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Stress psychology and its analysis from artificial intelligence

Psicología del Estrés y su análisis desde la Inteligencia artificial

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Abstract

A review of emotions from cognitive brain processes is presented, taking into account some experimental neurobiological studies. Sensory systems, physiological reactions of the central nervous system, and emotional stimuli are evaluated. It is evident that subjective expressions and physical reactions of emotions are conditioned to social patterns, family customs and mainly to moral stereotypes of people. Artificial intelligence is used, with state vector machine for the pro-cessing of human biological signals. Among the observed characteristics, self-sanctioning emo-tions and those of empathy with the other prevail. Finally it was possible to recognize that the restriction of emotions in people can cause psychological behaviors that harm their relationship with the environment. It produces antisocial behaviors, lack of understanding of the feelings of others, breach of the rules, among other activities and social standards.

key words: artificial intelligence, human emotions, medical application, occupational health

Resumen

Se presenta una revisión de las emociones de los procesos cognitivos del cerebro, teniendo en cuenta algunos estudios neurobiológicos experimentales. Se evalúan los sistemas sensoriales, las reacciones fisiológicas del sistema nervioso central y los estímulos emocionales. Es evidente que las expresiones subjetivas y las reacciones físicas de las emociones están condicionadas a los patrones sociales, las costumbres familiares y principalmente a los estereotipos morales de las personas. Se utiliza inteligencia artificial, con máquinas de vectores de estado para el procesamiento de señales biológicas humanas. Entre las características observadas, prevalecen las emociones autoritativas y las de empatía con el otro. Finalmente, fue posible reconocer que la restricción de las emociones en las personas puede causar comportamientos psicológicos que dañan su relación con el medio ambiente. Produce comportamientos antisociales, falta de comprensión de los sentimientos de los demás, incumplimiento de las normas, entre otras actividades y estándares sociales.

Palabras clave: inteligencia artificial, emociones humanas, aplicaciones médicas, salud ocupacional.

1. Introduction

Human emotional states are a condition of the sub conscious (Ibañez, 2009), and these emotions characterize people's behavior and condition them to their personali-ties, making them affective, or non-affective, expressive or non-expressive people, and thus defining the multiple personalities and the different characters. These par-

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ticularities are characteristic of the person, and depending on other social, cultural, family factors, among others, may condition them to their social or labor performance (Palacios, 2017).

Human emotions were studied for the first time in 1872 (Fernández-Berrocal, 2009), when the scientist Charles Darwin focused his studies on the analysis of emo-tions, managing to describe them as a non-exclusive condition of humans. In his research the scientist managed to define two important aspects of emotions; They are innate and they are universal. With this he confirmed that emotions do not de-pend on any specific geographical area, nor do they depend on race, or any other external aspect, but on human nature itself. The scientist stressed that the face is one of the greatest reflectors of emotions, and that this determines the reflection of the personality and individual characteristics of people (Montagu, 1959). Other authors (Harlow & Harlow, 1962), (Reite & Short, 1981) also affirm that crying and that the smile are exclusive to humans and that they are not present in the emotional mani-festations of other species.

Social and cultural conditions are decisive for the manifestation of emotions, and with that human expressions and attitudes are added, (Palacios, 2017), (Surrallés, 2005), (Bourdin, 2016), giving rise to people's behavior. Emotions can be intense or mild, and depending on how often they occur they can cause significant changes in behavior, which can significantly affect people's mental health, creating possible addictions that trigger stress (Moure, 2011).

The manifestations of stress are presented according to the self-demands of people or the demands of the environment, which may well be labor or social (Naranjo M., 2009). Modernism and new ways of focusing the quality of life influence people's expectations, which lead them to seek new social goals, which to be reached affect the emotional state of people and trigger health problems (Naranjo M., 2009), (Martínez & Díaz, 2007).

Stress stimuli are those situations that trigger stress, which can produce important changes in the body, activating the hypothalamic–pituitary–adrenal axis and the vegetative nervous system, (Sánchez & Vaquero, 2008), (González & Landero, 2006), this activation produces the release of hormones, which may well cause negative or positive aspects in people (Berrío & Mazo, 2011), (Maslach & Pines, 1977). The manifestations of stress are presented through emotions, which may be associated with the adaptation to the change that the person may have be-fore different stressors. (Berrío & Mazo, 2011), (Rubio, Guerrero, & Castro, 2003).

The adrenal gland is responsible for regulating many aspects of everyday life, including carbohydrate regulation, immune response, blood pressure, electrolytes and sexual characteristics. Thus, the adrenal cortex produces glucocorticoids, mineralo-corticoids and antrogens, which add to the body's daily regulatory process (Domínguez & Olvera, 2006), (Lavalle-González, y otros, 2011). In addition, the hypothalamic-pituitary-adrenal axis has an important participation in the stress res-ponse, achieving cortisol production.

The study of emotions is linked to stress, and their effects are involved at the same time with the mental and physical health of people (Lavalle-González, y otros, 2011), since situations of constant stress can trigger anxiety, depression, behavior disorders and situations personnel that can affect the relationship with the environment, caus-ing family, social and work problems (Naranjo M., 2009).

Companies and organizations are increasingly demanding, and increasingly expect greater results from their employees (Flórez, 2014), causing a significant level of stress on workers, trying to improve their profiles to give more effective and effective responses in their workplace However, this burden of occupations, performance, rapid responses, can cause occupational diseases, which will be explicit in the level of stress in people, and can be seen in problems of high blood pressure, skin problems, gastro-intestinal problems. (Moure, 2011), (Lavalle-González, y otros, 2011)- (Slipak, 1991). This type of health problems will not only affect the workers, but will also affect the productive process of the companies, product of labor absenteeism, the rotation of qualified

personnel, among others circumstances that will cause problems in the organizations, product of collective stress (Antón, 2013), (Naranjo M., 2009), (Sánchez & Maldonado, 2003).

This paper presents an analysis of the methods of detection of emotions by analysis of facial physiognomy and by voice analysis, both using artificial intelligence, with state vector machine. The developed methods allow to detect emotions with different human variables: face and voice spectrum. Artificial intelligence allows an effective and rapid detection of emotions, and each of them offers advantages for an effective classification of stress in people.

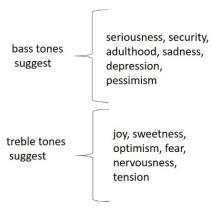
This work consists of four sections: In the first, a brief introduction to the process of human emotions has been described, in the second, theoretical aspects that underlie this research will be addressed, in the third, the methodology will be considered, and in the fourth, the results will be presented. Finally, the conclusions will be described.

2. Methodology

The organs of the respiratory and digestive systems, which are controlled by the central nervous system, are involved in the projection process of the voice (Sánchez & Pérez, 2007). The excitation generated in the vocal cords is propagated through the pharynx, the oral cavity and the nasal cavity (Sánchez & Pérez, 2007), (Vargas., 2003). These cavities determine the acoustic characteristics of the voice (Cárdenas., Ceballos., Shang-Hsueh., Pavez., & Terrisse., 2010).

Duque and Morales (Sánchez & Pérez, 2007) affirm that there is a close relationship between the characteristics of the voice and the emotional states of people. Some investigations (Roldan., 1998), (Núñez, Cortéz, & Suárez., 2006), have shown that various aspects of people's physical and emotional state, including age, sex, intelligence level, physical appearance and personality, can be identified only with the voice.

Figure 1Relationship of voice to emotional states



Source: Authors

Thanks to recent studies (J.González., T.Cervera., & J.Miralles., 2002), it has been possible to ensure that some of the components of the voice are characteristic for expressing emotions, among which we can mention:

- The fundamental frequency.
- The duration time.
- The quality of the voice.

Of the multiple methods that exist for the recognition of emotions (Acevedo & Luca, 2000), (Valderrama & Ulloa, 2011), (Ibañez, 2009) the one proposed in (Ibañez, 2009), presents more than a good performance, its accuracy analysis considering multiple stages of training and evaluation for 7 basic emotions such as: anger, contempt, disgust, fear, happiness, neutral, sadness.

3. Results

Stress can be caused by different causes, which are called stress stimuli. These stimuli can be presented in different areas of daily life, whether in work, academic, family or social situations in general (Karasek, 1979). Most of the manifestations of stress occur in work environments, this because of the demands of globalization, new technologies, new requirements to respond to spaces of business competence (Karasek, 1979)- (Peiró, Estrés Laboral, Liderazgo y Salud Organizacional, 2008). Social changes, socio-political conditions and new impacts on the labor market greatly influence the proliferation of stress situations in people (Peiró & Rodríguez, Estrés laboral, liderazgo y salud organizacional., 2008).

Some authors (Trianes, y otros, 2009), have developed different mechanisms to measure stress, using specific questionnaires with 25 items; referred to health, education, family. The results showed that emotions are significant in the manifestations of stress, and that this in turn significantly affects the health of people and of more diseases (Mcade, Lumley, & Casey, 2001).

Suárez and Rosales (Suárez & Rosales, Algoritmo para la evaluación del modelo dinámico del estrés, 2019) state that stress can be represented from a set of mathematical variables that relate stress and health status, and in these it is possible to determine that stressors and the frequency or permanence of they are triggers of stress, and that this condition is in turn triggers of serious health problems, so it is very important to pay attention to stressors to avoid their continuous presence.

Knowing the possible consequences of stress, electronic systems have been developed for analysis (García, Garzón, & Camargo, 2011), where physiological variables have been considered for their detection, including galvanic impedance of the skin, blood pressure, heart rate, blood volume, between others. The study revealed that non-invasive methods are the most appropriate for measurement, and are also simpler for data processing.

Other authors (Acevedo & Luca, 2000) have developed a digital system that in addition to considering the variables of muscle tension, skin conductance and respiration, has taken into account the inclusion of a stress test and cortisol and prolactin values. In this study, the importance of stress in people has been assessed and it has been concluded that the Stress-Disease scale must be validated by experts from the medical area, who demonstrate that the System does not present a total success in the determination of stress, but the results are relative.

Artificial intelligence in medical processes contributes significantly to signal processing, data processing, automated learning, among others (Martínez & Goddard, 2001), (Valderrama & Ulloa, 2011). These intelligent algorithms contribute to formulating more appropriate solutions, more optimal processing times and a smaller margin of error than in classical processes (Alfonso, Muñoz, López, & Caicedo, 2007).

The Support Vector Machine (SVM) is widely used for the classification of variables, which have been discriminated against for a specific purpose. This type of method constructs a hyperplane to delimit the space between both categories. Some authors (Mosquera, Castrillón, & Parra, 2018), have developed support vector machines for the psychosocial verification of teachers, these studies showed that the use of SVM is useful for the treatment of human physiological information, achieving efficiencies in 96.3%.

A comparative process of the detection of emotions in people is carried out, in order to estimate the stress present in them. The facial image characterization method was used, using a vector support machine, and evaluating the process with a population of 343 images with different emotions, which are linked to certain expressions such as anger, disgust, fear, joy, sadness, surprise and finally the absence of emotions. For this process, The Histogram of Oriented Gradients (HOG) and K-Nearest Neighbors (KNN) are used because it yields a precision value of 71.42%, however, with the HOG and SVM it yields a precision value of 75% with the KDEF base. The HOG algorithm with the KNN and with the MUG base yields 65.71%. Figure 2, describes the process performed for facial image classification.

Face Image Classification Process Database Face Feature Learning and Detection classification Extraction Trainina Emotion Face Video **Feature** Clasification **Estimation** Detection images Extraction Evaluation Source: Authors

Figure 2

The diagram shown in Figure 2, provides information on the process performed for the classification of emotions. It has been considered a set of images that represent different emotions, and these are processed by an image normalization algorithm, to then be worked with a HOG that allows the training process, finally the SVM classifier perform the last stage together with an algorithm KNN. The System showed a success rate of 96.13%.

On the other hand, an analysis of the emotions is carried out from the voice, for which we worked with the frequency of the voice spectrum. For this, the central coefficients in Mel frequencies are used, which focuses on human auditory perception. In the process, filtering is done to discard information that is not relevant, such as background noise, external noise, among others. This type of process was proposed by Davis and Mermelstein (Davis & Mermelteins, 1980) in the last century, since then they are the fundamental basis for scientific argumentation in voice recognition.

Figure 3, shows the human voice classification process. For this, 535 voice spectra were used, representing different emotions. For this, a reading process was used by the evaluated person, in order to perceive and discriminate the voice of the same.

Figure 3. Voice Processing Diagram SETS FOR LABELED INCREASED CLUSTERIZATION TRAINING - DATABASE TRAINING AND DATABASE VALIDATION Emotion 1 Emotion 2 Emotion 3 Fmotion 6 Emotion 5 Fmotion 7 EXTRACTION OF STANDARDIZATION STORAGE BUFFER LABELING ACCORDING VECTOR OF OF FEATURES 20 OVERL, 10 TO CHARACTERISTICS **FEATURES** VECTOR Fmotion 4 *** *** 10 ACTORS DEFINE NETWORK DEFINE NETWORK TRAIN NETWORK OPTIONS TEST AUDIO **IDENTIFIED EMOTION** TRAIN NETWORK **EVALUATION** DATABASE USER INTERFACE

Voice frequency analysis is one of the fundamental aspects in the determination of human emotions (Antón, 2013), (Cárdenas., Ceballos., Shang-Hsueh., Pavez., & Terrisse., 2010), (Giraldo. & Quintero., 2010). The central coefficients in Mel frequencies, allow to represent the voice spectrum and analyze it using the Fourier Transform by windows, which opens 30ms spaces without overlapping. A bi-directional dependent reading network is used for natural language recognition. The results showed that the percentage of success with voice recognition was 79.6%.

Source: Authors

It was evident that both methods are appropriate for the classification of emotions, these being fundamental for the determination of stress in people. However, face classification is 16.53% more effective, making this method provide a greater contribution to the stress validation process.

Once the process of classification of variables has been achieved, it is possible to suggest that a Stress Classification System will require a third variable that allows the equilibrium of appreciation of the results and the stress classification process to be more optimal. Stress can be classified taking into account some aspects of people, these in turn are innate, are not modifiable by choice.

4. Conclusions

Once the methodology for characterizing emotions is evaluated, using face classification and voice classification, both taking into account vector support machine, it is possible to mention the following conclusions:

Emotions are innate characteristics of people, which allow them to express themselves according to the conceptions of each individual, the same situation is not appreciable in the same way by two people, since the perception is conditioned to the individual particularities, which can be the culture, family formation, social conditions and the environment in general.

Emotions allow people to identify stress situations in people, which can manifest themselves at different levels and that can affect people's health status, their attitudes and their work and social performance.

Stress stimuli trigger stress situations and as long as these occur regularly and frequently, stress will continue to have an effect on people's health status, which can trigger severe illnesses, social attitudes that can cause behavioral problems, work problems and even death.

The classification of emotions by face, offers more optimal results for the determination of stress in people, the percentage of success results in 96.13% which could be improved if a broader database is considered with people not trained for it, who can Express your different emotions spontaneously. In addition, convergence with another additional method could be considered, for the limitation of the region of analysis for each figure, so that the non-relevant areas can be discriminated and the process made more effective.

The classification of emotions by voice, is known as a fairly effective method, however, it did not yield optimal results to ensure emotional state, creating spaces of uncertainty, which is not convenient as a single analysis in the classification of emotions and in consequence of stress

The results found reveal that it is essential that some additional variable be included for the assessment of emotions in people, so that it is possible to triangulate the information and achieve more optimal results.

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