

# XXIII



Universidad Nacional Jorge Basadre Grohmann  
Facultad de Ingeniería  
Escuela Profesional de Ingeniería en Informática y Sistemas



Proyecto  
**Concytec**  
Banco Mundial

# CONGRESO INTERNACIONAL EN INFORMÁTICA Y SISTEMAS

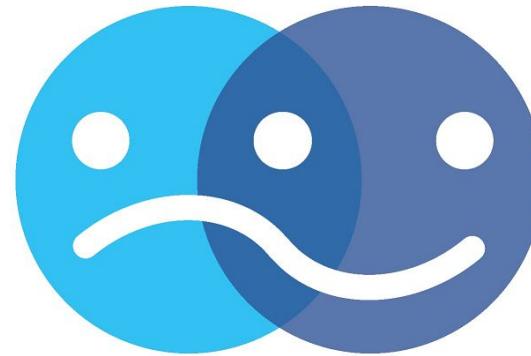


**KUSISQA:** Ayudando a  
Regular Nuestras  
Emociones en los Procesos  
de Enseñanza y Aprendizaje

**Dr (c). Wilber Ramos Lovón**  
Coordinador Técnico del proyecto

Del 07 al 11 de noviembre de 2022

Tacna - Perú



# **KUSISQA: Ayudando a Regular Nuestras Emociones en los Procesos de Enseñanza y Aprendizaje**

*Contrato N° 014-2019-FONDECYT-BM-INC.INV*

**Dr (c). Wilber Ramos Lovón**

Coordinador Técnico del proyecto

**Universidad Nacional de San Agustín de Arequipa**

Escuela Profesional de Ciencia de la Computación

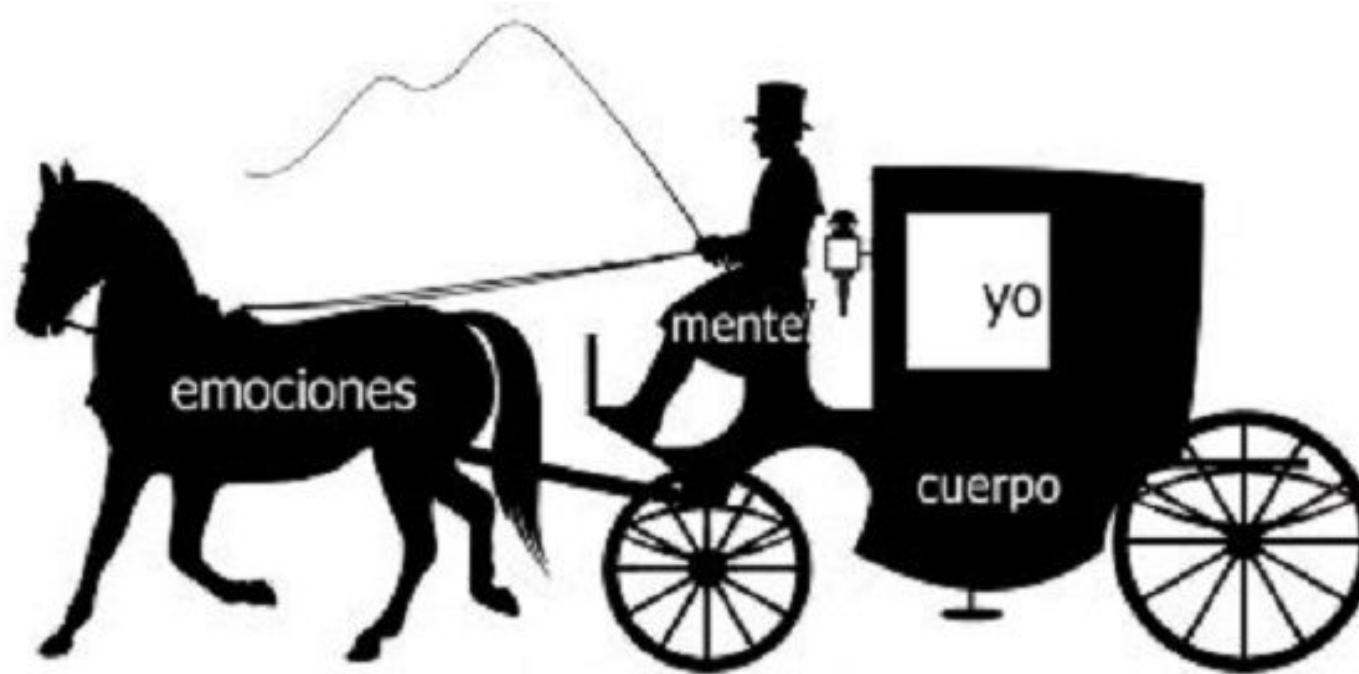
Noviembre de 2022

# Las emociones humanas



Fuente: [Gearstd/Shutterstock](#)

# La metáfora del carroaje



Fuente: [Haiki](#)

El **cuerpo** humano es el **carruaje**; el **yo**, el hombre que lo **conduce**; el **pensamiento** son las **riendas**, y los **sentimientos/emociones**, los **caballos**. *Platón*

# Las emociones humanas en la toma de decisiones



Decisiones de índole personal

Decisiones de índole organizacional/corporativo

# Medición de las emociones humanas

## Autoinforme



## Cuestionario



## Entrevista

Técnicas tradicionales

# Medición de las emociones humanas

Técnicas tradicionales

## Autoinforme

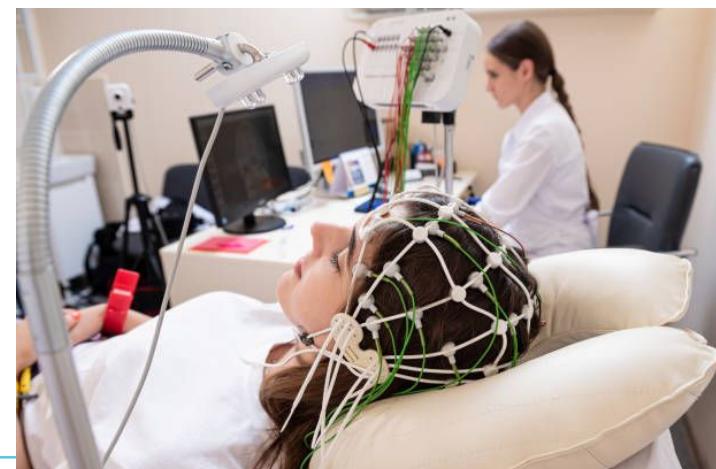
## Cuestionario



## Entrevista

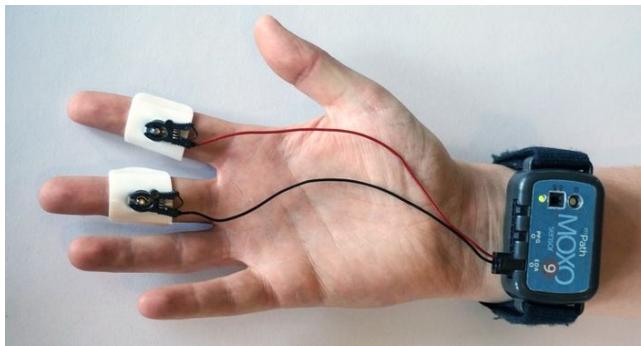


Neuroimagen

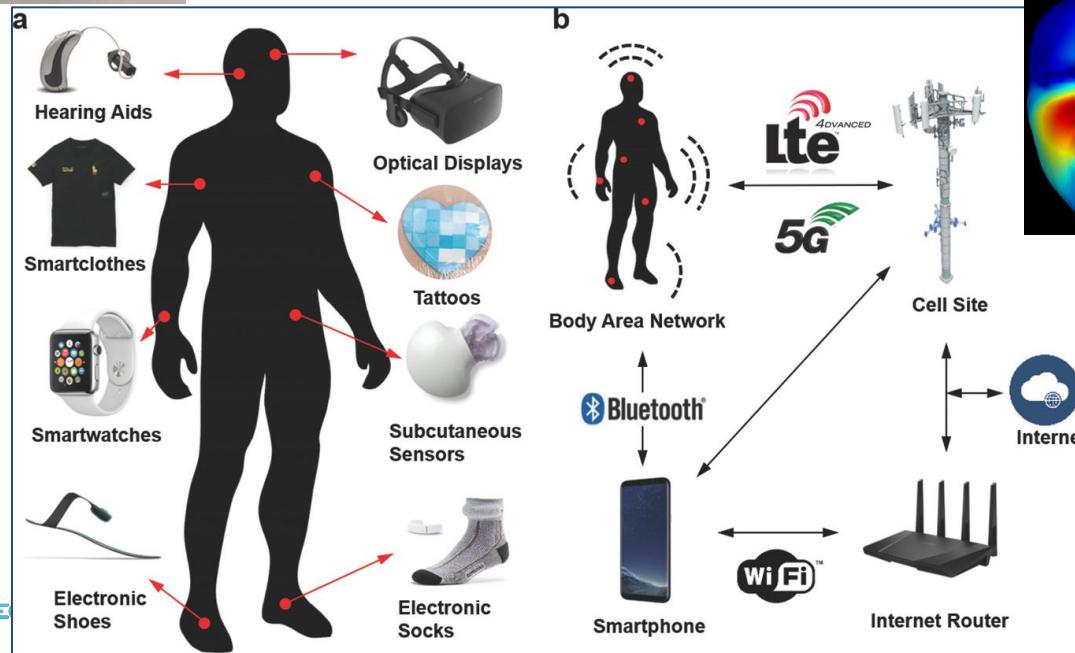
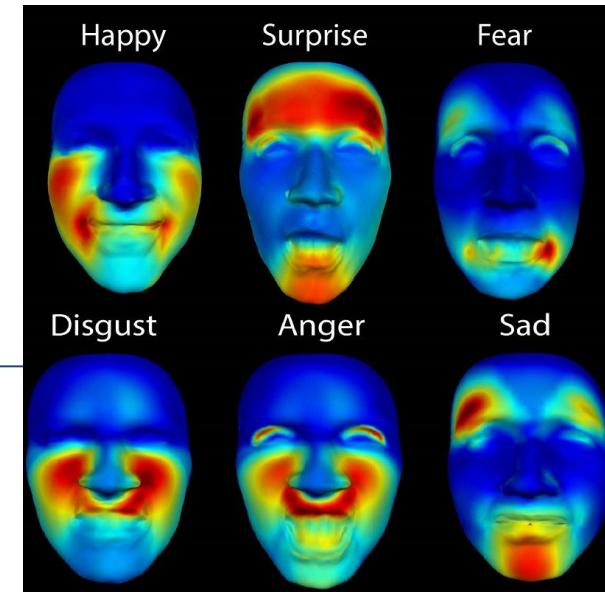


Electroencefalograma

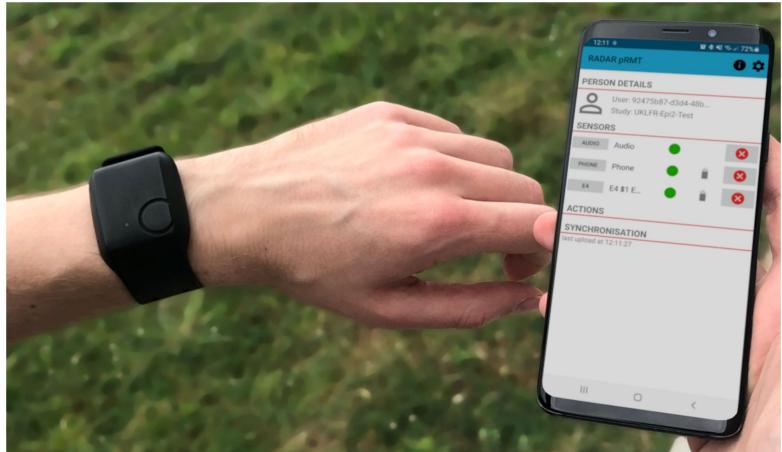
# Medición de las emociones humanas



Técnicas modernas



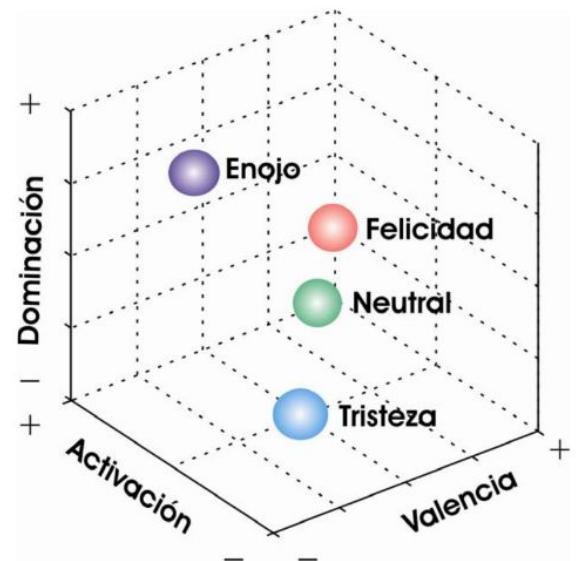
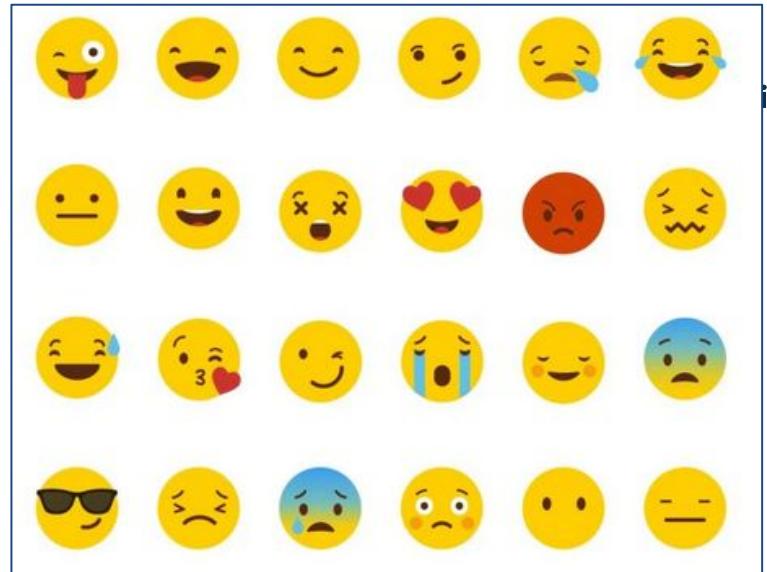
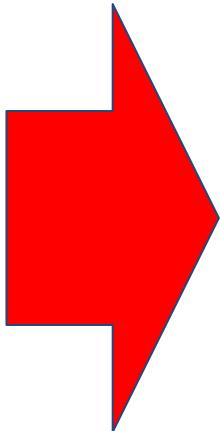
# El proyecto KUSISQA



Señales fisiológicas (EDA)



Señales de voz



# El proyecto KUSISQA: equipo técnico



# KUSISQA: resultados

Indicador	Comprometido	Alcanzado
Al menos dos (02) proyectos adicionales presentados para algún financiamiento que contribuya con la continuidad de la línea de investigación	3	3
Al menos un (01) prototipo a nivel de laboratorio	3	3
Al menos un (01) artículo científico, por investigador incorporado, presentado o aceptado para publicación en revistas indizadas en Q1 o Q2, preferentemente	5	5
Al menos dos (02) tesis de pregrado o posgrado presentados ante el jurado de tesis que conlleven a la obtención de títulos o grados académicos en universidades peruanas	4	4
Al menos dos (02) ponencias de alcance internacional	7	11
Al menos un (01) contrato de trabajo o prestación de servicios con un investigador incorporado con una vigencia mínima de tres (03) años, una vez culminado el proyecto	1	2
(Opcional) Para la línea de investigación a crear, fortalecer o consolidar: Al menos un (01) artículo científico, presentado o aceptado para publicación en revistas indizadas, de proyectos que tengan en curso o culminados	2	3

# KUSISQA: resultados

Al menos dos (02) proyectos adicionales presentados para algún financiamiento que contribuya con la continuidad de la línea de investigación

## Proyecto 1

**EUROPEAN COMMISSION**  
 Education, Audiovisual and Culture Executive Agency  
 Department A: Greenbase - EU Solidarity Corps  
**Erasmus+ A2: Erasmus Mundus, Sport**

Ref. Ans/2020/07/081-14/07/2020

Brussels,  
EACEA.A.3/LDF

Tanja Grönfors  
LAPPEENRANTA-LÄHDE TEKNILLINEN  
YLIOPISTO LUT  
YLIOPISTOKUAT SU  
LAPPEENRANTA  
FI - 53850

Programme EPPKA1 - KA1 - Key Action 1  
 Action JMD-MOB - Joint Master Degree  
 Proposal Call for proposals : EAC/MA2/2019 JMD-WOB  
 Application No. 619328-PP-1-2021-1-FI-EPPKA1-JMD-MOB  
 Title Educating software engineers with a sustainability mindset to improve the well-being of people through software solutions  
 Decision ACCEPTED

Dear Applicant,

You have submitted an application to the Erasmus+ programme, 2020 call for proposals for the action specified above. The call for proposals closed on 13/02/2020 - 17:00 (CET/CEST, Brussels time). The Education, Audiovisual and Culture Executive Agency (EACEA) received 97 eligible applications for this call.

I am writing to inform you about the selection decision taken by the Head of Department of the Agency, acting in her capacity as authorising officer, based on the recommendations of an Evaluation Committee assisted by external experts.

I am pleased to inform you that your application has been selected for EU co-financing.

It received 86/100 points, which is on or above the minimum threshold for funding of 81/100 points given the available budget.

The maximum amount of funding to be awarded to your project is 4.460.000,00 Euro.

For your information, out of the 100 applications submitted, 40 have been selected for funding.

The list of all selected projects will be published on the website of the Executive Agency when all applicants have been notified about the selection results:

[http://eacea.ec.europa.eu/erasmus-plus/selection-results\\_en](http://eacea.ec.europa.eu/erasmus-plus/selection-results_en)

Attached to this letter (Annex 3), you will find an evaluation report drawn up by the Evaluation Committee.

On behalf of the EACEA, the Research Executive Agency Validation Service (REA Validation Service) may contact you via the messaging system embedded in the Participant Register, requesting you to submit certain documentation within a specified deadline.

- \* This is in order to prove – in the event that your PIC number (Participant Identification Code) has not yet been validated – the legal existence and status of your organisation. In case of a multi-applicant proposal, each applicant will be contacted individually.
- The REA Validation Services may also contact your organisation requesting documentation to assess the financial capacity of your organisation.

Confirmation received from your Contracting Authority:  
 Avenue du Ravelin 1-109 - Bruxelles - Belgium - Tel. +32 22991111  
 Office: J50-01653 - Tel. direct fax +32 229-67108  
 619328-PP-1-2020-1-FI-EPPKA1-JMD-WOB

<http://eacea.ec.europa.eu> [Lutens.DI-PONZI@ec.europa.eu](mailto:Lutens.DI-PONZI@ec.europa.eu)

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## Proyecto 2

8/26/2021 Detección automatizada del estado emocional de los adultos mayores en procedimientos de atención sanitaria pos...

**IDB Inter-American Development Bank**

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**Detección automatizada del estado emocional de los adultos mayores en procedimientos de atención sanitaria post-pandemia COVID-19**

Language English

Institution name Universidad Nacional de San Agustín de Arequipa

Director Wilber Roberto Ramos Lovón

Department Escuela Profesional de Ciencias de la Computación

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Telephone +51 54 285298

E-mail epc@unsa.edu.pe

Web address <https://fis.unsa.edu.pe/cienciadeicomputacion/>

Country Peru

Proposal file FORMULARIO GDLAB Esp\_1.pdf

<https://research-proposals.idb.org/rede/2481>

1/2

## Proyecto 3

**UNSA** UNIVERSIDAD NACIONAL SAN AGUSTÍN DE AREQUIPA

**UNSA Investiga** VICERRECTORADO DE INVESTIGACIÓN

**BASES DE FONDOS CONCURSABLES PROYECTOS DE INVESTIGACION - 2022**

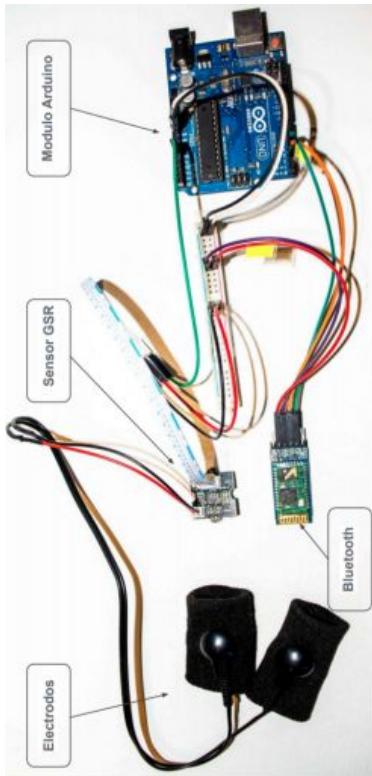
**SE FINANCIAN HASTA 280 000 NUEVOS SOLES**

Vicerrectorado de Investigación de la Universidad - UNSA  
<http://vri.unsa.edu.pe>

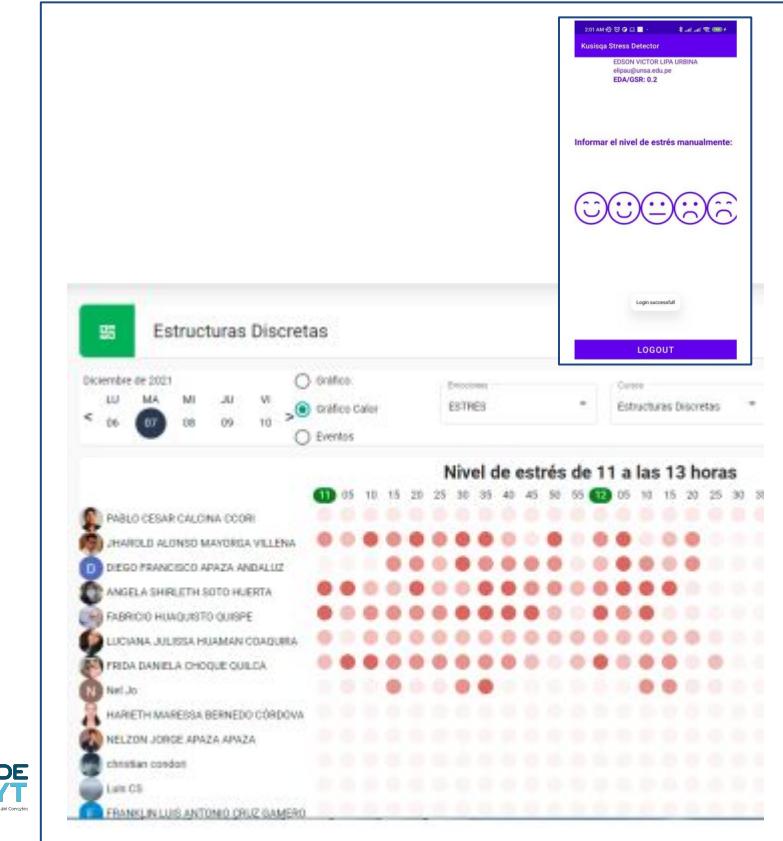
# KUSISQA: resultados

Al menos un (01) prototipo a nivel de laboratorio

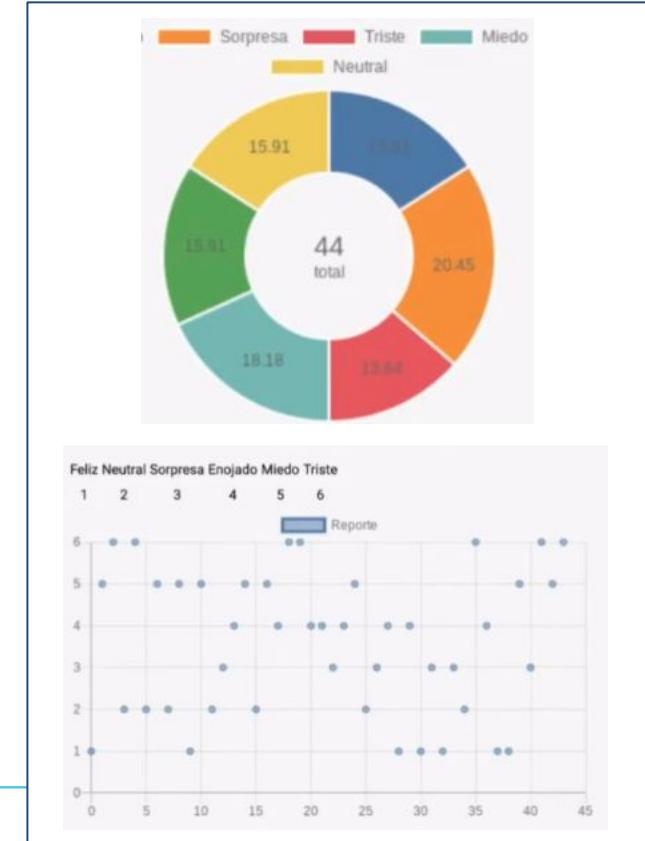
Prototipo 1



Prototipo 2

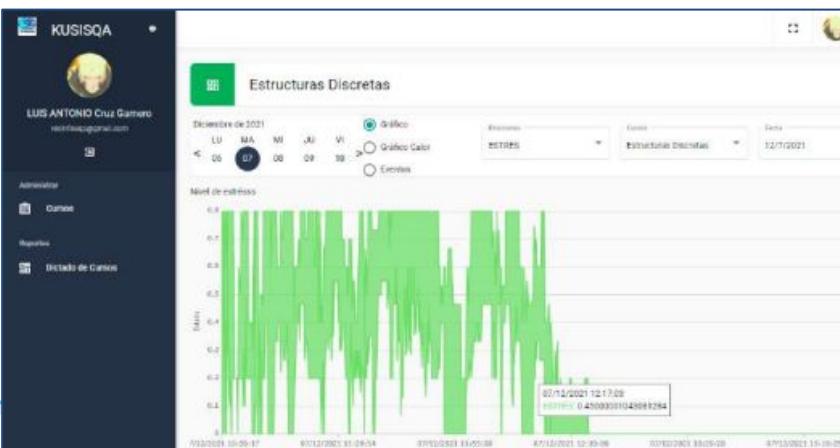
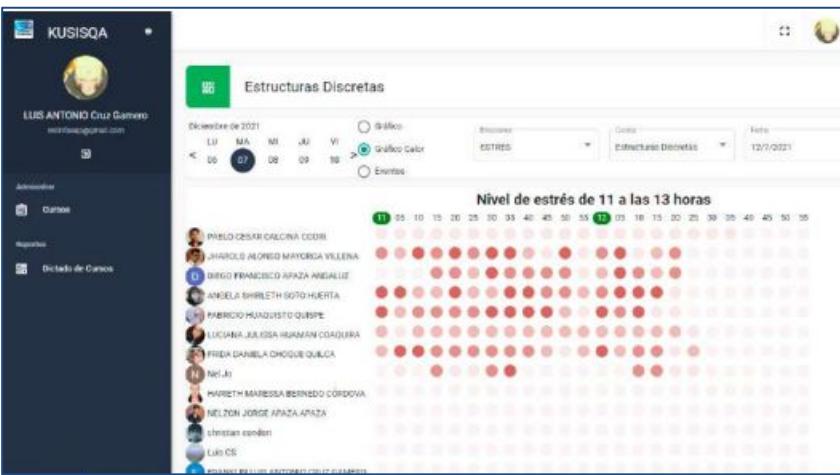


Prototipo 3



# El prototipo de software

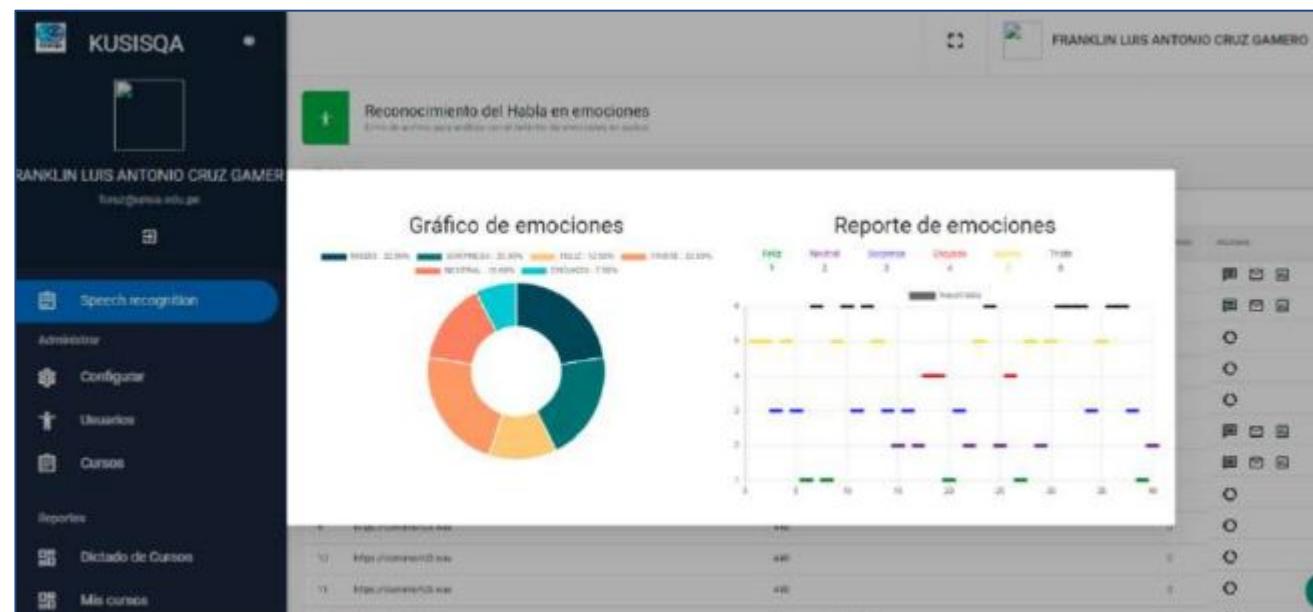
(detección de emociones a partir de señales fisiológicas - EDA)



# El prototipo de software

## (detección de emociones a partir del habla)

The screenshot shows a software interface for speech recognition. On the left is a sidebar with options like 'Speech recognition', 'Configurar', 'Usuarios', 'Cursos', 'Reportes', 'Dictado de Cursos', and 'Mis cursos'. The main area has a title 'Reconocimiento del Habla en emociones' and a sub-section 'Resumen de análisis para análisis con el detector de emociones en audio'. It displays a table with columns 'Nombre del archivo a emitir', 'Duración', 'Peticiones', 'Estado', and 'Acciones'. The table contains several rows of processed audio files, each with a green circular icon indicating successful processing.

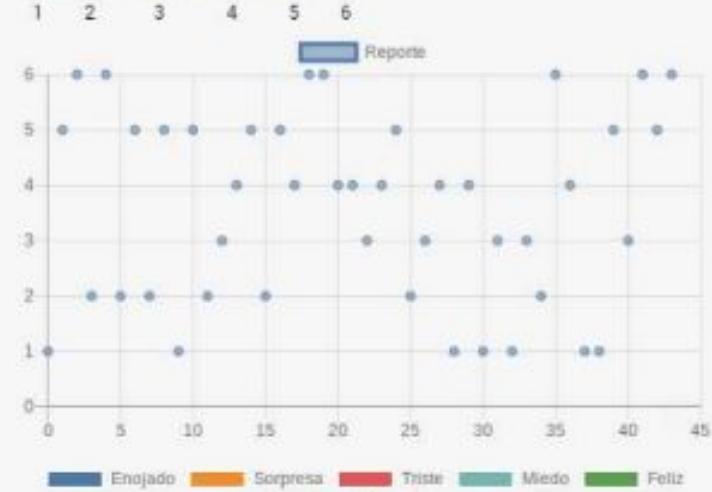


Buenos días Luis Chirinos!

El siguiente mensaje es con respecto a la información de tu última clase:

## Reportes del video

Feliz Neutral Sorpresa Enojado Miedo Triste



# KUSISQA: resultados

Al menos un (01) artículo científico, por investigador incorporado, presentado o aceptado para publicación en revistas indizadas en Q1 o Q2, preferentemente

## A Mapping Study on Anxiety Detection Using Wearable Technology

ANGELA MAYHUA-QUISPE, Universidad Nacional de San Agustín de Arequipa, Peru  
NELLY CONDOR-FERNANDEZ, Universidad de Coruña, Spain, Vrije Universiteit Amsterdam, The Netherlands, and Universidad Nacional de San Agustín de Arequipa, Peru

ALVARO CUNO, Universidad Nacional de San Agustín de Arequipa, Peru

Anxiety detection using wearable technology has become an interesting research field over the last years due to its potential for providing valuable information about anxiety levels. To date, little is still known about how the accuracy evaluation of anxiety detectors is carried out. To fill this knowledge gap, we conducted a systematic mapping study (SMS) that takes as inputs 15 primary studies. The SMS aims to answer several research questions that will allow us to (i) identify the measurement used to measure anxiety, as well as the experimental designs and data analysis; and (ii) determine the best methods to evaluate the performance and classification techniques used in anxiety detection. The review results show that anxiety detection has been studied as binary and continuous problems. We also found that several physiological signals were considered for detecting anxiety – in fact, most studies detected anxiety by heart rate – whereas the ECG signal is considered as the most reliable input because of the high accuracy (greater than 95%) achieved with different classifiers. Moreover, most of the studies created their own datasets to evaluate their classifiers, although a few studies validated the accuracy due to the small size of the datasets.

CSCS Concepts → **Hardware → Sensor devices and platforms → Applied computing → Psychology → Computing methodologies → Machine learning approaches**.

Additional Key Words and Phrases: wearable, anxiety detection, anxiety classification, physiological signal

ACM Reference Format: Angela Mayhua-Quispe, Nelly Condor-Fernandez, and Alvaro Cuno, 2020. A Mapping Study on Anxiety Detection Using Wearable Technology. *1*: 1 (November 2020), 23 pages. <https://doi.org/10.1145/immune.2020.1711>

1. INTRODUCTION  
Wearable technology was introduced successfully to the people in 1977 as a calculator watch [4]. After that year, this technology was evolving by the use of sensors, electronics, software, connectivity, and power sources in body-worn devices – e.g., watches, wristbands, glasses, bracelets, and garments – that known simply as wearables [41]. Nowadays, wearables can unobtrusively collect physiological signals applying the recognition of human emotions.

The first step, where a set of job that must be processed in an ordered sequence, is called scheduling. There are two main types of scheduling: the static scheduling, where the tasks are processed in the same order as they are given, and every machine processes the jobs in the same order (without job planning). Each job needs to be assigned to a specific machine. For example, if there are three machines and each machine can process one job at a time, it is said to be a single machine scheduling problem. The scheduling problem is a problem that is often solved using mathematical optimization models [28].

Currently, there are several challenges related to the availability of speech datasets (also known as corpus) usable in intelligent systems, particularly those to be used to recognize emotions. For example, recognize the emotions of speakers of different languages using a single language is a challenge. In this case, it is necessary to know that corpora can be grouped into two main groups: a majority group where several corpora of the same language can be found (English or German) and a minority group made up of corpus in languages such as Spanish<sup>1</sup>, Hebrew, Russian, Korean, Japanese, Shangai, and so on. One aspect that increases the complexity of the speech databases is the variety of dialects and accents used, and the variety of the speakers used, for example, it is not the same to have a native speaker as an experienced translator from Spain with a Spanish-speaking speaker from Peru. Studies by Hansen [42] and Yang X. et al. [43] have shown that variations in speech due to dialect is a factor that significantly affects the speech system as each dialect uses different phonetic features and words for the same meaning, as well as differences in grammar and vocabulary.

The work above, an ear corpus in Spanish but with a Peruvian accent is justified. Different machine learning systems from different manufacturers could use this corpus to obtain greater precision when recognizing the emotions of Peruvian speakers. Likewise, it could also be used in cross-language research which has shown that there is a lack of data to support a better selection among the best positions, such as [44] and [45].

In this paper, we propose a set of speech corpora for emotion detection in Spanish, as well as an effort to point out the main contributions of this paper. The last two proposals represent the best results when the students between Spanish speakers with a Peruvian accent. It contains audio segments labeled with three emotional attributes: valence (negative to positive), arousal (calm to active), and dominance (dominated to dominating). This corpus contributes to the current state of the art of automatic recognition of emotions, which according to our knowledge, is the first of its kind.

**ARTICLE IN PRESS**  
European Journal of Operational Research  
<https://doi.org/10.1016/j.ejor.2020.05.023>

Received 20 August 2019  
Revised 11 April 2020  
Accepted 19 April 2020  
**Keywords:** Job scheduling; Optimisation; Job sequencing; Discrete optimisation; Scheduling

**Abstract**  
This paper introduces a new algorithm proposed for the generalization of shop scheduling problem with multi-resource restrictions on the next lesson (M3) constraint heuristic and the shared capacity approach. The M3 constraint heuristic is a well-known method for discrete optimisation problems, involving the division of the scheduling problem into several smaller subproblems, evaluating the feasibility of the resulting schedule for all resource positions, and selecting the shortest path. The shared capacity approach is a methodology based on a shared resource that allows multiple machines to work simultaneously, thus improving the results produced by approximate methods when extended onto this. In this paper we propose a new algorithm named "Job Insertion Tiebreaker" (JIT) based on the M3 constraint heuristic and the shared capacity approach proposed by inserting the jobs in each slot (one-time procedure). They were designed considering the M3 constraint heuristic as the core of the algorithm, but it is also included a tiebreaker mechanism for improving the results of the M3 constraint heuristic. The experiments demonstrate that the proposed algorithm has obtained results slightly better than the best approximate methods or for the permutation flow shop scheduling problem.

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**Keywords:** Job scheduling; Optimisation; Job sequencing; Discrete optimisation; Scheduling

**1. Introduction**  
The JIT problem, where a set of job that must be processed in an ordered sequence, is called scheduling. There are two main types of scheduling: the static scheduling, where the tasks are processed in the same order as they are given, and every machine processes the jobs in the same order (without job planning). Each job needs to be assigned to a specific machine. For example, if there are three machines and each machine can process one job at a time, it is said to be a single machine scheduling problem. The scheduling problem is a problem that is often solved using mathematical optimization models [28].

**2. Background & Summary**  
Emotions are an essential element in communication between human beings, although they are also essential in communication systems and machines [1]. For this reason, in many areas of the scientific community, automatic emotion recognition systems are a latest research topic nowadays<sup>2</sup>. One of the schemes that have received the most attention from researchers due to their availability is those based on recognizing emotions from speech using techniques based on supervised deep learning. However, it is known that these techniques strongly depend on the existence of data for training, validation, and testing, as well as on the quality of the data used. Therefore, it is very difficult to obtain high-quality datasets with the required amount of data. Currently, there are several challenges related to the availability of speech datasets (also known as corpus) usable in intelligent systems, particularly those to be used to recognize emotions. For example, recognize the emotions of speakers of different languages using a single language is a challenge. In this case, it is necessary to know that corpora can be grouped into two main groups: a majority group where several corpora of the same language can be found (English or German) and a minority group made up of corpus in languages such as Spanish<sup>1</sup>, Hebrew, Russian, Korean, Japanese, Shangai, and so on. One aspect that increases the complexity of the speech databases is the variety of dialects and accents used, and the variety of the speakers used, for example, it is not the same to have a native speaker as an experienced translator from Spain with a Spanish-speaking speaker from Peru. Studies by Hansen [42] and Yang X. et al. [43] have shown that variations in speech due to dialect is a factor that significantly affects the speech system as each dialect uses different phonetic features and words for the same meaning, as well as differences in grammar and vocabulary. The work above, an ear corpus in Spanish but with a Peruvian accent is justified. Different machine learning systems from different manufacturers could use this corpus to obtain greater precision when recognizing the emotions of Peruvian speakers. Likewise, it could also be used in cross-language research which has shown that there is a lack of data to support a better selection among the best positions, such as [44] and [45].

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<sup>1</sup> Corresponding author.  
<sup>2</sup> See authors: <https://www.semanticscience.org/> (A.J. Bejarano).

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XXXXX-XXXX(2020)1-ART\_1-\$15.00  
<https://doi.org/10.1016/j.ejor.2020.05.023>

## ESCorpus-PE: A speech emotional database in Spanish with Peruvian accent

Alvaro Cuno<sup>1\*</sup>, Alessandra Delgado<sup>2</sup>, Wilber Ramos Lovón<sup>1</sup>, and Harieth Bermedo<sup>1</sup>  
<sup>1</sup> Universidad Nacional de San Agustín de Arequipa, Departamento Académico de Ingeniería de Sistemas e Informática, Arequipa, Peru  
<sup>2</sup> corresponding author: Alvaro Cuno (acunoa@unsa.edu.pe)

### ABSTRACT

Lastly, there has been considerable progress in speech emotion recognition systems based on deep learning. However, many of them are not generalizable to languages for which they were not trained. One of the reasons for this limitation is the absence of corpora in all existing human languages, dialects, and accents. This paper contributes to filling this gap, presenting ESCorus-PE, which is a speech emotional database in Spanish with Peruvian accent. It contains 153 audio samples of speech with three emotional attributes: valence, arousal, and dominance. Speech data originate from YouTube videos with interactions in their natural form. Our database can be a valuable resource for researchers studying emotion recognition in multilingual settings.

### Background & Summary

Emotions are an essential element in communication between human beings, although they are also essential in communication systems and machines [1]. For this reason, in many areas of the scientific community, automatic emotion recognition systems are a latest research topic nowadays<sup>2</sup>. One of the schemes that have received the most attention from researchers due to their availability is those based on recognizing emotions from speech using techniques based on supervised deep learning. However, it is known that these techniques strongly depend on the existence of data for training, validation, and testing, as well as on the quality of the data used. Therefore, it is very difficult to obtain high-quality datasets with the required amount of data. Currently, there are several challenges related to the availability of speech datasets (also known as corpus) usable in intelligent systems, particularly those to be used to recognize emotions. For example, recognize the emotions of speakers of different languages using a single language is a challenge. In this case, it is necessary to know that corpora can be grouped into two main groups: a majority group where several corpora of the same language can be found (English or German) and a minority group made up of corpus in languages such as Spanish<sup>1</sup>, Hebrew, Russian, Korean, Japanese, Shangai, and so on. One aspect that increases the complexity of the speech databases is the variety of dialects and accents used, and the variety of the speakers used, for example, it is not the same to have a native speaker as an experienced translator from Spain with a Spanish-speaking speaker from Peru. Studies by Hansen [42] and Yang X. et al. [43] have shown that variations in speech due to dialect is a factor that significantly affects the speech system as each dialect uses different phonetic features and words for the same meaning, as well as differences in grammar and vocabulary. The work above, an ear corpus in Spanish but with a Peruvian accent is justified. Different machine learning systems from different manufacturers could use this corpus to obtain greater precision when recognizing the emotions of Peruvian speakers. Likewise, it could also be used in cross-language research which has shown that there is a lack of data to support a better selection among the best positions, such as [44] and [45].

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<sup>1</sup> Corresponding author.  
<sup>2</sup> See authors: <https://www.semanticscience.org/> (A.J. Bejarano).



## Use of mobile application and persuasiveness in first year students of Computer Science at the Universidad Nacional de San Agustín

To carry out this work, two learning sessions about ordering and search were designed for the Discrete Structure II course, with the participation of 7 and 5 students respectively. During their development, they wore an E4 wristband to record their electrodermal activity and, depending on the level of stress found, a persuasive message was sent to them.

The results indicate that in the second session there was a decrease in the number of cognitive responses, and in the mean belief for both positive and negative thoughts. Meanwhile, the polarity index, in the second session had negative polarity indexes with an increase in negative thought responses. The difference between the two sessions could be due to the influence of a change in the persuasive messages or some internal factor of the participant.

**Keywords:** Persuasion; elaboration likelihood model; persuasive software; thought listing; cognitive responses.

### Introduction

The use of technological tools to improve the teaching and learning process has increased in recent years, especially those related to the development of specialized software for communication and the development of academic activities. Consequently, different projects and initiatives have emerged to take advantage of scientific development in areas such as software engineering, affective computing and artificial intelligence.

Among these initiatives is the *KUSISQA project: Helping to Regulate Our Emotions in the Teaching and Learning Processes*. This project focuses on the recognition of emotions for the development of a persuasive software system that allows students to improve their academic performance. It consists of the collection of physiological information of the student's electrodermal activity through a wearable system that records the student's electrodermal activity using a mobile application.

**A prototype for emotion detection in classroom based on speech analysis**

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**Abstract**  
During a class session there are many situations that generate different emotions in the participants. There are many works devoted to the analysis of the emotions in students, however, they have focused mainly on the analysis of the emotional response, but have not developed a prototype that extracts audio from video recordings of classes. From the extracted audio we use a classifier to identify the emotion contained in each video and generate a response message to help the student to regulate his/her emotion based on the implemented algorithm with a manual annotated dataset.

### 1. Introduction

In the COVID-19 pandemic scenario, remote education has played a transcendental role in maintaining the educational activity. However, the lack of interaction has limited the effectiveness of the educational process. In this context, remote education has faced new challenges related to the increase of stressful situations and other negative emotions in the students.

In this sense, it is necessary to be able to know the emotions of students and teachers to better understand their behavior. Recently, there is a range of interests in human emotion recognition in a wide range of situations, with several fields of applications, such as health, security, and entertainment.

Many previous efforts were devoted to the development of emotional recognition systems for real-time scenarios, such as speech emotion recognition and facial expression recognition. These systems are based on the implementation of various machine learning models, such as Support Vector Machines, Random Forests, and Convolutional Neural Networks. In this work, we propose an affective speech emotion recognition system based on a generative and discriminative model. The system is able to recognize emotions in speech recordings. The proposed model is based on the combination of the HMM and GMM models. In this work, the HMM and GMM models are used to recognize the speech emotion. The HMM is used to extract the speech features and the GMM is used to classify the speech emotion.

Particularly, the speech emotion recognition (SER) is a field of growing interest among the scientific community due to its applications in real-life scenarios. In [1], the facial and vocal emotion recognition systems are presented. Regarding the speech emotion recognition, there is a wide range of applications in human-computer interaction (HCI), such as speech emotion recognition in a wide range of situations, with several fields of applications, such as health, security, and entertainment.

In [2], the authors propose a framework for emotion detection in speech. In [3], the authors present a speech emotion recognition system based on a generative and discriminative model. In [4], the authors propose a speech emotion recognition system based on a generative and discriminative model for the Hindi language. In [5], the authors propose a speech emotion recognition system based on a generative and discriminative model for the Persian language. In [6], the authors propose a speech emotion recognition system based on a generative and discriminative model for the Arabic language. In [7], the authors propose a speech emotion recognition system based on a generative and discriminative model for the English language. In [8], the authors propose a speech emotion recognition system based on a generative and discriminative model for the Spanish language. In [9], the authors propose a speech emotion recognition system based on a generative and discriminative model for the German language. In [10], the authors propose a speech emotion recognition system based on a generative and discriminative model for the French language. In [11], the authors propose a speech emotion recognition system based on a generative and discriminative model for the Italian language. In [12], the authors propose a speech emotion recognition system based on a generative and discriminative model for the Portuguese language. In [13], the authors propose a speech emotion recognition system based on a generative and discriminative model for the Spanish language.

Finally, in [14], the authors propose a speech emotion recognition system based on a generative and discriminative model for the English language. In [15], the authors propose a speech emotion recognition system based on a generative and discriminative model for the Spanish language. In [16], the authors propose a speech emotion recognition system based on a generative and discriminative model for the English language. In [17], the authors propose a speech emotion recognition system based on a generative and discriminative model for the Spanish language.

In this work, we propose a speech emotion recognition system based on a generative and discriminative model for the Spanish language. In [18], the authors propose a speech emotion recognition system based on a generative and discriminative model for the Spanish language.

# KUSISQA: resultados

Al menos dos (02) tesis de pregrado o posgrado presentados ante el jurado de tesis que conlleven a la obtención de títulos o grados académicos en universidades peruanas

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# KUSISQA: resultados

Al menos dos (02) ponencias de alcance internacional

REFSQ 2020 Doctoral Symposium

## Anxiety as Implicit User Feedback in Serious Games

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**Abstract.** The application of serious games in different areas (*e.g.*, as education, training, health care) has been growing rapidly because of their main purpose of teaching or improving skills in end-users. A well-designed game maintains the players into their flow zones, avoiding negative emotions (*e.g.*, anxiety) to achieve the didactical and pedagogical goals. For that reason, good user experience is fundamental to engage the player with the game and support his learning process. In this context, this paper presents a general approach to use the player's anxiety levels as implicit feedback for validating UX requirements and improve the serious game.

**Keywords:** Anxiety detection · Implicit user feedback · Serious games.

## 1 Introduction

The emerging area of serious games has been growing during the last ten years in the research and the industry fields [10]. It is estimated that this market reaches \$9,167 million by 2023 due to serious games are being used in different contexts (*e.g.*, healthcare, education, government) with the main function to train, inform, or teach end-users [14]. Similar to other games, the entertainment is a component of serious games [10] and has an important role to maintain the user engagement in the training or learning process.

A well-designed game maintains the player into his flow zone, evoking positive emotions and feelings. According to Chen [3], the flow experience means what a user experiences playing a game so that it depends on maintaining a balance between the players' ability and the inherent challenge of the game. Fig. 1a displays the location of Flow Zone compared to anxiety and boredom. For example, if the challenge is higher than ability, the activity generates anxiety and the player could leave the game. On the other hand, Fig. 1b shows the different flow zones according to the different players (*e.g.*, hardcore, novice). This means that the flow experience for each user depends on his/her abilities.

The flow experience is related to user experience (UX) playing the game. Moreover, according to Condori-Fernandez *et al.* [4], the negative emotions perceived by end-users are useful information for discovering unsatisfied non-functional requirements (NFR), and improving the UX. In serious games, as negative emotions (*e.g.*, anxiety, stress) can also affect the processing of the learning content

Live study track at REFSQ 2020

## Exploring the influence of the emotional state on the efficiency for correcting defects in conceptual models: a live study

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### Abstract

The purpose of this live study is to investigate the impact of emotional states on the efficiency of correcting defects within a Conceptual Model verification context. We plan to use the CoSTest tool for verifying defects and measure the experienced emotions during the testing tasks through L-PANAS-SF and SUDS instruments. Furthermore, the design, overview and potential threats of this experiment are presented.

## 1 Introduction

### 1.1 Research problem

Emotions have an important role in our life and their influence on our work is not an exception. According to Weiss and Cropanzano [WC96], emotions and moods have effects on work performance, affecting the duration of activities and the perceived experience. This situation is not only present in industrial activities; some researchers have been analyzing the influence and importance of emotions in software engineering [Wer13, CSdB14, MP15, RFB19], being the software development process one of the principal centers of interest.

Wrobel [Wer13] analyzed the influence of the emotional state on the performance of programming tasks collecting data through a questionnaire and an interview for each participant, the Job Emotions Scale (JES) was applied to measure human emotions. Romano *et al.* [RFB19] also investigated in this area, analyzing emotions when novice developers apply the TDD (Test-Driven Development) approach on change tasks; in a similar way, they collected data through self-reported emotions by the participants. In the same way, we propose to analyze the influence of the emotional state on the work productivity based on correcting defects in conceptual models.

In line with this notion, according to Granda *et al.* [GCV15], defects (*e.g.*, missing, wrong and unnecessary elements) in conceptual models (*e.g.*, UML class diagram) can be located in several ways through Validation & Verification techniques, which use a detection mechanism (based on rules, metrics, and modelling conventions)

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ACM SAC 2020

## Towards a Non-Functional Requirements Discovery Approach for Persuasive Systems

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Nevertheless, to find the right way of persuasion is important to have first a clear understanding of requirements that factually help in changing the user behavior to achieve sustainability goals. This can be accomplished by discovering, from their actual use, relevant requirements that should be considered to redesign and maintain software interactive systems, and consequently contribute to a longer term usage.

We have found several approaches for discovering requirements such as scenario-based approaches [6][7][8], and user feedback-driven approaches [9][10]. One of the disadvantages of the scenario-based approaches is that the effectiveness of discovering requirement depends on how the scenarios are represented; whereas user-feedback driven approaches, depending on the available online data (in app store reviews, blogs, forums, etc.), can be biased [11] and consequently give fake feedback to software designers.

In contrast to these approaches, our proposal in this paper lies on the importance of understanding how user experiences with a software product and gaining insight into user needs. In this direction, this paper is concerned with (i) the theoretical and methodological aspects that should be considered in an UX assessment of BCSSs; (ii) the discovery of user needs emerging from UX assessment; (iii) and the importance of translating such user needs into NFR that must be addressed for improving UX.

We use the Persuasive System Design (PSD) model [12] as the theoretical framework for designing our empirical UX assessment and defining the requirements discovery process. With our approach we aim to increase awareness in designers/developers about NFRs that were discovered as a consequence of detecting negative UX.

The rest of this paper is organized as follows. Section 2 introduces the PSD model as a design instrument for BCSSs. The NFR discovery approach is presented in Section 3. Section 4 draws the conclusions and discusses future work.

## 2 BACKGROUND: THE PSD MODEL

In this paper, we consider the PSD model [12], as the theoretical framework for our research. The PSD model is a recent conceptualization for designing, developing and evaluating persuasive systems. It consists of the premise behind any persuasive system, the persuasion context and the persuasive

# KUSISQA: resultados

Al menos dos (02) ponencias de alcance internacional

SIMBIG 2020

## An evaluation of physiological public datasets for emotion recognition systems

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**Abstract.** [Background] The performance of emotion recognition systems depends heavily on datasets used in their training, validation, or testing stages. [Aims] This research aims to evaluate the extent to which public available physiological datasets created for emotion recognition systems meet a set of reference requirements. [Method] Firstly, we analyze the applicability of some reference requirements proposed for stress datasets and adjust the corresponding evaluation criteria. Secondly, nine public physiological datasets were identified from a previous survey. [Results] None of the evaluated datasets satisfy all the reference requirements in order to be considered as a reference dataset for being used in the construction of reliable emotion recognition systems. [Conclusion] Although the evaluated datasets do not support the whole reference requirements, they provide a baseline for further development. Also, a greater effort is needed to establish specific reference requirements that can appropriately guide the creation of physiological datasets for emotion recognition systems.

**Keywords:** Physiological datasets - Reference requirements - Assessment.

## 1 Introduction

In recent years, automatic emotion recognition systems have been undergoing considerable popularity growth because it can be applied in many critical areas such as safe driving [32], health care [19, 29] citizen security [21], and so on [25, 1, 17, 33]. In particular, those attracting the most attention are the systems based on physiological signals because of their involuntary nature and the difficulty involved in controlling them. According to Shi et al. [25], in emotion recognition systems, physiological signals (i.e., electroencephalogram, temperature, electrocardiogram, electromyogram, galvanic skin response, respiration, etc.) are having

SCCC 2020

## A FAIR evaluation of public datasets for stress detection systems

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**Abstract.**—Nowadays, datasets are an essential asset used to train, validate, and test stress detection systems based on machine learning. In this paper, we use two sets of FAIR metrics for evaluating five public datasets for stress detection. Results indicate that all these datasets comply to some extent with the (Findable, Accessible, and Reusable) principles, but none with the (Interoperable) principle. These findings contribute to raising awareness: (i) the need for the (FAIR) development and improvement of stress datasets, and (ii) the importance of promoting open science in the affective computing community.

**Index Terms**—FAIR principles, Stress detection, Datasets.

### I. INTRODUCTION

Open Science [1] is the international movement to make any research artifact (i.e., software source code, datasets, algorithms, analysis scripts, manuscripts, tools, and so on) available and accessible to society. One of the common justifications for open science is the growing concern that researchers have expressed about a “reproducibility crisis” in some scientific fields [2], [3], including artificial intelligence [4] and computer science [3]. Open Science seeks to increase the transparency and thus, **repeatability** (i.e., same team, same experimental setup), **reproducibility** (i.e., different team, same experimental setup) and **replicability** (i.e., different team, different experimental setup) of the scientific process and their results [6].

Several countries have taken the initiative to follow the Open Science movement. For instance, in 2018, the European Commission launched the European Open Science Cloud (EOSC) as a process of making research data in Europe accessible to all researchers under the same terms of use and distribution [7]. This initiative aims to push Europe towards a culture of open research artifacts that are Findable, Accessible, Interoperable, and Reusable. In other words, it aims to satisfy the FAIR principles [8]. In a broader global context, there are similar initiatives such as the USA NIH Data Commons, the Australian Research Data Commons, and the proposed African

The FAIR foundational principles introduced by Wilkinson et al. [8] provide guidelines to enable digital resources, such as datasets, source code, tools, workflows, and other scientific research artifacts, to become more Findable, Accessible, Interoperable, and Reusable for humans and computers. Jacobson et al. [16] describe the core objective of these principles as following:

### II. BACKGROUND

#### A. The FAIR principles

The FAIR foundational principles introduced by Wilkinson et al. [8] provide guidelines to enable digital resources, such as datasets, source code, tools, workflows, and other scientific research artifacts, to become more Findable, Accessible, Interoperable, and Reusable. In this context, the user feedback is interpreted as relevant information from any response about the interaction between the software system and the user [13]. During many years, user feedback has been mainly used by recommendation systems [35], depending largely on explicit feedback (e.g., filling questionnaires, sending a report about an error, suggestion of new functionality, or rating) that users can provide. Nonetheless, users are not interested to give feedback, or they could even give false feedback; for that reason, explicit feedback is sometimes difficult to obtain [36].

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## Understanding Implicit User Feedback from Multisensorial and Physiological Data: A case study

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### ABSTRACT

Evaluating the quality of user experience is very important for increasing the acceptance likelihood of software applications, which can be affected by several contextual factors that continuously change over time (e.g., maintained state of end-user). Due to these changes in the context, software continually needs to adapt for delivering software services that can satisfy user needs. However, to achieve this adaptation, it is important to gather and understand the user feedback. In this paper, we mainly investigate whether physiological data (or implicit user feedback) can be used as a kind of implicit user feedback. To this end, we conducted a field study involving a tourist traveling abroad, who used a wearable device for monitoring his physiological data, and a smartphone with a mobile app for reminding him to take his medication on time during four days. Through the case study, we were able to identify some factors and activities as emotional triggers, which were used for understanding the user context. Our results highlight the importance of having a context analyzer, which can help the system to determine whether the detected stress could be considered as actionable and consequently as implicit user feedback.

### KEYWORDS

Artificial emotion, Implicit user feedback, Physiological data, Context information, Case study

### 1 INTRODUCTION

Nowadays, software systems have become more complex but at the same time also more interactive. The main aim of interaction technologies is to enable the user to interact with the system. However, to have the (self-)adaptation of a system, it is required to obtain information about any response of the interaction between the system and the user (i.e., user feedback) [46]. According to Maeshiro and Vandervalk, the usefulness of this information depends on its consistency, credibility, and accuracy [34]. Given user feedback has become a source of valuable information for software developers, companies or users that are looking for improvements of the quality of a service or product (e.g., [32]). It has been largely investigated by different computer science fields, especially in Human-Computer Interaction [22, 28, 34, 39, 51] and Software Engineering [30, 38, 35].

Hence, one of the main issues is how to collect useful user feedback, which is both accurate and credible [34]. In this direction, different approaches have been investigated (e.g., [11, 18, 28, 36]). For example, some works argue that implicit user feedback is sometimes difficult to obtain [36].

# KUSISQA: resultados

Al menos dos (02) ponencias de alcance internacional

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## How Do Negative Emotions Influence on the Conceptual Models Verification? A live study proposal

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### ABSTRACT

The present live study is proposed with the objective of investigating the influence of negative emotions (i.e., stress) in the efficiency for verifying conceptual models. To conduct this study, we use a Model-driven Testing tool, named CoTest, and our own version of stress detector within a competition setting. The experiment design, overview of the empirical procedure, instrumentation and potential threats are presented in the proposal.

### 1 INTRODUCTION

#### 1.1 Research problem

Emotions and moods are inherent to all human experiences and consequently have effects on our work performance [15]. For example, emotions might affect the correctness and duration of activities, or the perceived experience. In the last years, some researchers have been focused on analyzing what emotions are present and their influence in the software engineering field (e.g., [2, 4, 11, 13, 18]), being the software development process one of the main topics of interest.

The influence of the emotional state on the performance of programming tasks was analyzed by Wedel [18], who applied a questionnaire and an interview for each participant to collect data. In this study, the Job Emotions Scale (JES) was applied to measure human emotions. Romano et al. [13] analyzed emotions when novice developers apply the TDD (Test-Driven Development) approach on change tasks; similarly, they collected data through self-reported emotions by the participants.

On the other hand, researchers have not focused only on analyzing self-reported emotions states, we have also found some attempts in using physiological data to determine automatically some emotions in a specific context. For example, Maller and Fritze [10] collected heart rate measures (e.g., electrodermal activity, electrocardiography, skin temperature, heart rate) using different devices (Empatica E4-wristband, Neurosky Mindband and Eye-Trak) to distinguish positive and negative emotions using machine learning techniques in the context of software change tasks; also, participants assessed their emotions answering periodically a short questionnaire. Granda et al. [4] replicated the work of Maller and Fritze in the same context, using similar devices (Empatica E4-wristband,

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Emotiv Insight and Tobii 4C), including more subjects in their experiment. Other work was proposed by Suni Lopez et al. [10], they were focused on detecting physiological stress in real-time in a quiet office workplace environment, using some well-known emotional triggers from the Psychology community. Authors used the E4-wristband for gathering the electrodermal activity (EDA) and applying an annual-based statistical approach for the stress detection.

Although the amount of research for understanding the influence of emotions in software development process is rising, the impact of emotions in model-driven development (MDD) has not been yet well investigated. We selected the MDD context, because its adoption in the industry is growing quite rapidly [12]. Additionally, we have access to CoTest, a model-driven testing tool. A conceptual model (CM) is a key asset in MDD because represents abstract concepts of the relationships between objects in a specific project. If a CM has defects, these are passed on to the following stages (e.g., coding) and could be more expensive the correction. According to Granda et al. [5], defects in conceptual models (e.g., missing, wrong and unnecessary elements) can be located in several ways through Validation & Verification techniques, which can be statically or dynamically supported by a tool and can have different scope and limitation depending on its purpose (i.e., detect, prevent and resolve).

For these reasons, we propose to analyze the influence of negative emotions (focusing mainly on detect stress from physiological data) on the conceptual models verification process. This verification process includes the correction of defects found in the conceptual model.

#### 1.2 Motivation to conduct the study

In this live study proposed, we start from the hypothesis that negative emotions of workers influences on their productivity when they develop model-driven tasks. As the emotional state could be influenced by the user profile (e.g., background, personality, experience in the task) or the way how the emotions are experienced (the task or the person way). For this reason, we propose to analyze the influence of negative emotions using machine learning techniques in the context of software change tasks; also, participants assessed their emotions answering periodically a short questionnaire. Granda et al. [4] replicated the work of Maller and Fritze in the same context, using similar devices (Empatica E4-wristband,

PERSUASIVE 2021

## Towards an Automatic Generation of Persuasive Messages

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**Abstract.** In the last decades, the *Natural Language Generation*(NLG) methods have been improved to generate text automatically. However, based on the literature review, there are not works on generating text for persuading people. In this paper, we propose to use the SentiGAN framework to generate messages that are classified into levels of persuasiveness. And, we run an experiment using the Microtext dataset for the training phase. Our preliminary results show 0.78 of novelty on average, and 0.57 of diversity in the generated messages.

**Keywords:** Text generation · SentiGAN · Persuasive Message

## 1 Introduction

The messages used to persuade a target audience in certain domains (e.g., marketing) have demonstrated to be very effective. However, the production of persuasive message requires a profound knowledge of receptor's characteristics as well as the involvement of a variety of stakeholders. For companies and organizations, this translates into time and money.

Automatic text comprehension and generation are very important in the areas of Natural Language Processing (NLP) and Natural Language Generation (NLG) respectively. In the literature, exist different methods for automatically generating text, which means that its operation does not require human intervention. Despite this growing interest in proposing techniques and models for the automatic text generation (e.g., [11], [14], [8]), there is not yet evidence on automatic message generation that can be used to persuade people. Currently there are some works that address the design of persuasive message for behavior change (e.g., [3]), and others focused on detection of persuasion in texts or conversations [6]. In this paper, we focus on the automatic generation of persuasive messages, through the application of SentiGAN[11], a short text generation framework based on Generative Adversarial Network(GAN). We consider that our work might contribute not only to reduce time, but also to enhance the novelty and diversity of the generated persuasive messages in certain domains. To do this, we exploit the Microtext corpus [13] as a training dataset, which

ICCE 2022

## An Emotion-Aware Persuasive Architecture to Support Challenging Classroom Situations

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**Abstract.**—Nowadays, students face stressful situations that affect their mental health and negatively impact their academic performance. This situation has been aggravated due to the COVID-19 pandemics. Teaching and learning process stakeholders face two main problems: the lack of awareness of the situations arising from students and the little available information that support the teaching activities. Here, we propose an emotion-aware architecture for persuasive systems in education, identifying challenging classroom situations and preventing mental health through multi-modal persuasive notifications. An important module is emotion detection, which reads real-time physiological data from students through sensors. In this architecture, we developed the Kanisqa persuasive system, to support students in regulating their negative emotions (They can all lead to stress) during challenging classroom situations. Preliminary results provide insights that teachers achieved an increased awareness of the situations of students and perceived the prototype tool as useful.

An emotion-aware persuasive system (E-APS) must be aware of the situations (e.g. student is feeling a negative emotion at any moment during the lecture) arising from the actors' context and provide relevant persuasive notifications to change actors' attitudes or behaviors, so that, they can use (read, hear or watch - media type) in a specific context [3]. An emotion-aware prototype system was developed in this architecture, we developed the Kanisqa persuasive system, to support students in regulating their negative emotions (They can all lead to stress) during challenging classroom situations. Therefore, an E-APS is a context-aware [3] and persuasive system [5].

To validate the architecture, we developed the Kusisqa persuasive system, a prototype tool to support students in regulating their negative emotions (stress responses often include negative emotions) in challenging classroom situations.

Preliminary results of evaluation provide evidence that teachers and educational supervisors achieved an increased awareness of the situations of students and perceived the prototype system as an adequate tool to support their coordination while attending to a challenging situation.

Despite low performance are two major issues faced by educational systems of all levels. The pandemic has deeply exacerbated existing education issues, including stress, sleep difficulty, and anxiety [1]. This emotional disturbance poses mental health challenges that negatively impact students' academic performance.

Though many efforts were devoted to monitoring and evaluating the learning process [2], the interventions occur after measuring learning outcomes and it is hard for teachers to be aware of the situations arising from students and to provide the right information, at the right time, in the right place, in the right way, and to the right person.

In this paper, we propose an emotion-aware architecture for persuasive systems in education using a powerful emotion detection module. An Empatica E4 wristband<sup>®</sup> device is used to capture physiological signals of a student. These signals are used as the input to the emotion detection module. An algorithm based on statistical learning is used to identify

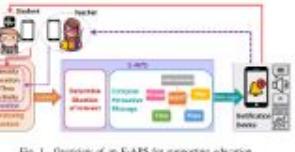


Fig. 1. Overview of an E-APS for supporting education

**II. EMOTION-AWARE PERSUASIVE ARCHITECTURE**  
An E-APS in education domain must consider contextual information about two actors: (i) the student with negative emotions, as the person to be persuaded (receive notification) and about whom the event (acquired context) is generated; and (ii) the teacher and his/her environment, as the person who

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<http://www.concytec.gob.pe/ciclo3/014-2019-FONDECYT-10410019/>

# KUSISQA: resultados

Al menos dos (02) ponencias de alcance internacional  
COMIA 2022

## Una Evaluación Comparativa de Modelos de Deep Learning para el Reconocimiento de Emociones a partir del Habla

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**Resumen.** Diversos modelos de reconocimiento de emociones a partir del habla vienen siendo propuestos en los últimos años. Sin embargo, las evaluaciones de desempeño de algunas propuestas podrían no ser lo suficientemente confiables. Este trabajo tiene como finalidad contribuir con abordar esta problemática, presentando una manera práctica de implementar evaluaciones comparativas cuyos resultados disten de ser atribuidos a la casualidad. La propuesta utiliza las pruebas de significancia estadística no paramétricas en base al test de Wilcoxon para la comparación del desempeño de modelos de Deep Learning. Se demostró la utilidad de la propuesta, realizando las comparación del desempeño de cinco modelos convolucionales, dos entrenados con la base de datos RAVDESS y tres entrenados con la base de datos IEMOCAP.

**Keywords:** Emociones, Evaluación.

### 1. Introducción

El interés en lograr reconocer las emociones humanas de manera automatizada ha sido motivado por diversas áreas de aplicación. Por ejemplo, para diseñar robots inteligentes que puedan interactuar de manera empática con las personas, para crear anuncios personalizados que tengan en cuenta el estado emocional de potenciales clientes, para mejorar los procesos de enseñanza/aprendizaje, para mostrar contenido más adecuado a una audiencia objetivo, entre otras aplicaciones [4].

Existen varios métodos para medir las emociones humanas, cada una con sus ventajas y desventajas. Para el caso de los sistemas automatizados, podemos clasificarlos en invasivos y no invasivos. Entre los invasivos, tenemos aquellos que utilizan dispositivos pegados al cuerpo para medir alguna señal fisiológica como electrocardiogramas, electroencefalogramas, respuestas galvánicas de la piel, temperatura, respiración, entre otros. Entre los no invasivos podemos encontrar a los que utilizan como señal de entrada expresiones faciales, gestos, posturas, voz, entre otras.

IAL  
DIAL

## GloTS 2022

## A Persuasive system for Stress Detection and Management in an Educational Environment \*

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**Abstract.** This paper addresses the development of a persuasive IoT system for stress detection and management in students during classroom situations. An emotion-aware persuasive architecture is developed with four modules: Context Acquisition, Context Manager, Persuasion Manager and Context-Aware Applications. By using the galvanic skin response biomarker, the real-time stress level is measured by the wearable wristband Empatica E4. The data, processed and classified on discrete stress levels from 0 to 5, is sent to the context module that identifies situations of interest where the students need positive reinforcement from the persuasive system. Based on the situation of interest and the user's profile, the persuasive module composes personalized persuasive messages displayed in a mobile application. The persuasive system was evaluated through an exploratory study during a class session, with encouraging results in detecting stress levels and the positive effect of persuasive messages on students.

**Keywords:** Persuasive system - Internet of Things - wearable electronics - student's stress management - emotion-awareness - affective computing

### 1 Introduction

Students are exposed to various stressful situations, from homework to their relationships [11]. Continuous stressful situations can cause psychological and physiological problems in students, leading to reduced academic performance and even school dropout. In the pandemic scenario, conditions causing stress have been accentuated, while others have appeared, such as confinement and remote education [18, 3].

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# KUSISQA: resultados

(Opcional) Para la línea de investigación a crear, fortalecer o consolidar: Al menos un (01) artículo científico, presentado o aceptado para publicación en revistas indexadas, de proyectos que tengan en curso o culminados

## A speech corpus of Quechua Collao for automatic dimensional emotion recognition

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### ABSTRACT

Automatic speech emotion recognition is an important research topic for human-computer interaction and affective computing. Over ten million people speak the Quechua language throughout South America, and one of the most known variants is the Quechua-Collao one; however, this language does not have a speech corpus for automatic emotion recognition, which creates a barrier for Quechua speakers who want to use this technology. Therefore, the contribution of this work is the creation of a speech corpus in Quechua-Collao variant, which is publicly available to the scientific community. The corpus is created from a set of words and sentences explicitly collected for this task, divided into nine categorical emotions. The annotation was performed on a discrete scale according to 3 dimensions: valence, arousal, and dominance. To demonstrate the usefulness of the corpus, we have performed speech emotion recognition using machine learning methods and neural networks.

### Background & Summary

Studies have shown that AI is able to understand and express emotions, improving the quality and effectiveness of Human-Machine Interaction<sup>1,2</sup>. Speech emotion recognition (SER) is useful in affective computing applications such as patient-in-home monitoring<sup>3</sup>, early detection of psychiatric diseases and disorders<sup>4,5</sup>, support diagnosis and treatment in military healthcare<sup>6</sup>, recognize deceptive speech<sup>6,9</sup>, or stress on it<sup>6,11</sup>. Furthermore, SER is used in call center conversations to categorize voice mail messages. Humans naturally can recognize emotions from facial expressions and speech. Experts claim that facial expressions are universal. However, speech is not; there are variations in the vocal signature for some emotions, e.g., anger. This situation is due to differences in the expression of emotions in different cultures and languages<sup>12,13</sup>.

Quechua is a language spoken in seven countries in South America (Peru, Ecuador, Colombia, Bolivia, Argentina, Chile, and Brazil). In Peru, Quechua is considered a vital language. This means it was spoken before the diffusion of the Spanish language and it is used across the country. However, many variants of Quechua are in danger, given that there is a significant decrease in the importance of this language in Peru<sup>14</sup>. This is reflected in the lack of speech datasets of this language when compared, for example, to Spanish or English. There is no Quechua corpus focused on the emotion recognition field. Languages with data scarcity are known as low-resource languages (LRLs)<sup>15</sup>; in this sense, Quechua is a low-resource language.

To the best of our knowledge, there are no works in the literature that focus on creating a corpus in Quechua for SER. However, there are few works that involve the creation of speech corpora for Quechua that deal mainly with automatic speech recognition (ASR). One of these works is Siminich<sup>16</sup>, a speech corpus for the preservation of Southern Quechua, intended to be used in speech recognition systems. It was created from radio programs and volunteers (native Quechua speakers) who made audio transcriptions. A Hidden Markov Model was used to perform ASR. Although Siminich has a larger version<sup>17</sup> that was created by adding more data sources such as recordings of repetitions of audio prompts, readings of text prompts, and free speeches of native speakers, it still seems to be a work in progress<sup>18</sup>. Chacra et al.<sup>19</sup> used Mel-Frequency Cepstral Coefficients (MFCC), Dynamic Time Warping (DTW), and K-Nearest Neighbor (KNN) in an ASR system to recognize numbers in Quechua. For this, isolated words were recorded from 30 native speakers, between men and women, counting from 1 to 10. However, this database was limited as it contained only 300 audios.

Therefore, there is an emotion recognition gap for Quechua speakers. We aim to fill this gap by creating a speech corpus of Quechua Collao for automatic emotion recognition, evaluating it using machine learning techniques and deep neural networks, and making it publicly available. We selected the dimensional approach, based on valence, arousal, and dominance proposed in<sup>20</sup> to indicate people's state of feeling, as categorical emotions can be recovered from dimensional values<sup>21</sup>. Evaluation with machine learning techniques and neural networks will be useful as a baseline for other researchers who want to study the

## Kusisqa: An IoT-based Emotion-Aware Persuasive Architecture to Support Challenging Classroom Situations

Pablo Calicina Ccori, Eduardo S. Rodriguez-Canales, Alvaro Cuno Parati and Edgar Sarmiento-Calisaya

**Abstract—**Nowadays, students face stressful situations that affect their mental health and negatively impact their academic performance. This situation has been exacerbated due the COVID-19 pandemic. Teachers and parents are also vulnerable to two main problems: the lack of awareness of the situations arising from students and the little available information that support the teaching activities. Here, we propose an emotion-aware architecture for supporting teachers during challenging classroom situations and preventing mental health through multi-modal persuasive notifications. An important module is emotion inference, which reads real-time physiological data from students using the Empatica E4 wristband. Based on this architecture, we developed the Kusisqa persuasive system to support students in regulating their negative emotion during challenging classroom situations through automatically generated persuasive messages. The effectiveness of the system at the student level and context. The effectiveness of the system at the student level is evaluated through an exploratory study, where the stress levels of students separated into two groups (Experimental and Control) are measured using the SISCO Academic Stress Inventory. The results show a positive attitude of the students towards the persuasive messages and lower stress levels compared to the control group.

**Index Terms—**Context-Aware Systems, emotion-aware, Persuasive systems, Internet of things, software architecture

### I. INTRODUCTION

The pandemic scenario has compromised the mental well-being of students, increasing cases of stress, sleep difficulty, and anxiety [1]. Changes related to confinement, remote classrooms and interpersonal relationships have caused negative impacts on students' academic performance.

Although many efforts were devoted to monitoring and evaluating the learning process [2], interventions occur after measuring learning outcomes, being hard for teachers to be aware of the situations arising from students and to provide the correct information, at the right time, as the right place, in the right way, and to the right person.

In this paper, we propose an emotion-aware architecture (EA-APS) for persuasive systems in education using a powerful

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<sup>a</sup>https://www.empatica.com/researchkit/



### II. RELATED WORKS

Several works address the development of IoT-based health-care systems for the users' monitoring and managing negative

<https://www.ncbi.nlm.nih.gov/pmc/articles/PMC8698156/>

## Engineering Context-Aware Persuasive Systems: A Scenario-based Approach for Understanding and Modeling Context in Persuasion

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In the last decades, persuasive systems have been developed and applied in different domains (e.g., e-learning, marketing, e-health) in order to influence the feeling, attitude or behavior of users through persuasion-related messages. Special effort has been put on characteristics related to the application domain, psychological strategies used for persuasion and explicit user interaction with the system. However, there is a relatively limited discussion in the literature on how take advantages of context-awareness to support the system to be aware of the environment and adapt its behavior to the user (context-awareness) and personal and intervention, i.e. allowing the persuasion system to be aware of the persuasion context and intelligently adapt the persuasion strategy (foreground and mechanism) to this context at runtime - personalization [26].

The aim is the integration of context-awareness into persuasive systems engineering approaches. However, context-awareness raises a variety of software engineering challenges. In this paper, we address these challenges by proposing a scenario-based approach to tackle them. Specifically, we propose to: 1) understand persuasion contexts (context comprehension); 2) determine what elements of the persuasion environment are relevant to sense context; 3) determine the rules for how the system should adapt the persuasion strategy as different persuasion situations, and 4) base the design on this insights. We also present an engineering approach that can be used to implement with these tools, models and templates.

The proposal is finalized through the development of a persuasive system that allows teachers to detect and mitigate challenging classroom situations. Additionally, the application of the proposal has been evaluated by a team of software engineers experienced with the field of intelligent context-aware or IoT systems design.

The proposed approach will help practitioners and researchers to integrate context-awareness into persuasive systems in a systematic way, enabling personalization and improving persuasion.

**CCS Concepts** → Human-centered computing → Ubiquitous and mobile computing design and evaluation methods; Software and its engineering → Requirements analysis. Software design engineering.

**Additional Key Words** and Phrases: persuasive system, context-aware system, context-awareness, emotion-awareness, requirements engineering, software architecture.

**ACM Reference Format:**  
Edgar Sarmiento-Calisaya, Pablo C. Calcina-Ccori, and Alvaro Cuno. 2022. Engineering Context-Aware Persuasive Systems: A Scenario-based Approach for Understanding and Modeling Context in Persuasion. In ACM, NLP, UNSC, L2 pages (https://doi.org/10.1145/3510953.3511000).

\*Both authors contributed equally to this research.

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Despite e-CPAs have successfully developed and deployed in industry and research, there is still dearth in research about how to change user's attitudes, as Fig. 3.

Despite e-CPAs have successfully developed and deployed in industry and research, there is still dearth in research about how

# KUSISQA: transferencia (equipos)

S/. 195,750



Concytec  
Mundial



# Conclusiones

- Se ha cumplido a cabalidad todas las metas comprometidas entre el FONDECYT y la UNSA
- Se han formado recursos humanos capacitados en Inteligencia Artificial para el reconocimiento de emociones humanas
- Se ha promovido el desarrollo interdisciplinario (computación, psicología, educación, electrónica)

# Recomendaciones

- Formalizar la creación de las líneas de investigación desarrolladas en el proyecto Kusisqa:
  - Ingeniería de Software experimental, convenio COPPE-UFRJ
  - Paralingüística Computacional
  - Internet de las Cosas
  - Machine Learning
- Crear laboratorios de investigación, que establezca el vínculo entre estudiantes, de las Escuelas de Pregrado y Posgrado que dirige la SCC.
- Difundir las investigaciones realizadas de la EPCC, SCC a la comunidad universitaria Agustina y a la comunidad regional.

# Agradecimientos



Vicerrectorado  
de Investigación

# Logistica

# UNSA investiga

# Tesoreria

Integrantes

Monitoras Fondecyt

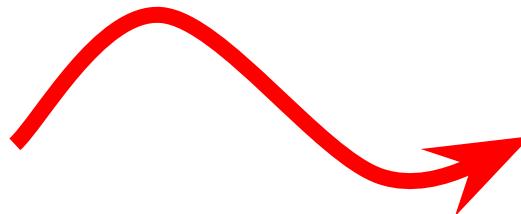


# Próximos pasos

# Próximos pasos (corto plazo)

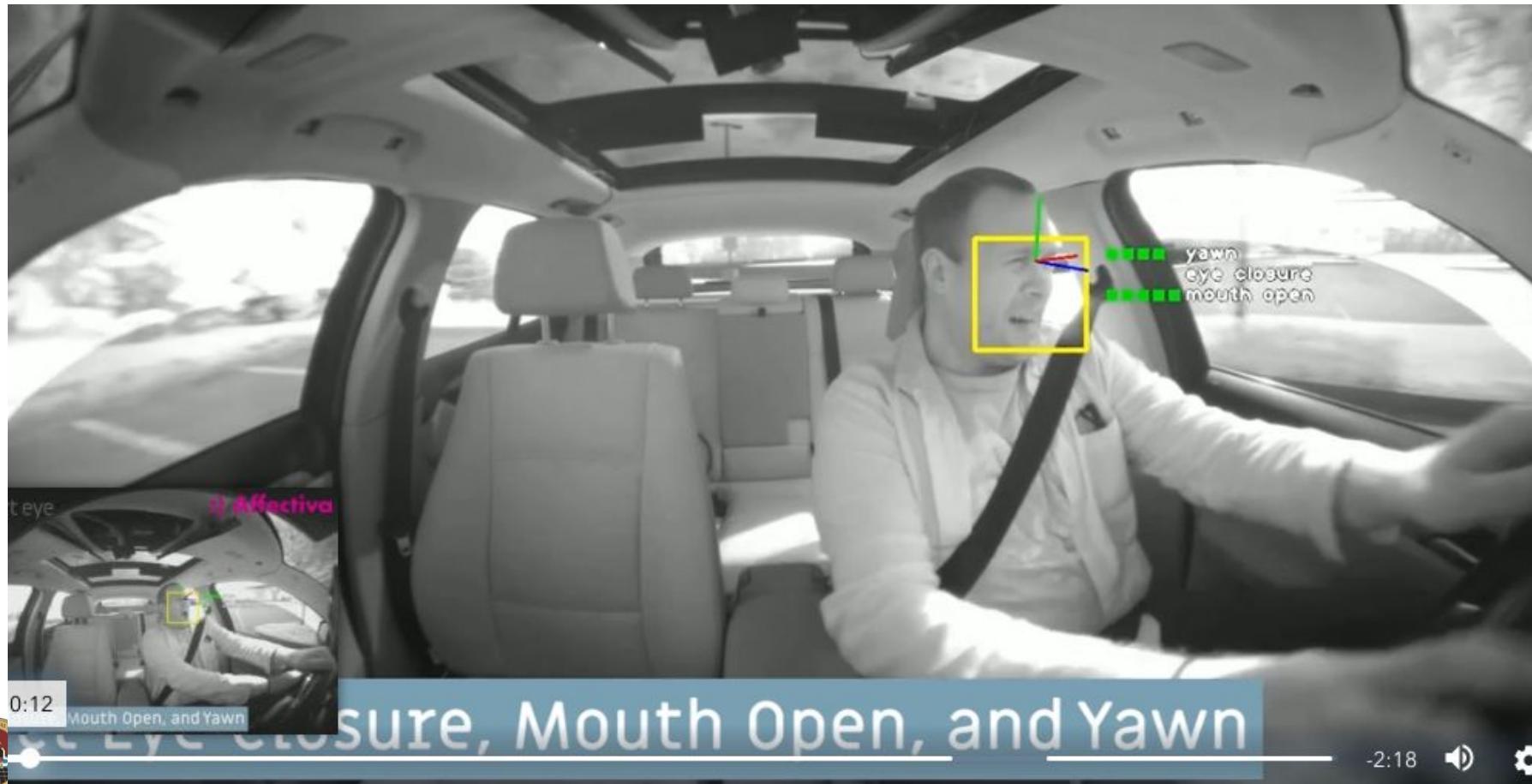


Pulseras E4 (US\$ 2000 aprox.)

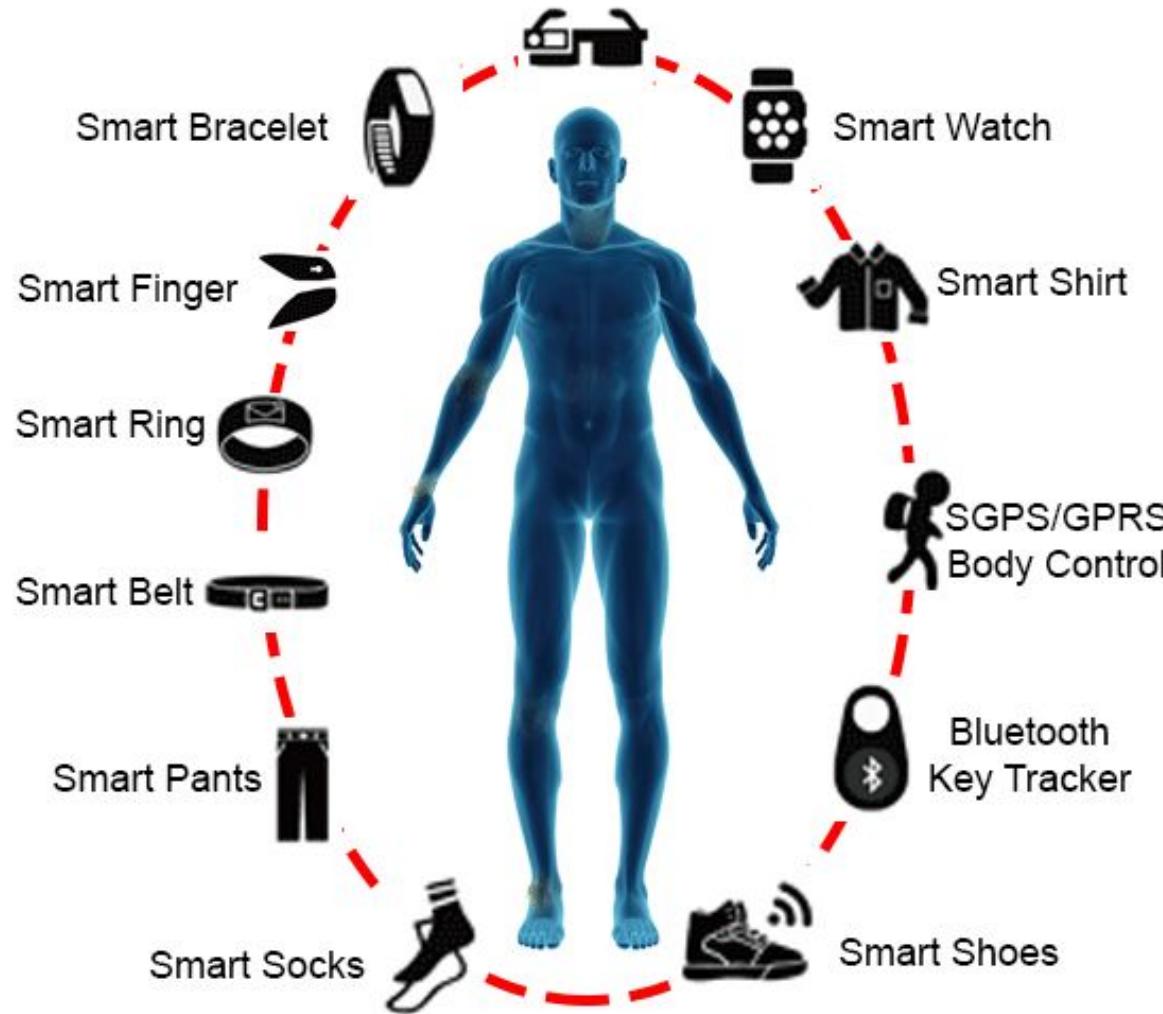


Smartwatch (US\$ 350)

# Próximos pasos (mediano plazo)



# Próximos pasos (largo plazo)



# JUNTOS ESTAMOS TRANSFORMANDO EL PERÚ



Proyecto  
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