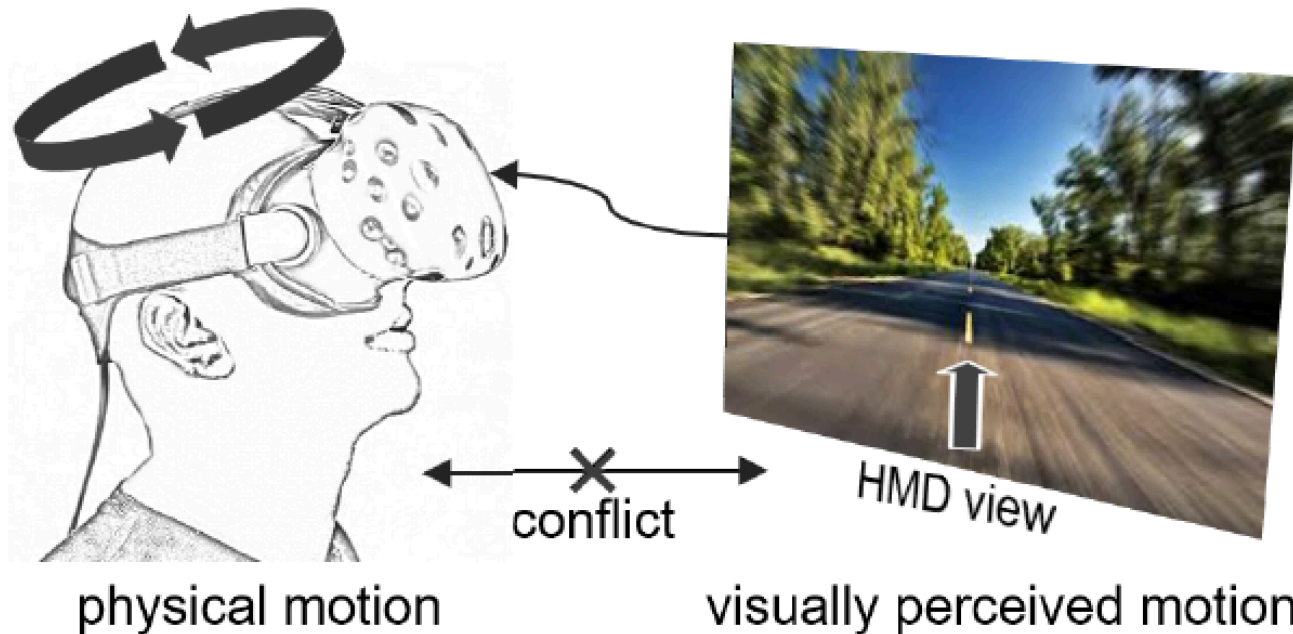


Introduction

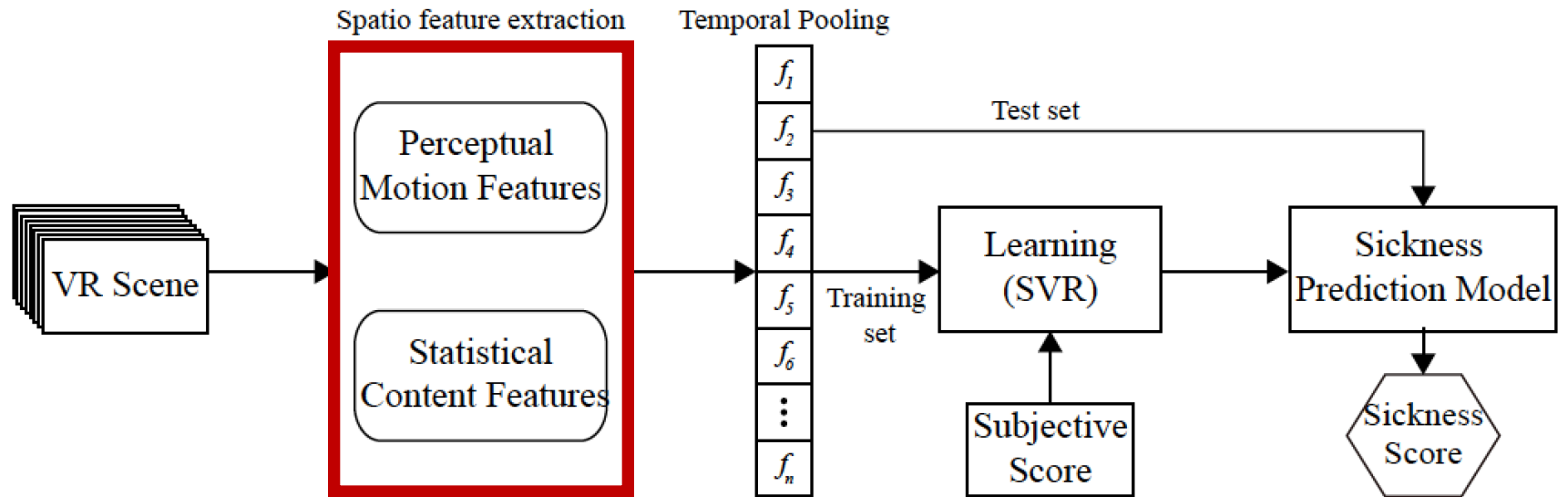
- Possible factors causing sickness
 - Wide FoV, time latency, low frame rate and rendering speed
 - sensory conflict between the visual perception by the virtual stimuli and the vestibular perception by actual head motion



Goal

- Explore an objective VR sickness prediction framework dealing with the recent HMD based VR contents
 - Perceptual motion features: user's head rotation and virtual camera rotation in virtual space
 - Statistical content features: distribution of the spatial texture, motion of objects and background
- ⇒ objective VR sickness predictor (VRSP)
- consisting of 36 contents with 10 scenarios
 - 80 subjects

VR Visual Sickness Predictor



VR Visual Sickness Predictor

- Perceptual motion features
 - User head angular velocity ω_{vest}
 - Visual angular velocity ω_{vis}
 - Perceived angular velocity ω_{per}

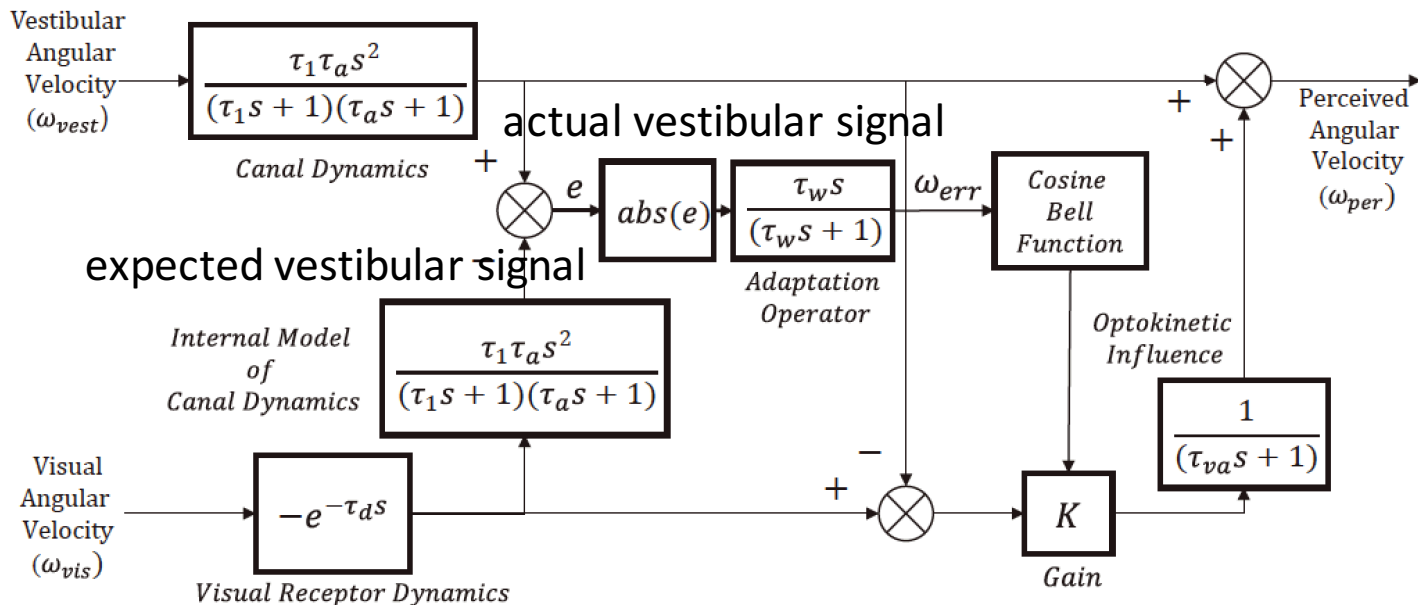
$$\omega_{vest}^t = (P_{vest}^t - P_{vest}^{t-1}) \times \gamma \text{ (degree/sec),}$$

$$P_{vest}^t = [r_{vest}^t, p_{vest}^t, y_{vest}^t],$$

$$\omega_{vis}^t = \omega_{vest}^t - \omega_{cam}^t \text{ (degree/sec),}$$

$$\omega_{cam}^t = (P_{cam}^t - P_{cam}^{t-1}) \times \gamma,$$

$$P_{cam}^t = [r_{cam}^t, p_{cam}^t, y_{cam}^t].$$



VR Visual Sickness Predictor

- Statistical content feature

- Texture features: the distribution of the spatial texture has a great effect on the motion perception of the HVS

$$E^t = \sqrt{\int_{\Omega} CSF(\Omega) |B(\Omega)| d\Omega},$$

- Motion features: the distribution of motion affects visual sickness considerably

$$V^t(x, y) = \frac{1}{N_x N_y} \sum_{x=1}^{N_x} \sum_{y=1}^{N_y} \sqrt{u^t(x, y)^2 + v^t(x, y)^2},$$

- Temporal Pooling: the human perception of the video content is influenced by the scene characteristics in numerous consecutive frames in the time domain

$$\mu_m = \frac{1}{T} \sum_{t=1}^T f_m^t,$$

$$\sigma_m = \frac{1}{T} \sum_{t=1}^T (f_m^t - \mu_m)^2,$$

$$l_m = \frac{1}{d_{max}} \cdot \left(\frac{1}{N_p^l} \sum_{n < N \cdot p / 100} d(n) \right),$$

$$h_m = \frac{1}{d_{max}} \cdot \left(\frac{1}{N_p^h} \sum_{n > N \cdot (100 - p) / 100} d(n) \right)$$

VR Sickness Assessment Dataset

- create a reference scene which includes diversified textures using a Unity 3D engine
 - 36 VR scenes
 - HTC Vive
 - 400x300 pixels frame resolution

Object Movement	Camera Movement	Content Component
<ul style="list-style-type: none">- Horizontal- Vertical- Diagonal- Forward- Backward	<ul style="list-style-type: none">- Roll- Pitch- Yaw- Horizontal- Vertical- Forward- Backward	<ul style="list-style-type: none">- Texture- Field of View- Visual Guidance

Experiment Results

- Subjective test environment
 - 80 subjects, 21-50 years
 - 1-5
 - Individual scores and MOS
- Setup
 - SVR with linear kernel
 - 80% for training and 20% for testing
 - Repeated 1000 times
- Metrics
 - Pearson linear correlation coefficient (LCC) and Spearman rank-order correlation coefficient (SROCC)

MOS and Individual Scores

- MOS

LCC	Mean	Median	Std.
F_{vest}	0.484	0.492	0.063
F_{per}	0.682	0.688	0.054
F_{con}	0.421	0.439	0.079
$VRSP$	0.724	0.728	0.081

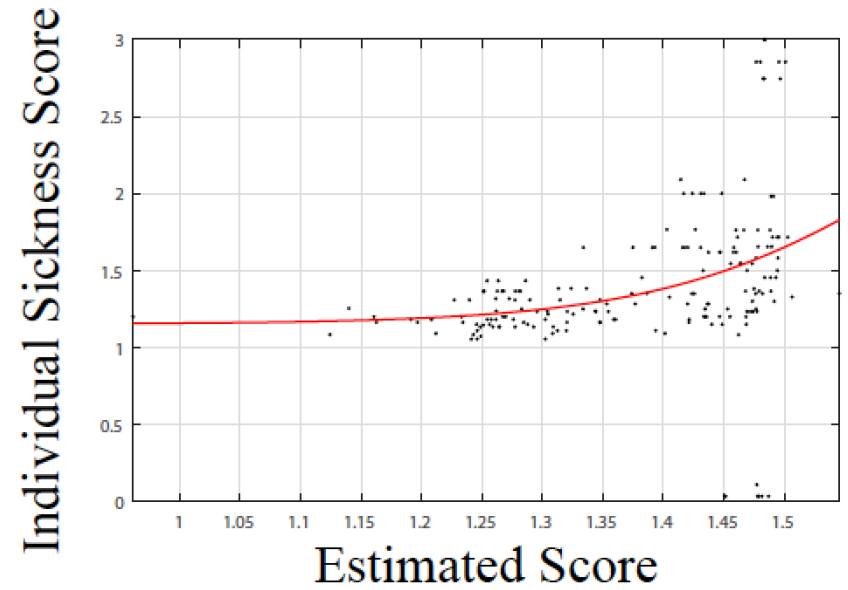
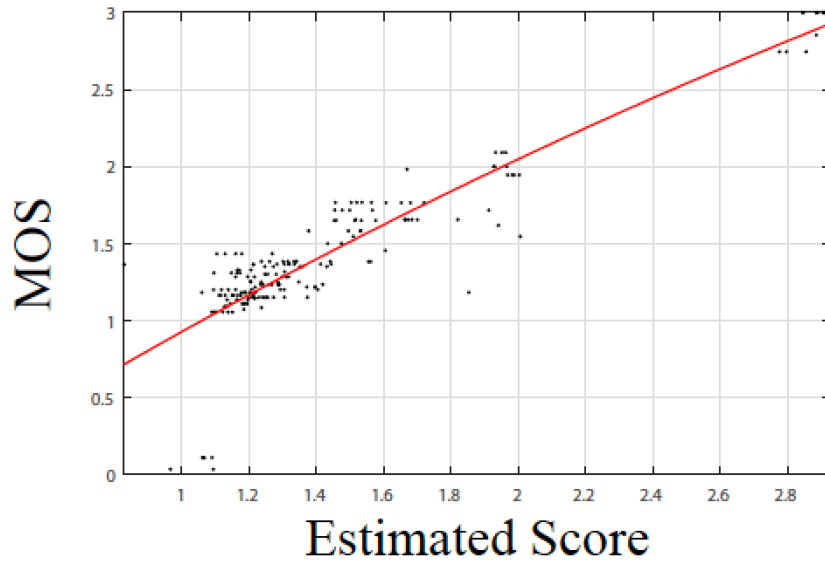
SROCC	Mean	Median	Std.
F_{vest}	0.434	0.451	0.110
F_{per}	0.643	0.647	0.050
F_{con}	0.434	0.457	0.084
$VRSP$	0.710	0.733	0.048

- Individual scores

LCC	Mean	Median	Std.
F_{vest}	0.413	0.420	0.068
F_{per}	0.601	0.612	0.091
F_{con}	0.322	0.343	0.070
$VRSP$	0.671	0.676	0.103

SROCC	Mean	Median	Std.
F_{vest}	0.473	0.477	0.047
F_{per}	0.627	0.634	0.087
F_{con}	0.389	0.391	0.094
$VRSP$	0.688	0.692	0.098

MOS and Individual Scores



Conclusion

- predicting VR sickness on HMD viewing
 - by suitably characterizing the visual-vestibular interaction model and contents features
 - predicted from human head movement and the perceived scene in the HMD
- Future work
 - develop a database with gaming scenario, natural experience (360 videos) and human subject scores on them