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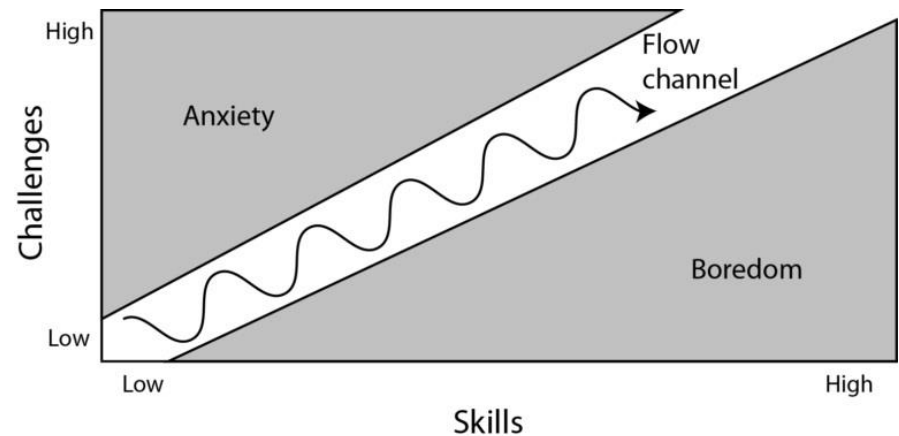
Stress Assessment in Video Games Using Players' Behavioural and Physiological Data

Susanna Brambilla, Giuseppe Boccignone,
N. Alberto Borghese and Laura A. Ripamonti

Introduction

Goals:

- **Entertainment:** improve the players experience by keeping them in the **flow state** (CSIKSZENTMIHALYI, 1990)
- Support the development of **emotion-driven** video game adaptation



Idea:

Detect in **real-time stress level** of the players by collecting their motion behavioural and physiological data.

State-of-the-Art

Past works (e.g.):

- data considered: **pressure** on the game pad (Sykes and Brown, 2003), **pressure and sliding velocity** on a graphic tablet (Frommel et al. 2018)

Results:

- **Pressure** is **harder** when people are stressed
- **Sliding velocity** on tablet **higher**
- **good accuracy** in classification tasks

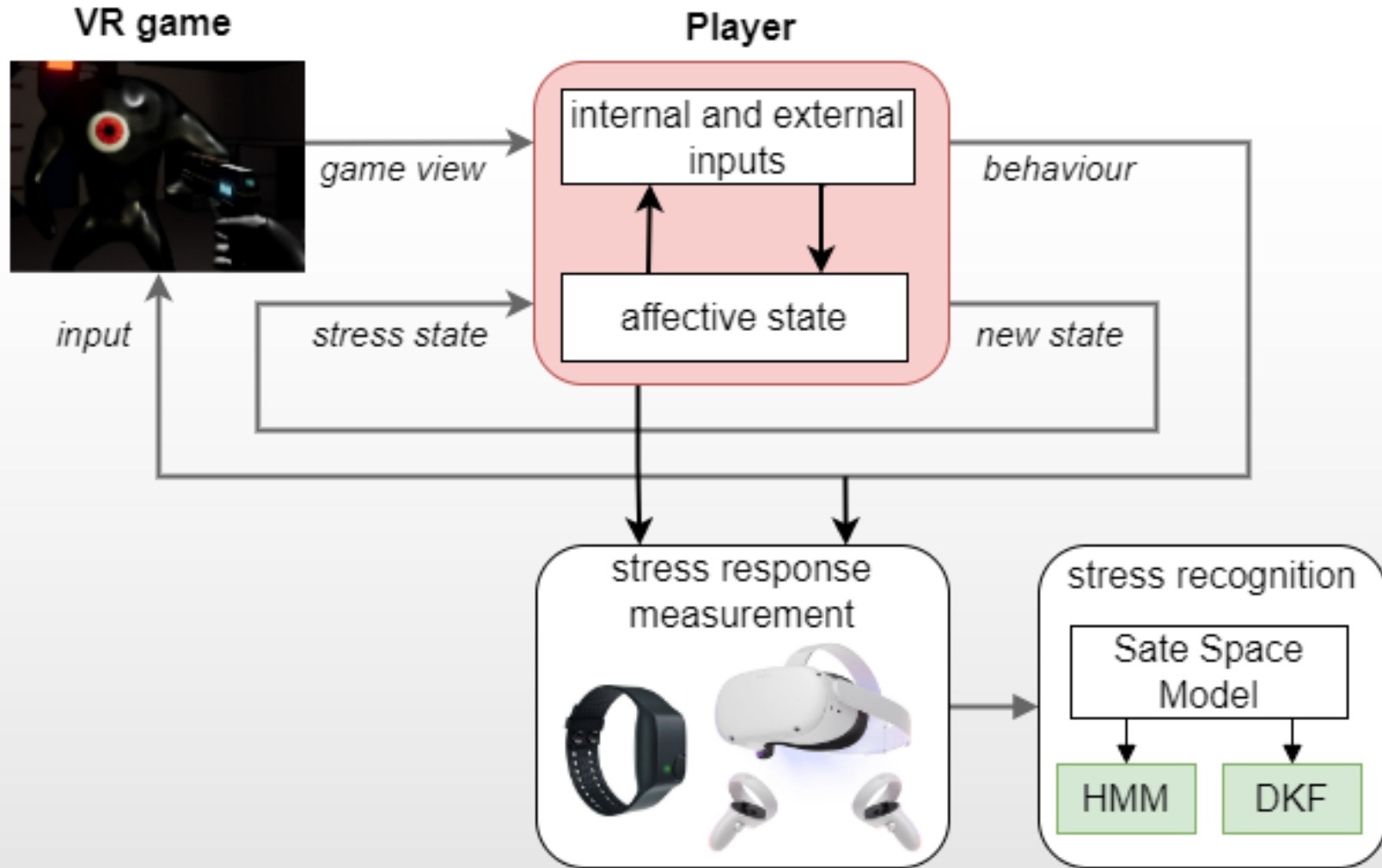
Limitations:

- **Stress** not characterized in its **evolution over time**
- No **VR environment** considered

Our contributions:

- **Evolution of the stress** level considered in terms of **state-space dynamics**
- **VR environment**: high level of **immersivity** and richer **interaction data**

System Structure



Data Collected



EDA

velocity
angular velocity
acceleration
angular acceleration
grip pressure
trigger pressure
thumbstick position x
thumbstick position y
grip pressed
trigger pressed



velocity
angular velocity
acceleration
angular acceleration

velocity
angular velocity
acceleration
angular acceleration
grip pressure
trigger pressure
thumbstick position x
thumbstick position y
grip pressed
trigger pressed

Design of the Video Game

Players' goal: escape from an abandoned space station, facing enemies in a hostile environment, surviving with few resources.

Lebois et al., 2016

expectation violation

perceived self-threat

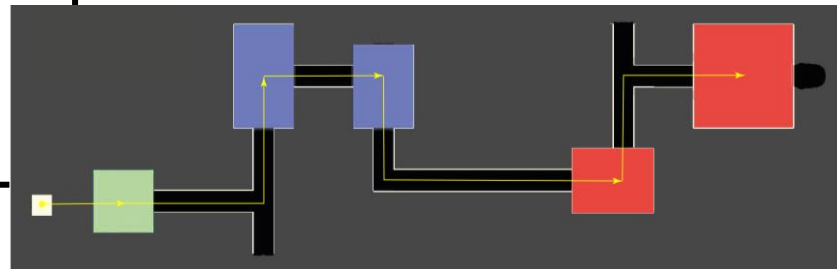
perceived inefficacy

□ Baseline

■ Low

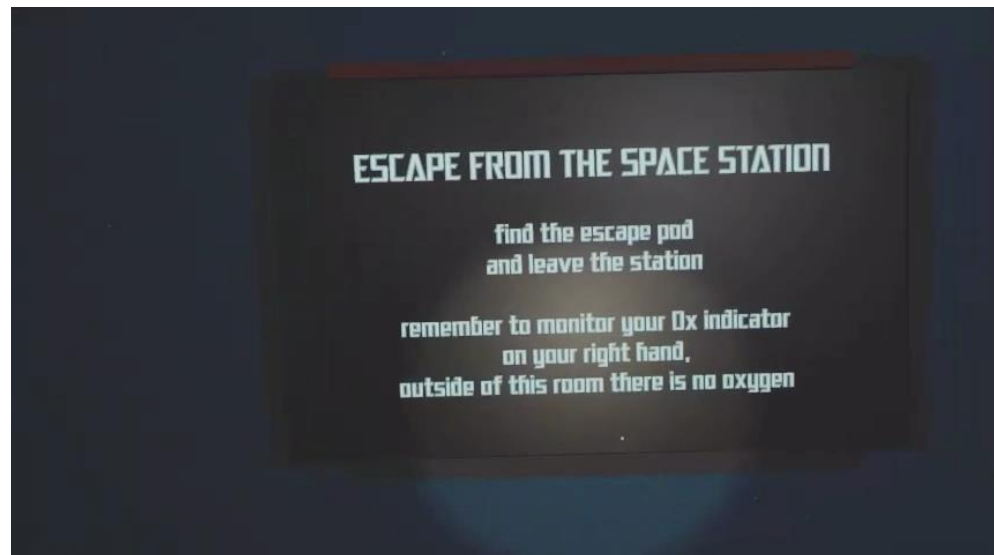
■ Medium

■ High



Stress elements:

- Monstrous enemies
- Scary sounds and music
- Dim lights
- Lack of oxygen
- Scarcity of resources

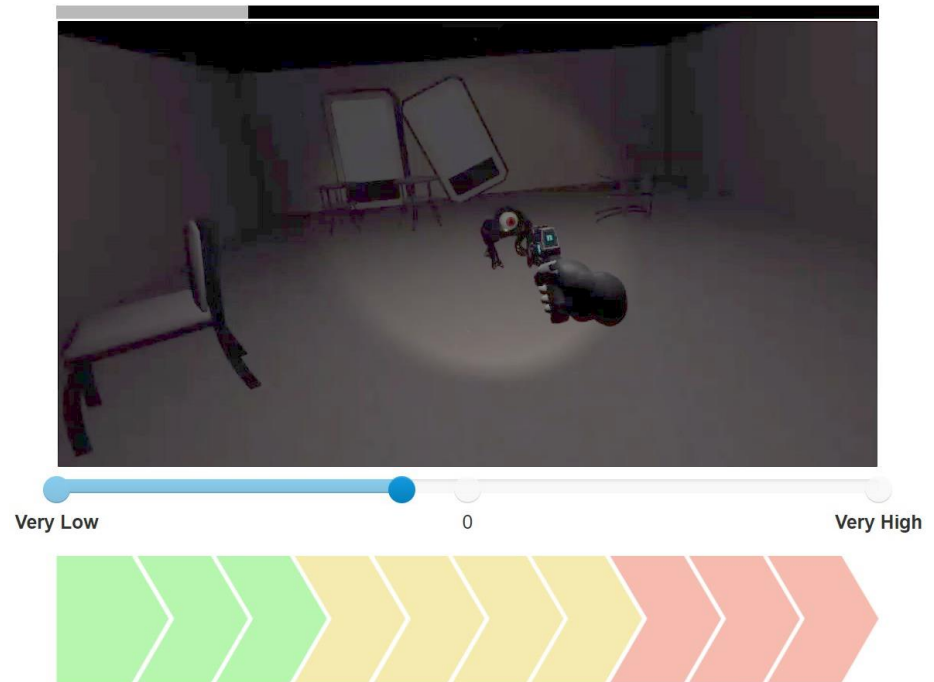
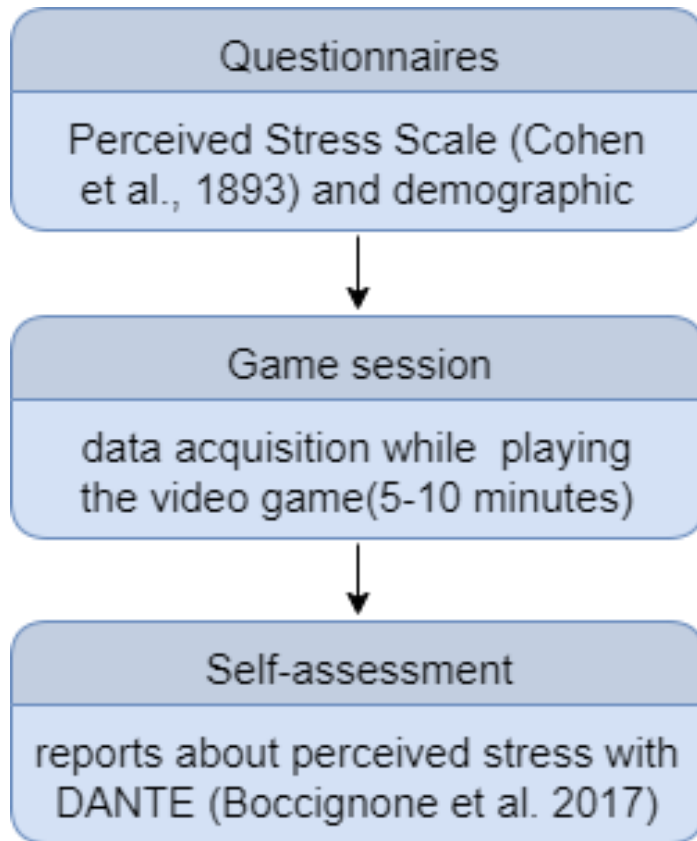


Testing

Data Collection

16 volunteer testers, in the range of 22 and 29 years (mean age of 24.7).

- Participants read and signed a **consent form** prior to the test.
- **No motion sickness**



Data Analysis

Data preparation:

- **filtering and segmentation** using a sliding window (6s size, 1s stride)
- **feature extraction** of statistical features:
 - **VR**: mean, standard deviation, minimum and maximum values, and mean of the number of times a button is pressed
 - **EDA**: mean, standard deviation, minimum and maximum values, slope, dynamic range, mean of the SCR/SCL components, and correlation between SCL and time

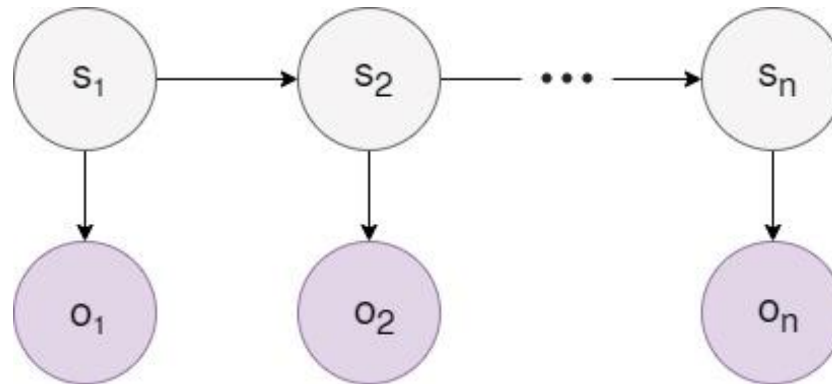
Statistical analysis:

- **Pearson correlation with matrices** computation
- **Selection** based on correlations (73 VR features and 7 EDA features input to the models)

Modeling Stress Level Dynamics

State Space Model (SSM) for stress level dynamics in which each hidden state s_t generates physiological and behavioral observations (vector o_t) at each time step.

SMM to perform posterior inference about hidden states (stress level state estimation).



Problems:

- **classification**, where the stress level is a discrete variable → **Hidden Markov Model (HMM)**
- **regression**, where the stress level is a continuous variable → **Discriminative Kalman Filter (DKF)**

Stress Assessment Models

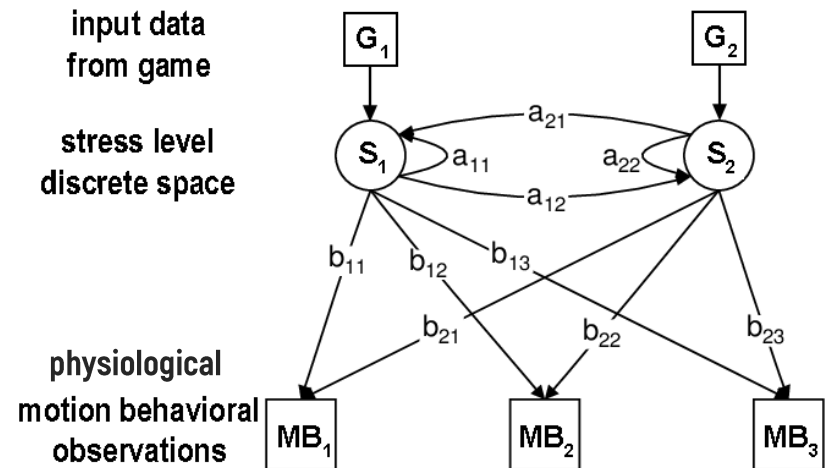
Hidden Markov Model (Bishop, 2006)

Parameters:

- initial probabilities of each state, π
- state transition probabilities A
- emission probabilities B

gaussian distribution of observations

Training: Baum-Welch algorithm



- Label values were **discretized** (between 0 and 1) and different experiments with different number of **discrete states**
 - **Binary:** stress vs no-stress (0-0.3, 0.3-1)
 - **Three classes:** no-stress vs low stress vs high stress (0-0.25, 0.6-1)
 - **Four classes:** no-stress vs low stress vs medium stress vs high stress (0-0.25, 0.25-0.5, 0.5-0.75, 0.75-1)
- **Leave-One-Out Cross-Validation (LOOCV)**
- **Log-likelihood** computation and classification of each test sequence
- **Accuracy** comparing the predicted labels to the actual labels

Stress Assessment Models

Discriminative Kalman Filter (Burkhardt et al., 2020)

Recursive algorithm

- **prediction phase:** next state and errors estimation using current state and error covariance
- **update phase:** estimates correction using the measurement observation

DKF:

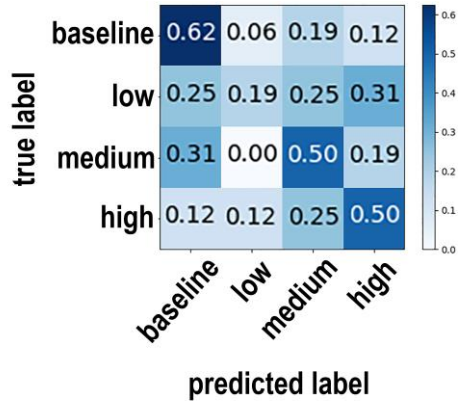
An **approximation** for models with linear, gaussian dynamics but nonlinear, nongaussian observations

- **Leave-One-Out Cross-Validation (LOOCV)**
- **normalized Root Mean Squared Error (nRMSE)**

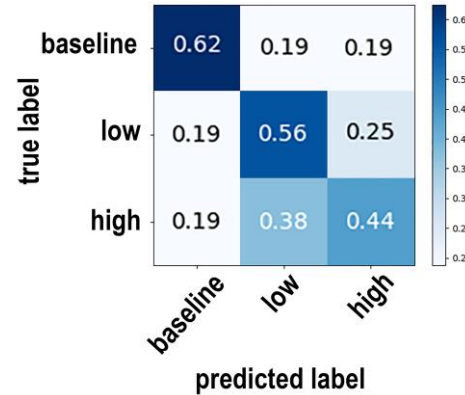
Discrete SSM Results

EDA:

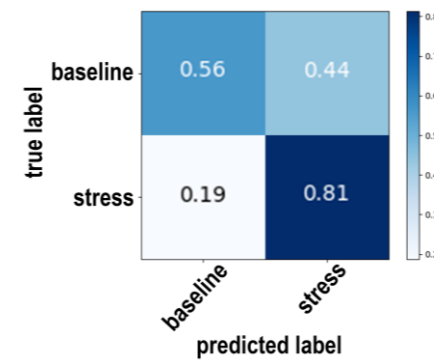
Four classes



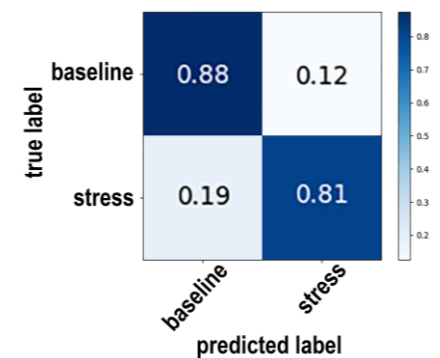
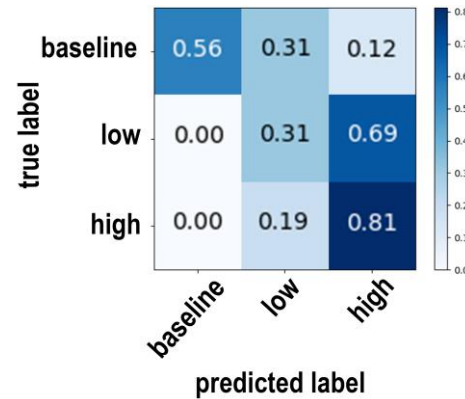
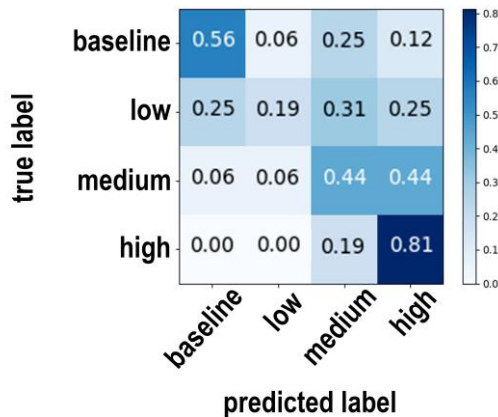
Three classes



Two classes



VR:



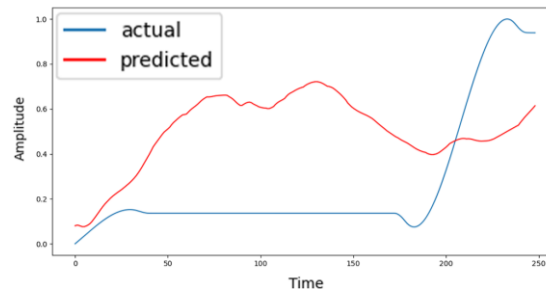
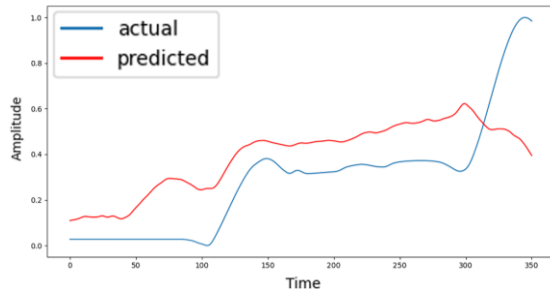
	four classes	three classes	two classes
data	ACCURACY SCORE		
EDA	45.3	54.1	68.8
VR	50.0	56.2	84.4

Continuous SSM Results

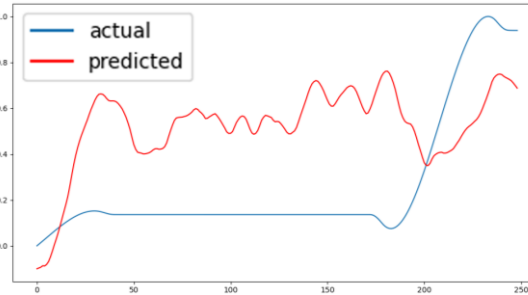
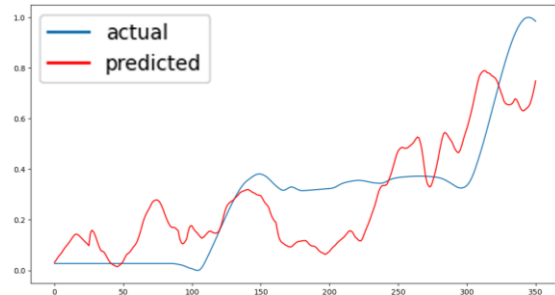
subject 13

subject 15

EDA



VR



time

time

nRmse computed for each trial:

	EDA	VR
subj	nRMSE	
01	0.35	0.44
02	0.65	0.39
03	0.45	0.59
04	0.68	0.42
05	0.39	0.61
06	0.44	0.49
07	0.69	0.89
08	0.48	0.57
09	0.48	0.60
10	0.67	0.73
11	0.36	0.50
12	0.43	0.45
13	0.54	0.43
14	0.54	0.44
15	1.0	1.0
16	0.56	0.80
AVG	0.54	0.57



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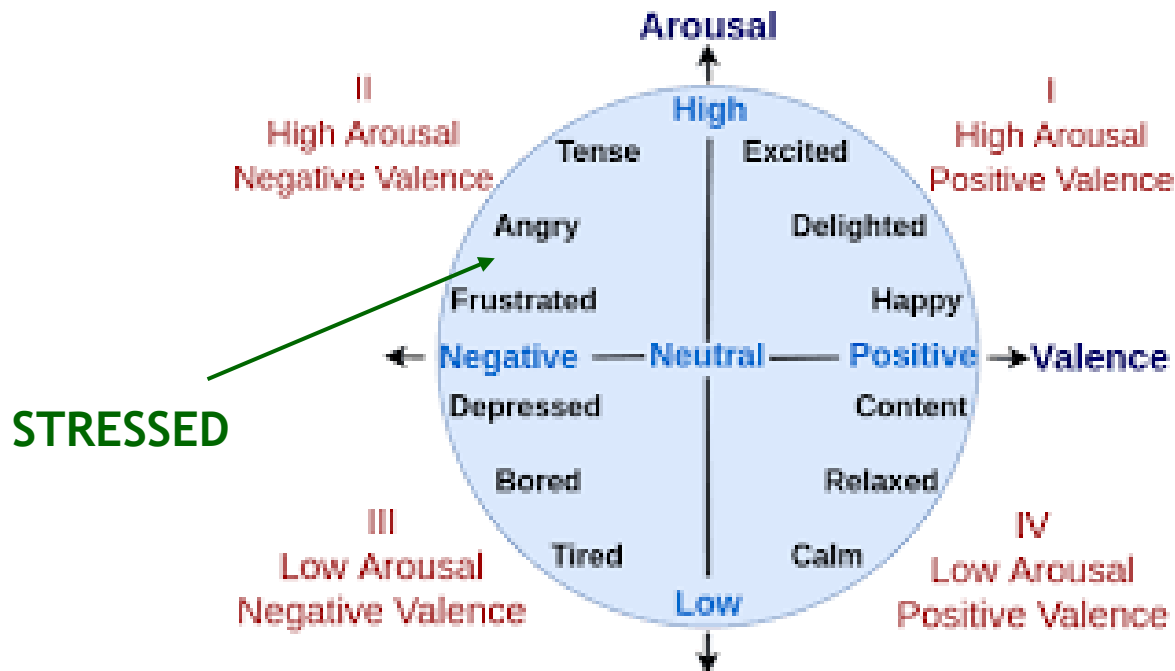
Tracing Stress and Arousal in Virtual Reality Games Using Players' Motor and Vocal Behaviour

Riccardo Lombardi, Giuseppe Boccignone, N. Alberto Borghese, Susanna Brambilla, Eleonora Chitti, Riccardo Lombardi, Laura A. Ripamonti

Key Concepts

(Russell, 2003; Csikszentmihalyi, 1990)

- *Valence*: emotional pleasure (from negative to positive)
- *Arousal*: emotional excitement (from low to high)



Introduction

Idea: integrate **stressors and voice interaction** in a Virtual Reality game to assess player arousal and stress

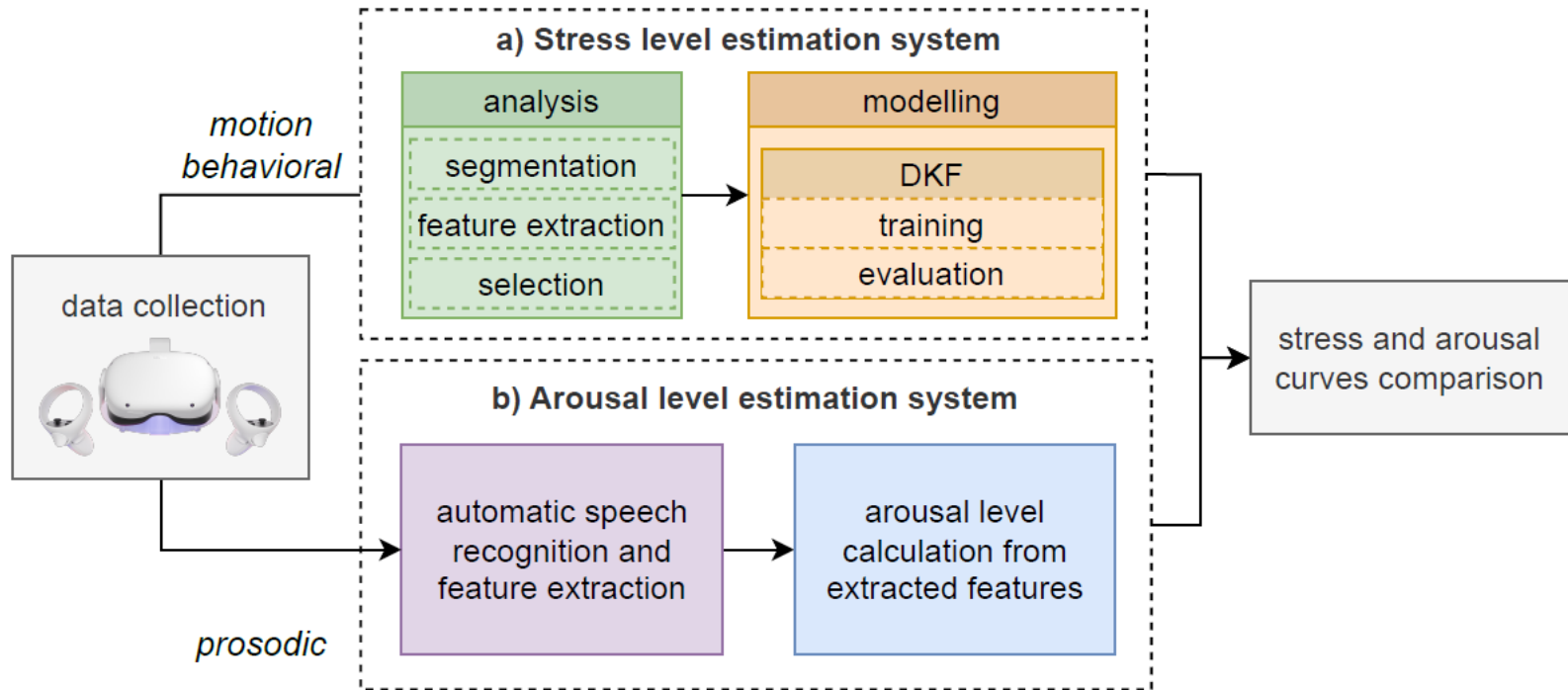
Research questions:

1. Evaluate the **relation of players' actual arousal level with their actual stress level**
2. understand if **stress and arousal can be estimated** from players' behaviours (players' movements/action and voice)

Goals:

- **Entertainment:** improve the players experience (flow)
- Support the development of **emotion-driven video game adaptation**

System Overview



Video Game Voice Commands

A layer of **vocal interactions** added to the game

Minerva -> artificial intelligence of the space station

- **Analyzes the sentence**
- **Extract the needed keywords**

Two types of commands:

1. **Players ask Minerva for resources.**

1. Available at any time in the game
2. by pronouncing «Minerva» and one specific command, Minerva provides what it is requested (e.g., ammunition, oxygen, ...)

2. **Minerva asks a question**

1. Available in different stages of the game
2. The AI asks the players to answer a specific question (e.g., at the beginning of the game to collect baseline data players are asked to introduce themselves)

Speech recognition

Automatic Speech Recognition (ASR) or *Speech-to-text*: feature that allows a program to process human speech into a written format.



DictationRecognizer: functionality of the *Unity Speech Recognition* library, which exploits the *Windows Online Speech Recognition* technology (Cortana) to perform the Speech-To-Text operation

Values:

- **Fundamental frequency F_0 (Hz)** -> calculated from frequencies, over time on a specific audio channel
- **Intensity (dB)** of players' voice -> calculated from amplitude of the sound over time on a specific audio channel

Arousal from Voice

Fundamental frequency and intensity values collected used to calculate:

- μ : the mean of the values,
- σ : the standard deviation,
- *min*: the minimum value of the voice recording,
- *max*: the maximum value of the voice recording,
- *R*: the range, namely, the difference between max and min.

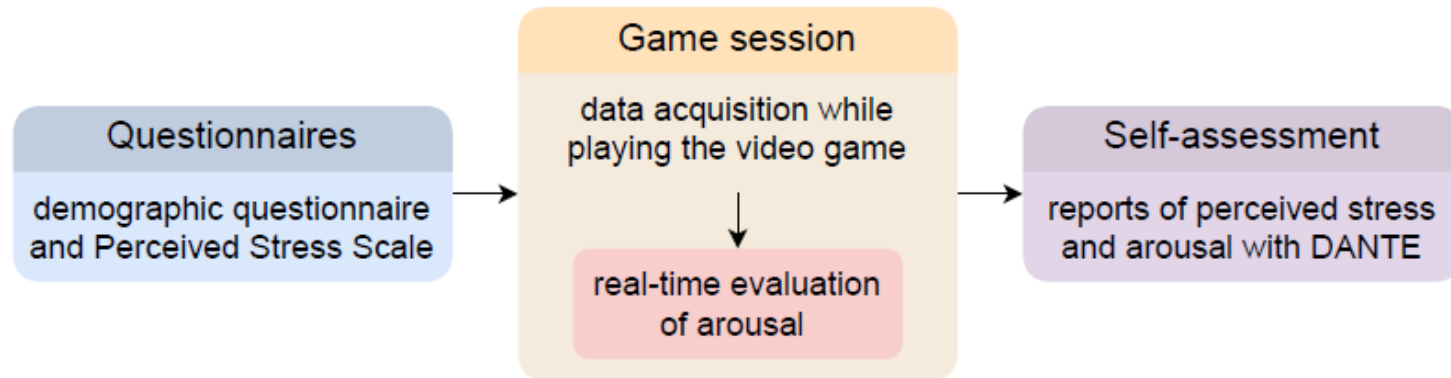
The **arousal** obtained is a number between -1 and 1 -> compared to baseline and evaluated in **real-time**

$$a_i(n) = \text{sgn}(x_i(n) - \underline{x_i}) \quad \overline{a(n)} = \frac{1}{N_x} \sum_{i=1}^N a_i(n)$$

Testing

16 volunteer testers, average age 25.

- Participants read and signed a **consent form** prior to the test.

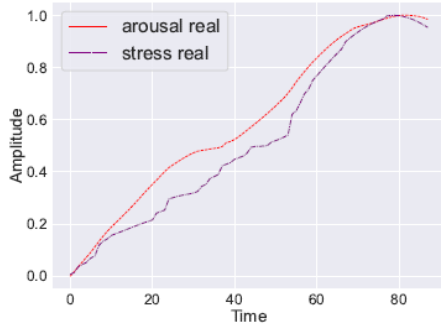


DANTE (Dimensional Annotation Tool for Emotions) [Boccignone et al. 2017]:

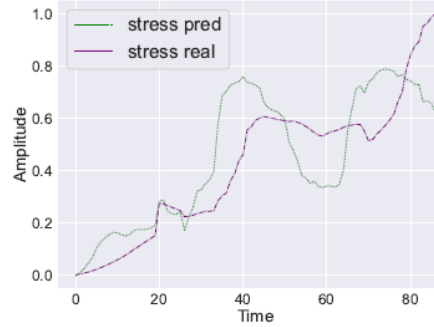
- watch gameplay video
- continuous self-assessment of stress and arousal



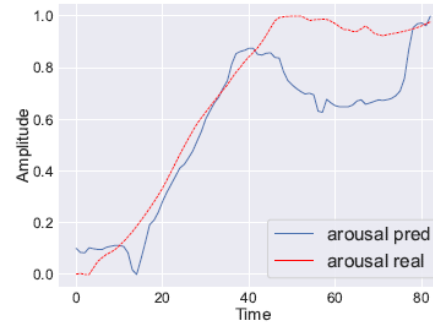
Results



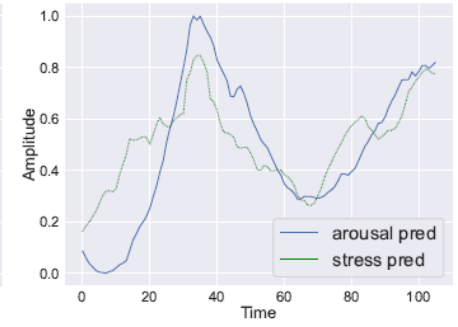
(a) Curves related to subject 08.



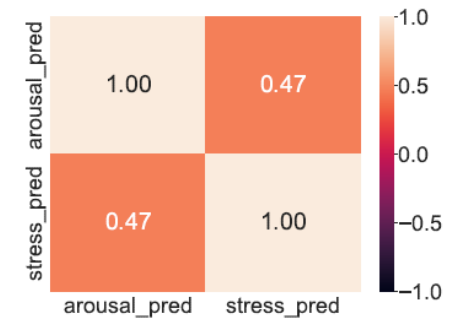
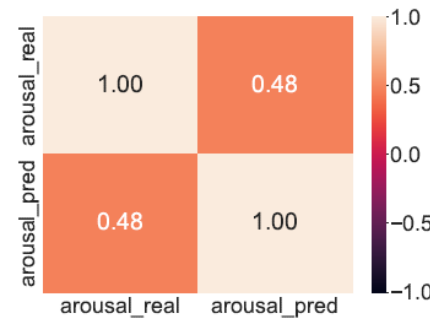
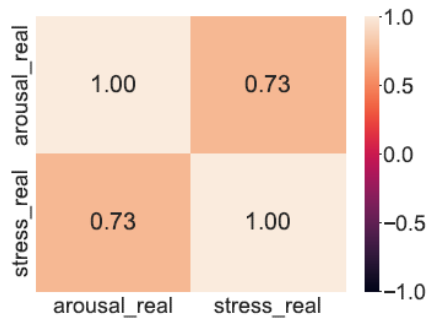
(b) Curves related to subject 12.



(a) Curves related to subject 07.



(b) Curves related to subject 16.





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StrEx: Towards a Modulator of Stressful Experience in Virtual Reality Games

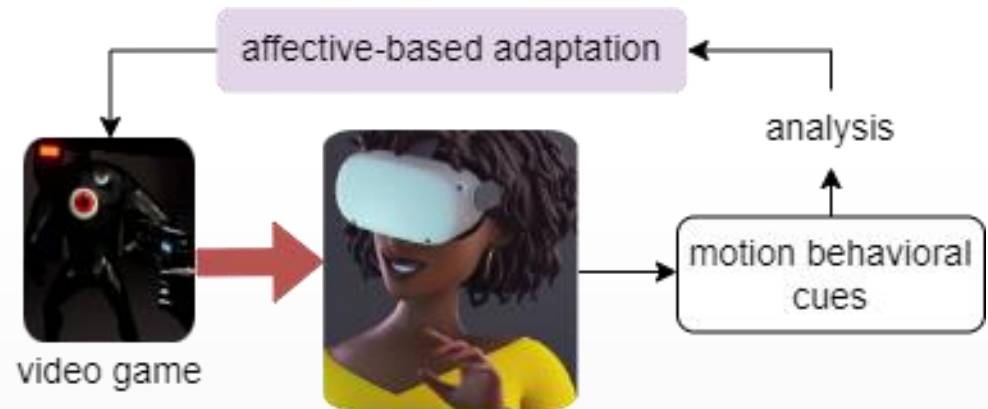
Daniele Croci, Giuseppe Boccignone, N. Alberto Borghese, Susanna Brambilla, Laura A. Ripamonti



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Introduction

Affective Gaming (AG):

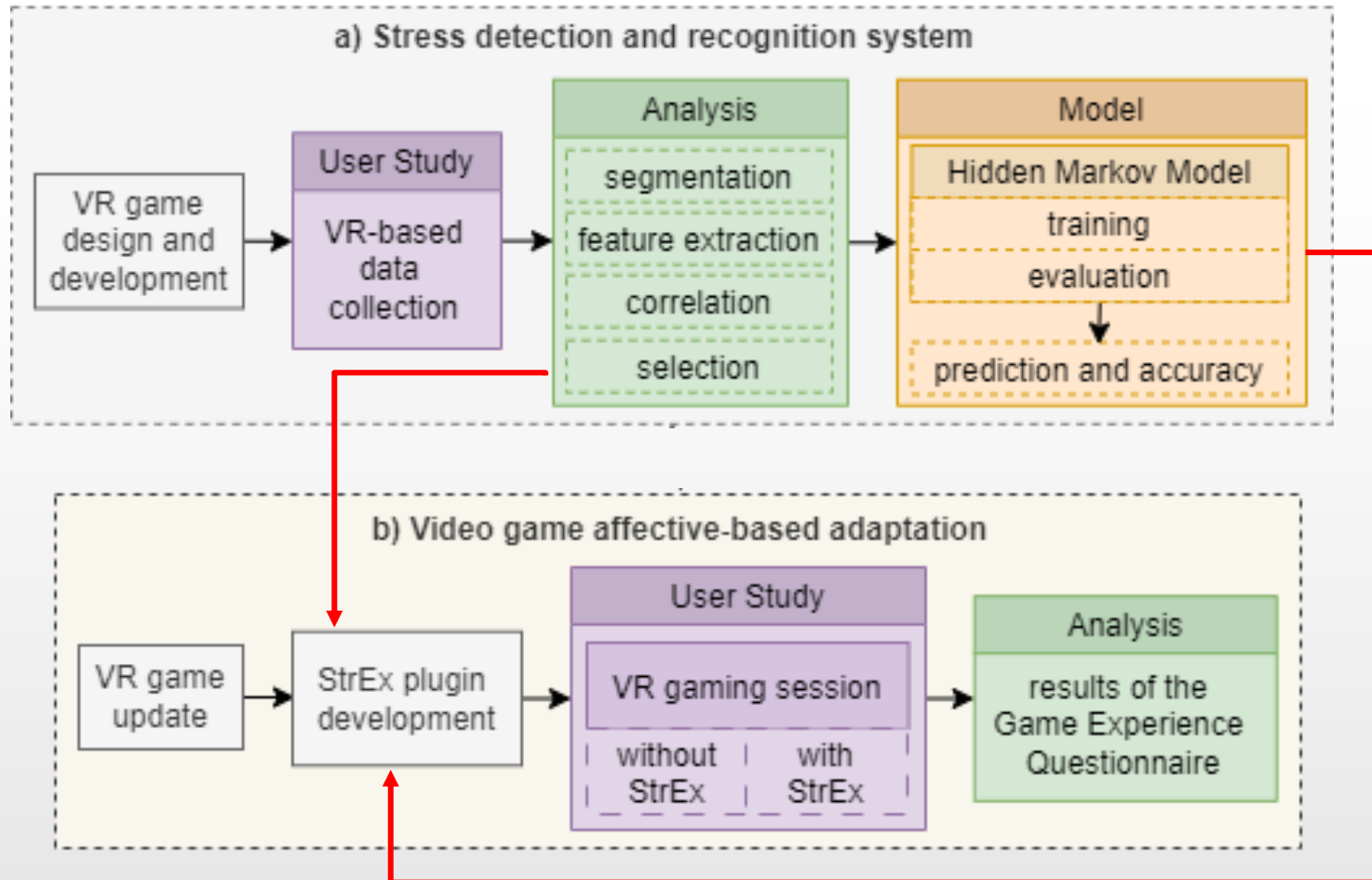


Idea: design and develop a system that **dynamically adapts** a virtual reality game content based on **players' stress**

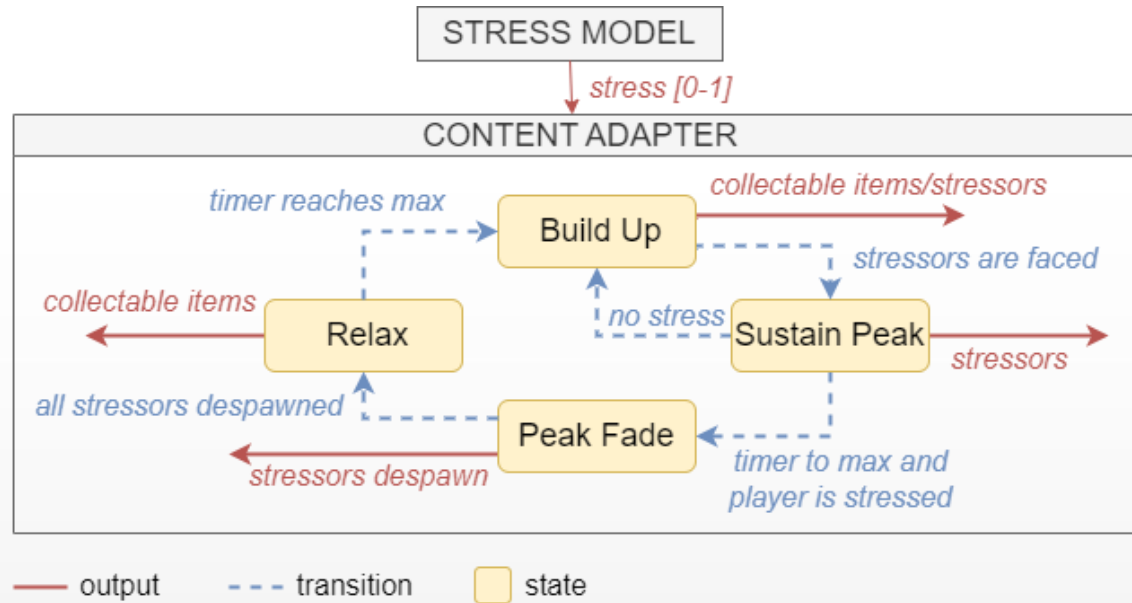
Goals:

- Previous **model validation**, HMM binary model trained with **motion behavioral data** (Brambilla et al., 2022)
- **Evaluate** impact on players' **experience**

Overall System



StrEx Structure



Build up: the AI generates stressful elements and items near the player; when the player faces an external stressor -> Sustain Peak

Sustain peak: if a threshold of the stress level is exceeded by players (in a predefined time period) -> Peak Fade; else -> Build Up

Peak fade: all stress elements the players are not facing are removed -> Relax

Relax: no stressful elements are generated, only items; after a recovery time (30-45 seconds) -> Build Up

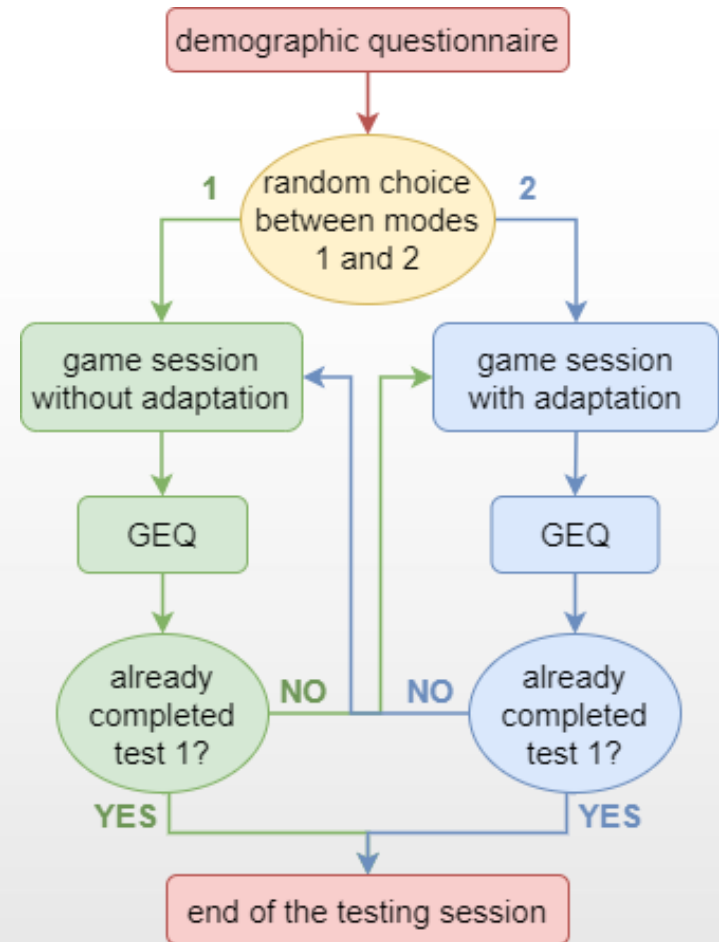
Testing

13 volunteer students, age 18-29

Game Experience Questionnaire (GEQ)

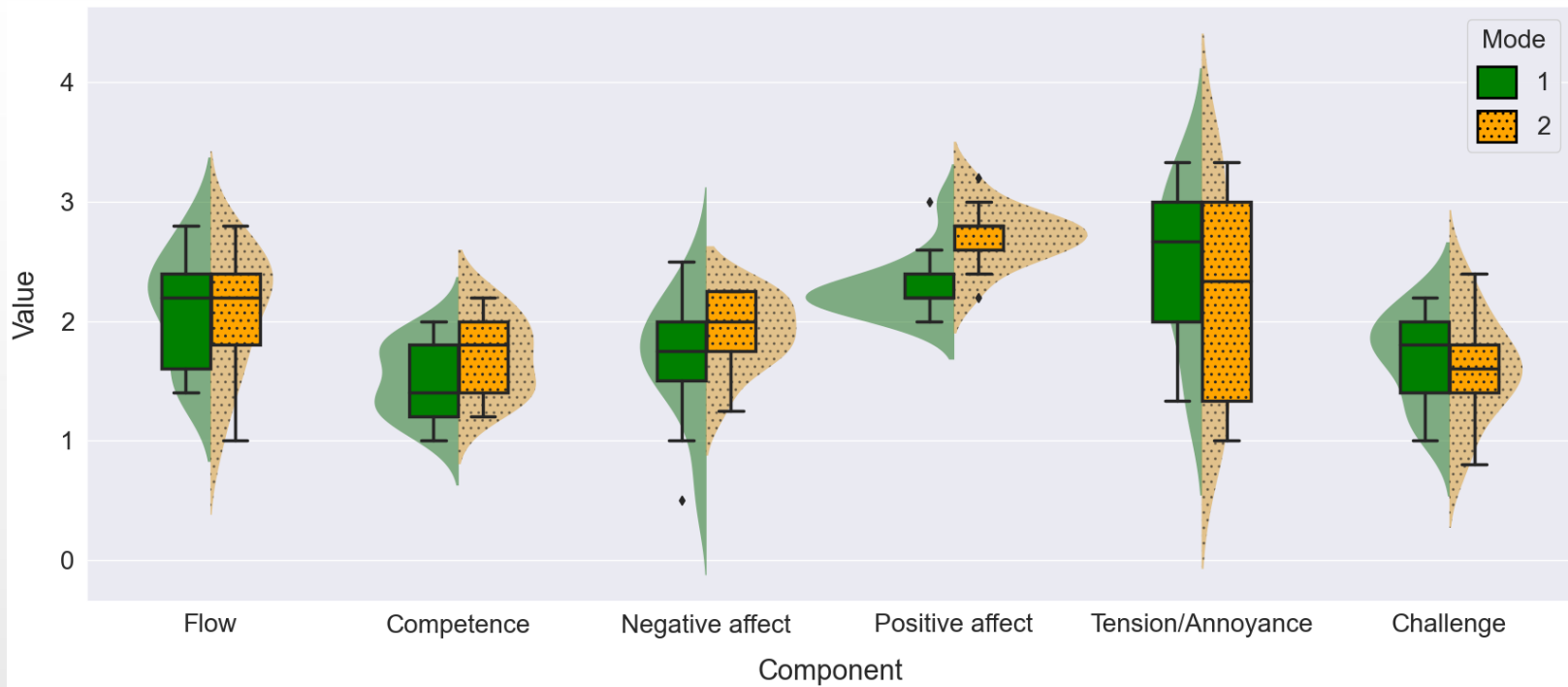
Components:

- Sensory/imaginative immersion ❌
- Flow ✓
- Competence ✓
- Positive affect ✓
- Negative affect ✓
- Tension/annoyance ✓
- Challenge ✓



Results

Violin plots: **distribution** of the average score for the different components using all subjects, **comparing the two game modes**



CONCLUSIONS

Promising results:

- **potential** for using **voice interaction** and **motion behavioral data** as methods to assess players' **affective states** during VR game play
- Stress-based adaptation could **improve** players' **perceived competence** and the game becomes **less frustrating and challenging** which results in a **decrease in tension/annoyance**

Future works:

- gather **data** from more subjects
- include information for players **personalization**
- inclusion of other **channels**

Thanks for your attention