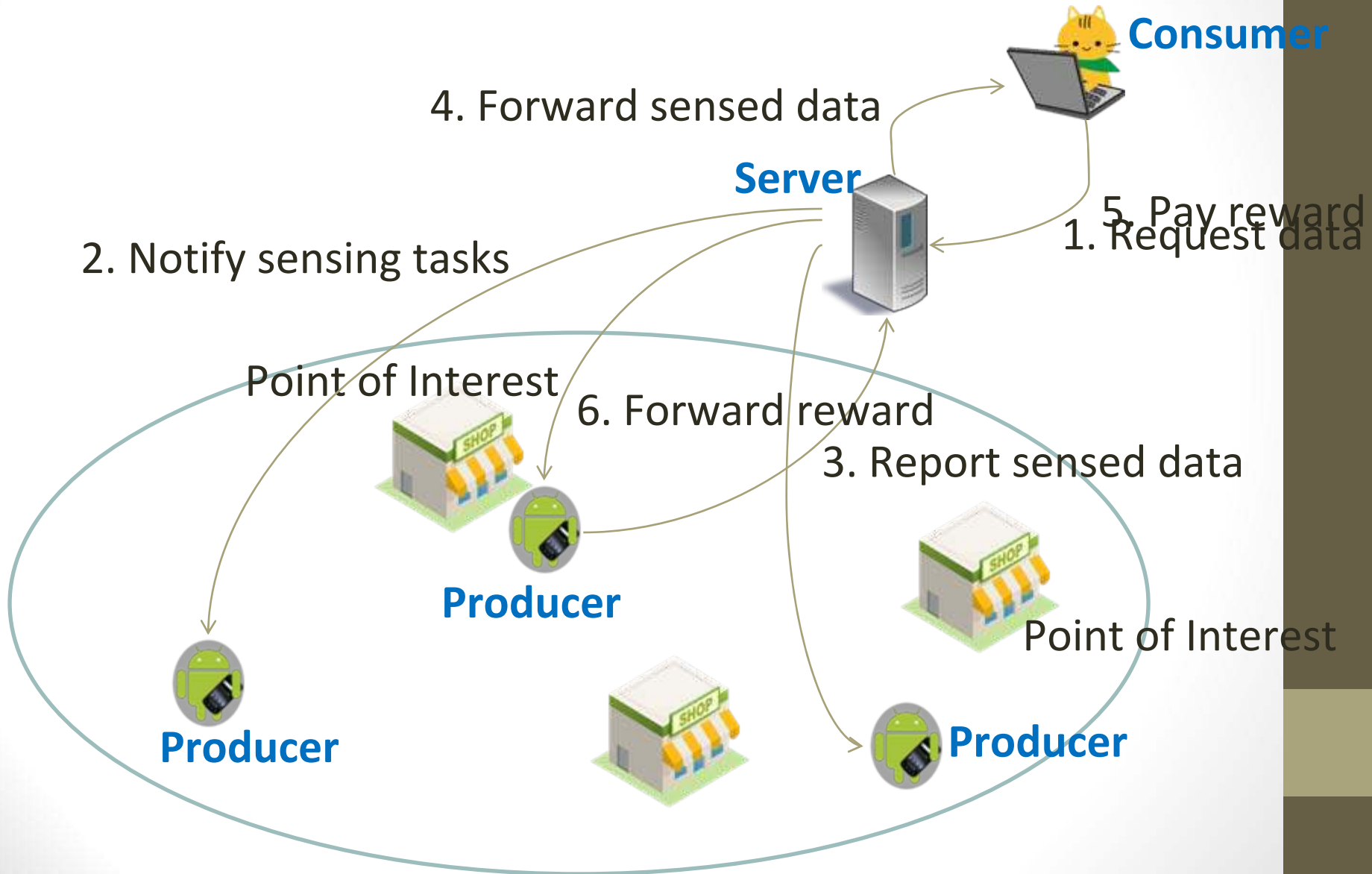
- Thresholds are optimized?
 - Use machine learning technique
- Avoid detection error
 - Use features in frequency domain



Transmission of Data

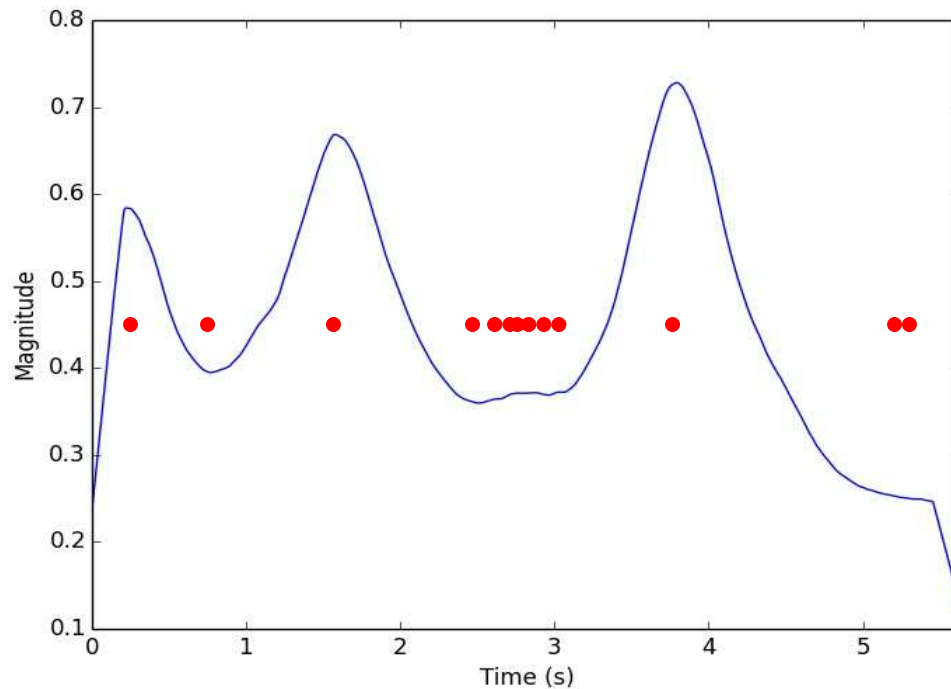
- # of packets affects energy consumption
- Wait until the maximum packet size and send

SenseUtil System: Server, Consumer & Producer



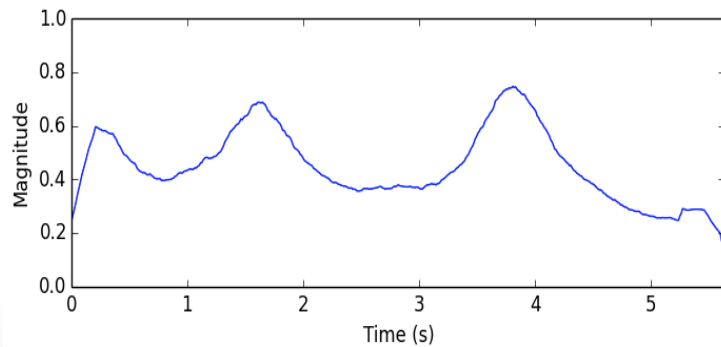
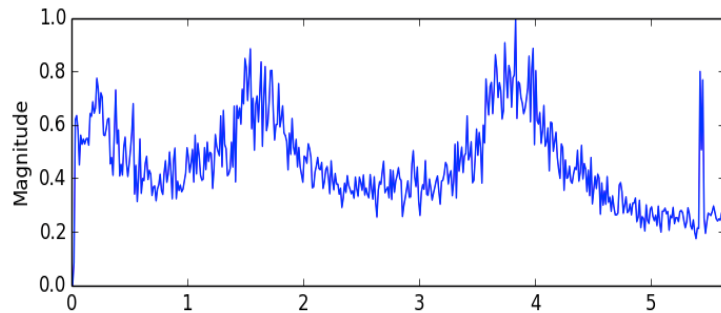
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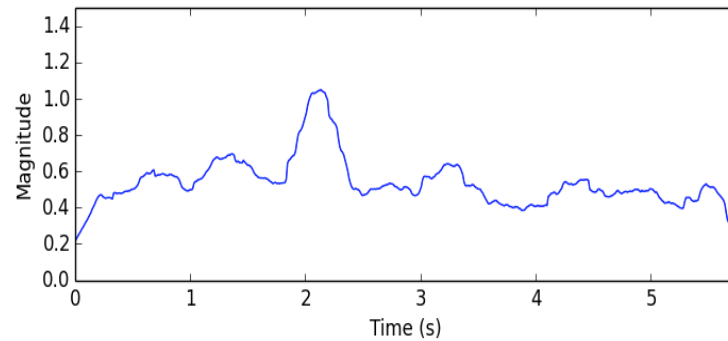
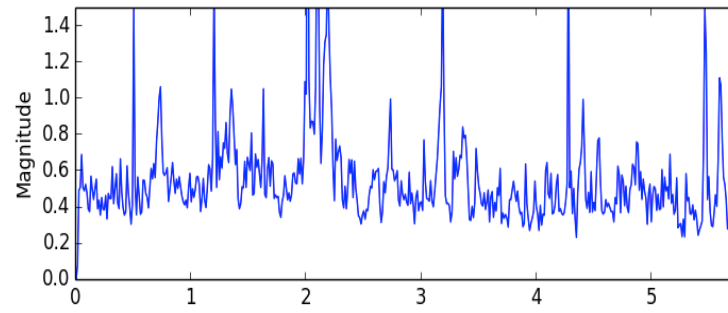


Filtering

- Non-car sound may still be contained in candidate arches



In a case where 3 cars pass



In a case of other environmental sound