# Human Augmented Cognition based on Human and Environment Understanding

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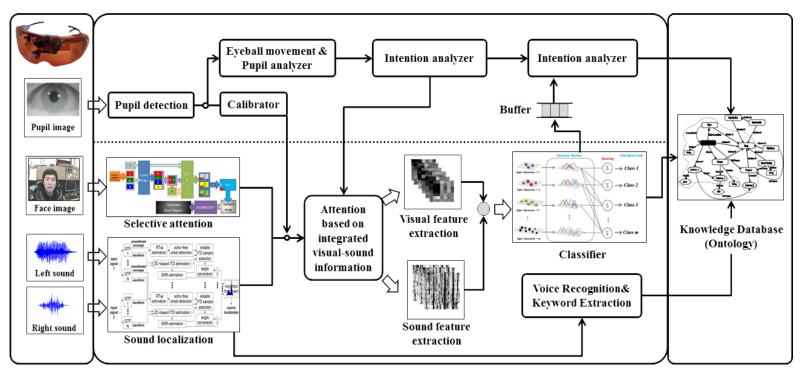
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# Proposed model of augmented cognition system

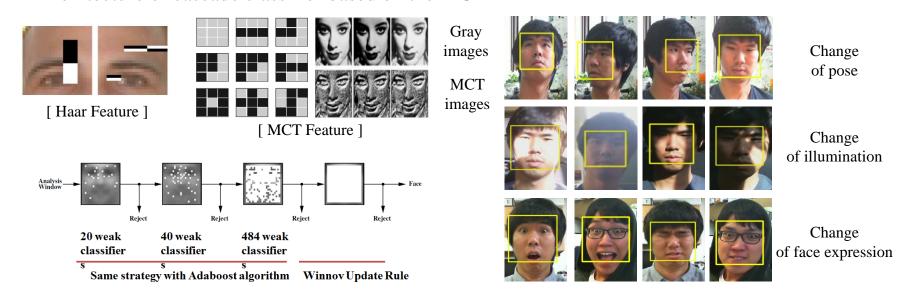
- Development of human augmented cognition system aims to actively provide appropriate information to normal and abnormal people.
- Basically, this system has 3 purposes memory capacity expansion, decision support and inattention blindness compensation, for the cognition augmentation.



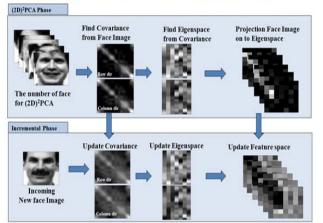
[ Overall system block diagram ]

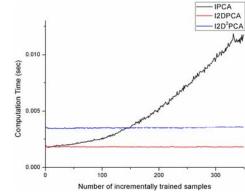
## Environment understanding: Visual feature extraction

Architecture of cascade classifier based on the MCT



Incremental two dimension and direction principal component analysis



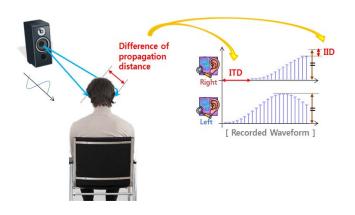


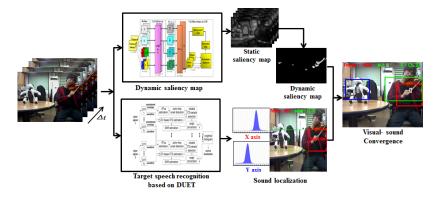
[ Performance comparison using ORL DB ]

Method	Accuracy (%)	Dimension
IPCA	79.8	185
I(2D)PCA	81.6	512
I(2D) <sup>2</sup> PCA	82.4	234

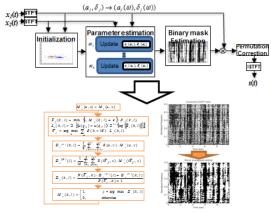
## Environment understanding: Auditory feature extraction

Localization of sound sources





- The selective attention model based on the convergence of visual and auditory information
- Improved performance of speech recognition

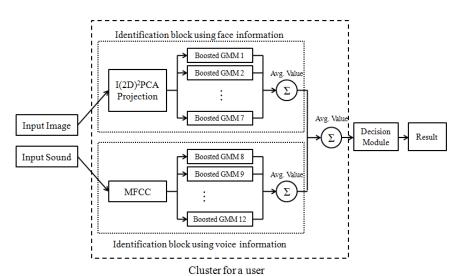


[ DUET algorism ]

[ Improved speech recognition performance ]

Performances (%) of word					
SNR	Clean	20dB	15dB	10dB	5dB
Before processing	90.32	88.17	68.82	31.18	13.98
After processing		90.32	87.10	64.52	20.43

## Environment understanding: Personal authentication

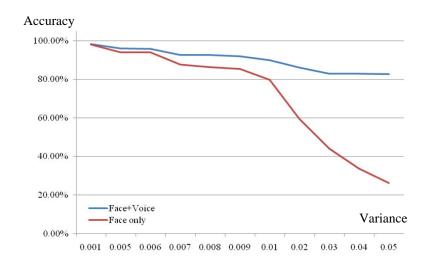


Integrated feature data: Face + Voice signal (100 people)

Num.	5	10	20	30	40	50
Face only	0.05	0.099	0.198	0.294	0.391	0.486
Face+Voice	0.05	0.099	0.198	0.295	0.395	0.489

Num.	60	70	80	90	100
Face only	0.576	0.673	0.759	0.853	0.937
Face+Voice	0.586	0.685	0.778	0.873	0.962

Performance with respect to the noise of face images



- Add Gaussian random noise to face images
  - Set mean zero and change variance
- Trained with 30 people (30 classes)
- Examples of noisy images (variance)







Original

0.001

0.01

0.05

Artificial Brain Research Lab.

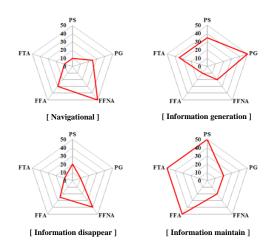
## Human understanding: Human intent recognition

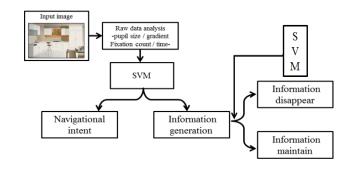
## Improved performance of speech recognition





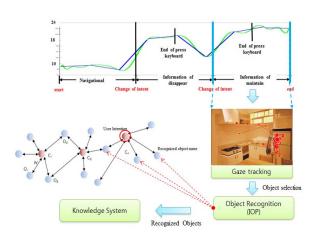






	Navigational to informational	Maintain to disappear	
Train per(%)	96.89 ±0.29 (%)	98.71 ±0.19 (%)	
Test per(%)	90.02 ±0.48 (%)	90.36 ±0.59 (%)	

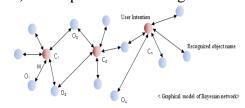
### Human intent probing

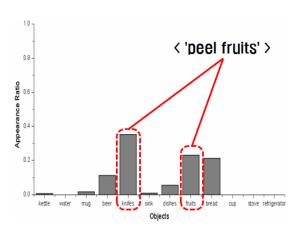


Probabilistic approach : Naïve Bayes

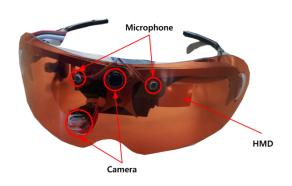
$$P(C \mid O_1, ...O_n) = \frac{1}{Z} P(C) \prod_{i=1}^n P(O_i \mid C)$$

Measurement of semantic relatedness by (LSA) of the probabilistic weights

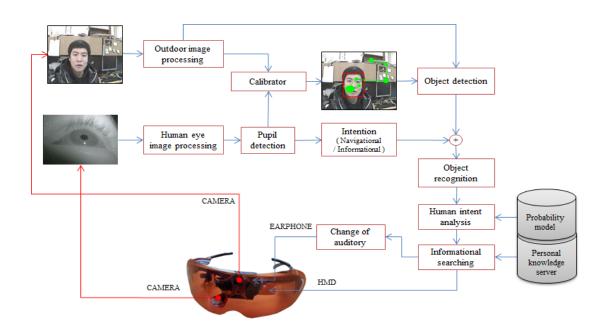




## Human augmented cognition of embedded prototype system

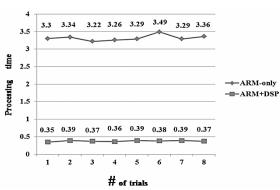


[ Prototype platform of human augmented cognition system ]





[ OMAP3530 Platform ]



[ Face detection processing time of ARM & ARM+DSP]

#### [Average processing time for each of functions]

Functions	Processing time average (ms)
Saliency Map	77.29
Face Detection	61.02
Face Recognition	13.02
Information Retrieval	220.58
Total(Whole Process)	Min: 272.17 / Max: 574.69

## Conclusion & further works

- A system designed for augmented cognition should account for human cognitive limitations and user conditions in order to retrieve information appropriately from the external repository
- This probabilistic approach can make normative predictions about how humans implicit intention combines prior knowledge (object's functions) with sensory reaction (gaze).
- We measured the performance of the proposed system to assist human cognitive capability, especially, memory capacity expansion, memory retrieval and attention span.
- Currently, we concentrate on developing a large scale knowledge system which can automatically construct knowledge from user's experience and structured general knowledge for human augmented cognition system.

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