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Anomaly Detection Network for Video Surveillance Using CNN

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ABSTRACT: Video analytics is the method of processing a video, gathering data and analysing the data for getting domain specific information. In the current trend, besides analysing any video for information retrieval, analysing live surveillance videos for detecting activities that take place in its coverage area has become more important. Such systems will be implemented real time. Automated face recognition from surveillance videos becomes easier while using a training model such as Artificial Neural Network. Hand detection is assisted by skin colour estimation. This research work aims to detect suspicious activities such as object exchange, entry of a new person, peeping into other's answer sheet and person exchange from the video captured by a surveillance camera during examinations. Nowadays, people pay more attention to fairness of examination, so it is meaningful to detect abnormal behaviour to ensure the order of examination. Most current methods propose models for particular cheating behaviour. In this system, we extract the optical flow of videodata and propose a 3D convolution neural networks model to deal with the problem. This requires the process of face recognition, hand recognition and detecting the contact between the face and hands of the same person and that among different persons. Automation of 'suspicious activity detection' will help decrease error rate due to manual monitoring.

KEYWORDS:convolution networks; optical flow; abnormal behaviour; examination surveillance video.

I. INTRODUCTION

Human face and human behavioural pattern play an important role in person identification. Visual information is a key source for such identifications. Surveillance videos provide such visual information which can be viewed as live videos, or it can be played back for future references. The recent trend of 'automation' has its impact even in the field of video analytics.

Video analytics can be used for a wide variety of applications like motion detection, human activity prediction, person identification, abnormal activity recognition, vehicle counting, people counting at crowded places, etc. In this domain, the two factors which are used for person identification are technically termed as face recognition and gait recognition respectively. Among these two techniques, face recognition is more versatile for automated person identification through surveillance videos. Face recognition can be used to predict the orientation of a person's head, which in turn will help to predict a person's behaviour. Motion recognition with face recognition is very useful in many applications such as verification of a person, identification of a person and detecting presence or absence of a person at a specific place and time.

In addition, human interactions such as subtle contact among two individuals, head motion detection, hand gesture recognition and estimation are used to devise a system that can identify and recognize suspicious behaviour among pupil in an examination hall successfully. This paper provides a methodology for suspicious human activity detection through face recognition. Video processing is used in two main domains such as security and research. Such a technology uses intelligent algorithms to monitor live videos. Computational complexities and time complexities are some of the key factors while designing a real-time system. The system which uses an algorithm with a relatively lower time complexity, using less hardware resources and which produces good results will be more useful for time-critical applications like bank robbery detection, patient monitoring system, detecting and reporting suspicious activities at the railway station, exam holes etc.

II. METHODOLOGY

In this section, we define the abnormal behaviour, introduce the proposed CNN model in detail, and describe the algorithm of detecting the abnormal behaviour in examination surveillance video. The abnormal behaviour can be identified as irregular behaviour from normal one. During examination, we pay more attention to the abnormal

behaviours which mainly include leaning, reaching out hand, turning around, entering and leaving the classroom halfway. If there exist frequent abnormal behaviours in a period, it is very likely that some problems appear in the examination. For example, if students can't hear English listening clearly due to equipment problem, they would lean and turn around to confirm the problem. Meanwhile, the electronic monitor would detect the abnormal behaviours and notify the supervisor to deal with the problem in time. So it is essential and significant to detect abnormal behaviour in examination surveillance video.

surveillance video. Table I shows the architecture of our 3D CNN model. The model has 2 convolution layers, 2 pooling layers, 2 fully-connected layers and 1 SoftMax layer. The number of filters for 2 convolution layers are 64 and 128. Two fully connected layers each have 256 and 128 outputs. The kernel size of the first convolution layer is $7 \times 7 \times 7$ which represents that height, width and temporal depth of the feature map are 7. The kernel size of the second convolution layer is $7 \times 7 \times 3$ which represents that height and width of the feature map are still 7 but temporal depth is 3. All convolution kernels have 1 stride. All pooling layers are max pooling with kernel size $2 \times 2 \times 2$ and have 2 strides. At last, the soft max layer has 2 outputs, which means the proposed 3D CNN is a binary classification model. We train the model from scratch using mini-batches of 10 clips, with initial learning rate of 0.0005.

III. PROPOSED ALGORITHM

Algorithm 1 Negation phrases identification

Input: Tagged Sentences, Negative Prefixes

Output: NOA Phrases, NOV Phrases

```

1: for every Tagged Sentences do
2:   for  $i/i + 1$  as every word/tag pair do
3:     if  $i + 1$  is a Negative Prefix then
4:       if there is an adjective tag or a verb tag in next pair then
5:         NOA Phrases  $\leftarrow (i, i + 2)$ 
6:         NOV Phrases  $\leftarrow (i, i + 2)$ 
7:       else
8:         if there is an adjective tag or a verb tag in the pair after next then
9:           NOA Phrases  $\leftarrow (i, i + 2, i + 4)$ 
10:          NOV Phrases  $\leftarrow (i, i + 2, i + 4)$ 
11:        end if
12:      end if
13:    end if
14:  end for
15: end for
16: return NOA Phrases, NOV Phrases

```

Occurrence not worth NOA
 not go wrong NOA
 not bad NOA
 not be happier NOA
 not good NOA
 didn't work NOV
 don't recommend NOV

IV. SIMULATION RESULTS

we propose a unified deep learningbased framework for abnormal event detection from exam hall. The proposed Anomaly Net consists of three blocks which are designed to achieve three keys of anomaly detection in neural networks. In short, the motion fusion block is designed to keep the temporal and spatial connection between the motion and appearance cues.



V. CONCLUSION AND FUTURE WORK

In this project, we can detect and recognize faces of the criminals in a video stream obtained from a camera in real time. We have used Haar feature-based cascade classifiers in Open CV approach for face detection. It is a machine learning based approach where a cascade function is trained from a lot of positive and negative images. It is then used to detect objects in other images. Also, we have used Convolutional Neural Network (CNN) for face recognition. Several advantages of this algorithm are: Efficient selection of features, Scale and location in-variant detector, instead of scaling the image itself, we scale the features. CNN recognizer can recognize faces in different lighting conditions with high accuracy. Also, CNN can recognize efficiently even if single training image is used for each person. The real-time automated

this proposed system, we've got investigated the functionality for CNN to analyze capabilities from video frames. We recommend a unified deep reading primarily based definitely framework for abnormal event detection and criminal identification in public. The proposed system includes 3 blocks that are designed to advantage 3 keys of extraordinary detection

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