Covid-19 Cough Classification Using Deep Learning

Dr. K Rama Abirami^[1], Gazala Fathima^[2], Anam Sayeeda S^[3], Natasha C^[4]

^{1,2,3}BE Students, ⁵Professor, Department of Information and Science, Davananda Sagar Academy of Technology and Management, Bengaluru – 560082

Abstract-COVID-19 was declared a pandemic on February 11, 2020, with the support of the World Health Organization (WHO). The old COVID-19 predictions obtained exposure to COVID-19 with coughs when the process of analyzing cough samples was performed using different voice output methods and different phases of machine learning. Breath sounds need to be analyzed. .. The main focus of this paper is on the classification of important cough-related audio lessons learned in in cough detection and chain disease, based on the amount, duration, and intensity of cough samples.

Keywords-COVID-19, cough classification, cough analysis, cough signals, MFCC.

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I. INTRODUCTION

COVID-19 (2019 coronavirus disease) caused by Acute Respiratory Syndrome 2 (SARS-CoV2) was declared a global epidemic by the World Health Organization (WHO) on February 11, 2020. It has SARS-CoV (severe acute respiratory pattern coronavirus) and MERS-CoV (Middle East respiratory pattern coronavirus) and occurred independently in 2002 and 2012 compared to other coronaviruses . Common symptoms of COVID-19-Fever, fatigue and dry cough [3]. Other symptoms include shortness of breath, general pain and pain, myalgia, abdominal symptoms, and loss of smell and taste [4]. Many attempts have been made to identify the initial symptoms of COVID-19 using artificial intelligence. Cough is one of the most serious signs and symptoms of COVID-19 [5], a symptom of over 100 different illnesses, and its respiratory effects [6]. The number of cases (31.7 million) and the number of deaths are reported to be the highest [7]. The spread of this epidemic outweighs other healthcare systems that require experimental and surgical cases [8]. Cough is always a symptom of many illnesses. It is possible to distinguish between cough and diagnosis by examining logical features with an interdisciplinary approach. An overview of how to obtain Covid-19 can be included. When a person becomes infected with Covid-19, the interaction of breathing speech, voice sounds, and verbal neural movements changes. Therefore, our audio processing model has been introduced to track both Covid-19s . Voice processing of breath sounds and cough sounds can also be monitored with the Covid-19 idea. For the high-risk crop COVID-19, early detection plays an important role because affected cases can already be classified according to opinion the need for interaction between COVID-19 and non-COVID-19. Given that-a case of COVID 19 by pattern analysis Cough is a normal kick of the immune system, cleaning the airways and preventing contaminants from entering the airways [10].Dry cough is one of the main symptoms of COVID-19, so high body temperature can be used as a diagnostic tool. Cough is associated with the sound of traits, and in this article, the cough / pattern of COVID-19 causes congenital malformations at various times. Other methods include lab tests for aging tar, nasal tar, and natural blood tests, but results are up to 2 days in other tests such as advanced RTCRT (Rear Recap Polymerase Chain Response) tests and antigen tests. It takes. 30 blinks will give you results Use DSP chipsets and equipment with built-in clamp algorithm and firm grip to reduce chemical laboratories and avoid the production of chemical waste and toxins You can then process the coughing sound and view the results in seconds.

This article introduces the concept of cough analysis under conditions and classifies waves by various parameters to distinguish them from COVID-19 and healthy human sounds [10]. The data series for COVID-19 is complex and the dataset usually does not exist without difficulty in identifying it. However, much research has been done to find comparable datasets.

II. METHODOLOGY

• DATA

The dataset used : Kaggle dataset

Kaggle is the world's largest data science community with powerful tools and resources to help you reach your data science goals. Participants coughs were collected were collected. This dataset, also known as CoughVid, contains over 25,000 crowd source cough records that represent a wide range of participants' age, gender, geographic location, and COVID-19 status. A subset of 2,800 records were labeled by four experienced

physicians for the medical diagnosis of cough. Abnormal. It contributes to one of the largest expert-labeled cough datasets in the current literature, which can be used for various cough speech classification tasks. There were several covid and non-covid cough audio files which were available and achieved very good number of audio files .For this work , a huge number of cough audio is been used.

III. FEATURE EXTRACTION

Sound waves have a set of parameters called audio properties. Determining these properties is an important step that affects the accuracy of your system. Figure 2 shows the flow chart of the proposed system.

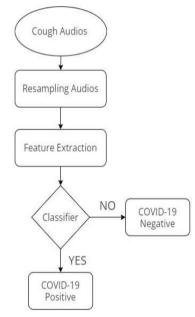


Fig.2 Flowchart of the proposed system

• Mel-Frequency Cepstral Coefficients (MFCC):

MFCC is an important function mainly used in the field of emotion recognition for highly expressing human hearing [7, 8, 9, 10]. In addition, the MFCC was calculated using a psychologically motivated filter bank, followed by logarithmic compression and the Discrete Cosine Transform (DCT). Feature extraction is considered an important step because one of our goals is to find the right feature extraction method to improve the accuracy of the system.

IV. NETWORK ARCHITECTURE

• Deep Learning :

Deep learning models are created using neural networks. There is an input layer, a hidden layer, and an output layer. The input layer receives the inputs, the hidden layer processes these inputs with weights that can be fine-tuned during training, and the model outputs predictions [11].

Deep learning models are primarily used to predict future data sequences using previous data samples.

V. RESULTS AND EXPERIMENTS

After reading all the research papers, I understood how to analyze cough samples and what can be used as the best way to detect COVID-19. Therefore, the analyzed the cough sounds from the and concluded that it was an extensible speech processing platform that categorized them into different categories. You can configure multiple parameters for the sound processing platform. This includes different feature extraction and classification methods. The results of this paper provided new perspectives and insights for the classification of cough. The goal is to improve the current test system. Future work will focus on testing various platforms for the classification of COVID-19 cough sounds.

VI. CONCLUSION

The paper presented a new approach in classifying COVID19 cough samples that could be the first step in introducing contactless identification of COVID-19 patients. Here the machine learning model uses deep learning to predict as well as and give good accuracy of the proposed system. Given that the work initiated is part of a large project to integrate machine learning classification or COVID-19 recognition software, the results will be interesting from the point of view of the underlying model choice.

REFERENCES

- [1]. https://www.who.int/emergencies/diseases/novel-coronavirus-2019
- [2]. [3]. https://www.kaggle.com/himanshu007121/coughclassifier-trial
- Pahar, M., Klopper, M., Warren, R., & Niesler, T. (2021). COVID-19 cough classification using machine learning and global smartphone recordings. Computers in Biology and Medicine, 135, 104572.
- [4]. https://github.com/virufy/virufy_data/tree/main/clinical/segmented
- Van Hulse, J., Khoshgoftaar, T. M., & Napolitano, A. (2007, June). Experimental perspectives on learning from imbalanced data. [5]. In Proceedings of the 24th international conference on Machine learning (pp. 935-942).
- [6]. Krawczyk, B. (2016). Learning from imbalanced data: open challenges and future directions. Progress in Artificial Intelligence, 5(4), 221-232.
- [7]. Shahin, I. (2012). Studying and enhancing talking condition recognition in stressful and emotional talking environments based on HMMs, CHMM2s and SPHMMs. Journal on Multimodal User Interfaces, 6(1), 59-71.
- [8]. Shahin, I. M. A. (2013). Employing both gender and emotion cues to enhance speaker identification performance in emotional talking environments. International Journal of Speech Technology, 16(3), 341-351.
- [9]. Shahin, I. (2010). Employing second-order circular suprasegmental hidden Markov models to enhance speaker identification performance in shouted talking environments. EURASIP Journal on Audio, Speech, and Music Processing, 2010, 1-10.
- [10]. Shahin, I. (2016). Employing emotion cues to verify speakers in emotional talking environments. Journal of Intelligent Systems, 25(1), 3-17.
- [11]. https://www.educba.com/deep-learning-model/