Multidimentional View of Automatic Video Classification : An Elucidation

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Abstract — Media is one of the foremost roles in human daily life activity. Multimedia is the integration of multiple forms of media, which includes text, image, audio, and video. Most of the people are always working with their Personal Digital Assistant (PDA) that provides computing, information storage and retrieval capabilities for personal or business use. Images and videos engage more space than other kinds of data on their PDA or electronic device. There are many kinds of videos available in day to day life, so we need an efficient tool to classify the videos with sky-scraping accuracy. The main goal of video classification is to help the people to find video of their interest. In this paper we study multi dimensional view of video classification methods and techniques, compare them and also conclude with opinion for further research.

Keywords—

I. INTRODUCTION

Images and videos are shared all the way through social Medias such as facebook, youtube, linkedIn etc in day to day life. As per the survey there are 2 billion monthly active users (June 2017). Especially video has one of the major part in humans routine life.

A. Video:

Video is an electronic medium for the recording, copying, playback, broadcasting, and display of moving visual media. In general video is nothing but sequence of images, which may be sometimes call it as frames. Video has different set of properties such as size, format, bit rate, frame rate etc...

B. Frames:

A frame is one of the many still images which may produce the complete moving picture. In general video has 24 frames per second (fps). Video size can be calculate based on the following properties such as frame, image size, color and so on. The quality of video is based on the number or frames per second used in a video.

C. Video Classification:

There are huge volumes of video data available in day to day life for personal and business use. So we need an automatic video classification tool to categorize the video for processing it. The main intention of video classification is to categorize the video whether is coming under which variety. There are few major groups are there such as movies, short films, sports, cartoons and animation movies, news, weather report etc... Video classifications can be done by the following techniques [1][2]. 1. Text based approaches

- a. Visible text on screen
- b. Text Extracted from speech
- c. Text on object
- 2. Audio based approaches
- 3. Video based approaches
 - a. Frame Selection
 - b. Frame Majority vote
 - c. Temporal Feature Pooling (TFP)
 - d. 3D Convolution (C3D)
 - e. Color based
 - f. Object based
- 4. Other approaches



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D. Applications of video classification:

Many videos are available in day to day life. So we need an efficient tool to search a specific video from the video database. The main goal of video classification is to help the people to find video of their interest from the internet. There are huge varieties of applications available. Here we have mentioned few examples. In social web site we are receiving unwanted video that we don't want at most. A high jump sport video consists of two different actions, running and high jump, which shared with other videos such as running or hurdling sport video.

II. EARLY TECHNIQUES AND METHODS

In this section, the author describes the previous research works in the form of title, problem statement, objectives, not repeat the information discussed in Introduction [2].

A. Text Based Approach:

One of the video classification approaches is text based approach, which is least common approach. There are four common methods for text based classification visible text on screen, text extracted from speech [3], text on object and closed caption.

a. Visible Text on screen:

This is one of the approaches to classify video on text based approach. Text displayed on the bottom of the screen relevant to conversations of the frames in a video. If the text is not readable then the text is converted into readable text using OCR software. Features are extracted from the text and then given as input for any classification tool like SVM. Finally machine will label the video as specific category.

b. Text Extracted from speech:

This is also one the approaches to classify video on text based approach. Sample audios are separated from the given video and it convert into text using speech recognizing software. Based on the text features video can be classified.

c. Text on Object:

One more method to classify the video based on text displayed on object. Texts are displayed on object are captured and converted into viewable text by using OCR (Optical Character Recognition) tool. Example for text on object is vehicle registration number plate, Building name, city name and so on.

B. Audio Based Approach:

Audio based approach needs little computation and processing resource compared with video based approach. This method is tricky to distinguish with multiple sound audio file. There is multiple level of audio based feature. Time and frequency domain features are the low level features [4]. Volume standard deviations and volume

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dynamic range can be used to detect whether the video has constant level of noise. Zero crossing rate (ZCR) is the number of signal amplitude sign changes in the current frame. If the loudness and ZCR are both below thresholds, then this frame may represent silence. [4] Bandwidth is a measure of the frequency range of a signal. Some types of sounds have more narrow frequency ranges than others. Speech typically has a lower bandwidth than music.

C. Frame Based Approaches:

A frame is one of the many still images which may produce the complete moving picture. A frame is sometimes called as image. Frame based video classification itself is one of the major research area. There are few methods to classify the video based on frame and frame based properties [2][5].

a. Frame selection:

Selecting the frame itself is one of the crucial works to be done before classification of given video. There are huge amount of frames for a single video. From these frames we have to select the frame by implementing some classifier algorithms. Depending upon the algorithm the classification accuracy may increase or decrease and time complexity may also increase and decrease. Once the frames are selected from the video these set of frames are called as keyframes. After creation of the keyframe, features are extracted from the keyframe and give as input for any classifier.

b. Frame Majority Vote:

Video classification is similar to image classification. Every frame in a single video can be labeled as same category and features are extracted. Then train any ANN or CNN classifier using the dataset of the frames. In test mode, every frame per video is predicted as a video category so that the video type can be obtained from the majority vote of frame predictions. For example there are 30 frames in a video among these 25 frames are labeled as spots and the remaining 5 frames are labeled as advertisement. Now this video is classified as sports category since it has 25 majority frames labeled as sports.

c. 3D Convolution Approach:

C3D is one of the simplest and effective approaches, used to deep 3-dimensional convolution neural networks trained on a large scale video dataset. C3D operates on stacked video frames and extends the original 2D convolution kernel and 2D pooling kernel into 3D kernel to capture both the spatial and temporal information. [Class-caption] However, training a 3D CNN is very time consuming and the spatial temporal structure in videos may be too complex to capture.

d. Color Based:

Video is nothing but sequence of related images sometimes called as frames. Frames are composed of lines. Each line is

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sampled to create number of pixels per line. Resolution of the frame is dependent on number of lines per frames and number of pixels per line. RGB, CMYK are two common color spaces . The color pixel is represented by combination of the individual colours red, green and blue in some amount. In the HSV colour space, colours are represented by hue (i.e., the wavelength of the colour percept), saturation (i.e., the amount of white light present in the colour), and value (also known as the brightness, value is the intensity of the colour) [5].

e. Object Based Approach:

This is one of the complicated methods to classify the video. Tracking and detecting the object is more crucial work. Object tracking is one of the complicated research area to track the object from the given video. Tracking a ball from the fo for c featu

D. Other Approach:

- STUW Spatiotemporal Uncertainty Weighting [6]
- \geq RWRS – Random Walk with Restart Saliency [6]

The above methods and approaches are used to classify the given video based on the category. Based on the approaches and methods features are extracted from the given video and then classified into specific genre. Following categories are the most frequently used genre.

Movie

News

Sports

Movie Trailer

i.

ii.

iii.

iv.

he foot ball sports video for further process is an example or object based. Once the object has been tracked the eatures are extracted and given to trained CNN.		v. Advertisements
		vi. Short Films
		vii. Animated Movies
eatures are entracted and g		viii. Cartoons, etc
Techniques/Methods	Pros	Cons
Text Based Approach		Computationally expensive,
 Visible Text 	Low computational cost	Special software needed.
on Screen	More Accurate, High Dimensionality	
 Text 	Easy to convert speech to text	High computational cost, High error rate,
Extracted	-	Speech Recognition software needed
from Speech		
 Text on 	High Dimensionality,	Computationally more expensive,
Object	Error rate may decrease	Time complexity may increase for
		separating text from an object.
Audio Based	Low computational resources	Needs to handle more number of
Approach	Low computational resources	properties Difficult to separate multiple
Арргоасн		sound samples
		Sound sumpton
Video Based		Algorithms needed to select frames,
Approach		Needs to handle many properties
 Frame 	High accuracy	Difficult to identify frames.
Selection		
 Frame 	High accuracy	Algorithms needed to select frames,
Majority	Easy to implement	Needs to handle many properties
Vote		Difficult to identify frames.
	Low time consuming	Only handle 3D videos
• 3D	Expect to get more stable result	
Convolution	Simple to implement and process	Unsophisticated representation
 Color Based 		Larger size.
	Moderate accuracy	Computationally expensive
 Object Based 		Difficult to track objects
Other Approach	STIW algorithm aghistras	Acouracy depends on Spatial Terrarel
outer Approach	performance: Method may be extended	Salienov value
	in many ways	Sancicy value.
• KWK5	in many ways.	

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III. CONCLUSION AND FUTURE DIRECTION

There are huge volumes of video data available in day to day life for personal and business use. So we need an automatic video classification tool to categorize the video for further processing. We have reviewed various video classification methods and techniques and explored more ideas on it. Most of the methods and techniques are compared along with features and their pros and cons are tabulated. Each and every technique use to handle different set of features to classify the given video in a pre defined specific genre. All the features are drawn from three major modalities; they are text, audio and video. The main goal of video classification is to classify the data set on a specific category with high accuracy in order to help the people to find video of their interest. With this details we conclude frame selection and feature extractions is one of the major role to classify the video with sky-scraping accuracy.

To improve the accuracy of video classification we need to draw on some better method to select the frames and its features. Increasing accuracy and efficiency of video classification is our extended work along with existing methods and techniques like keyframe extraction and feature selection. To have better performance, extracting more frames with some diverse methods and techniques from video are likely to be helpful.

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