

Intelligent CCTV Surveillance Based on Sound Recognition and Sound Localization

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Abstract— CCTV is used for many purposes, especially for surveillance and for traffic condition monitoring. This paper proposes an intelligent CCTV system that tracks sound events based on sound recognition and sound localization. From the experimental results, it is evident that the proposed method can be successfully used for the intelligent CCTV system of CCTV.

Keywords - intelligent CCTV, surveillance, sound recognition, sound localization

I. INTRODUCTION

To eliminate blind spot, the need for intelligent CCTV has been increasing. This paper focuses on sound recognition and sound localization to make intelligent CCTV. To increase the accuracy of sound recognition, it is necessary to consider the appropriate noise-canceling technology. The spectral subtraction is commonly used for noise reduction in environments with various noises [1]. We use end-points detection algorithm to find the interval of the signal [2]. Then, mel-frequency cepstral coefficients (MFCCs) are calculated [3]. The K-means algorithm is used for vector quantization (VQ) codebook [4, 5]. Finally, we use root mean square (RMS) to distinguish the left and right positions of the sound, and then find the correct angle and control the CCTV [6].

II. SPECTRAL SUBTRACTION

Spectral subtraction algorithms are widely used in speech enhancement. This method is to obtain the original signal by subtracting the spectrum of the noise estimated from the noise-added signal [1]. However, when estimating a noise signal, a problem may arise in which an important part of the spectrum can be regarded as noise. Therefore, to use the spectral subtraction technique, a sufficient interval is required to estimate the noise spectrum. Spectral subtraction results are shown in the figure 1.

III. ENDPOINT DETECTION

Endpoint detection is used to distinguish speech from noise and is required in many speech applications, such as speech recognition, speech coding and communication. In a speech recognition system, for example, accurate endpoint detection can improve the recognition accuracy under various types of background noise and reduce the computing power induced by incorrect speech detection [2].

IV. SOUND LOCALIZATION

Sound localization is to find where the sound is located [3]. For the sound localization, we used interaural

time difference (ITDs), the time difference in which a signal arrives in two ears. In this experiment, two microphones are used to record sound. After selecting the location of the microphone, we used the time difference of the sound reaching the microphones, then calculate the angle of the direction of the sound and find out where the sound is coming from.

V. SOUND RECOGNITION

A. Mel-frequency cepstral coefficients

MFCC feature is commonly used for sound recognition [4]. MFCC is one of the methods of expressing the power spectrum of the short-time signal. The frequency band of the MFCC can be evenly divided on the Mel-scale. So we can better express the sound.

B. K-means algorithm

The K-means clustering algorithm is used for constructing VQ codebook [5, 6]. The algorithm divides the data set into several clusters. The K-means algorithm determines the sum of squares of the distance between the centroid of each cluster and updates the clusters in the direction of minimizing a cost function. We used the resulting VQ codebook for the sound recognition. Figure 2 is an example of the codebook with 16 centroid vectors.

VI. EXPERIMENTS

We conducted the experiments with various sample sounds. Each sample has a mix of different types of noise, such as wind sounds and cell phone ring tones. The experimental results are shown in Table. 1.

TABLE I. SOUND RECOGNITION RESULTS (%)

sound type	recognition rate (%)
horn sound	92
alarm sound	83
collision sound	83
brake sound	75

