

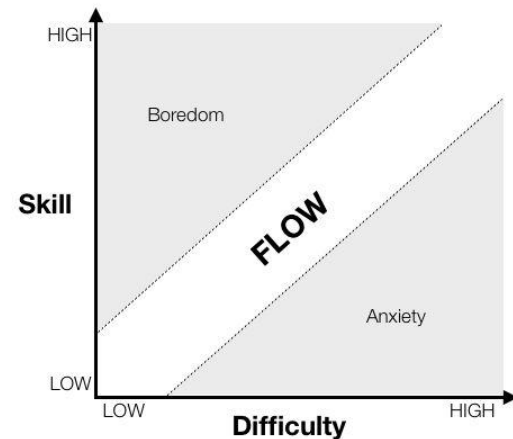
Physio and emotional signals for better Game Flow



Stefan N, Shao-Yen T, Dillon K, Qinyi L

Introduction

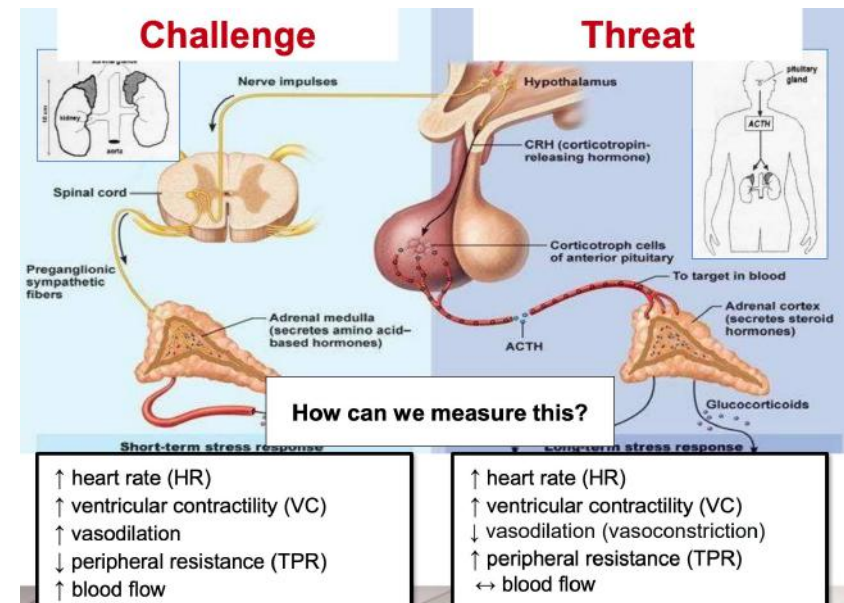
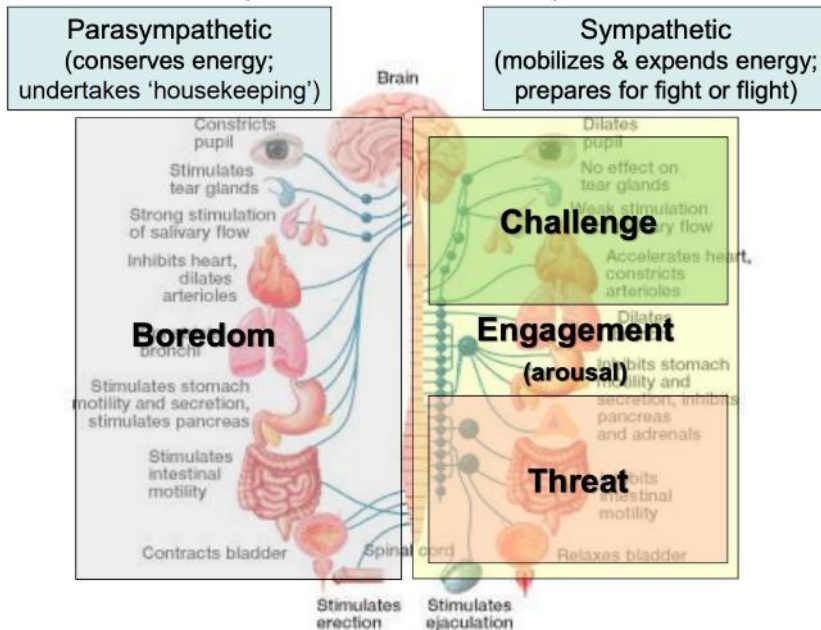
- **What:** make video-game more enjoyable
- **How:** Dynamic Difficulty Adjustment (DDA) based on affective signals
 - Flow: psychological state of being completely involved in an activity
 - Not too boring, nor too frustrating
- **Why:**
 - recreation is essential to a balanced life
 - Good video-games can be hindered by small details such as difficulty
 - Applicable to E-learning
 - Past studies on DDA do not involve facial expressions as feedback
 - Physio signals are relatively un-fakable



Theoretical perspectives built upon

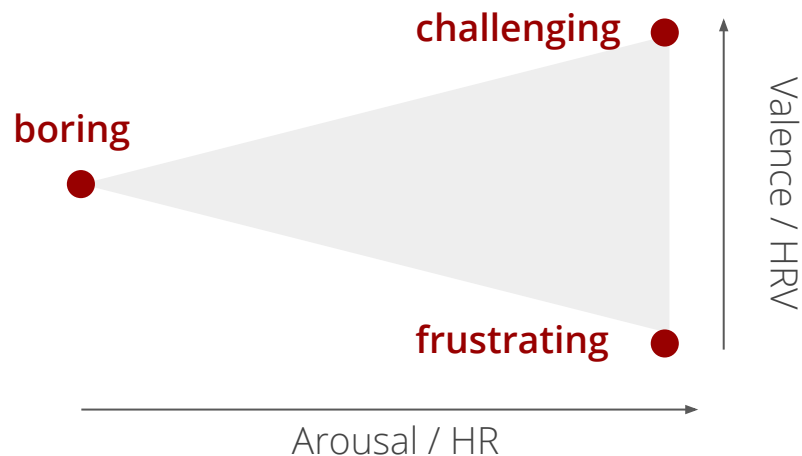
- **General area:** affective feedback
- Basic emotions displayed through facial expressions
- Sympathetic and parasympathetic relation to arousal
- Increased blood flow = increased challenge

Emotion	Action Units
Happiness	6+12
Sadness	1+4+15
Surprise	1+2+5+26
Fear	1+2+4+5+7+20+26
Anger	4+5+7+23
Disgust	9+15+16



Hypothesis

- Players enjoy the game the most when they are engaged
- By adapting the game's difficulty we can impact valence and arousal
- Player is more aroused \Rightarrow higher heart-rate
- When aroused, more engaged \Rightarrow higher heart-rate variability
- Described adaptive version is more enjoyable
 - Than non-adaptive versions
 - Than adaptive versions without affective signals



Current state-of-the-art

Table 1. Properties of affective video games

N	References	Game name	Game purpose	Game genre	Delivery platform	Mode	Adapted game features
1	[92]	PhysiRogue (on Rogue Signals game)	applied	2D action	desktop, GPS-location-aware	multi-player	Seekers' desirability of predators, player visual representation
2	[10]	Half-Life 2	entertainment	3D FPS	desktop	single-player	Avatar speed, sound volume, environmental density, gravity, and transparency; weapon damage; red and b/w filters; NPC creation
3	[63, 96]	EMO-Pacman (on Pac-Man game)	entertainment	2D arcade	desktop	single-player	Objects' speed
4	[83]	Tetris	entertainment	2D puzzle	desktop	single-player	Speed of falling Tetris blocks
5	[16, 97]	Bug-Smasher (a Playware ambient game)	entertainment	physical platform	Playware playground platform	single-player	Bugs' speed and the entropy of bug-visited tiles
6	[9]	Pong	entertainment	2D arcade	desktop	two-players	Speed and size of both ball and paddle; sluggish or overresponsive keyboard
7	[86, 93]	MindTactics (Unity3D game)	applied	3D strategy	desktop	single/multi-player	Territorial control and distractors
8	[87]	A 3D games on TORCS ¹	entertainment	racing	desktop	single-player	Opponent skill
9	[32]	2D Xbox360 FPS game	entertainment	2D side-scrolling shooter	desktop with Microsoft Xbox 360 controller	single-player	Enemy target size, flame length, speed and jump height, weather conditions and boss speed
10	[20]	Tetris	entertainment	2D puzzle	desktop	single-player	Speed of falling Tetris blocks
11	[84]	Archery game	entertainment	shooting	desktop	single-player	Archery focus level
12	[88, 91]	VANISH	entertainment	3D FPS	desktop	single-player	Event probabilities, PCG and tunnel vision effects; character's movement speed, stamina, and sanity level
13	[34]	Car racing game	entertainment	3D racing	desktop, Logitech G27 racing wheel	single-player	Visibility, steering, and speed
14	[94]	BioPong (based on the Pong game)	entertainment	2D arcade	desktop	two-players	Ball speed, paddle size

Bontchev, Boyan. "Adaptation in affective video games: A literature review." *Cybernetics and Information Technologies* 16.3 (2016)

Current state-of-the-art

No	FT ²	Measured signals	F ³	Biofeed-back device	Rate (Hz)	Time window (s)	Calibration (min)	Signal filtering	Recognized emotions	Classification / estimation	Accuracy (%)
1	+	Phasic EMG Phasic EDA Tonic EMG Tonic EDA	1 1 1 1	ProComp2 (Thought Techn.)	10 0,33 10 0,33	1 30 6 300		Kalman filter	Stress level	Direct mapping	
2	+	HRV (ECG) HR EDA	2 2 2	Lightstone (Wild Divine)		2	5	Average downsampling at 2 s	Horror	Direct mapping	
3	—	BVP EDA EMG RESP KEYB	4	NeXus and Biotrace+ (Mind Media)	128 32 1024 32 100	180		High-pass Chebyshev 20Hz filter; smoothing filter	Boredom, frustration and enjoyment	Direct mapping	
4	C	EDA BVP HR RESP TEMP	4 2 3 3 2	Biosemi Active 2	1024	20	1.5	Moving average filters	Boredom, anxiety, and engagement	SVM	53.33
5	+	HR BVP EDA	13 7 13	ProComp Inf. (Thought Techn.)	256	45	0	Discrete Fourier Transform (DFT) filter	Entertainment value	ANN	79.76
6	C	ECG PPG EDA EMG TEMP	17 3 5 16 3	Biopac (BIOPAC Systems)	1000	0.025 0.1		Low-pass filters, DFT, wavelets	Anxiety level	RT kNN BNT SVM	88.5 80.4 80.6 88.9
7	+	fNIR	4	Bespoke fNIR device	2048	16.5	0.33	Low-pass cut-off frequency filter (0.14 Hz)	Attention level	kNN NBC	73.77 57.37
8	+	BVP ECG EDA RESP TEMP	11 11 11 11 11	ProComp Infinity (Thought Techn.)	2048 2048 256 256 256	60	1	Horror	Preference level	LDA	74
9	+	BVP EDA ECG EMG RESP TEMP	1 1 1 1 1 1	Flexcomp Infinity (Thought Techn.), TTLAPI	2048		10	Chebyshev type II filters, downsampling by 64	Preference level	Direct mapping	
10	C	Alpha & theta (EEG)	2			2	2		Boredom, engagement, flow, overload	Linear estimation	
11	C	Alpha (EEG)	3	Bespoke device	256	2	3	FFT band-pass filter (0.5~50 Hz)	Focus level	Direct mapping	
12	C	BVP EDA EMG	2 2 1	Nexus-10 (Mind Media) and BioTrace+	32 32 1024	2 5 0.125	5	Smoothing filter (moving average)	Arousal and valence levels	Linear / nonlinear regressions	85
13	+	EDA	1	FlexComp (Thought Techn.)	1	30	8		Arousal	Direct mapping	
14	C	EDA HR	1 1	Arduino			0		Arousal	Direct mapping	

Bontchev, Boyan. "Adaptation in affective video games: A literature review." *Cybernetics and Information Technologies* 16.3 (2016)

Methodology

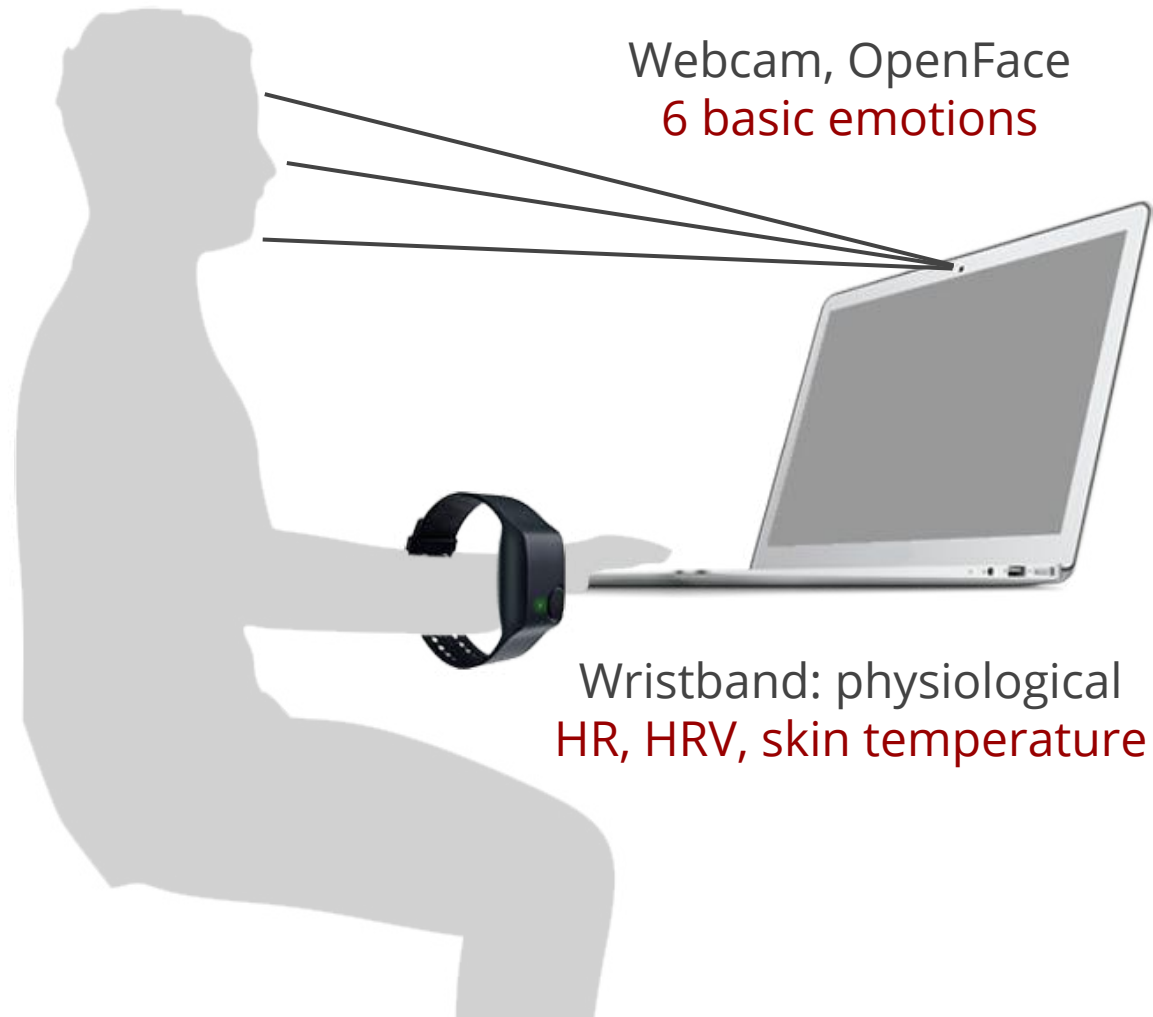
Study design

- Ask basic info about competitiveness/video game skill
 - a. For baseline and later analysis
- Play four versions of the game
 - a. Two for baseline: very easy, very hard
 - System baseline and participants accommodation
 - b. Two adaptive ones: using a classic strategy and using our strategy
 - Counterbalanced design: ordered randomly
- Each version replayed multiple times (because it is very short)
 - a. Last two versions can be played for as long as desired
- After each version, participant rates it (out of five stars):
 - a. Difficulty: easy to hard
 - b. Valence: frustrating to challenging — whether it was hard in an enjoyable way
 - c. Enjoyment: how much they liked it overall
- After playing all versions, participants are asked to pick their favorite
- Participants don't know what the game adapts to
 - a. But are (obviously) aware that they wear a measurement watch
 - b. And that they are being recorded

Variables

Independent variable	Whether to adjust difficulty based on affective feedback
Dependent variable	User enjoyment: ratings, favorite version, play duration
Measurements	<p>Affective feedback:</p> <ul style="list-style-type: none">- emotions through facial expressions- HR, HRV, skin temperature <p>Supplementary info:</p> <ul style="list-style-type: none">- participant's video game skill / competitiveness- self-reported difficulty perception- self-reported valence (frustrating vs challenging)

Capturing affective signals



Affective signals

- Emotion signals
 - OpenFace: generate facial Action Units
 - Subtract each participant's AU baseline
 - Compute the 6 Basic Emotions by summing AUs
- Physiological signals
 - Heart rate
 - HRV = inter-beat interval standard deviation
 - Skin temperature
 - Baseline collected during first variant



Classic adaptive strategy

Goal: maintain player performance at below peak level

After every turn

- Adjust the difficulty based on performance of **previous turn**
- Increase/decrease **base difficulty** using percentage overlap
 - > 50% : increase difficulty scaled by % overlap (maximum 5)
 - < 50% : decrease difficulty scaled by % overlap (maximum -5)
- The maximum change is also affected by reported user skill

Every 500 ms

- Add **random fluctuations** ± 3 (constrained by base ± 13)

Affective adaptive strategy

Goal: maintain player affective at a certain level of anxiety

- Physiological signals (cardiovascular and skin temperature):
 - Adapt difficulty in order to increase physio values
 - If any drop below the user's baseline: **harder**
- Emotional signals
 - Adapt difficulty in order to invoke emotions (measured through facial expressions)
 - If player displays happiness, surprise or sadness: **harder**
 - If player displays disgust, anger or fear: **easier**
 - (emotions in order of magnitude)

Rohrmann, Sonja, Jürgen Hennig, and Petra Netter. "Changing psychobiological stress reactions by manipulating cognitive processes." *International Journal of Psychophysiology* 33.2 (1999)

Liu, Changchun, et al. "Dynamic difficulty adjustment in computer games through real-time anxiety-based affective feedback." *International Journal of Human-Computer Interaction* 25.6 (2009)

Experiment

Study Phases

- Participants: USC students
- Participants don't know
 - Hypothesis
 - If/how the difficulty is adjusted
- Phase I
 - 10 participants
 - Discarded: used for learning lessons
- Phase II
 - 13 participants
 - Improve UI and mechanics
 - Analyzed and reported on

Know how
far along
you are



Welcome!

You will play a game while wearing a measuring watch and being recorded.

Game: stack platforms on top of each other, as many and as neatly as you can.

Scoring: the higher the difficulty, the larger the reward. Easy games increase score very slowly.

Setup: you will play 4 versions (color-coded, as indicated at the top). You will replay each version a couple of times.

Feedback: you will rate each version after you play it. Try to remember them, at the end you will pick your favorite.

Got it



Preliminary info about yourself:

Name

How skilled are you at video-games?



How competitive are you, in general?



Start

Game: Stack

- Mechanics
 - Stack platforms on top of each other
 - Platform moves horizontally, tap to drop it
 - The less precise your drop, the harder it'll be
 - Game ends if you drop the platform outside the tower
- Difficulty adjustments
 - Platform lateral movement
 - Higher drop position

Intro > Green Game > Red Game > **Purple Game** > Blue Game > Outro

Score accounts for difficulty to reduce "easy-is-best" bias

8.32

Play as long as you wish

Press **X** whenever you'd like to advance

After every version

Intro Green Game Red Game **Purple Game** Blue Game Outro

Amazing! You were in the top 1.2% players for this version.

Rate your previous game:

☒ ☐ ☐ ☐ ☐

easy hard

☒ ☒ ☒ ☒ ☐

frustrating challenging

☒ ☒ ☒ ☒ ☒

unfun fun

Next

Reduce comparison across versions

Difficult can be either good or bad

At the end

Intro Green Game Red Game Purple Game Blue Game **Outro**

Closing thoughts:

Favorite version?

☐ Green

☐ Red

☒ Purple

☐ Blue

Comments:

What you liked, what we can improve?

☒ Ok to use my data for this class project

Done

Recognition rather than recall

Results

Versions ratings

85% of participants rated it higher or equal to the classic version

77% of participants rated it (their) highest out of all versions



Order bias

Affective first

Classic	0	1	1	3	3
Affective	0	2	1	2	3

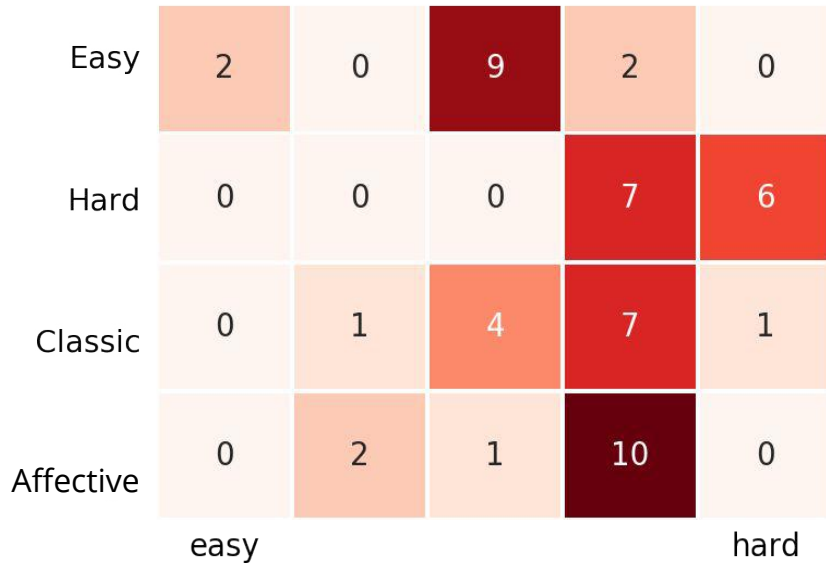
Affective last

Classic	0	0	0	3	2
Affective	0	0	1	0	4
	1 (unfun)	2	3	4	5 (fun)

Enjoyment rating

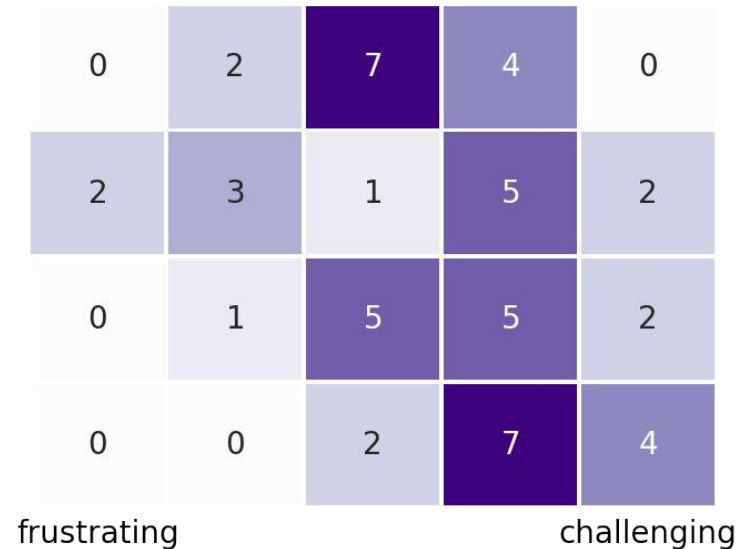
Specialized ratings

Difficulty



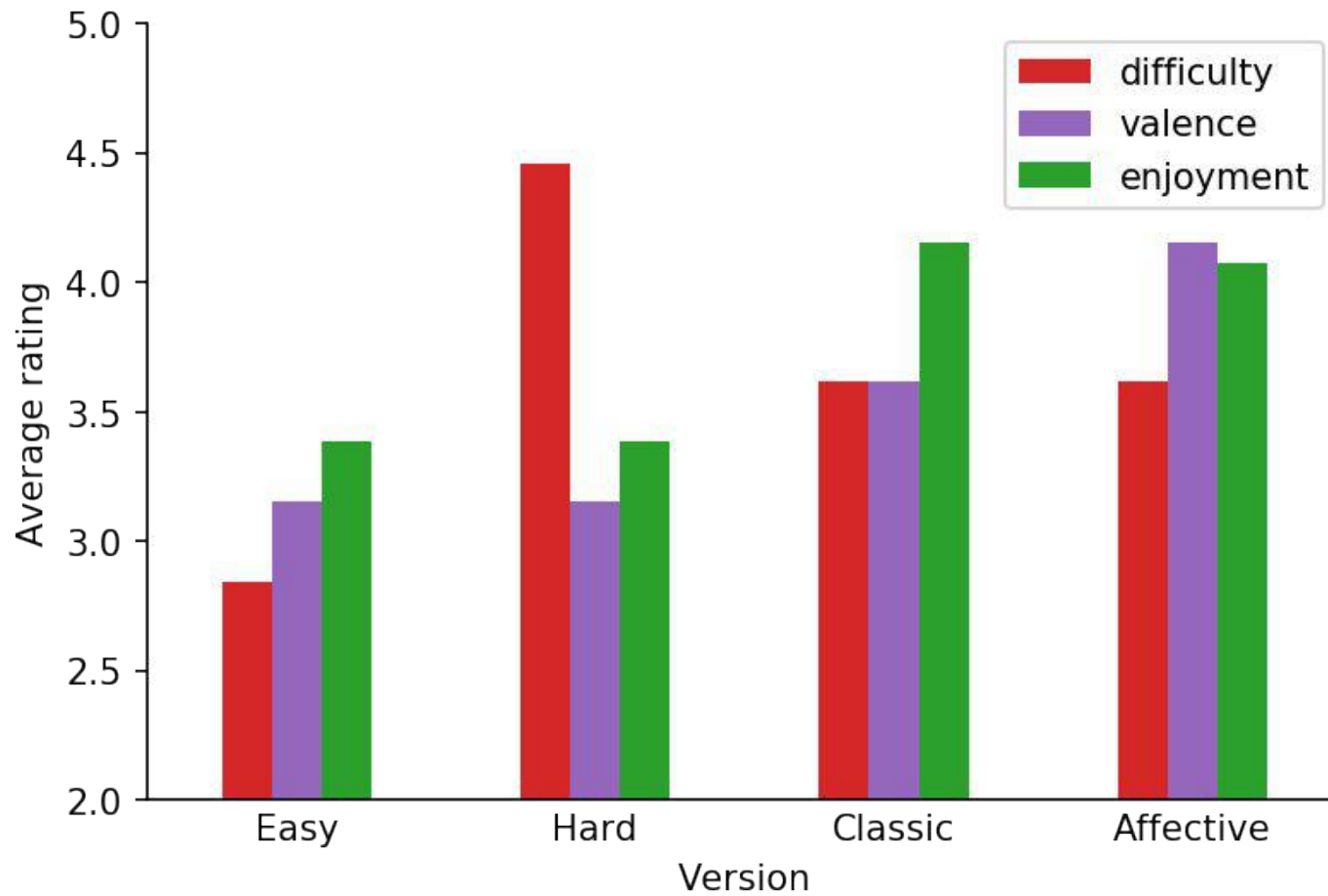
Ours was harder...

Valence

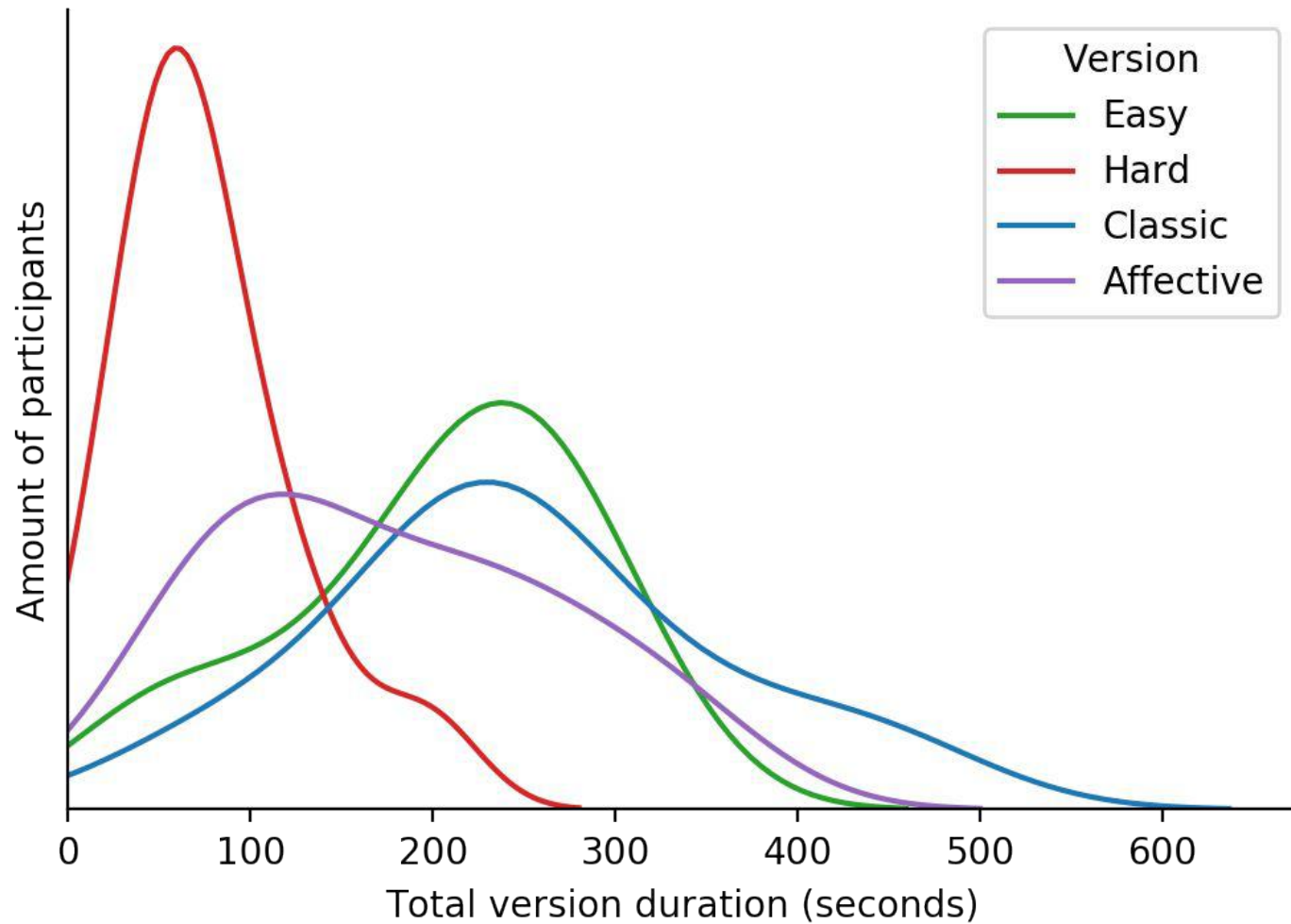


... but in a pleasant way

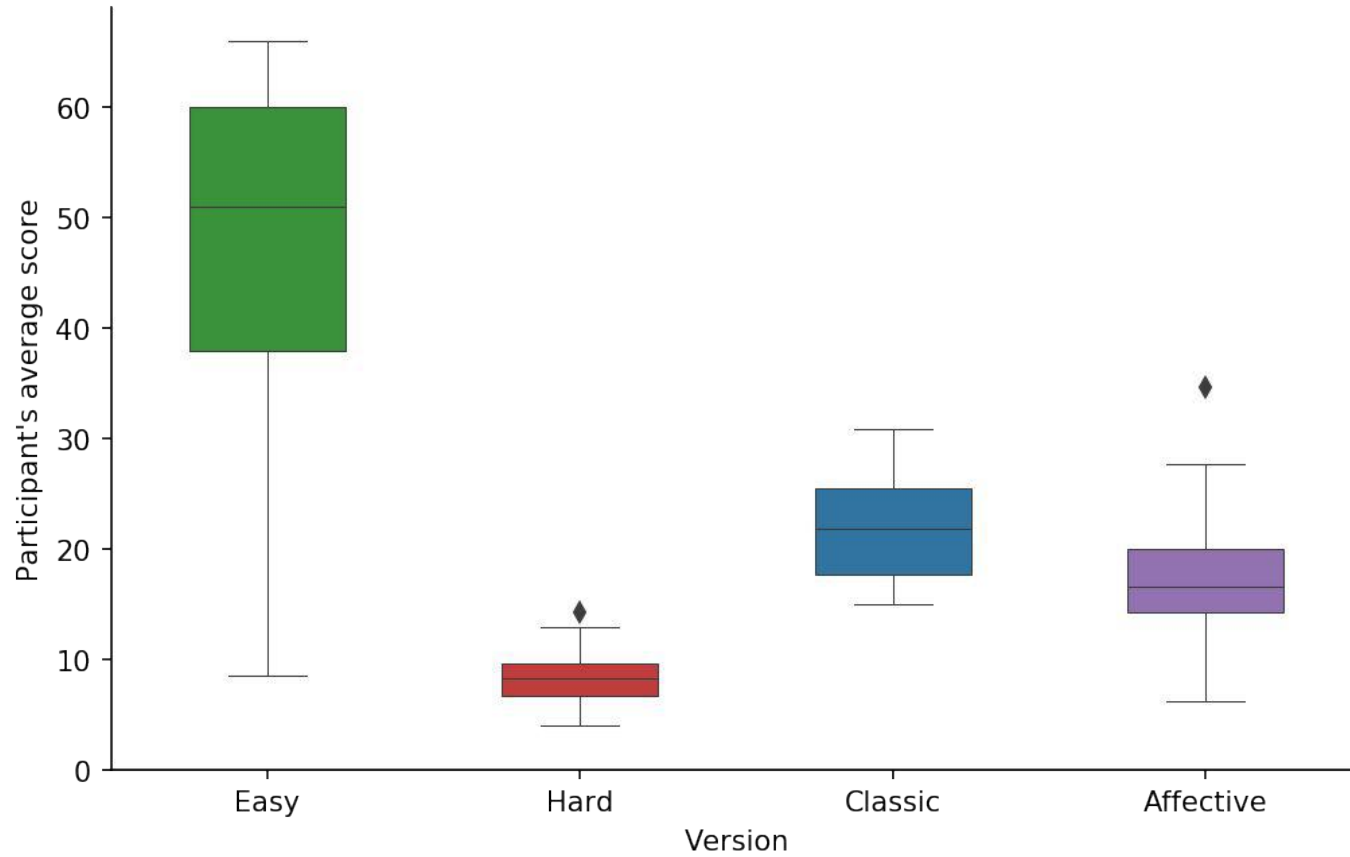
Rating Averages



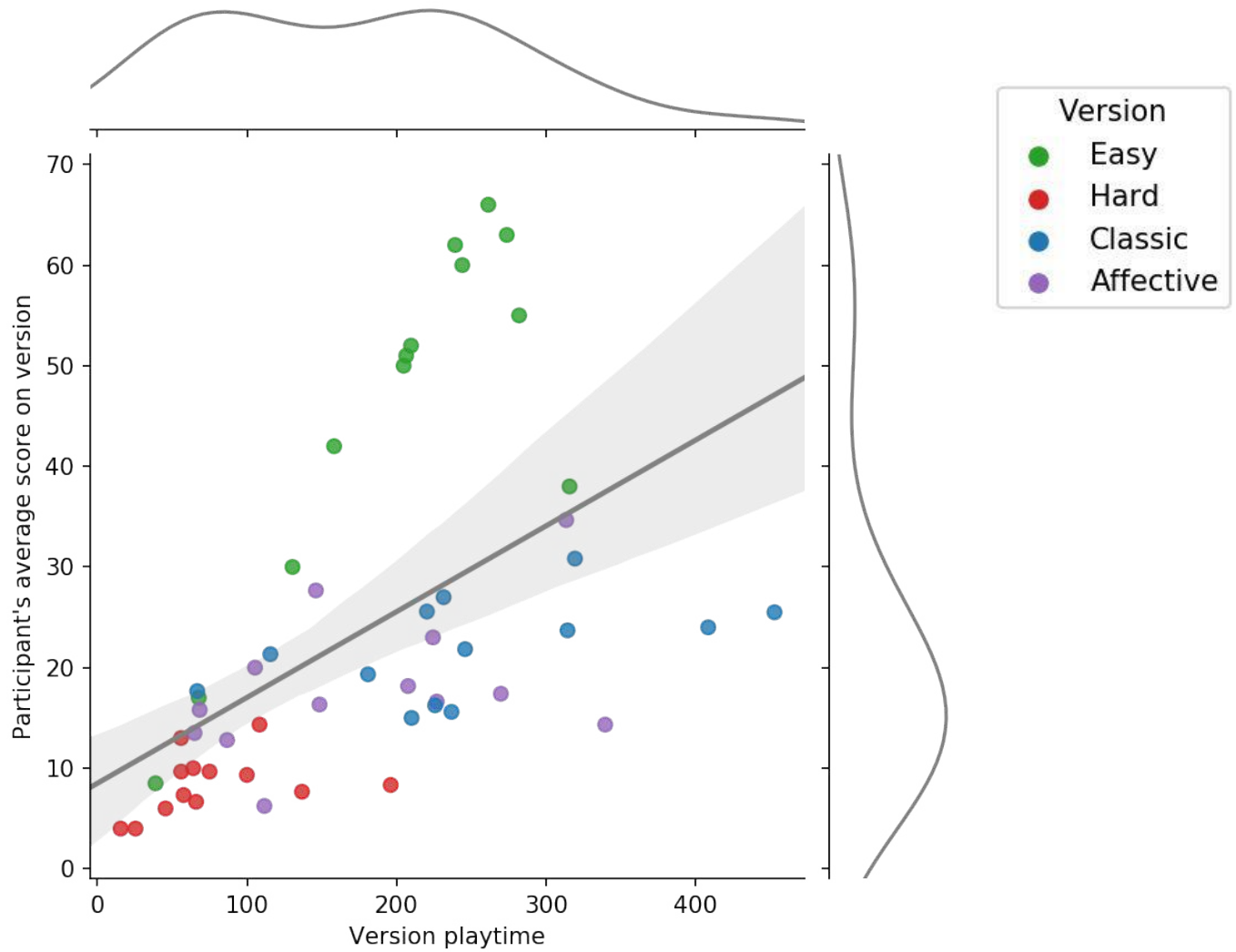
Version play-time



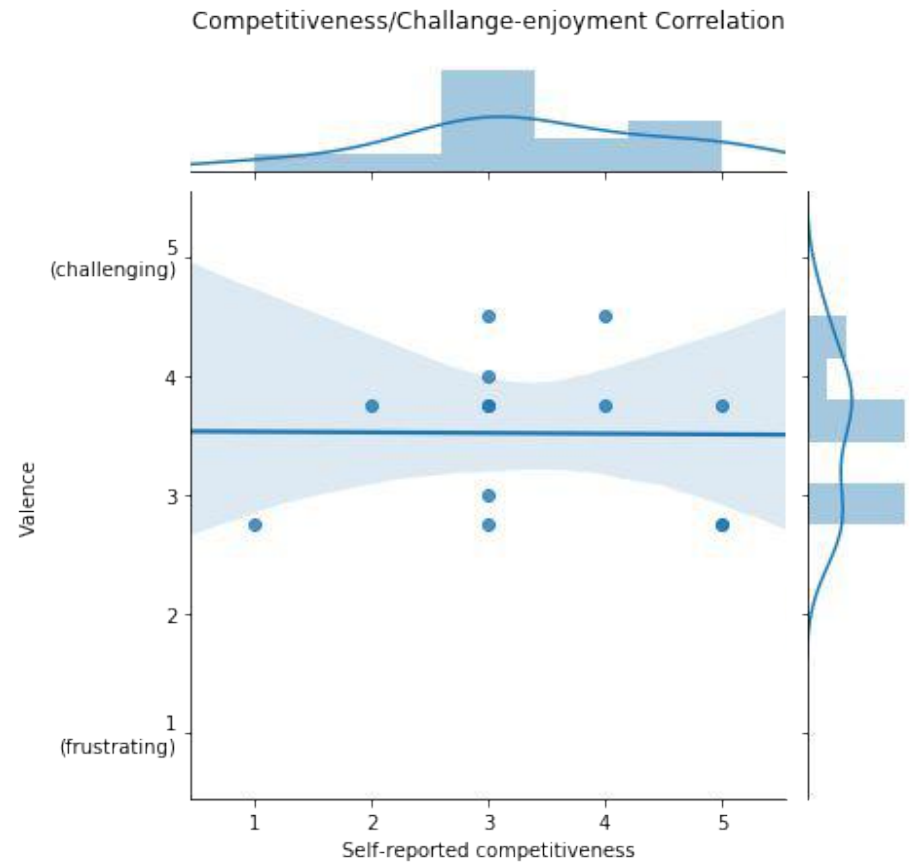
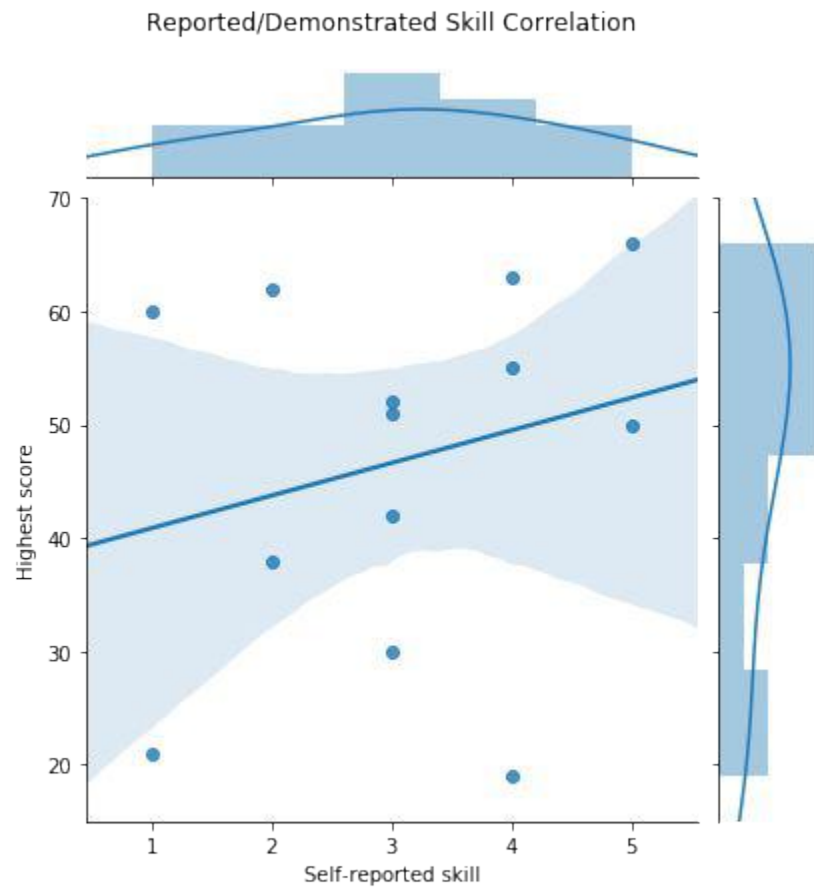
Scores across versions



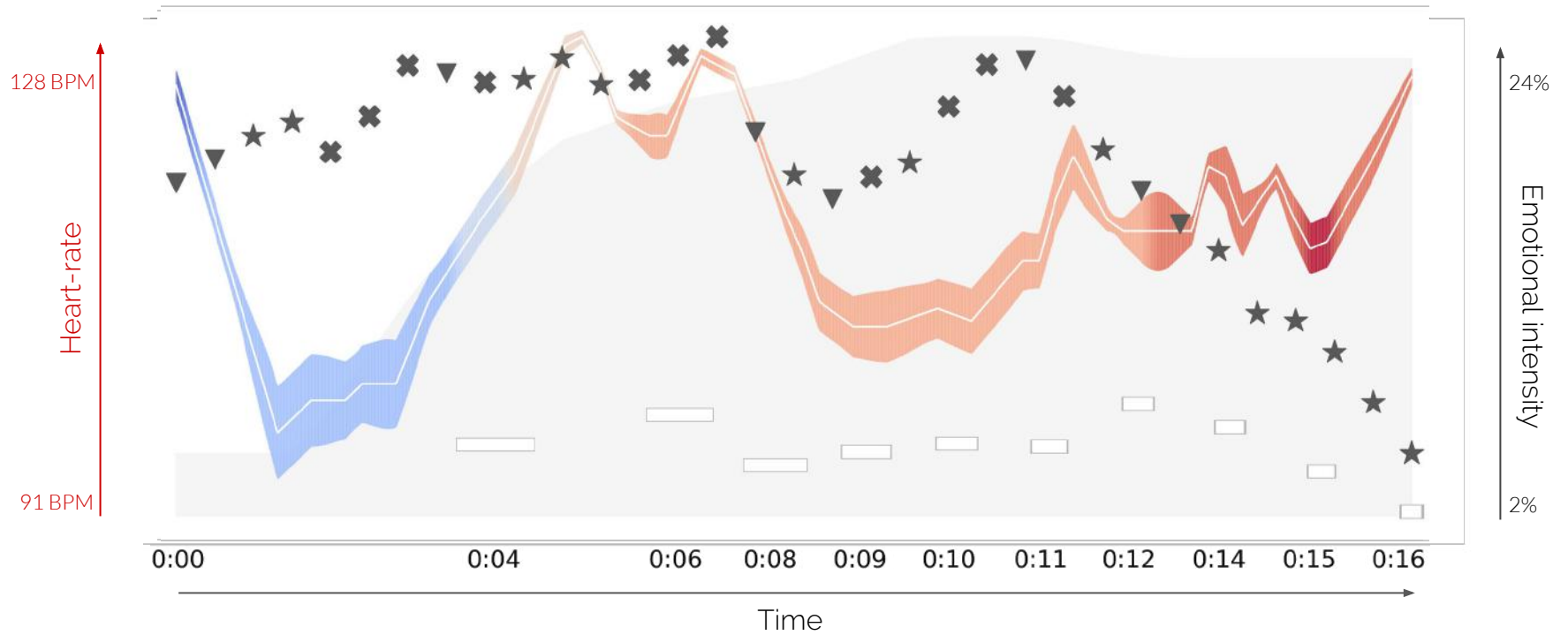
Practice makes perfect



Self-reported measures accuracy



Analysis



.66 IBI
.47 IBI

Heart-rate variability

Skin temperature
32.2°C 32.4°C

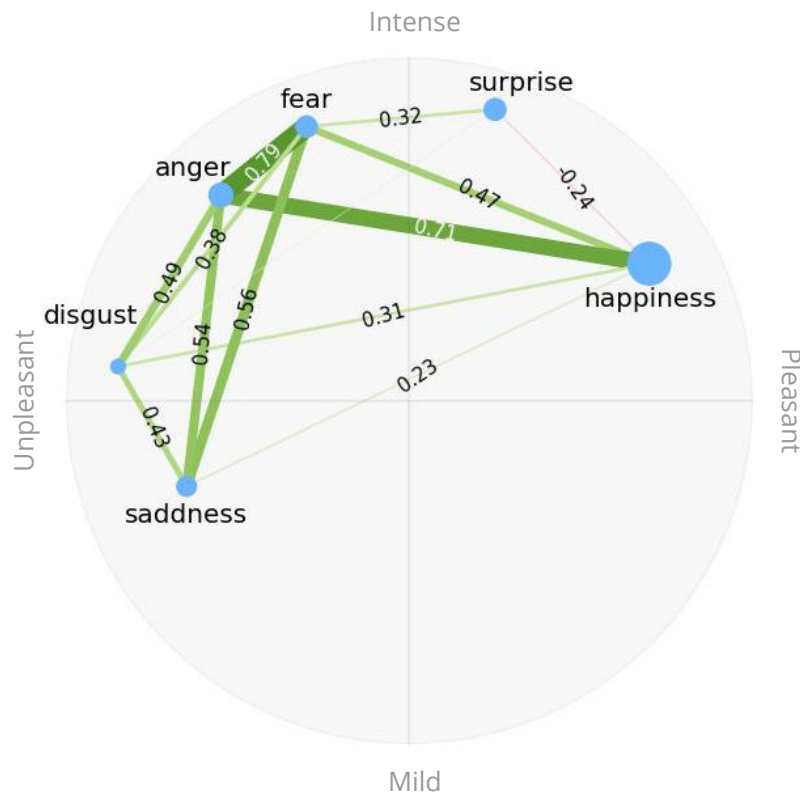
Difficulty (65 - 84)
Drop precision

Dominating factor

- ★ determined
- ✕ taken aback
- ▼ disappointed

Emotion aggregation

Linear correlation

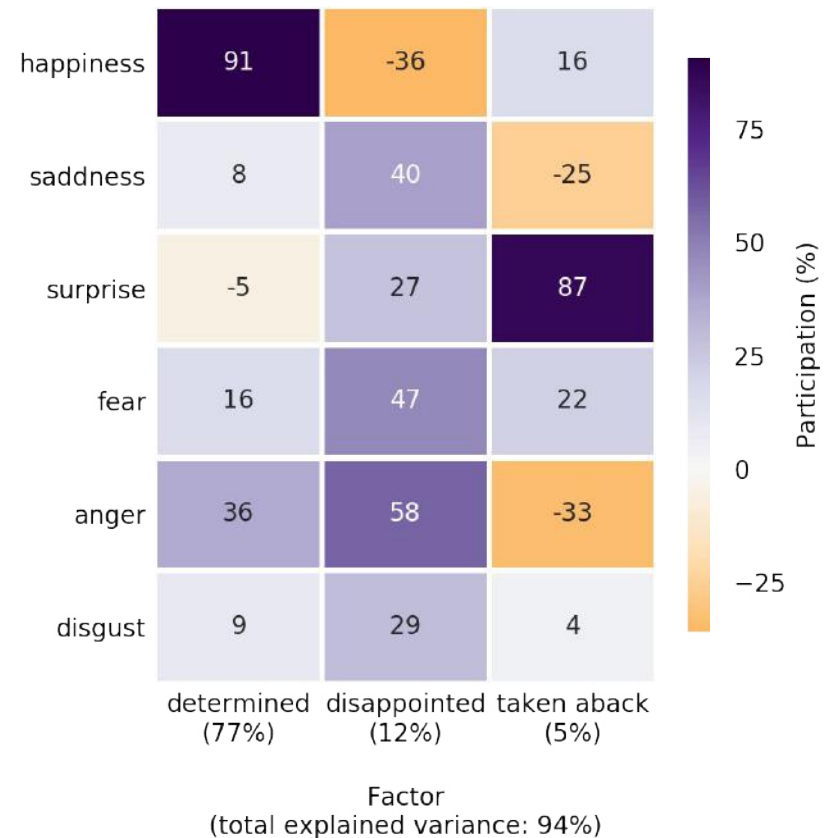


Node position: valence/arousal

Node size: average intensity

Edge width: pair-wise correlation

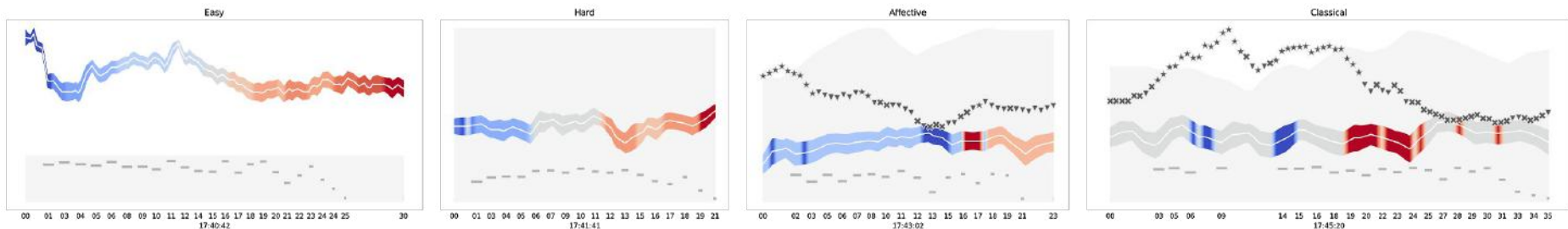
Linear dependencies



Yet more dimensions

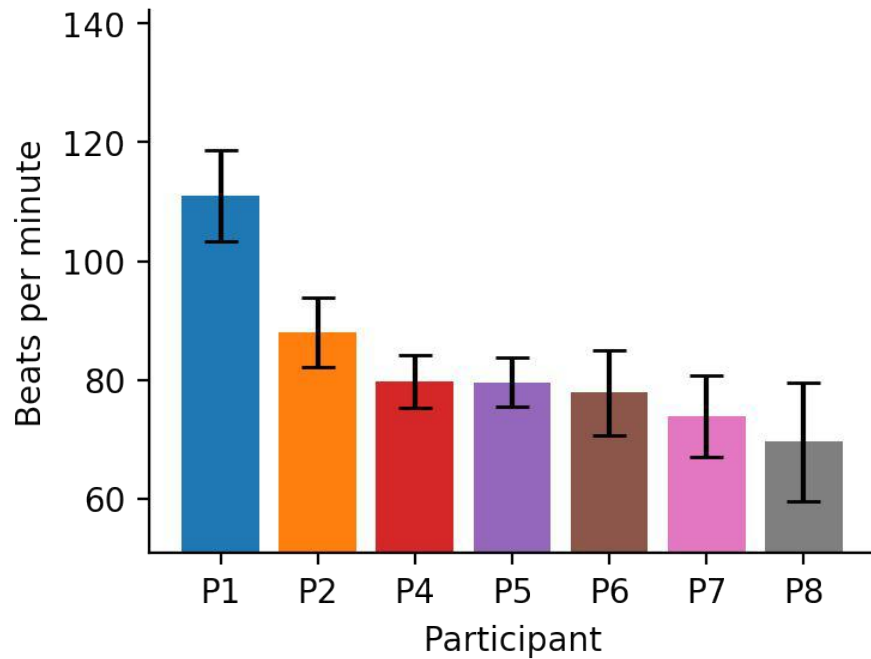
- Version
- Overall time and order (human body does not reset)
- Replay: one with max score plotted below here
- User preferences and ratings

⇒ affective signals unalignable across users, have to look at averages

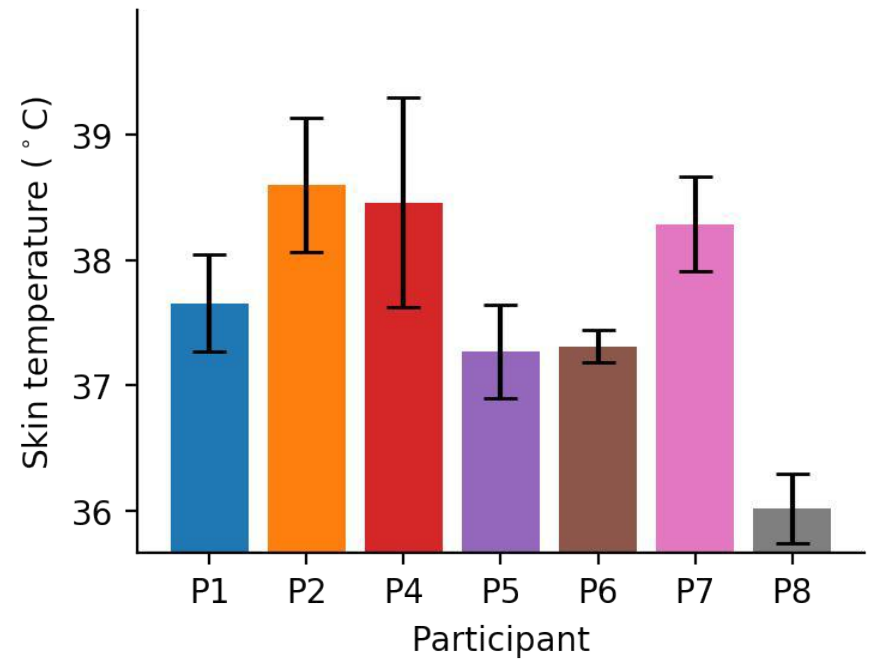


Physio variation among participants

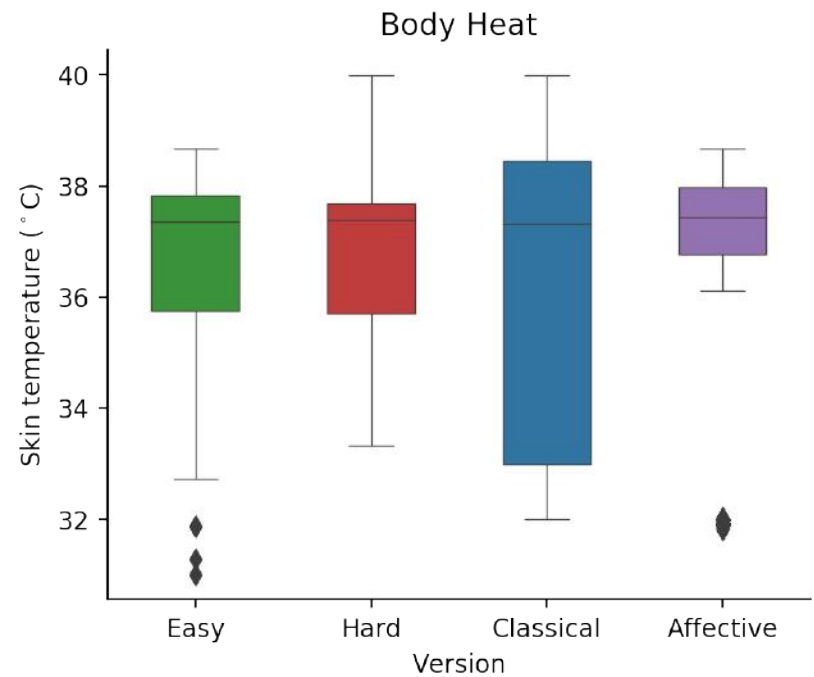
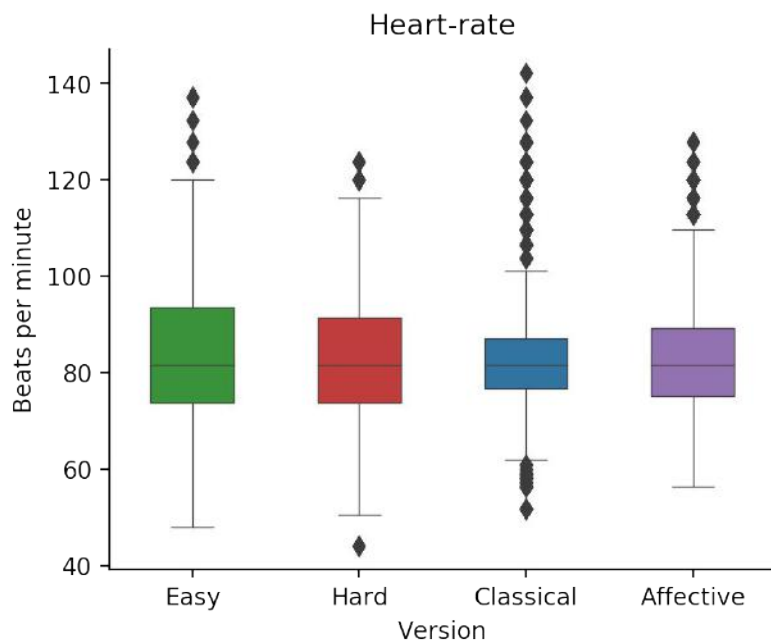
Heart-rate



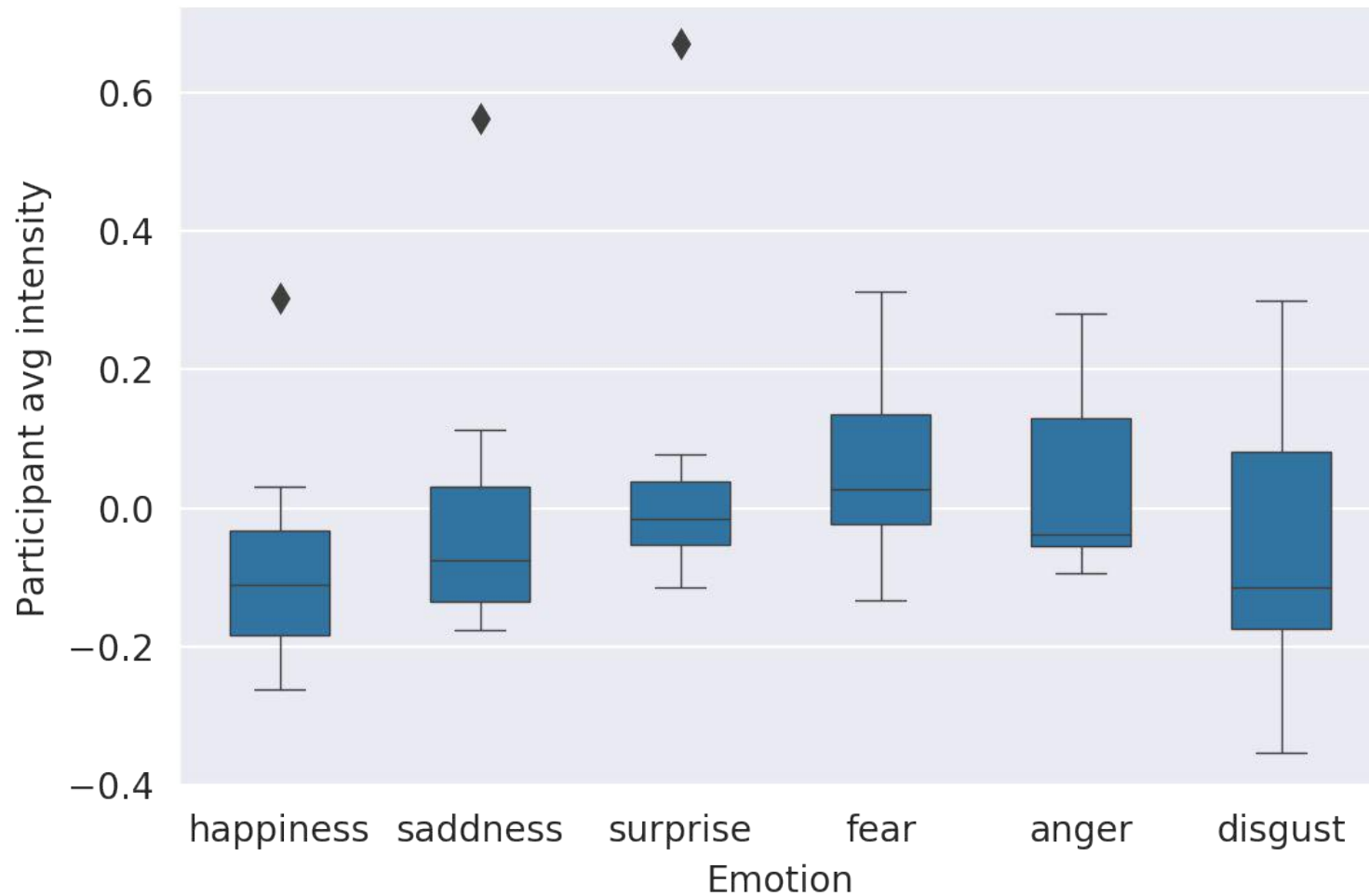
Body Heat



Physio variation across versions



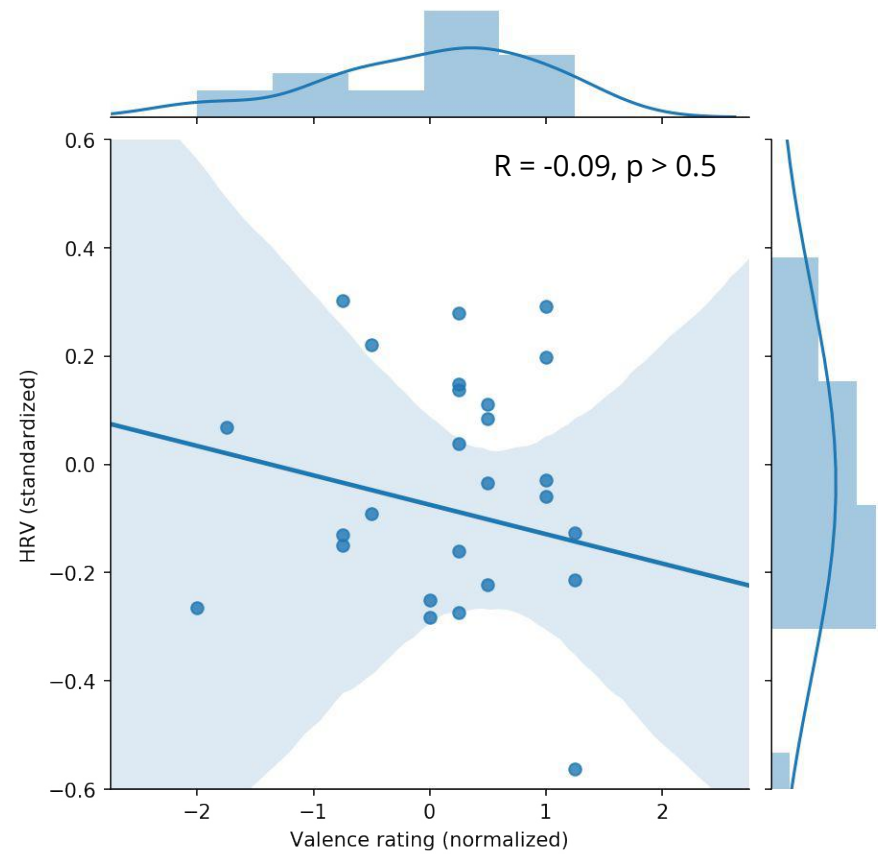
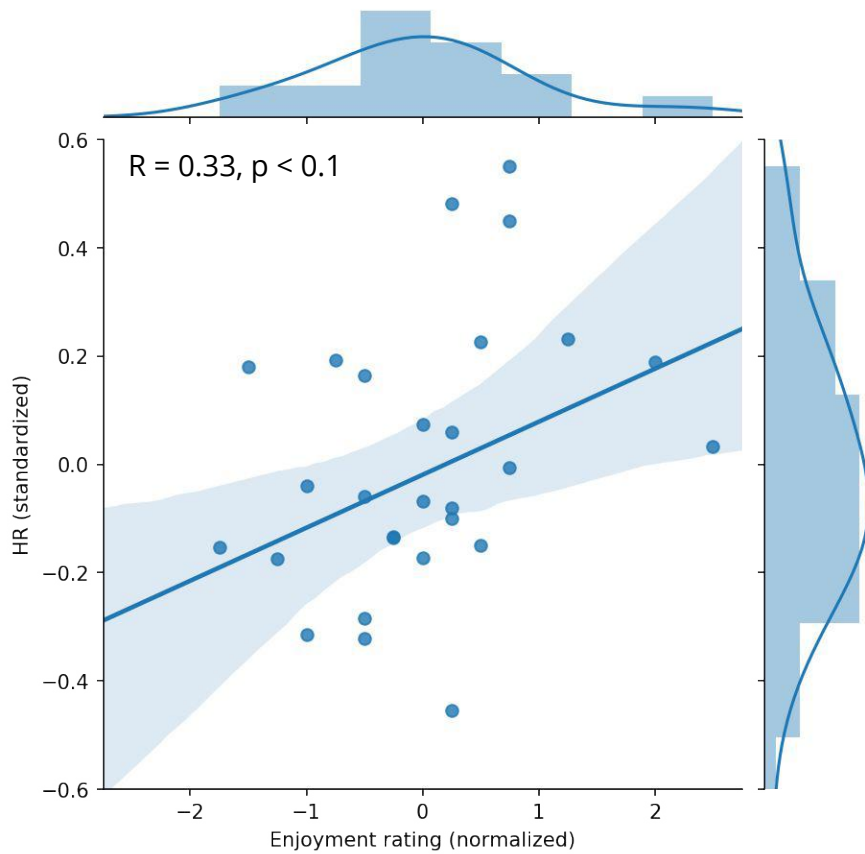
Emotional variation among participants



Emotional variation across versions



Cardiovascular/rating correlation



Conclusions

Conclusions

- Affective version at least as enjoyable as the classically-adaptive one, but more challenging, pleasant way
- Positive correlation between heart rate and game enjoyment
- Physio, emotional, and their combination is valuable
(express complementary information)

Study design lessons learned

1. Correct incentives
 - a. “Try to get a high score” => focus on absolute value & cross-version comparison
 - b. Fix: larger rewards for more difficult drops
 - c. Further isolate versions: version-wise “leaderboard”
2. Participants don’t pay attention (unless specifically instructed)
 - a. Name and color code versions
 - b. Tell from beginning what’s required of them
3. System performance is critical
 - a. Lag: makes it harder, unintentionally
 - b. Interpretation
4. Participant’s ratings are unreliable
 - a. Rate (nearly) everything the same
 - b. Not much agreement between highest rating and favorite (or play time)
5. Affective data is hard to isolate
 - a. Climb stairs before coming to experiment room => high HR
 - b. Emotions from one version/replay carry over to the next
 - c. Measurements need to be fixed in post-processing

Possible next steps

- Less intrusive: HR & HRV from video
- More affective signals:
 - galvanic skin response
 - eye gaze
 - voice tone
 - Sentiment analysis on comments
- More “real” incentives: global scoreboard, monetary reward
- Modeling of player type: risk-takers, high-scorers, fast-players etc.
- Transparency to participants
 - More enjoyment if they understand the mechanics
 - (without being able to game the system)
- Immersion:
 - sound effects, VR
 - large scale game
- Machine Learning for adaptive difficulty

Demo:

cetus.usc.edu/physio-difficulty

(Firefox only, no physio)

Division of labor

■ Dillon

■ Stefan

■ Shaoyen

■ Qinyi



Q & A

Moments from our journey...



Validating frameworks



Eliciting physio responses



Meeting ad-hoc



Running studies

Backup : Detailed Affective Dynamic Difficulty

Δ -difficulty = baseInc + physioInc + emoInc

```
int baseInc:
    if (cur_difficulty >= 85) return -3
    if (cur_difficulty >= 60) return -2
    if (cur_difficulty <= 15) return 3
    if (cur_difficulty <= 40) return 2
```

```
int physioInc:
    diff=0
    foreach physio_signal
        if (ph < baseline)    diff++
    return diff
```

Backup : Detailed Affective Dynamic Difficulty

```
int emoInc:
switch(emotion)
case HAPPINESS: return math.floor((100-cur_difficulty) / 10 )

case SADNESS:   return math.floor((100-cur_difficulty) / 10 * 0.5)

case SURPRISE:  return math.floor((100-cur_difficulty) / 10 * 0.75)

case FEAR :     return math.floor(cur_difficulty / -10 * 0.5)

case ANGER :    return math.floor(cur_difficulty / -10 * 0.75)

case DISGUST :  return math.floor(cur_difficulty / -10 * 0.8)
```