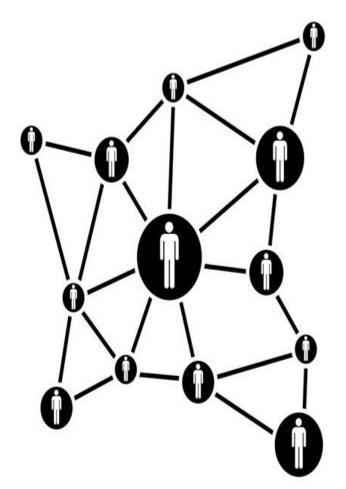
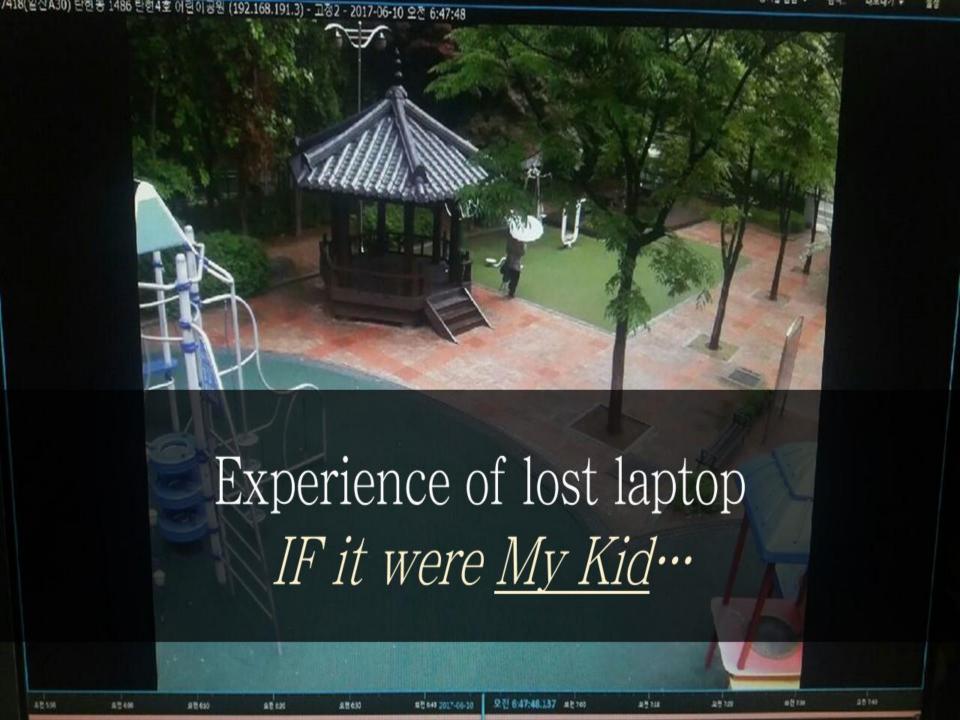
### IOT DATA ALCHEMIST HACKATHON

### REAL-TIME HUMAN TRACKING SYSTEM



INDIE-VIDUAL Shishi, 정기수, 조은희



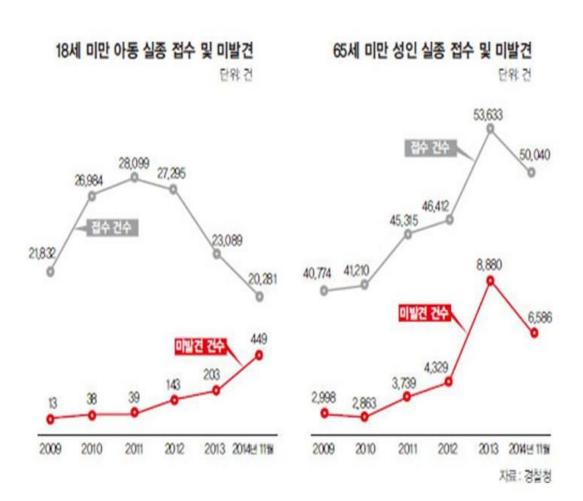
### Even Right now missing cases are being occurred

(Children, Mentally challenged people, patients with dementia) The number of missing person per year 40,000 person 4~5 people disappear per a hour

## Adult missing persons are regarded as simple runaway from home and they are exposed to crime

(Adult) The number of missing person per year 65,000 person 185 people disappear per a day

# The disappearance problem is a social issue that has not been solved yet and it is becoming a bigger issue due to the increase in the number of patients with dementia in the aged society.



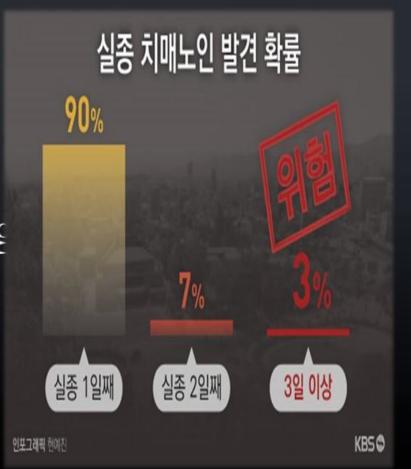


The most important thing in the disappearance case is 'fast finding'.

# GOLDEN TIME Within 24 hours

Probability of not finding a missing

After 12 hours 58% After 24 hours 68% After a week 89%



On the day when the "LEE young-hak Incident" occurred recently, 4 missing persons were reported to Seoul District Police. However, none of them investigated and two of them were killed on this day.



# Currently, the long-dated missing issues are not resolved because the police are relying solely on police workforce and low efficiency methods.









The National Assembly also recognizes the seriousness of the disappearance problem and trying to include adults in the Missing Children Act. However, the National Police Agency is suffering from limitations in the workforce.





홍길동님을 찾았습니다.

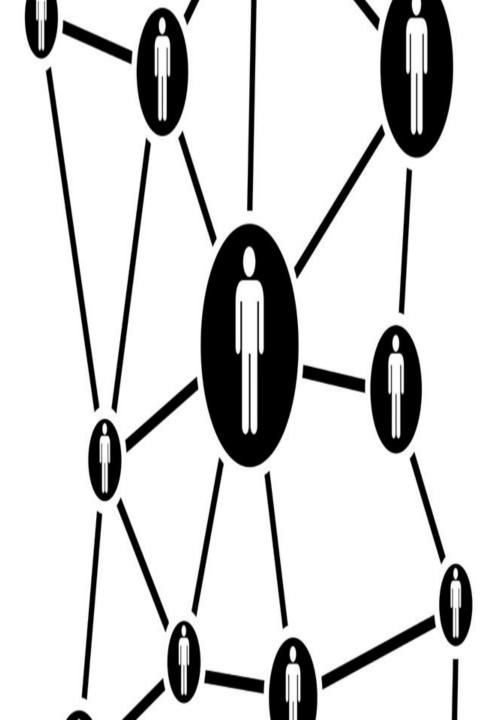
매칭률 90%

찾는 사람이 맞습니까?

YES

NO

REAL-TIME HUMAN TRACKING SYSTEM



There was good feedbacks on the idea of finding missing persons using a mobile phone



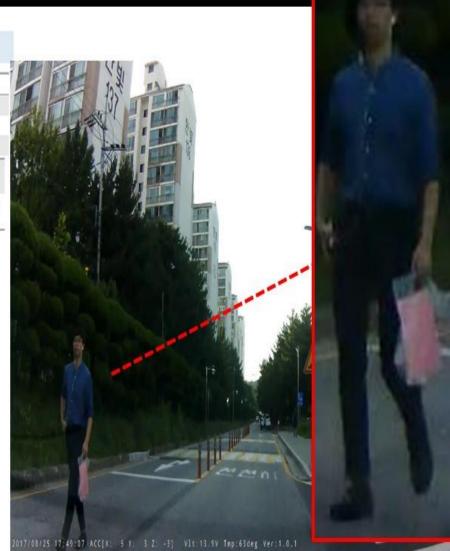
We have tried to analyze the image of mobile Urban Sensing Data by using driving image and GPS information, but faced difficulty to apply face recognition technology because of fast moving image and blurry pixel.

#### Data Introduction Output Type Field data TIME time Output in year-month-day hour: minute: second longitude LNG Korea longitude LAT latitude Korea latitude Four black boxes around the road collect images around Image data Black box the road Four webcams around the road capture images around the Image data Webcam road

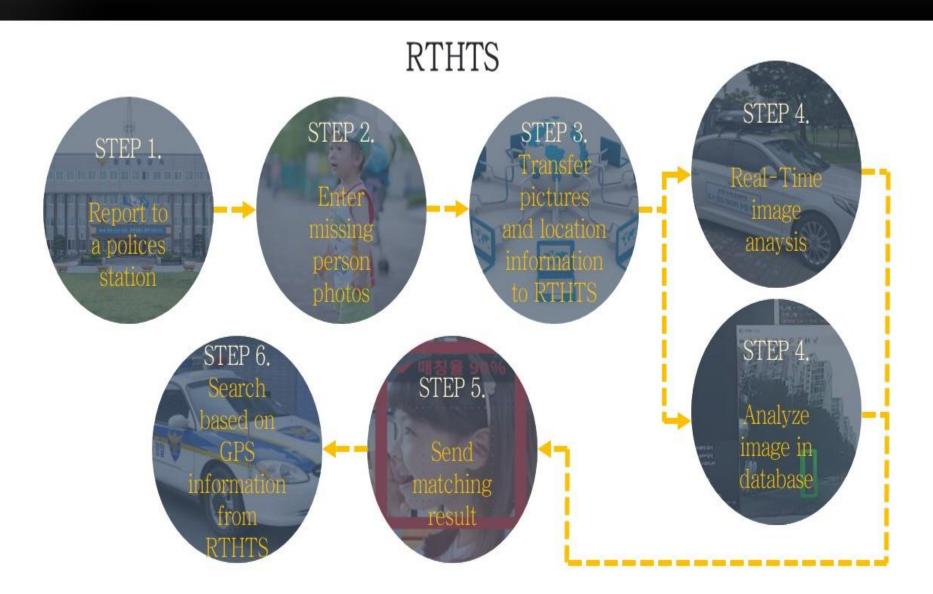
#### 2). 6 major metropolitan cities Mobile Urban Sensing Data Introduction

≥ 2H	
Explanation	Based on the 6 metropolitan cities, the vehicle for the sensing collects the air, traffic, life, and flow environment data of the city center for 8 hours a day for 3 months
keyword	32 elements (Air Quality, Fine dust, Humidity, VOC, NO2, SO2, Noise, Vibration)
Collection area	Daejeon, Daegu, Busan, Ulsan, Gwangju, and Incheon. When a vehicle equipped with a sensor module travels eight hours a day, it collects 32 kinds of sensing data

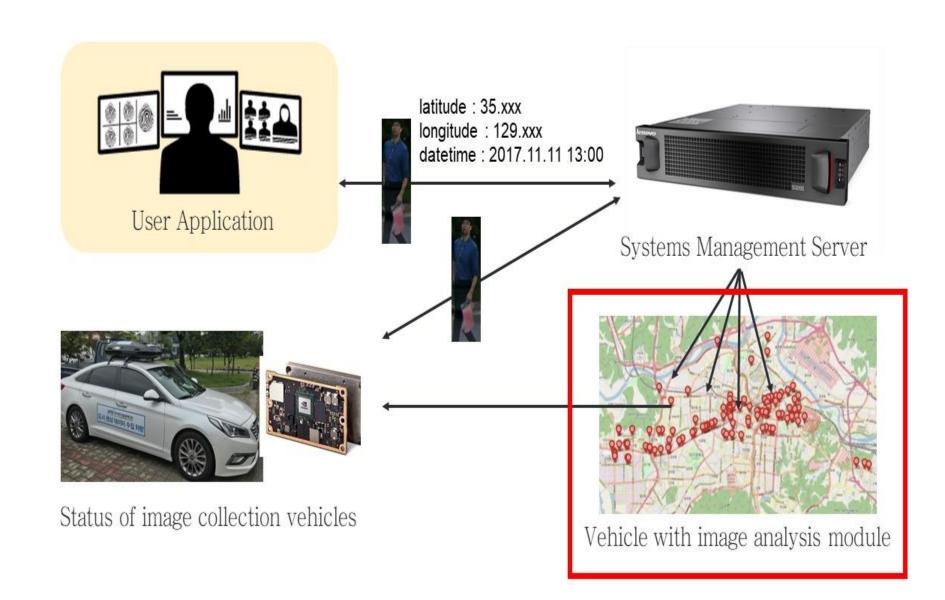




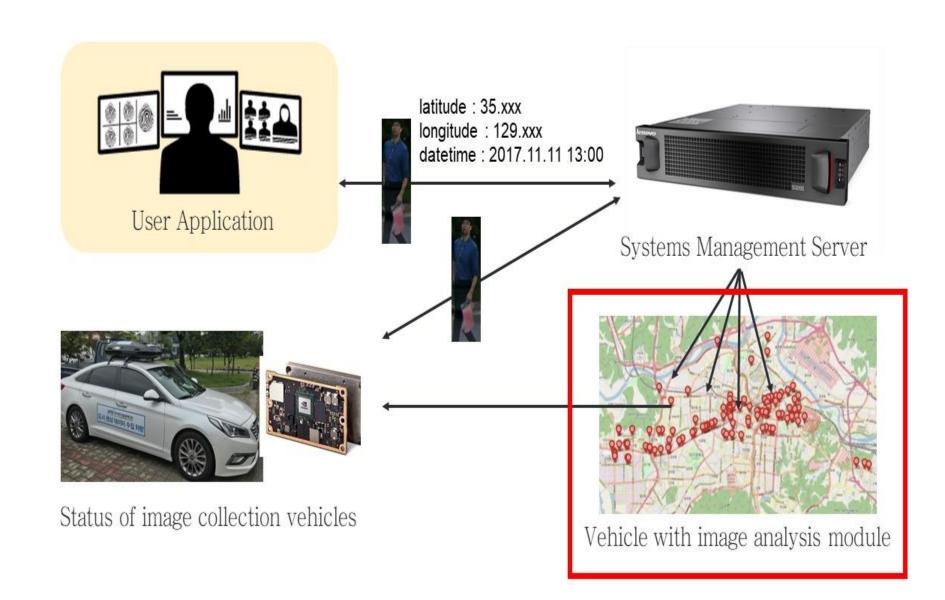
### Overall operating process



### Real-time tracking system configuration diagram

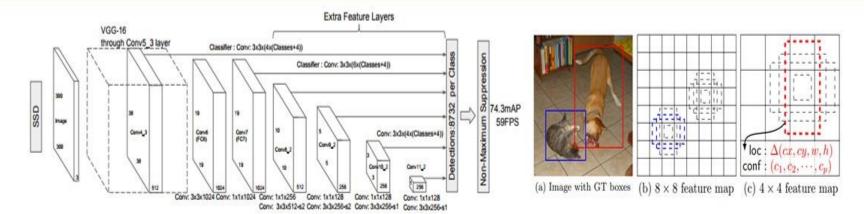


### Real-time tracking system configuration diagram



### Main algorithms - human detection

### Deep learning model for human detection – SSD(Single Shot MultiBox Detector)



### Used data set - MSCOCO(Common Objects In Context

Method	data	mAP	aero	bike	bird	boat	bottle	bus	car	cat	chair	cow	table	dog	horse	mbike	person	plant	sheep	sofa	train	tv
Fast [6]	07	66.9	74.5	78.3	69.2	53.2	36.6	77.3	78.2	82.0	40.7	72.7	67.9	79.6	79.2	73.0	69.0	30.1	65.4	70.2	75.8	65.8
Fast [6]	07+12	70.0	77.0	78.1	69.3	59.4	38.3	81.6	78.6	86.7	42.8	78.8	68.9	84.7	82.0	76.6	69.9	31.8	70.1	74.8	80.4	70.4
Faster [2]	07	69.9	70.0	80.6	70.1	57.3	49.9	78.2	80.4	82.0	52.2	75.3	67.2	80.3	79.8	75.0	76.3	39.1	68.3	67.3	81.1	67.6
Faster [2]	07+12	73.2	76.5	79.0	70.9	65.5	52.1	83.1	84.7	86.4	52.0	81.9	65.7	84.8	84.6	77.5	76.7	38.8	73.6	73.9	83.0	72.6
Faster [2]	07+12+COCO	78.8	84.3	82.0	77.7	68.9	65.7	88.1	88.4	88.9	63.6	86.3	70.8	85.9	87.6	80.1	82.3	53.6	80.4	75.8	86.6	78.9
SSD300	07	68.0	73.4	77.5	64.1	59.0	38.9	75.2	80.8	78.5	46.0	67.8	69.2	76.6	82.1	77.0	72.5	41.2	64.2	69.1	78.0	68.5
SSD300	07+12	74.3	75.5	80.2	72.3	66.3	47.6	83.0	84.2	86.1	54.7	78.3	73.9	84.5	85.3	82.6	76.2	48.6	73.9	76.0	83.4	74.0
SSD300	07+12+COCO	79.6	80.9	86.3	79.0	76.2	57.6	87.3	88.2	88.6	60.5	85.4	76.7	87.5	89.2	84.5	81.4	55.0	81.9	81.5	85.9	78.9
SSD512	07	71.6	75.1	81.4	69.8	60.8	46.3	82.6	84.7	84.1	48.5	75.0	67.4	82.3	83.9	79.4	76.6	44.9	69.9	69.1	78.1	71.8
SSD512	07+12	76.8	82.4	84.7	78.4	73.8	53.2	86.2	87.5	86.0	57.8	83.1	70.2	84.9	85.2	83.9	79.7	50.3	77.9	73.9	82.5	75.3
SSD512	07+12+COCO	81.6	86.6	88.3	82.4	76.0	66.3	88.6	88.9	89.1	65.1	88.4	73.6	86.5	88.9	85.3	84.6	59.1	85.0	80.4	87.4	81.2

### What is COCO?



COCO is a large-scale object detection, segmentation, and captioning dataset. COCO has several features;

- Object segmentation
- Recognition in context
- Superpixel stuff segmentation
- 1.5 million object instances
- 80 object categories
- 91 stuff categories
   5 captions per image
- ✓ 250,000 people with keypoints

### Main algorithms - human detection

### Tuning Deep Learning Architecture for Performance: MobileNet

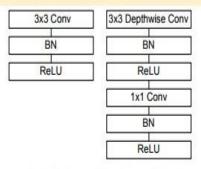


Figure 3. Left: Standard convolutional layer with batchnorm and ReLU. Right: Depthwise Separable convolutions with Depthwise

Model name	Speed	COCO mAP	Outputs	
ssd_mobilenet_v1_coco	fast	21	Boxes	
ssd_inception_v2_coco	fast	24	Boxes	
rfcn_resnet101_coco	medium	30	Boxes	
faster_rcnn_resnet101_coco	medium	32	Boxes	
faster_rcnn_inception_resnet_v2_atrous_coco	slow	37	Boxes	

Table 13. COCO object detection results comparison using different frameworks and network architectures. mAP is reported with COCO primary challenge metric (AP at IoU=0.50:0.05:0.95)

Framework Resolution	Model	mAP	Billion Mult-Adds	Million Parameters
	deeplab-VGG	21.1%	34.9	33.1
SSD 300	Inception V2	22.0%	3.8	13.7
	MobileNet	19.3%	1.2	6.8
Faster-RCNN	VGG	22.9%	64.3	138.5
300	Inception V2	15.4%	118.2	13.3
	MobileNet	16.4%	25.2	6.1
Faster-RCNN	VGG	25.7%	149.6	138.5
600	Inception V2	21.9%	129.6	13.3
	Mobilenet	19.8%	30.5	6.1



Figure 6. Example objection detection results using MobileNet SSD.

### Main algorithms - Image matching

### Feature Extraction

- GMS (Grid-based Motion Statistics for Fast, Ultra-robust Feature Correspondence)
- SIFT (Scale-Invariant Feature Trasform









Difficult to extract feature points due to noise and resolution problems

Sample data : 대전시 - 2017-08-25(FRI) - Car1 - blackbox front - Rec\_20170825\_174904\_D.avi

### Main algorithms - Image matching

### Histogram Comparison

- Correlation
- Chi-Square
- Intersection
- Bhattacharyya distance

### Correlation equation

$$d(H_1,H_2) = \frac{\sum_I (H_1(I) - \bar{H}_1)(H_2(I) - \bar{H}_2)}{\sqrt{\sum_I (H_1(I) - \bar{H}_1)^2 \sum_I (H_2(I) - \bar{H}_2)^2}}$$

Correlation shows the best performance.

We decided to set the threshold to 92 ~ 95% when compared with Target.

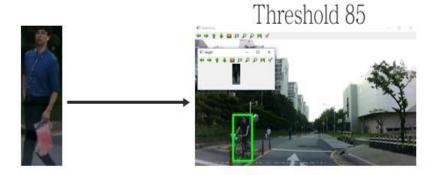


Sample data : 대전시 - 2017-08-25(FRI) - Car1 - blackbox front - Rec\_20170825\_174904\_D.avi

### Further improvement

### Automatic image matching threshold setting

- Error detection according to threshold
- Automatically set the histogram of the extracted histogram by machine learning





### Bounding box optimization

Only the target object is obtained accurately through background removal.



## Home CCTV allows you to find missing persons by comparing full-length photos.

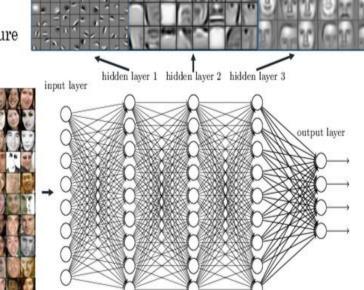


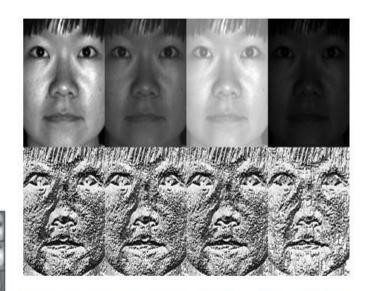
### Further improvement

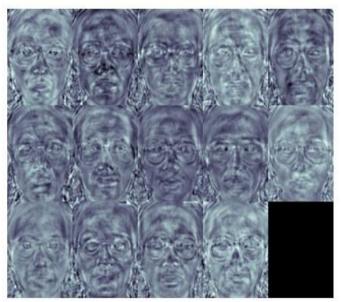
### face recognition technology

- Eigenface, Fisherface, LPB
- Deep learning

Deep neural networks learn hierarchical feature representations







## When the animal disappears, the range of movement is wide. quick response is most important.



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010,1234,5678/010,4321,8765

Facial recognition and feature point extraction technology using deep learning enables tracking of companion animals.

