

A Novel Approach for Detection of Depression Using Speech Analysis by Applying Convolutional Neural Networks (CNN)

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Abstract: Mental illness has now become more prevalent in the world. Depression is one such illness. As per World Health Organization (WHO), many individuals are likely to put up with depression, and that rate is globally increasing, especially at progressive age. The absence of objective measures and use of traditional techniques are not much effective in predicting mental health of an individual. Hence depression to be usually under-diagnosed but it is also most curable illness. Recent studies have revealed that speech is a sensible indicator of depression syndrome, this giving us an incentive to carry out depression diagnosis by using speech to form an associate degree objective measure. Building on the ideas, a supervised machine learning (ML) model using ensemble is built in identifying whether person is depressed or not by using audio attributes or features of audio datasets. The CNN is being used to train the useful attributes for depression classification from speech. The datasets used for the purpose of model training and testing are taken from Surrey Audio-Visual Expressed Emotion (SAVEE) and Toronto Emotional Speech Set (TESS). The features like MFCCs, spectrograms of the audio recordings and related depression criterion are extracted for audio classification using CNN model.

Keywords: Convolutional Neural Network, Artificial Neural Networks, Mel Frequency Cepstral Coefficient, spectrograms, Speech, Affect, SAVEE, TESS.

REFERENCES

- [1]. M. Hamilton, The Hamilton rating scale for depression, In Assessment of Depression. Springer: Berlin/Heidelberg, Germany, pp. 143–152, 1986.
- [2]. A.T. Beck, R.A. Steer, M.G Carbin, “Psychometric properties of the Beck Depression Inventory: Twenty-five years of evaluation,” Clinical Psychology Review, 8, pp. 77–100, 1988. [CrossRef]
- [3]. N. C. Andreasen, Scale for the Assessment of Negative Symptoms (SANS), The British Journal of Psychiatry, 1989.
- [4]. N. W. Hashim, M. Wilkes, R. Salomon, J. Meggs, & D. J. France, “Evaluation of Voice Acoustics as Predictors of Clinical Depression Scores,” Journal of Voice, 2016.
- [5]. F. Tolkmitt, and K. Scherer, “Effect of experimentally induced stress on vocal parameters,” J. Exp. Psychol, vol. 12, pp. 302–313. 1986, doi: 10.1037/0096-1523.12.3.302.
- [6]. J. Mundt, A. Vogel, D. Feltner, and W. Lenderking, “Vocal acoustic biomarkers of depression severity and treatment response,” Biol. Psychiatry, vol. 72, pp. 580–587, 2012, doi: 10.1016/j.biopsych.2012.03.015.
- [7]. E. Kraepelin, “Manic depressive insanity and paranoia,” The Journal of Nervous and Mental Disease, vol. 53, no. 4, pp. 350, 1921.
- [8]. T. F. Quatieri and N. Malyska, “Vocal-source biomarkers for de-pression: A link to psychomotor activity,” 13th Annual Confer-ence of the International Speech Communication Association, pp. 1058–1061, 2012.

- [9]. Girija Deshmukh, Apurva Gaonkar, Gauri Golwalkar, Sukanya Kulkarni, “Speech based Emotion Recognition using Machine Learning”, Institute of Electrical and Electronics Engineers, Mar. 2019.
- [10]. N. Cummins, S. Scherer, J. Krajewski, S. Schnieder, J. Epps, & T. F. Quatieri, “A Review of Depression and Suicide Risk Assessment Using Speech Analysis,” *Speech Communication*, vol. 71, pp. 10–49, 2015.
- [11]. Afshan, Amber, Jinxi Guo, Soo Jin Park, Vijay Ravi, Jonathan Flint, and Abeer Alwan, “Effectiveness of voice quality features in detecting depression,” in *Proceedings of Interspeech*, pp. 1676-1680, 2018.
- [12]. S. Scherer, G. Stratou, G. Lucas, M. Mahmoud, J. Boberg, J. Gratch, A. S. Rizzo, and L.-P. Morency, “Automatic audiovisual behavior descriptors for psychological disorder analysis,” *Image and Vision Computing*, vol. 32, no. 10, pp. 648–658, Oct 2014.
- [13]. N. Cummins, S. Scherer, J. Krajewski, S. Schnieder, J. Epps, and T. F. Quatieri, “A review of depression and suicide risk assessment using speech analysis,” *Speech Communication*, vol. 71, pp. 10–49, 2015.
- [14]. K. J. Piczak, “Environmental sound classification with convolutional neural networks,” in *Proceedings of the 2015 IEEE 25th International Workshop on Machine Learning for Signal Processing (MLSP)*, Boston, MA, USA, 17–20 September 2015, pp. 1–6.
- [15]. T. Nguyen, F. Pernkopf, “Acoustic scene classification using a convolutional neural network ensemble and nearest neighbor filters,” in *Proceedings of the Detection and Classification of Acoustic Scenes and Events 2018 Workshop (DCASE2018)*, Surrey, UK, 19–20 November 2018, pp. 34–38.
- [16]. X. Ma, H. Yang, Q. Chen, D. Huang, Y. Wang, “DepAudioNet: An Efficient Deep Model for Audio Based Depression Classification,” in *Proceedings of the 6th International Workshop on Audio/Visual Emotion Challenge, AVEC '16*, Amsterdam, The Netherlands, 16 October 2016; ACM: NY, USA, pp. 35–42, 2016.
- [17]. Y. Bengio, “A connectionist approach to speech recognition,” *International Journal of Pattern Recognition and Artificial Intelligence*, vol. 07, no. 04, pp. 647–667, Aug 1993.