

International Journal of Research Publication and Reviews

Journal homepage: www.ijrpr.com ISSN 2582-7421

Detection and Analysis of Human Emotions Through Voice and Speech Pattern Processing

Dr.A.S.Prabaharan¹,M.Kabilan²,R.Karthi³,K.Kathiravan⁴, S.Poornachandaran⁵

¹Supervisor,^{2,3,4}Student(B.E)

DepartmentofInformationTechnology,MuthayammalEngineeringCollege,Rasipuram,TamilNadu,India. kabilanmalayan@gmail.com,poornachandaranfb2000@gmail.com,karthirk9944@gmail.com,kathir2251@g mail.com, prabaharan.as.it@mec.edu.in,

ABSTRACT

The ability to modulate vocal sounds and generate speech is one of the features which set humans apart from other living beings. The human voice can be characterized by several attributes such as pitch, timbre, loudness, and vocal tone. It has often been observed that humans express their emotions by varying different vocal attributes during speech generation. Hence, deduction of human emotions through voice and speech analysis has a practical plausibility and could potentially be beneficial for improving human conversational and persuasion skills. This presents an algorithmic approach for detection and analysis of human emotions with the help of voice and speech processing. The proposed approach has been developed with the objective of incorporation with futuristic artificial intelligence systems for improving human-computer interactions. These systems aim to facilitate the natural interaction with machines by direct voice inter- action instead of using traditional devices as input to under- stand verbal content and make it easy for human listeners to react. Some applications include dialogue systems for spoken languages such as call center conversations, onboard vehicle driving system and utilization of emotion patterns from the speech in medical applications. Emotion recognition from speech signals is an important but challenging component of HCI. In the literature of SER, many techniques have been utilized to extract emotions from signals, including many well-established speech analysis and classification techniques. Deep Learning techniques have been recently proposed as an alternative to traditional techniques in SER. This paper presents an overview of Deep Learning techniques and discusses some recent literature where these methods are utilized for speech-based emotion recognition. The review covers databases used, emotions extracted, contributions made toward speech emotion recognition and limitations related to it.

KEYWORDS:Deep learning, emotional analysis, human emotions, speech processing, voice processing.

Introduction

Deep Learning is continuously amusing us with its modern possibilities like self driving cars, fraud detection, and many more. Earlier we never imagined the things which are possible today and now we cannot even imagine a day without using it. Thus, in this blog, we are going to discuss this very interesting topic "Deep Learning" in much more detail.Deep Learning is at the beginning of what machines can do and developers and business leaders totally need to comprehend what it is and how it functions. Deep learning models are sufficiently competent to focus on the exact features themselves by requiring a little direction from the programmer and are useful in taking care of the issue of dimensionality. Therefore, deep learning algorithms are used, particularly when we have a vast number of inputs and outputs. It is a kind of machine learning that prepares a computer to perform human-like errands, for example, perceiving speech, distinguishing pictures, or making forecasts. Rather than arranging information to go through predefined conditions, deep learning sets up essential boundaries about the information and trains the computer to learn on its own by perceiving designs using numerous layers of processing. Deep learning has networks worthy of learning unsupervised from information that is unstructured or unlabeled. In simple language, deep learning is a type of algorithm

LITERATUREREVIEW

One of the most important information that speech acoustics provide is the expression of emotions. The purpose of this research is to identify the pitch differences between two basic emotions: anger and joy. In order to find answers to this question vocal data have been collected from small group of participants. Results from Friedman" s Two-Way Analysis of Variance by Ranks revealed difference in pitch levels when expressing anger and joy as well as jitter (rap). It is well known that speech is an acoustically rich signal that provides a lot of information about the speaker during vocal

interaction. The expression and recognition of emotions are extremely important steps for human communication process for this reason voice recognition is useful for detecting and identifying specific affective characteristics between the speakers.

PROPOSEDSYSTEM

Deduction of human emotions through voice and speech analysis has a practical plausibility and could potentially be beneficial for improving human conversational and persuasion skills. This paper presents an algorithmic approach for detection and analysis of human emotions on the basis of voice and speech processing. Three test cases have been examined, corresponding to the three emotional state: normal emotional state, angry emotional state, and panicked emotional state. Each case demonstrates characteristic associated vocal features which can help in distinguishing the corresponding emotional state. We examine the effectiveness of applying machine learning techniques to the sentiment classification problem. A challenging aspect of this problem that seems to distinguish it from traditional topic-based classification is that while topics are often identifiable by keywords alone, sentiment can be expressed in a more subtle manner. For example, the sentence "How could anyone sit through this movie?" contains no single word that is obviously negative. Thus, sentiment seems to require more understanding than the usual topic-based classification. So, apart from presenting our results obtained via machine learning techniques, we also analyze the problem to gain a better understanding.

Approach To Detection Of Human Emotions Algorithm

This section describes an algorithmic approach for deducing human emotions through voice- and speech-pattern analysis. In order to achieve this objective, three test cases have been examined, corresponding to the three emotional states: **Normal** emotional state, **Angry** emotional state, and **Panicked** emotional state. \Box For carrying out the analysis, four vocal parameters have been taken into consideration: pitch, SPL, timbre, and time gaps between consecutive words of speech. In order to quantitatively represent timbre, its temporal envelope for advance and decay times has been considered. Its a different emotional states by analyzing the deviations in the aforementioned four parameters from that of the normal emotional state. The proposed analysis was carried out with the help of software packages such as MATLAB and Wavepad.

SYSTEM SPECIFICATIONS

HARDWARE REQUIREMENTS

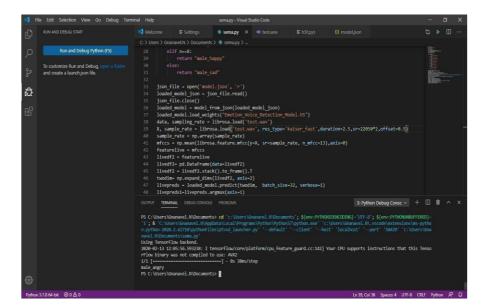
| Processor | : Intel processor 3.0 GHz |
|--------------|---------------------------|
| RAM | : 1GB |
| Hard disk | : 40 GB |
| Compact Disk | : 650 Mb |
| Keyboard | : Standard keyboard |
| Mouse | : Logitech mouse |
| Monitor | : 15-inch color monitor15 |
| | |

SOFTWARE REQUIREMENTS

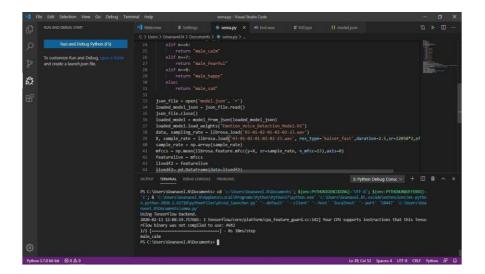
| Operating System | : Windows OS | | | | |
|--------------------|-------------------------------------|--|--|--|--|
| System type | : 32-bit or 64-bit Operating System | | | | |
| IDE | : Python 3.5 and above | | | | |
| Install Dependancy | : pip install pandas, | | | | |
| | pip install matplotli, | | | | |
| | Pip install keras, | | | | |

pip install tensorflow

1. RESULT



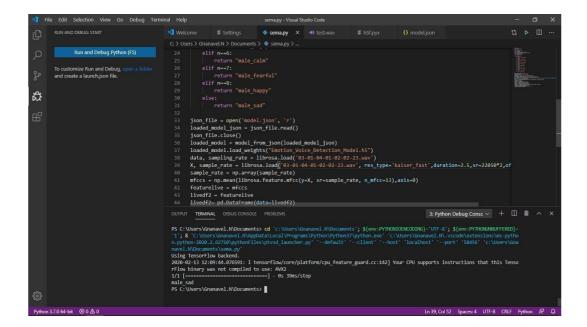
OUTPUT 1



OUTPUT 2

| 0 م | RUN AND DEBUG: START Run and Debug Python (F5) | | E Settings nanavel.N > Documents | 🔹 sema.py 🛛 🗙 | | ≣ h5f.pyx | | | | | |
|----------|---|---|--|---------------|--|-----------|--|--|--|--|--|
| | Run and Debug Python (F5) | | nanavel.N > Documents | | | = пэтрух | | | | | |
| Q | Run and Debug Python (F5) | | | > 🕏 sema.py > | | | | | | | |
| | | | elif n6: return "male_ca | | | | | | | | |
| ኑ ድ | To customize Run and Debug, open a folder and create a launch.json file. | | elif n==7: return "male_fe elif n==8: return "male_he | | | | | | | | |
| ler ₽ | | 31 32 33 json_ 34 loade | else: return "male_sa _file = open('model ed_model_json = jso file closo() | | | | | | | | |
| | | <pre>35 json_file_lose() 36 loaded_model = model_from_json(loaded_model_json) 37 loaded_model.load_weights("Emotion_Voice_Detection_Model.h5") 38 data, sampling_rate = lihrosa.load("d3-d1-d2-d3.vav") 39 X, sample_rate = linrosa.load("d3-d1-d2-d3.vav") 40 sample_rate = np.array(sample_rate) 41 mfccs = np.man(librosa.feature.mfcc(y-X, sr=sample_rate, n_mfcc=13),axis=0) 42 featurelive = mfccs 43 livedf2 = featurelive 44 livedf2 = featurelive 41 livedf2 = featurelive</pre> | | | | | | | | | |
| | 7704-ња ⊗9∆0 | OUTRUT TERMANAL DEBUG CONSOLE PROBLEMS 3: Python Debug Consol + III * < × PS C:\USers\\Gnamavel.N\Documents> cd 'c:\Users\\Gnamavel.N\Documents'; \$ {env:PYTHONIDENCEDING)-'UT-6'; \$ {env:PYTHONIDENFERED}- '1'; 4' C:\Wers\\Gnamavel.N\Documents> cd 'c:\Users\\Gnamavel.N\Documents'; \$ {env:PYTHONIDENCEDING)-'UT-6'; \$ {env:PYTHONIDENFERED}- '1'; 4' C:\Wers\\Gnamavel.N\Documents> cd 'c:\Users\\Gnamavel.N\Documents'; \$ {env:PYTHONIDENCEDING)-'UT-6'; \$ {env:PYTHONIDENFERED}- '1'; 4' C:\Wers\\Gnamavel.N\Documents> cd 'c:\Users\\Gnamavel.N\Documents'; \$ {env:PYTHONIDENFERED}- '1'; 4' C:\Wers\\Gnamavel.N\Documents\users\Documents'; \$ {env:PYTHONIDENFERED}- '1'; 4' C:\Wers\\Gnamavel.N\Documents\Users\Gnamavel.N\Documents'; * env: 'c:\Users\\Gnamavel.N\Documents'; * env: 'env: 'users\\Gnamavel.N\Documents'; * env: 'env: 'en | | | | | | | | | |

OUTPUT 3



OUTPUT 4

CONCLUSION

Building the model was a difficult undertaking as it included parcel of trail and mistake strategies, tuning and so on. The model is very much prepared to recognize male and female voices and it recognizes with 100% exactness. The

model was tuned to recognize feelings with over 70% precision. Exactness can be expanded by including more sound records for preparing.

Acknowledgements

Iwouldlikehumblyacknowledgethesupportofthemanagement, seniorlecturers, GuideandECE departmentof Muthayammal Engineering College, Rasipuram , Tamilnadu rendering the thankfulness which enabled us in proceeding with researchwork, in-line with the present social needs and trends.

REFERENCE

Poorna Banerjee Dasgupta, Detection and Analysis of Human Emotions through Voice and Speech Pattern Processing, International Journal of Computer Trends and Technology (IJCTT), 2017.

Li Deng, Jinyu Li, Jui-Ting Huang, Recent advances in deep learning for speech research at Microsoft, Microsoft Corporation, 2017.

3. Elka Popova, Ilona Isaeva, Emotion recognition through voice analysis, Tulborg University, 2015.

4. Jason Lee Wright, Voice to Text Conversion, US Patent application publication, 2015.

5. Andrew L. Maas, Raymond E. Daly, Peter T. Pham, Dan Huang, Andrew Y. Ng., Learning Word Vectors for Sentiment Analysis, 49th Annual Meeting of the Association for Computational Linguistics, 2011.

6. Theresa Wilson, Johanna Moore, Twitter Sentiment Analysis: The Good the Bad and the OMG!, Proceedings of the Fifth International AAAI Conference, 2011.

7. Michael L. Asmussen, Advanced set top terminal having a voice-to text conversion, IEEE, 2007.

8. Thomas E. Creamer, Peeyush Jaiswal, Voice-to-text reduction for real time in/chat/ sms, US Patent application Publication, 2004.

9. Gary L. Griffith, Communication device having voice and text message, United States Patent, 2002.

10. Bo Pang and Lillian Lee, Thumbs up? Sentiment Classification using Machine Learning Techniques, Proceedings of the Conference on Empirical Methods