

A Review on Intelligent Pathological Voice Disorders Detection Using Neural Network

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Abstract: In this paper we present the voice signal is taken of different patients for eg. 100 patients, male and female both than the output of this patients is passed through processing in which preemphasis filter is present which remove low frequency signal than we flatten the spectrum due to this features comes. As voice signal is non stationary so we have to divide into small frames due to this data is not lost than we overlap this frames due to this the data comes in continuous way and we get best features. In present we have to analysis disorder like Dysarthria and Balbuties which is also known as Stuttering Than the output of this is given to feature extraction in which MFCC, LFCC, ZCR, Energy is present all this features give informative characteristic of patients and also in arithmetic values. Than the output of this is given to classifier which is a neural network, this classifier does all similar diseases patients

Key Words: Pathological voices, MFCC, LFCC, Voice Disorder, Artificial Neural Network.

1. INTRODUCTION:

Voice pathology can be caused by various reasons which can be psychological or organic. The available diagnosis of voice or speech pathologies depends on experienced medical practitioners, Psychological or Organic disease affects speech of human being and frequent and all the practitioners are not that expert so to get confirmation and second opinion clinicians can use such automation. Medical expert system helps doctors to determine definitive diagnosis or a range of alternative diagnosis Computer still finds difficult to automatically analyze human voice so this automated program requires the features of human speech which can easily be obtained by recording the speech of a patient. The conventional approach to build medical expert system requires the formulation of rules by which the input data can be analyzed. But the formulation of such rules is very difficult with large sets of input data. Realizing the difficulty, pattern classification technique such as artificial neural networks (ANN) has been applied as an alternative to conventional rule based expert system. The system which we are going to design is given in the block diagram.

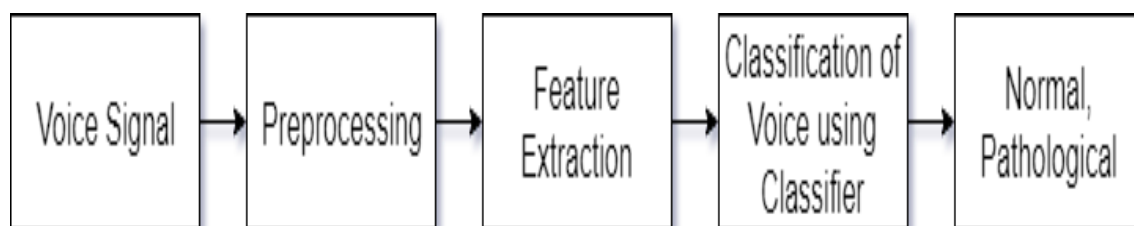


Fig.1.1 Block diagram of intelligent detection of voice pathology using classifier

Hence ANN has been found to be more helpful than a traditional medical expert system in the diagnosis of diseases. In the system various needs to be designed.

There are various disease in voice of patients is affected. We will be choosing pathological & normal voice signal according to need of society from German voice disorder database. This German voice disorder database developed by Putzer contain healthy and pathological voice, where each one pronounced vowels [i, a, u] for 12 s in wav format. [12]. These files are sampled at 50 KHz..

In pre-processing chosen voice signals are passed through pre-emphasis filter which will remove noise and flatten voice signal spectrum thus helps in getting informative features of voice signal. Voice signals are non stationary so they are processed by making small frames of 20ms to 100 ms. Such frames will give more correct information. These framed voice signal is passed through window as window smoothens the signal.

Features like MFCC, LFCC, Energy, ZCR, etc. may be extracted. Mel Frequency Cepstral Coefficients (MFCC) & Linear Frequency Cepstral Coefficients are most commonly used feature extraction method in speech recognition system. Pitch is nothing but frequency of voice signal & Formants are resonance in the voice spectrum. Selection of appropriate feature is important task which is also called principal component analysis.

Appropriate feature will be given to ANN classifier for classification.

2 . LITERATURE REVIEW:

D. Razzouk, , et al., 2006] have suggested a Clinical decision support systems, a useful tools for assisting physicians to diagnose complex illnesses. Schizophrenia is a complex, heterogeneous and incapacitating mental disorder that should be detected as early as possible to avoid a most serious outcome. These artificial intelligence systems might be useful in the early detection of schizophrenia disorder. The results showed a relatively low rate of misclassification (18-34%) and a good performance by the diagnosis of schizophrenia, with an accuracy of 66-82%.

According to [Vahid Majidnezhad, Igor Kheidorov,2013] A new type of feature vector, based on wavelet packet decomposition and Mel-Frequency-Cepstral-Coefficients (MFCCs), is proposed. Also Principal Component Analysis (PCA) is used for feature reduction. An Artificial Neural Network is used as a classifier for evaluating the performance which yielded accuracy of 91.54% [J.Nayak, et al., 2005] classified the normal, abnormal and hyper function using continuous wavelet transform. The author has used three layered feed forward network with sigmoid activation function and trained the neural network for three classes giving efficiency in the range of 80-85% accuracy.

[Lofti Salhi, et al., 2010] have proposed a method to classify normal and pathological voice using multilayer neural network and using energy coefficients of wavelet transform with accuracy of classification 80 to 100 %.

According to [M.Hariharan, et al., 2010] time domain features based on energy variations are extracted from speech to form feature vector. They have used probabilistic neural network(PNN) for the classification of two class normal Pathological with classification accuracy of 96.42 % and 99.4 % respectively. [Jung-Won Lee, et al., 2013] proposes an efficient feature extraction method for automatic diagnosis systems to detect pathological subjects using continuous speech. Experimental results show that the projected method improves the classification error rate by 11.2% (relative) compared to the conventional method using HNR. [Karthikzeyan Umapathy, et al., 2005] have proposed joint time-frequency approach for classifying pathological voices using continuous speech signals that removes the need of segmentation. In his research speech signals were decomposed using an adaptive time-frequency transform algorithm. The classification was done by linear discriminant analysis .and have classified normal and pathological speech with the classification accuracy of 93.4%.

Research work of [Nicolas saenz-Lechon, et al., 2006] described some methodological concerns to be considered when designing systems for automatic detection of voice pathology, in order to enable comparisons to be made with previous future experiments. In their work tthey have classified two classes pathological and normal signal using 18 MFCC coefficients with percentage classification as 92.73 and 86.34 respectively. The system has accuracy of 89.6%. [Vanitha S., Bharthi.D, 2016] have classified parkinson disease form normal .The classifier used was support vector machine with linear kernel. According to them the accuracy of the classification depends on the voice data samples, voice features and their number. Six voice features are used and 99.6% accuracy is achieved.

According to [Abdullah Caliskan, et al, 2017] classified parkinson disease with accuracy of 93.79% . For this purpose, a deep neural network classifier, which contains a stacked autoencoder and a softmax classifier, is proposed. The several simulations are performed over two databases to demonstrate the effectiveness of the deep neural network classifier. The results of the proposed classifier are compared with the results of the state-of-art classification method. The experimental results and statistical analyses are showed that the deep neural network classifier is very efficient classifier for Parkinson disease diagnosis.

According to research work of [Alexandra Konig, et al., 2015] The classification accuracy of automatic audio analyses were as follows: between Healthy elderly control HCs and those with Mild cognitive impairment(MCI), 79% between HCs and those with Alzheimer disease(AD), 87% and between those with MCI and those with AD, 80% demonstrating its assessment utility.

3. DETAILS OF DISEASES:

Dysarthria : Dysarthria is a speech disorder caused by muscle weakness. Dysarthria is a motor speech disorder resulting from neurologic impairment affecting mainly the control and execution of movements related to speech production. Occurrence of dysarthria in adult age is commonly manifested as a consequence of degenerative disorder such as Parkinson's disease (PD), Huntington's disease (HD), multiple system atrophy (MSA), progressive supranuclear palsy (PSP) or cerebellar ataxia (CA). Interestingly, identification of specific deviant speech characteristics can provide important clues about the underlying pathophysiology and localization of neurological diseases. Speech may also serve as a valuable marker of disease onset or treatment efficacy. Therefore, the main aims of this doctoral thesis were (a) to design the feasible algorithms, methodologies or measurements that would be sensitive and accurate enough to capture pathological changes in speech, (b) to objectively quantify the effect of neurological disorder on speech production and (c) to relate the potentially observed speech changes to overall motor performance or medication doses in order to provide deeper insight into the pathophysiology of speech disturbances.

Stuttering (Balbuties) :Stuttering is a speech disorder characterized by repetition of sounds, syllables, or words; prolongation of sounds; and interruptions in speech known as blocks. An individual who stutters exactly knows what

he or she would like to say but has trouble producing a normal flow of speech. These speech disruptions may be accompanied by struggle behaviors, such as rapid eye blinks or tremors of the lips. Stuttering can make it difficult to communicate with other people, which often affects a person's quality of life and interpersonal relationships. Stuttering can also negatively influence job performance and opportunities, and treatment can come at a high financial cost.

Symptoms of stuttering can vary significantly throughout a person's day. In general, speaking before a group or talking on the telephone may make a person's stuttering more severe, while singing, reading, or speaking in unison may temporarily reduce stuttering.

4. METHOD:

Symptoms of Generally a Disease Detection includes the following components.

Speech waveform capture (analog to digital conversion)

The a-to-d conversion is generally accomplished by digital signal processing hardware on the computer's sound card (a standard feature on most computers today). The typical sampling rate, 8000 samples per second, is adequate. The spoken voice is considered to be 300 to 3000 Hertz. A sampling rate 8000 gives a Nyquist frequency of 4000 Hertz, which should be adequate for a 3000 Hz voice signal.

Pre-emphasis filtering

Because speech has an overall spectral tilt of 5 to 12 dB per octave, a pre emphasis filter of the form $1 - 0.99z^{-1}$ is normally used. This first order filter will compensate for the fact that the lower formants contain more energy than the higher.

Feature Extraction

MFCC: For short time process, preprocess is again full applied for speech signals. Another MATLAB voice toolbox function "melbankm" is used for mel scale filter bank wrapping in Eq. (7). For a chosen 256 points FFT, 32 filters are used in filterbank. Then Fast Fourier transformation is carried out use "fft" command and "log" for the logarithm. After the discrete cosine transform carries out, wipe out the first and the last frame.

Classification with Neural Network Enter "nnstart" to start the MATLAB GUI of neural network. It has several functions and choose pattern recognition tool then the general working to load input data and target data. Input data here which indicates the state

5. CONCLUSION:

We will implement whole methodology in matlab using signal processing and neural network toolbox with extracting the feature we mention .

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