

PLAY EXPERIENCE ENHANCEMENT USING EMOTIONAL FEEDBACK

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ABSTRACT

Abstract less than or equal to 1 page one page stating what the thesis is about highlight the contributions of the thesis

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LIST OF ABBREVIATIONS

TT	Thought Technology
GSR	Galvanic Skin Response
EMG	Electromyography
HR	Heart Rate
BVP	Blood Volume Pulse
EDA	Electrodermal Activity
HCI	Human Computer Interaction
AV	Arousal/Valence
NPC	Non-Player Character
Mod	Modification
CSV	Camma Separated Values

CHAPTER 1

INTRODUCTION

Since computers are playing a significant role in our daily life, the need for a more friendly and natural communication interface between human and computer has continuously increased. Making computers capable of perceiving the situation in terms of most human specific factors and responding dependent to this perception is of major steps to acquire this goal. If computers could recognize the situation the same way as human does, they would be much more natural to communicate. Emotions are of important and mysterious human attributes that have a great effect on people's day to day behavior. Researchs from neuroscience, psychology, and cognitive science, suggests that emotion plays critical roles in rational and intelligent behavior [16]. Apparently, emotion interacts with thinking in ways that are nonobvious but important for intelligent functioning [16]. Scientists have amassed evidence that emotional skills are a basic component of intelligence, especially for learning preferences and adapting to what is important [13, 8] People used to express their emotions through facial expressions, body movement, gestures and tone of voice and expect others understand and answer to their affective state. But sometimes there is a distinction between inner emotional experiences and the outward emotional expressions [15]. Some emotions can be hard to recognise by humans, and inner emotional experiences may not be expressed outwardly [9]. Recent extensive investigations of physiological signals for emotion detection have been providing encouraging results where affective states are directly related to change in inner bodily signals [9]. However whether we can use physiological patterns to recognise distinct emotions is still a question [16, 4].

Although the study of affective computing has increased considerably during the last years, few have applied their research to play technologies [21]. Emotional component of human computer interaction in video games is surprisingly important. Game players frequently turn to the console in their search for an emotional experience [18]. There are numerous benefits such technology could bring video game experience, like: The ability to generate game content dynamically with respect to the affective state of the player, the ability to communicate the affective state of the game player to third parties and adoption of new game mechanics based on the affective state of the player [21]. This work concentrates on developing a real-time emotion recognition system for play technologies which can quantify player instant emotional state during a play experience The rest of the paper is organized as follows: in Section 2 we outline different emotion recognition theories with an overview of physiology sensors. In Section 3 we demonstrate some implementation details of the system. We then describe the experimental setup in Section 4 before giving

our results in Section 5. Finally, we give conclusions in Section 6.

CHAPTER 2

HUMAN AND PLAY TECHNOLOGIES

Using emotional responses to increase the level of users interaction with a real-time play technology requires an effective technique to identifying specific emotion states within an emotional space. Major existing emotion models in the psychology literature includes: basic emotion theory [6, 7], dimensional emotion theory [10, 19] and models from appraisal theory (e.g., [17]) [23]

Basic emotion theory identifies anger, disgust, fear, happiness, sadness, and surprise [14] as the concise set of primary emotions. These are actually the least six universal categories researchers agreed upon [22]. It also claims these primary emotions are distinguishable from each other and other affective phenomena [5]. On the other hand dimensional emotion theory argues that all emotional states reside in a two-dimensional space, defined by arousal and valence.

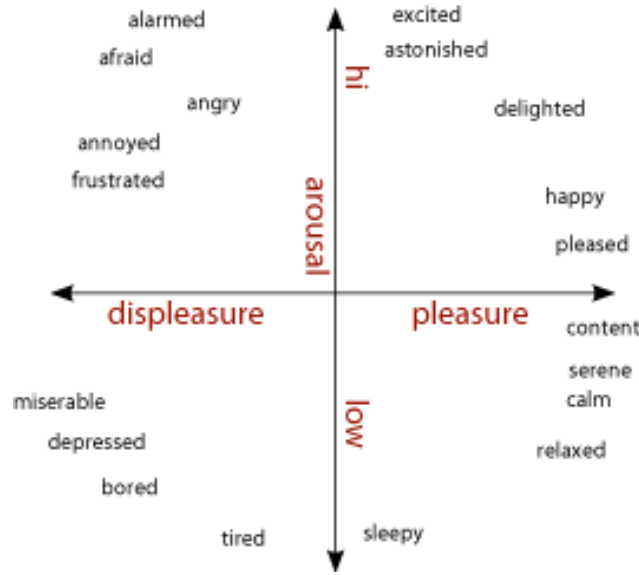
While there are various opinions on identifying emotional states, classification into discrete emotions [5], or locating emotions along multiple axes [20, 10], both had limited success in using physiology to identify emotional states [3].

Lang used a 2-D space defined by arousal and valence (pleasure) (AV space) to classify emotions [10]. Valence can be described as a subjective feeling of pleasantness or unpleasantness while arousal is the subjective state feeling activated or deactivated [1]. Using an arousal-valence space to create the Affect Grid, Russell believed that arousal and valence are cognitive dimensions of individual emotion states. Affect is a broad definition that includes feelings, moods, sentiments etc. and is commonly used to define the concept of emotion [15]. Russell's circumplex model has two "axes" that might be labeled as displeasure/pleasure (horizontal axis) and low/high arousal (vertical axis) It is not easy to map affective states into distinctive emotional states, However these models can provide a mapping between predefined states and the level of arousal and valence [22], Figure 2.1.

Both mentioned models for identifying emotions convey some practical issues in emotion measurement. In a HCI context, the stimuli for potential emotions may vary less than in human-human interaction (e.g., participant verbal expressions and body language) [23] and also the combination of evoked emotions [14]. However with help of physiological signals and the fuzzy logic in the model we are going to use, such issues with our dimensional emotion models would be minimized. Though it is anticipated to observe different range of evoked emotions while interacting with play technologies compared to interacting with other humans in

¹Photo credit: <http://imagine-it.org/gamessurvey/>

Figure 2.1: Russell’s circumplex model with two axes of arousal and valence ¹.



daily life. [23]. However our dimensional emotion models suffers some other problems. One problems is that arousal and valence are not independent and one can impact the other [12]. Continuously capturing emotional experiences in this applied setting is of its other halmarks. Subjective measures based on dimensional emotion theory, such as the Affect Grid [20] and the Self-Assessment Manikin [2], allow for quick assessments of user emotional experiences but they may aggregate responses over the course of many events [23]. This work uses Mandryk et al. version of AV space [12].

2.1 Affect and Emotion

2.2 Measuring Affect

2.3 Affect and Play Technologies

2.4 Real Time Game Adaptation

Playing video games as a kind of entertainment would help people to have new internal experiences. The virtual world of video games let adults to play as new rolls and enjoy filling their heads with new thoughts and emotions. Some people value the sensation

Games are opportunities for development and design of environemnts therefore the player can interactively experience various emotions and mental conditions. This interactive experience in contrast to movies and other major types of digital entertainment is what makes them exceptional

2.5 Play Technologies and Human Body

2.6 Adaptive Play Technologies

CHAPTER 3

CUSTOMIZING PLAY EXPERIENCE IN REAL-TIME

3.1 Adaptive Game Design

3.1.1 Player

3.1.2 NPCs

3.1.3 Environment

3.2 Physiological Signals

3.2.1 Heart Rate

3.2.2 Galvanic Skin Response

3.2.3 Facial Electromyography

3.3 Real-Time Affect Engine

3.3.1 Fuzzy Logic for Space Transformation

3.3.2 Physiological Signals to Arousal and Valence

CHAPTER 4

IMPLEMENTATION AND INTEGRATION

4.1 Recognising Emotion

Heart rate (HR), blood pressure, respiration, electrodermal activity (EDA) and galvanic skin response (GSR), as well as facial EMG (Electromyography) are of physiological variables correlated with various emotions most. Interpreting physiological measures into emotion state can be difficult, due to noisy and inaccurate signals, however recent on-going studies in this area by Mandryk and Atkins [12] presented a method to continuously identifying emotional states of the user while playing a computer game. Using the dimensional emotion model and the fuzzy logic, based on a set of physiological measures, in its first phase, their fuzzy model transforms GSR, HR, facial EMG (for frowning and smiling) into arousal and valence variables. In the second phase another fuzzy logic model is used to transform arousal and valence variables into five basic emotion states including: boredom, challenge, excitement, frustration and fun. Their study successfully revealed self-reported emotion states for fun, boredom and excitement are following the trends generated by their fuzzy transformation. The advantage of continuously and quantitatively assessing user's emotional state during an entire play by their fuzzy logic model is what makes their model perfect to be incorporated with real-time play technologies. Therefore exposing user's emotional state as a new class of unconscious inputs to the play technology.

4.2 Affect Engine

This project has developed a real-time emotion detection system which can continuously detect and recognise user's emotional state. The system uses Blood Volume Pulse (BVP), Galvanic Skin Response (GSR) and Electromyography (EMG; for frowning and smiling), to classify human affective states in 2-dimensional valence/arousal space, Figure 2. The system has three modules, Figure 4: The Blood Volume Pulse (BVP) signal is a relative measure of the amount of blood flowing in a vessel. From BVP we calculated heart rate and heart rate variability. The heart rate is known to reflect emotional activity and has been used to differentiate between both negative and positive emotions as well as different arousal levels [11] The Galvanic Skin Response (GSR) sensor to measure the skin's conductance (between two electrodes and is a function of sweat gland activity and the skin's pore size). As a person becomes more or less stressed, the skin's conductance

increases or decreases proportionally [15].

The sensor module includes a Thought Technology ProComp Infinity encoder [11] connected to PC with a USB cable. SensorLib is the basic module which receives and filters raw physiological inputs. Then filtered signals are fuzzified by the use of 22 fuzzy rules in the first phase of transformation. Then generated arousal and valence values are transformed into emotion values using another 67 fuzzy rules in the second pass [12]. Applications such as games can easily integrate the system where emotion recognition can offer adaptive control to maintain user interest and engagement. Once connected via sensors to the emotion recognition system, the affective state of the user can be captured continuously and in real-time and it can be monitored on a displayed 2-dimensional graph of valence and arousal, Figure 1.

4.3 Inegration With Valve Source Engine

4.3.1 Level Design

4.3.2 The Director

CHAPTER 5

EXPERIMENTATION

5 to 10 pages talk about the experimentation adequacy, efficiency, productiveness, effectiveness (choose your criteria, state them clearly and justify them) be careful that you are using a fair measure, and that you are actually measuring what you claim to be measuring if comparing with previous techniques those techniques must be described in Chapter 2 be honest in evaluation admit weaknesses

CHAPTER 6

DISCUSSION

5 to 10 pages talk for the discussion State what you've done and what you've found Summarize contributions (achievements and impact) Outline open issues/directions for future work

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APPENDIX A

FUZZY FUNCTIONS

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