

POLITECNICO DI MILANO

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Affective Preference From Physiology in Videogames: a Lesson Learned from the TORCS Experiment

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A lesson learned...

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Long Term Research Goal

Developing games that maximize the user enjoyment

What makes a video game enjoyable? Can we measure user enjoyment? Can physiology answer these questions?



Why Physiology?



The TORCS Experiment

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TORCS (The Open Racing Car Simulator):

- Easy enough to have homogeneous game experience among users with different gaming background
- Emotionally different situations characterized by similar physical activity
- Custom logging and opponent AI (open source project)





The races

- 3 minutes races
- Easy oval track
- 1 vs 1
- 3 custom AI opponents:
 - •L (Loser)
 - C (Challenging)
 - W (Winner)





Protocol (~30 minutes)

- General questionnaire on user experience in games <u>before</u> <u>the races</u>
- 7 races against one of 3 opponents tested with <u>different</u>
 <u>ordering</u> (W C L W L C W)
- 2 Alternative Forced Choice preference questionnaires <u>after</u> <u>each pair of races</u>
- No external interaction after setup phase

Physiological Acquired signals

- Blood Volume Pulse
- Galvanic Skin Response
- Electrocardiogram
- Respiration
- Temperature

In-game measurements

- Brake/acceleration
- Steering
- Speed/position on the track
- Position and distance from the opponent car

Camera recordings

- Frontal
- Environmental



Published Papers

Preference Learning

• CIG2010, Modelling enjoyment preference from physiological responses in a car racing game

Enjoyment Classification by Ranking Preferences

• **AFFINE2010**, Enjoyment Recognition From Physiological Data in a Car Racing Game

Generalization power of physiological features

• ACII2011, Generic Physiological Features as Predictors of Player Experience

Linear and non-linear methods comparison

 Poster @ ACII2011, Learning General Preference Models from Physiological Responses in Video Games: How Complex is it?

Relationship between stimuli, questionnaires and measurements

Poster@ACII2011, The affective triad: stimuli, questionnaires and measurements

Published

- Different approaches: preference learning, classification
- Different methodology: linear, non-linear
- Cross analysis with a different dataset

Not published generic issues

- Participants
- Questionnaire
- Measurements

Participants

Participants - Recruiting

Recruiting: a common problem

 Hard to find a reasonable amount of volunteers



• 75 subjects is OK but...

	year	#part.
Picard et al.	2001	1
Wagner et al.	2005	1
Yoo et al.	2005	6
Choi & Woo	2005	1
Healey & Picard	2005	9
Liu et al.	2006	14
Rani et al.	2006	15
Zhai & Barreto	2006	32
Jones & Troen	2007	13
Leon et al.	2007	8
Liu et al.	2008	6
Katsis et al.	2008	10
Yannakakis & Hallam	2008	72
Kim & André	2008	3
Chanel et al.	2009	13

van den Broek et. al. 2009

Participants – Sample Population



Participants - Motivation

Good motivation is fundamental!

Competing for a prize

- Keeps motivation high
- The genuine concept of enjoyment was distorted
- The preference is not related to "fun" but rather to "winning"

Questionnaire

Questionnaire: Preference



Questionnaire: 2AFC

Two Alternative Forced Choice

- Between pair of races
- "Which race did you like most?"
 - 1. The last race
 - 2. The previous race

Pros

- Normalizes the personal conception of enjoyment
- Allows the comparison between subjects

Cons

- Noisy answers when the player is not sure
- Strength of the preference is unknown



Questionnaire: 4AFC

Four Alternative Forced Choice

- "Which race did you like most?"
 - 1. The last race
 - 2. The previous race
 - 3. I liked them both
 - 4. I did not like them both

Pros

• Ignore uncertain preferences \rightarrow better model estimation



Questionnaire: Additional Question

Ask the player the reason of the choice

- "What did you enjoy most/least during the last game?"
- Helps in clusterizing players
 - Inexperienced (preference evolves with learning)
 - Experienced (preference related to the performance)

Main drawback

- The flow is interrupted
- Is OK with independent pairs of games

Is physiology the right way to go?

- High intra-subject variability
- Law of initial values
- Habituation
- Invasive sensors (for now...)



Can in-game measurements do better?

- Fun is not always related to the performance
- No additional hardware required
- Transparent to the players



We tried both physiology & in-game metrics

- Same methodology
- Same experiment
- Combined analysis

And the winner was...

- In-game measurement 80% CCR
- Physiology 74% CCR

But, wait a second! Are we measuring the task?



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Task typology (a priori hypothesis)

- 2 boring races: Losing and Winning opponent
- 1 challenging race: Challenging opponent

Number of overtakes

- 100% CCR in task prediction
- 80% CCR in enjoyment prediction

80% of the players preferred the challenging race

Given the task, we can predict users' preference at 80%

Race preference is biased toward a class of race

- · We need a method to better select features that are
 - Highly correlated to the preference
 - Loosely correlated to the task

We proposed a solution

Ask me how at poster session!



In Conclusion

How far are we from a videogame that successfully exploits the affective feedback?

Which is the right methodology to follow?

Are physiological measurements and generic complex model the right way to go?

Any Answers?

Thank you for your time

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