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# A Critical Review on Automatic Speaker Recognition

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**Abstract:** Automatic Speaker Recognition (ASR) is used to recognize persons from their voice. Since the voice of every human is not the same because their vocal tract shapes, larynx sizes and other parts of a human voice production system. Automatic Speaker recognition is a procedure to automatically recognize a speaker or who is speaking by the individual information contained in speech signal/waves. Automatic speaker recognition technique makes it possible to use the speaker's speech to verify their identity. It has many applications for example control access to services such as voice mail, voice dialing, banking by telephone, remote access to computers, telephone shopping, information services, database access services and security control for confidential information areas.

**Keywords:** Speaker Recognition, Prosodic, MFCC, Pre-Processing

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## 1. Introduction

Today's biometric information is used to distinguish between different persons' identities. Normally it may be considered that face is one of the most important features to recognize someone but through biometric there are some other unique features, such as retina, voice, finger-prints, iris, are often used [1]. A different way to recognize a person is from the voice/acoustic since each person's voice is different, and this is the one area of speech processing, automatic speaker recognition [2]. The fundamental research to discover innovative speaker individual features is to encode them into more affluent and more informative speaker models. Also to evaluate the effectiveness of these speech features and speaker models for speaker recognition [3]. There are many ways to extract the features from speech signal such as pursuing mutually a knowledge-based, building on existing linguistic constructs and directed by perceptions from psycholinguistics and human performance studies also a more speculative data-driven approach, seeking idiosyncratic vocal performances; spectro-temporal patterns with high speaker characterizing power, independent of linguistic constraints etc. [4][5][6]

In recent speech and speaker recognition have an important research and development area in biometric. The development here is the desire to create a natural form of communication between human and machine. As it is well accepted that speech is our most natural appearance of

communication, ASR has the capability to impact on numerous fields of research and development [6]. Prosodic features play an important role to retrieve features from human speech. The information contained by prosodic features is different from information contained in cepstral features. Prosodic features are extracted by larger frame size while acoustical features are extracted by small window. Since prosodic features exist over a long speech segment such as syllables. Prosody can be defined as it is the structure that organizes sound for example Pitch (tone), Energy (loudness) and normalized duration (rhythm) are the main components of prosody for a speaker. Prosodic features can vary from speaker to speaker and depend on long-term information of speech signal [7] [8] [9].

In this paper describes how to build a simple & complete automatic speaker recognition system. Such type of speaker recognition systems have been used in many security applications. [10]. ASR Systems can be classified as Automatic Speaker Verification (ASV) and Automatic Speaker Identification (ASI) systems. Speaker verification systems are used to verify whether an input speech signal matches to the claimed identity where as speaker identification objective to identify an input speech by selecting one speaker model from a set of enrolled speakers models. Sometimes speaker verification will follow speaker identification in order to validate the identification results [11].

## 2. Development of Speaker Recognition

The first type of speaker recognition system comes in to the existence at 1960's, which uses spectrogram of voices for identification and this system known as voiceprint analysis. Acoustic spectrum of the speech is like to the fingerprint however such type of analysis could not fulfill the objective of automatic recognition system. In the 1980's numerous methods was proposed to extract features from speech signal for speaker recognition [6] [8]. These features are represented in time domain and frequency domain or in both domains. Result Shows those acoustic features of speech signal be different between individuals and these acoustic features consist of both learned behavioral features such as pitch, accent and anatomy such as shape of the vocal tract and mouth [12]. The most commonly extracted features are the prosodic, Mel-Frequency Cepstral Coefficients (MFCC), Linear Predictive Coding (LPC) and Linear Predictive Cepstral Coefficient (LPCC) these features extracted to the short time analysis of speech signal which provides information on the vocal tract [13][14][15].

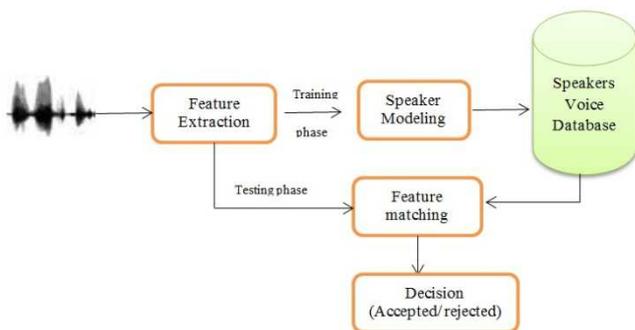


Fig. 1. Block diagram of speaker recognition system.

After feature extraction the main steps to create models for speaker's speech, different modeling techniques are available to model voiceprint extracted from speech signal. There are many models were introduced for example pattern matching i.e. Dynamic Time Warping (DTW) which does direct template matching between training and testing voice database. But study shows that direct template matching is time consuming, as the number of feature vectors increases [8]. To overcome from this clustering is a process to decrease the number of features vector through using a codebook to represent centers of the feature vectors e.g. Vector Quantization (VQ). For Vector Quantization (VQ) the LBG (Linde, Buzo and Gray) algorithm and the k-means algorithm are the most familiar algorithms. Also some other methods was proposed for speaker modeling such as neural networks (NN) and stochastic models that uses probability distribution for example Hidden Markov Model (HMM) and Gaussian Mixture Model (GMM) etc [11] [13] [16]. The following methods for speaker recognition will be aimed specifically at text-independent speaker identification system. The system involves following steps:

- Speech acquisition
- Pre-Processing

- Feature Extraction
- Speaker Modeling

### A. Speech Acquisition

In this case voice database are recorded by using different head sets and microphones [18].

### B. Pre-Processing

Pre-processing is a first step in automatic speaker recognition system. It is critical process performed on speech signal input in order to develop a robust and efficient speaker recognition system [14] [19]. In this process the task such as A/D, end point detection and pre-emphasis are completed.

### C. Feature Extraction

Feature extraction from speech signal is also called front end processing and is performed in both recognition (testing) and training phase. Feature extraction use to converts digital speech signal into sets of numerical descriptors or feature vectors that hold significant characteristics of the speaker's voice [20].

### D. Mel Frequency Cepstrum coefficient (MFCC)

Mel-frequency Cepstral coefficient (MFCC) is one of the most common method used for the voice feature extraction from speech signal. The basic difference among the MFC and cepstral analysis is that the MFC have frequency components with a Mel scale modeled which is based on the human ear perception of sound instead of a linear scale [21]. The Mel-frequency cepstrum characterizes the short-term power spectrum of a speech signal using a linear cosine transform of the log power spectrum of a Mel scale [19]. To found Mel scale the following formula can be used:

$$M = 2595 \log_{10} \left( \frac{f}{700} + 1 \right)$$

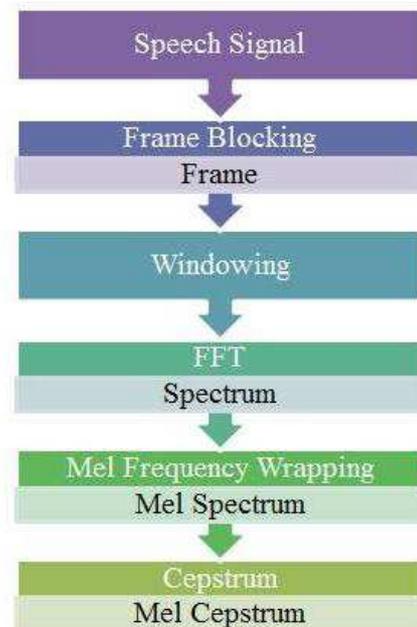


Fig. 2. Steps of MFCC.

MFCC is generally used for feature extraction since the advantage of MFCC is the robustness towards noise and

spectral estimation inaccuracies under various conditions during voice recording [22]. The study shows that the comparison of different feature extraction techniques it is found that the MFCC provides better performance than other feature extraction techniques [23].

#### E. Speaker Modeling

The purpose of modeling techniques is to create speaker models for feature matching of the speaker's voice. Speaker models can be defined as, it is the models that contain enriched speaker specific information with a compressed amount [14] [24]. At the time of training or enrolling phase speaker models are created using the specific features extracted from the current speaker and in the recognition phase, the speaker model is used to compare with the current speaker model for identification or verification. Gaussian Mixture Model (GMM) is generally used for speaker modeling [25].

### 3. Practical Approach to ASR

The development in the field of speaker recognition technology is going on rapidly. But still there are a few inherent difficulties occurs which is needed to be solved. As the reliability of the speaker recognition system drops drastically when a huge voice database is used or when data acquire under a noisy environment. Still research is in progress and moving fast and it may be possible to improve the robustness of the existing recognition systems to solve some of the issues [14]. There are many factors which affect the speaker recognition i.e. speaker identification and speaker verification. The required Practical implications of using a speaker identification system have to be assessed are designed and implemented [6] [27]. The following are some factors-

- Environment (where voice acquired)
- Quality of microphones
- Emotional state of speaker
- Channel differences
- Artifacts of the Speech Signal Silent removal
- Difference between male & female speakers
- Voice quality

The above mentioned features keep in mind during the voice data recording and for implementation of ASR system. Such type of ASR systems will also be used in a numerous fields, similar in design and acoustic properties. The speech category of recordings in meeting environments will be of high quality because recordings also used for speaker recognition [6].

### 4. Conclusion

Automatic speaker recognition technology is in full growing now days. A number of algorithms have been developed to improve the performance and robustness of Automatic Speaker Recognition systems. Automatic Speaker recognition systems are increasingly common and used in very different acoustic conditions. The use of Mel frequency

cepstral coefficients (MFCCs) for feature extraction from speech signal and it is one of the standard methods used in ASR systems for information retrieval.

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