

Deep Learning-based VR Sickness Assessment

IEEE SA WG3079 Meeting
: HMD based 3D Content Motion Sickness Reducing Technology

Hak Gu Kim
hgkim0331@kaist.ac.kr

Image and Video Systems Lab.
School of Electrical Engineering
Korea Advanced Institute of Science and Technology (KAIST)

July. 2018

Virtual Reality (VR)

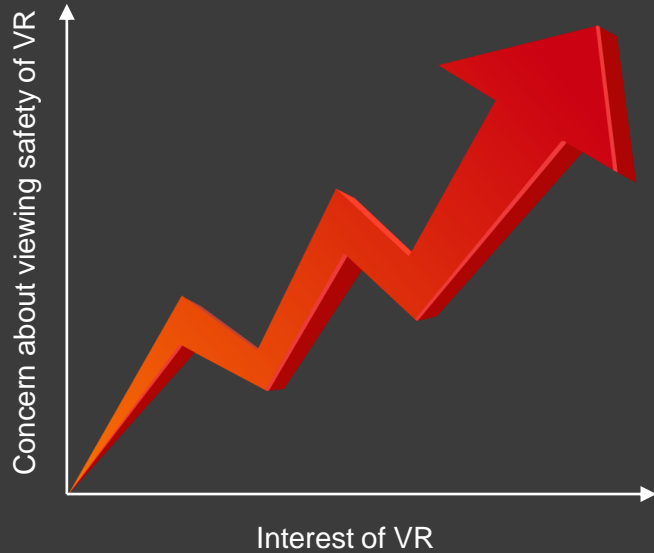


Immersive 360-degree VR Content



AirDano

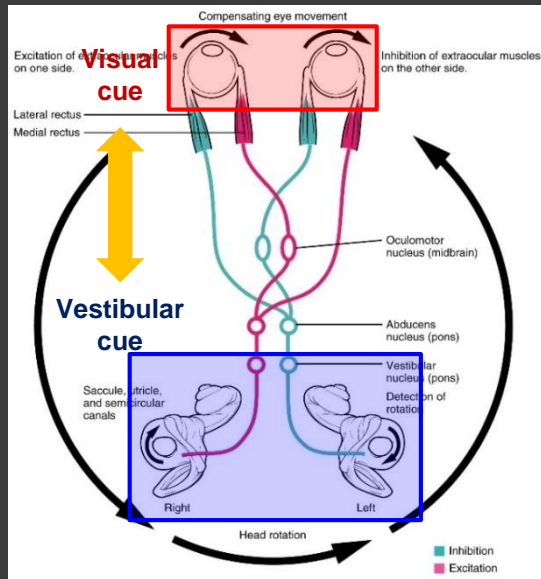
VR Sickness (\approx Cybersickness)



Main Factors of VR Sickness

Visual-Vestibular Conflict

- Mismatches between simulation motion of VR content and viewer's motion



Visual-vestibular interaction^[1]



Visual-vestibular conflict in VR experience^[2]

[1] <http://oerpub.github.io/epubjs-demo-book/content/m46557.xhtml>

[2] Mayo Clinic Research

VR Sickness Assessment (VRSA)

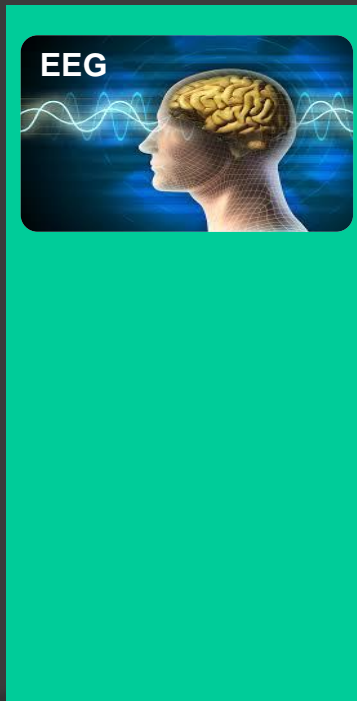


Objective
assessment

Subjective
assessment

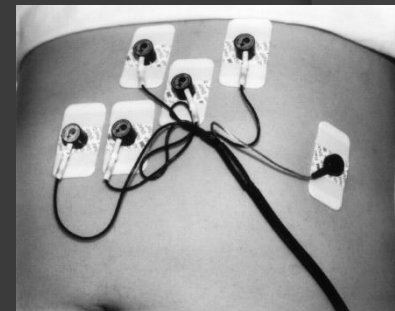
Objective VRSA: Physiological Measurements

Physiological
measurements



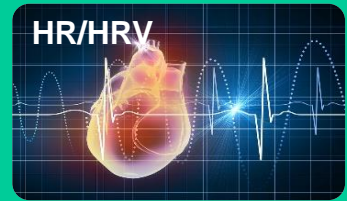
Objective VRSA: Physiological Measurements

Physiological
measurements



Objective VRSA: Physiological Measurements

Physiological
measurements



Subjective VRSA: Subjective Questionnaires



Physiological
measurements

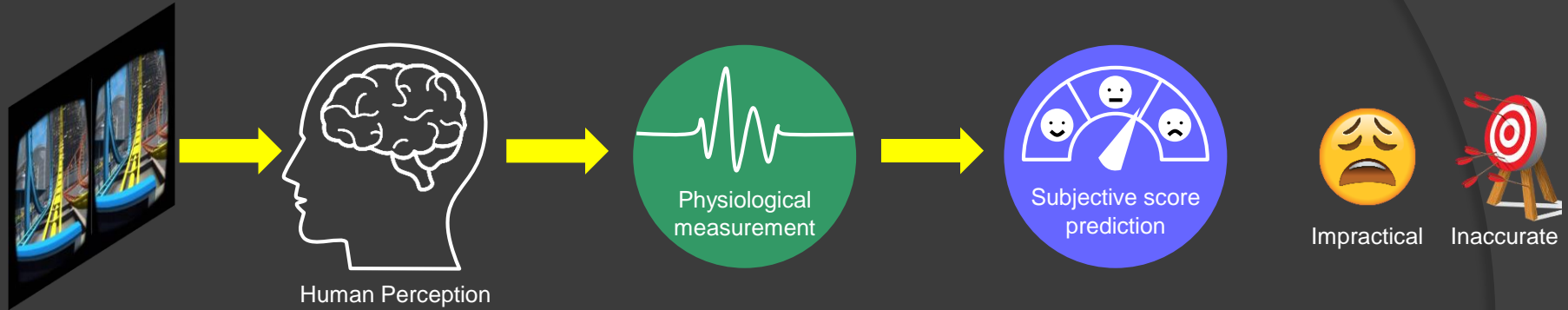


Subjective
questionnaires

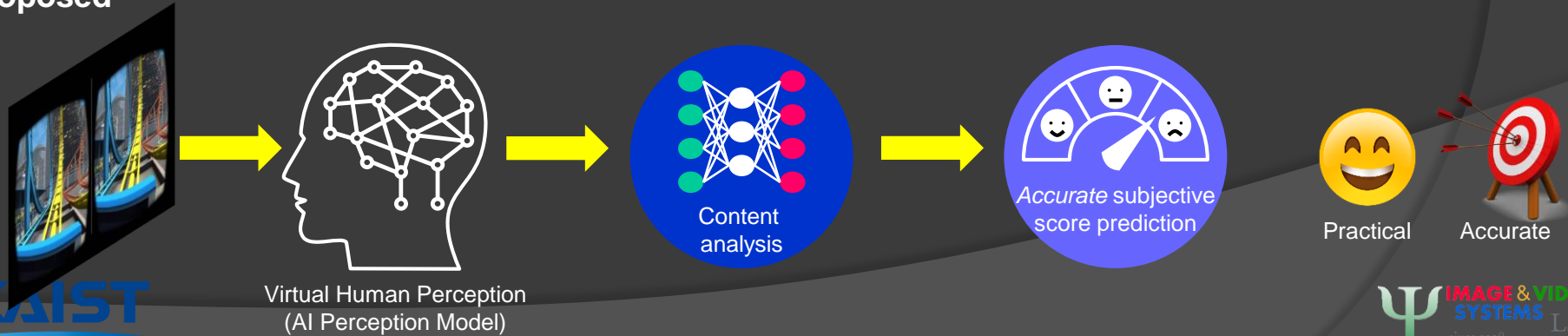


Proposed Framework for VRSA

Conventional



Proposed



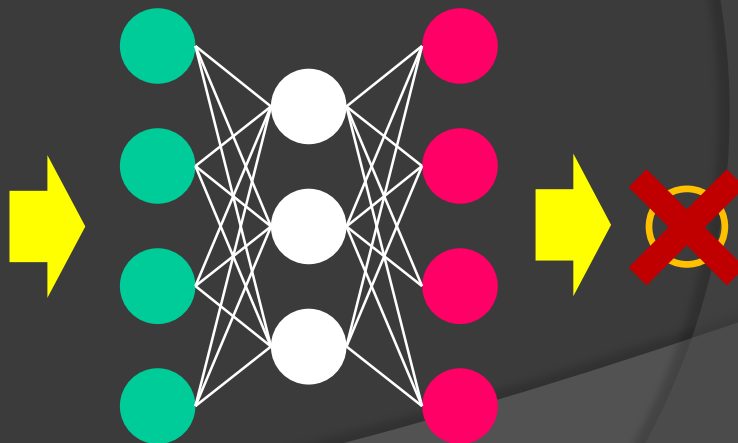
Challenges

○ Lack of labeled datasets

- It is difficult to collect a large-scale of fully labeled datasets
 - VR content and the corresponding subjective scores



A large number of target images



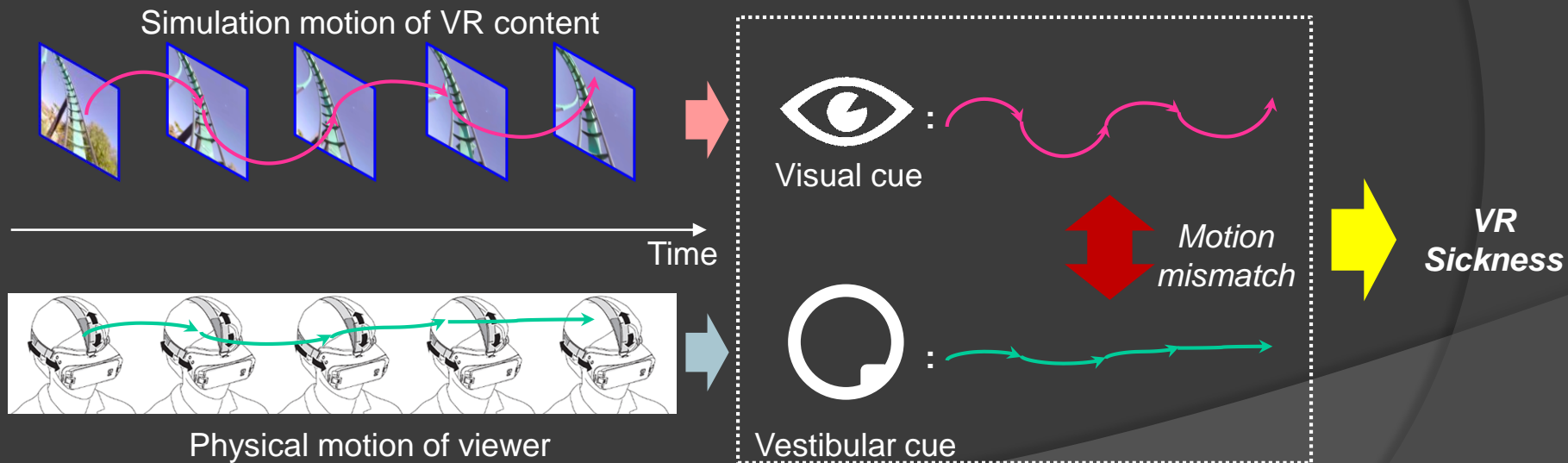
Regression

Corresponding subjective score

Proposed Method (1/5)

Visual-vestibular conflict

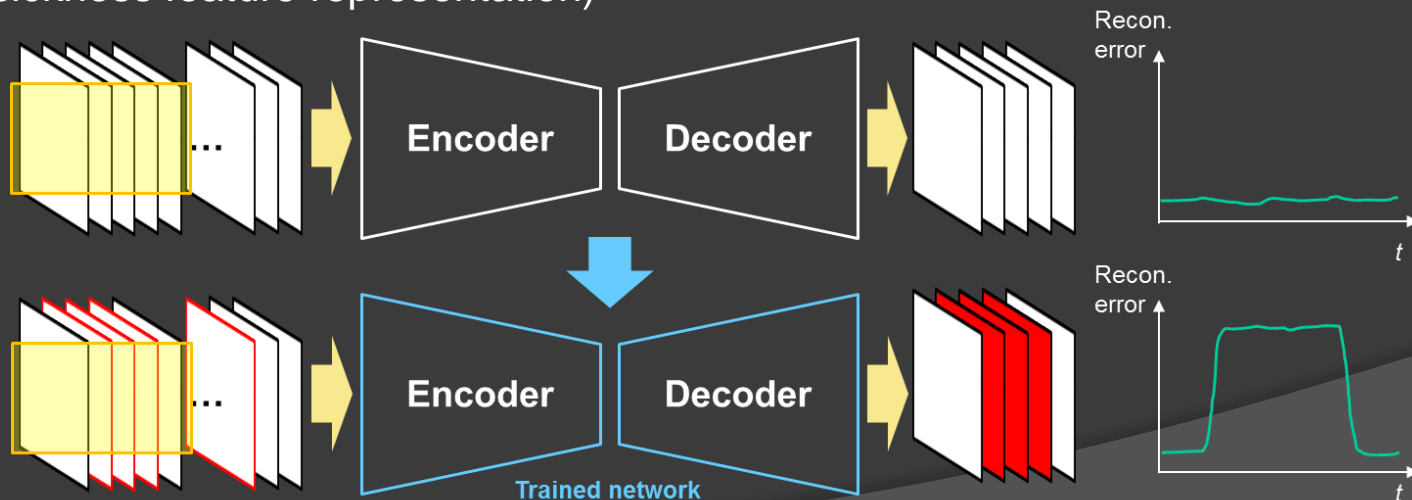
- Caused by exceptional motion of VR content



Proposed Method (2/5)

○ Main Idea of Our Research

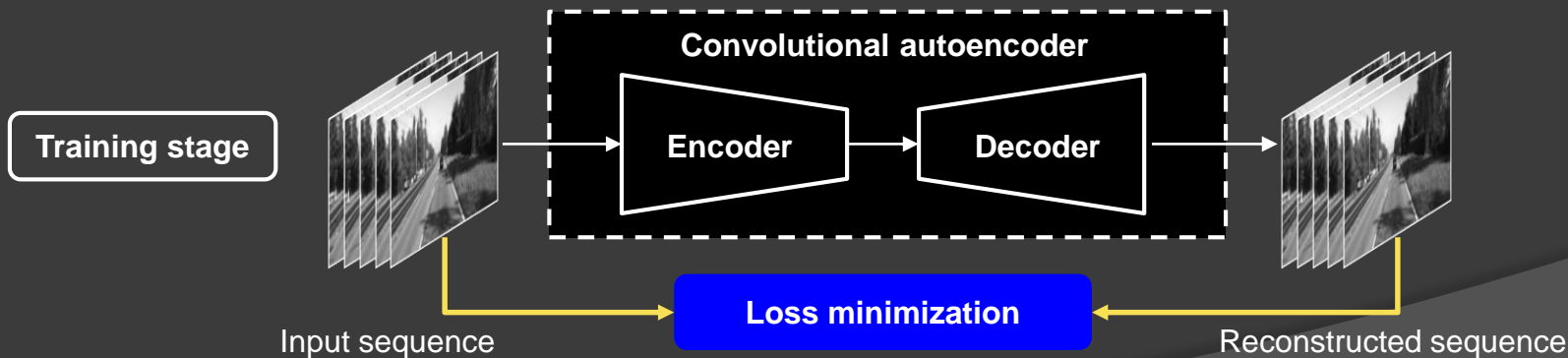
- Deep learning based generative model
- Non-excessive VR sickness feature representation (instead of excessive VR sickness feature representation)



Proposed Method (3/5)

Overall Procedure of the Proposed VRSA Framework

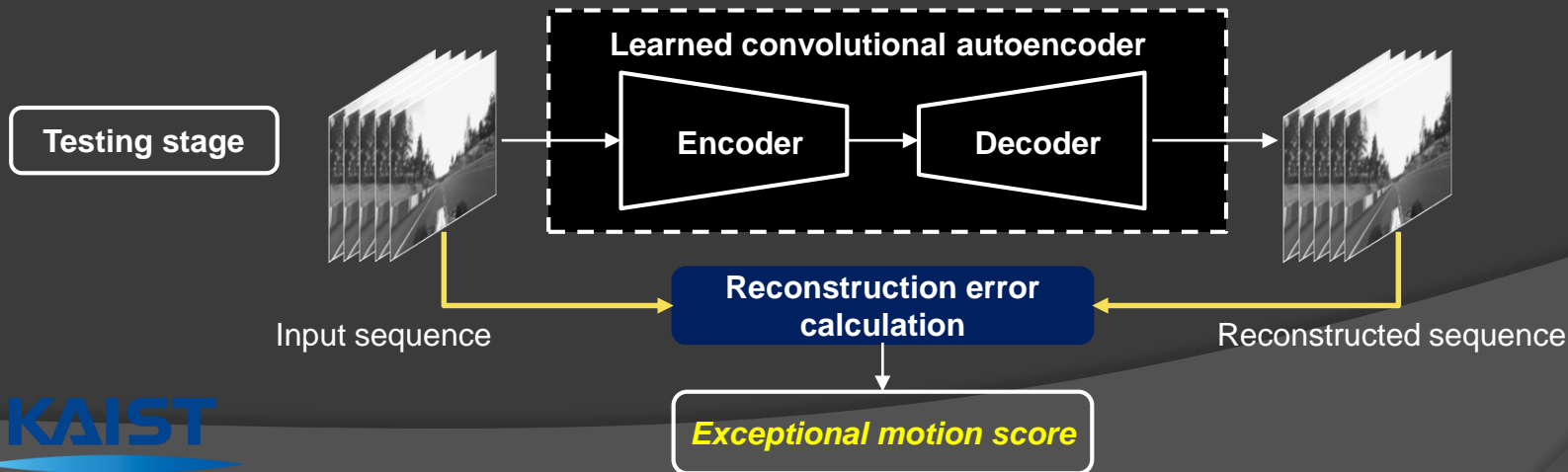
- In training, the convolutional autoencoder is trained to reconstruct original VR video sequences with non-exceptional motion such as slow and moderate motion velocity.
- In testing, by measuring the reconstruction error of the motion information in VR video content, the exceptional motion of VR video content can be detected and measured.



Proposed Method (3/5)

Overall Procedure of the Proposed VRSA Framework

- In training, the convolutional autoencoder is trained to reconstruct original VR video sequences *with non-exceptional motion* such as slow and moderate motion velocity.
- In testing, by measuring the reconstruction error of the motion information in VR video content, the exceptional motion of VR video content can be detected and measured.

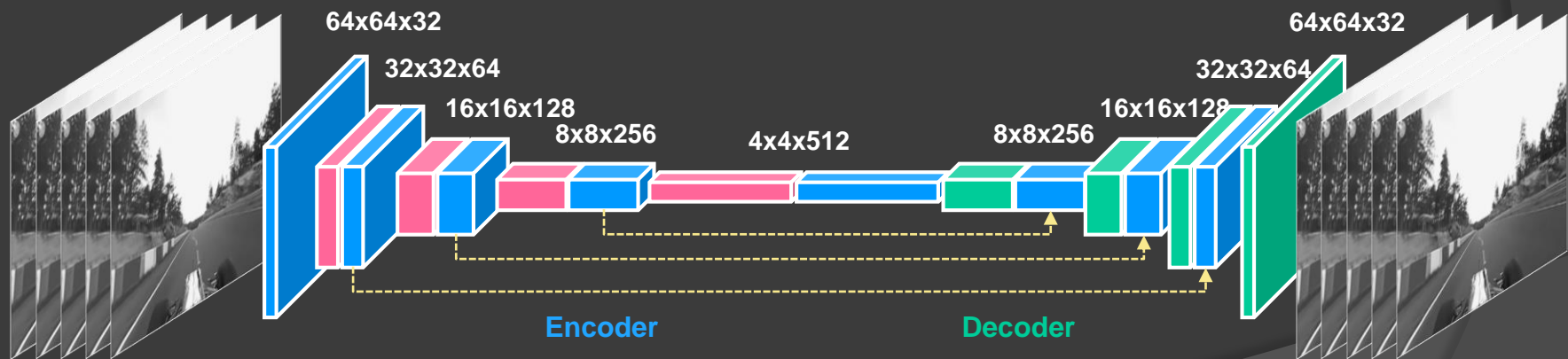


Proposed Method (4/5)

Deep Convolutional Autoencoder for Normal Motion Patterns Learning

- Encoder for representing the latent spatio-temporal feature of input sequence
- Decoder for reconstructing the original sequence from the encoded features

128x128x5



128x128x5

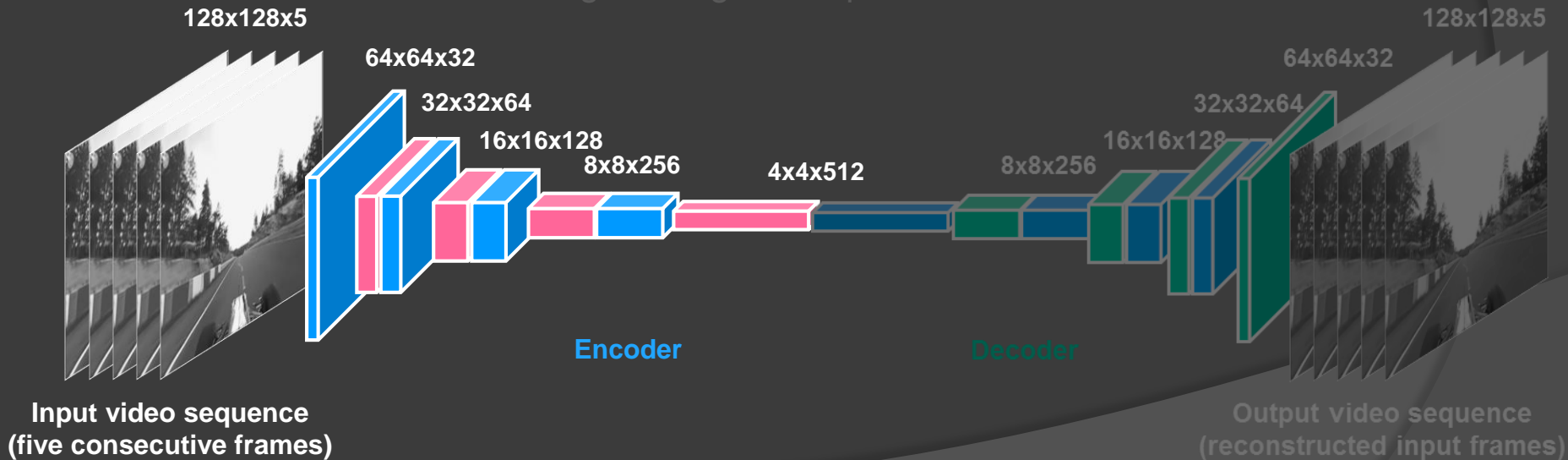
Input video sequence
(five consecutive frames)

Output video sequence
(reconstructed input frames)

Proposed Method (4/5)

Deep Convolutional Autoencoder for Normal Motion Patterns Learning

- Encoder for representing the latent spatio-temporal feature of input sequence
- Decoder for reconstructing the original sequence from the encoded features



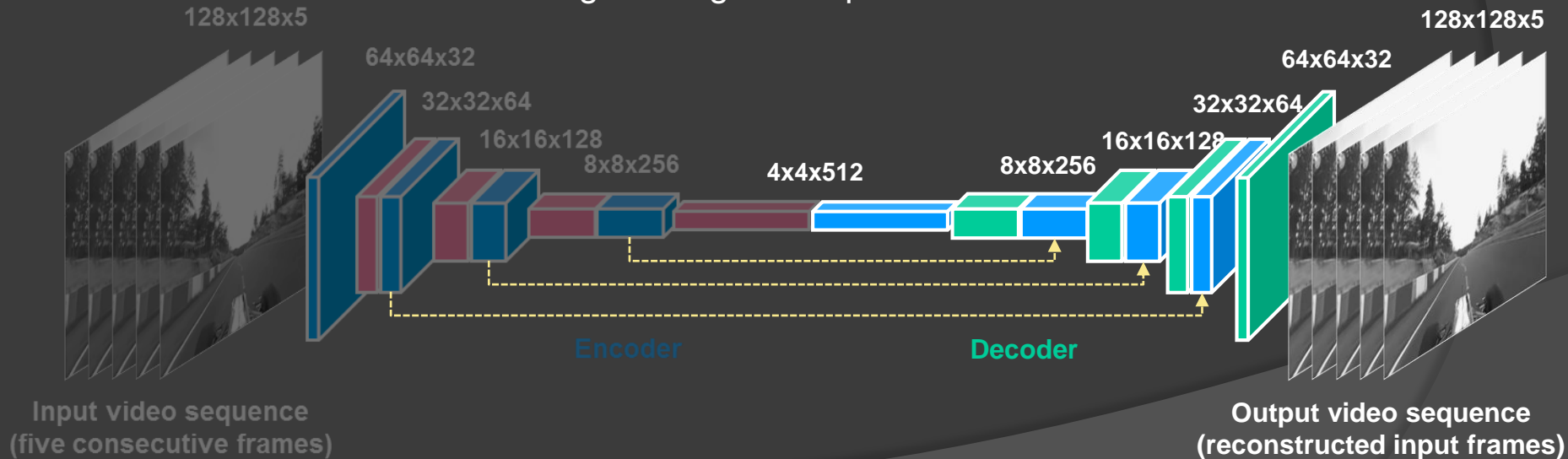
Input video sequence
(five consecutive frames)

Output video sequence
(reconstructed input frames)

Proposed Method (4/5)

Deep Convolutional Autoencoder for Normal Motion Patterns Learning

- Encoder for representing the latent spatio-temporal feature of input sequence
- Decoder for reconstructing the original sequence from the encoded features



Proposed Method (5/5)

○ Exceptional Motion Pattern Score

- Reconstruction error at t -th frame of test dataset

$$e(t) = \sum_{i=1}^W \sum_{j=1}^H \|I(i, j, t) - \hat{f}_W(I(i, j, t))\|^2$$

$I(i, j, t)$: Original t -th frame

$\hat{f}_W(I(i, j, t))$: Reconstructed t -th frame

W and H : width and height of the frame

- Proposed exceptional motion pattern score

$$s_m(t) = \frac{e(t)}{\sqrt{W \times H}}$$

$e(t)$: Reconstruction error at t -th frame

W and H : width and height of the frame

Experiments and Results (1/7)

○ Datasets for Training

- UCSD Ped 1 and Ped 2
 - Ped 1: 34 training video clips (200 frames)
 - Ped 2: 16 training video clips (120~180 frames)
- Avenue
 - 16 training video clips (180~360 frames)
- KITTI benchmark
 - 61 video clips



UCSD Ped1



UCSD Ped12



Avenue



KITTI Benchmark

Experiments and Results (2/7)

○ Datasets for Test

- Three 360-degree VR video contents, collected from Youtube^{[3],[4],[5]}
 - Video 1: Slow velocity
 - Video 2: Moderate velocity
 - Video 3: Fast velocity



Video 1



Video 2



Video 3

Experiments and Results (3/7)

○ Subjective Assessment Experiment

- Equipment for displaying VR content
 - Oculus Rift CV1 HMD
 - 2160 x 1200 pixels @ 90 Hz
 - FoV: 110 degree
 - Intel Core i7-4770@3.4 GHz, 32GB RAM, and NVIDIA GTX 1080TI

- Subjects
 - 15 subjects, ranging between 20 to 30 years old
 - Normal or corrected-to-normal vision
 - Minimum stereopsis: 60 arcsec.



Experiments and Results (4/7)

Subjective Assessment Experiment

16-item SSQ^[6]

SIMULATOR SICKNESS QUESTIONNAIRE
Kennedy, Lane, Berbaum, & Libenthal (1993)**

Instructions : Circle how much each symptom below is affecting you right now.

1. General discomfort	<u>None</u>	<u>Slight</u>	<u>Moderate</u>	<u>Severe</u>
2. Fatigue	<u>None</u>	<u>Slight</u>	<u>Moderate</u>	<u>Severe</u>
3. Headache	<u>None</u>	<u>Slight</u>	<u>Moderate</u>	<u>Severe</u>
4. Eye strain	<u>None</u>	<u>Slight</u>	<u>Moderate</u>	<u>Severe</u>
5. Difficulty focusing	<u>None</u>	<u>Slight</u>	<u>Moderate</u>	<u>Severe</u>
6. Salivation increasing	<u>None</u>	<u>Slight</u>	<u>Moderate</u>	<u>Severe</u>
7. Sweating	<u>None</u>	<u>Slight</u>	<u>Moderate</u>	<u>Severe</u>
8. Nausea	<u>None</u>	<u>Slight</u>	<u>Moderate</u>	<u>Severe</u>
9. Difficulty concentrating	<u>None</u>	<u>Slight</u>	<u>Moderate</u>	<u>Severe</u>
10. « Fullness of the Head »	<u>None</u>	<u>Slight</u>	<u>Moderate</u>	<u>Severe</u>
11. Blurred vision	<u>None</u>	<u>Slight</u>	<u>Moderate</u>	<u>Severe</u>
12. Dizziness with eyes open	<u>None</u>	<u>Slight</u>	<u>Moderate</u>	<u>Severe</u>
13. Dizziness with eyes closed	<u>None</u>	<u>Slight</u>	<u>Moderate</u>	<u>Severe</u>
14. *Vertigo	<u>None</u>	<u>Slight</u>	<u>Moderate</u>	<u>Severe</u>
15. **Stomach awareness	<u>None</u>	<u>Slight</u>	<u>Moderate</u>	<u>Severe</u>
16. Burping	<u>None</u>	<u>Slight</u>	<u>Moderate</u>	<u>Severe</u>

* Vertigo is experienced as loss of orientation with respect to vertical upright.

** Stomach awareness is usually used to indicate a feeling of discomfort which is just short of nausea.

Last version : March 2013

Procedure

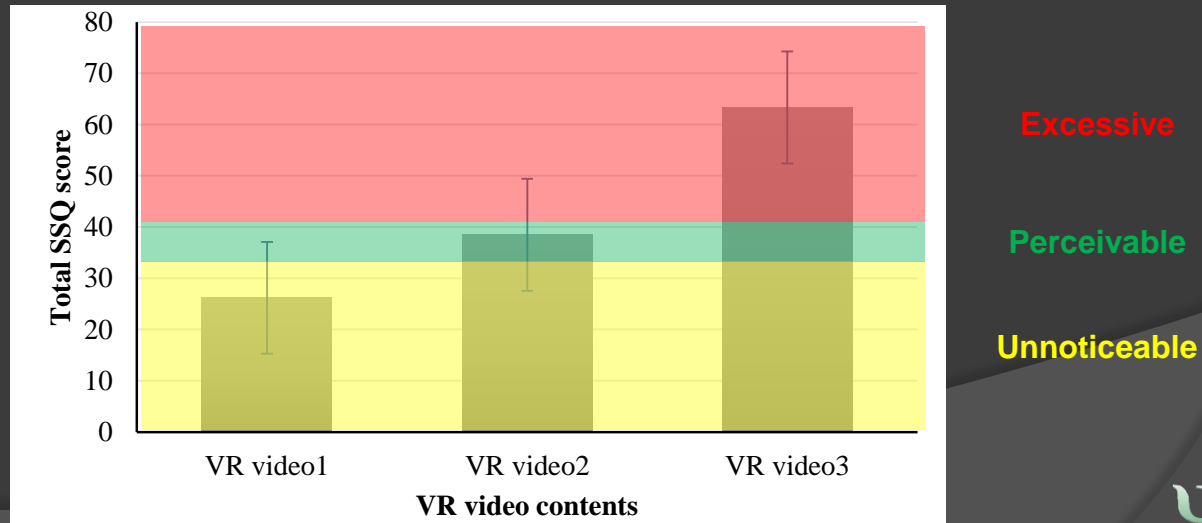
- A week before the actual subjective assessment experiments, we had subjects experience a variety of VR contents with Oculus Rift in order to allow them familiarize with VR environment.
- Every VR video content was displayed for 2 minutes through Oculus Rift CV1.
- After watching the VR video contents, subjects rated their perception of the VR sickness for each symptom in SSQ sheet .

[6] Kennedy et al. Simulator sickness questionnaire: an enhanced method for quantifying simulator sickness. The International Journal of Aviation Psychology, 1993

Experiments and Results (5/7)

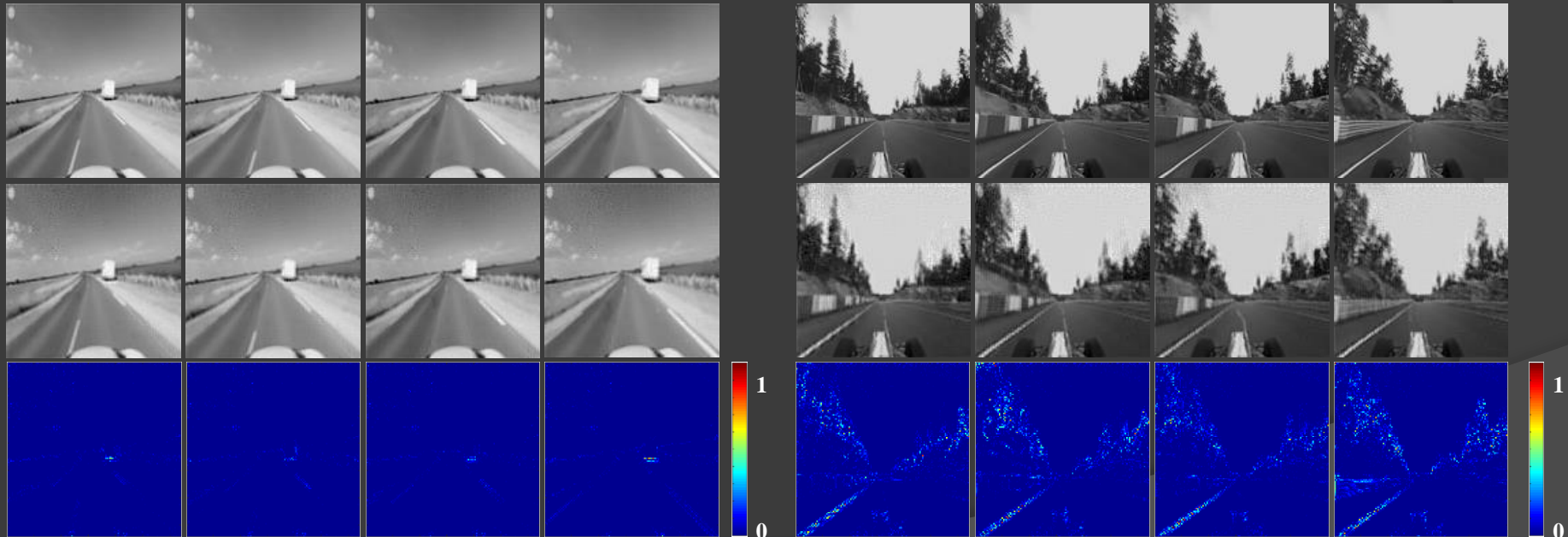
◉ Subjective Assessment Results

- Degree of VR sickness subjects felt was proportional to the motion magnitude.
- The subjective results show that subjects felt some symptoms of VR sickness when watching VR video 2 and 3. In particular, VR video 3 could lead to excessive VR sickness



Experiments and Results (6/7)

○ Performance of the Proposed Network



Reconstruction results of the proposed deep convolutional autoencoder network for VR video 1 with slow motion (left) and VR video 3 with exceptional motion (right)

Experiments and Results (7/7)

Performance of the Proposed Network

360 degree video with slow motion pattern



Original seq.

Reconstructed seq.

Recon. error

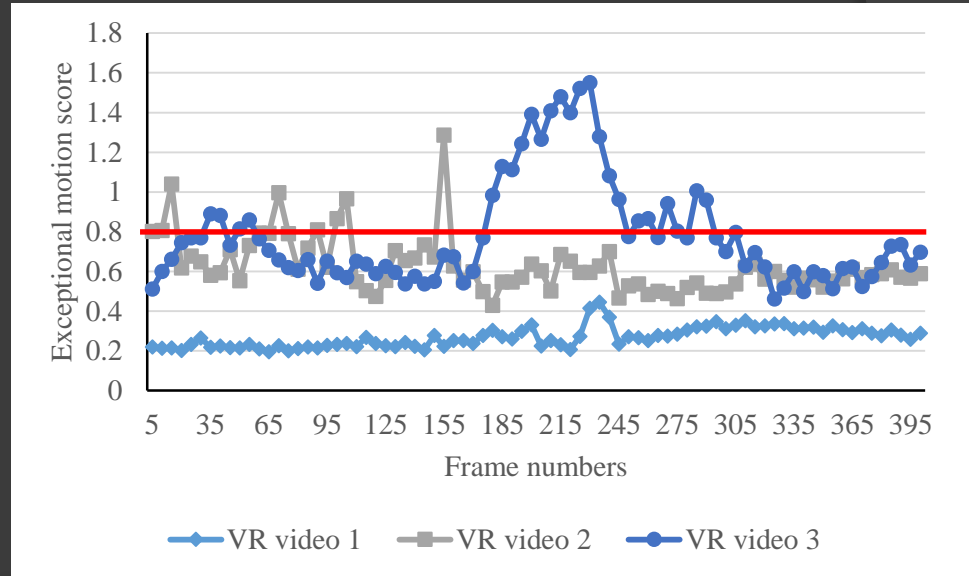
360 degree video with fast motion pattern



Original seq.

Reconstructed seq.

Recon. error



Conclusions

- This paper presented a novel measurement of exceptional motion using deep convolutional autoencoder network for assessing the VR sickness of VR video content.
- The convolutional autoencoder learned by normal datasets with slow and moderate motion could reconstruct the non-exceptional motion patterns but it could not recover VR video content having exceptional motion.
- Based on the fact that exceptional motion led to high reconstruction errors in the deep autoencoder network, the level of VR sickness of the input VR video content due to exceptional motion could be predicted.
- The results of our subjective assessment experiments showed that the proposed objective measure strongly had a high correlation with human subjective quality scores, SSQ of our test datasets (PLCC was 0.92).

Thank you

<http://ivylab.kaist.ac.kr>
hgkim0331@kaist.ac.kr

Appendix

○ Total SSQ score

- 1. Nausea score
- 2. Oculomotor score
- 3. Disorientation score
- Total SSQ = 3.74 x (1 + 2+ 3)

Table 1: SSQ used in our subjective assessment

SSQ-Symptoms	Nausea	Oculomotor	Dis-orientation
General discomfort	0	0	0
Fatigue	0	0	0
Headache	0	0	0
Eye strain	0	0	0
Difficulty focusing	0	0	0
Increased salivation	0	0	0
Sweating	0	0	0
Nausea	0	0	0
Difficulty concentrating	0	0	0
Fullness of head	0	0	0
Blurred vision	0	0	0
Dizzy (Eyes open)	0	0	0
Dizzy (Eye closed)	0	0	0
Vertigo	0	0	0
Stomach awareness	0	0	0
Burping	0	0	0