

Detection of Helmet Wear from Surveillance Videos for ATM Centre Security

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ABSTRACT In the modern economy, the deployment and usage of Automated Teller Machines (ATMs) are increasing manifold, when they left unattended, they pose a serious security risk. ATM machines are widely used for banking services such as deposit and withdrawal of funds nowadays. Despite the fact that ATM centres are equipped with surveillance cameras, there is still a security risk. Fraudsters may wear a helmet to conceal their faces from surveillance cameras, allowing them to engage in illegal activities. ATM machines providing banking services for 24x7 for easy deposits and withdrawals. Although monitoring cameras are used for automatic surveillance in remote places also, but security threat still exists. Fraudsters may wear helmets to conceal their faces from surveillance cameras, allowing them to engage in abnormal activities. The investigations will be more complicated by the fact that the fraudster's face is completely hidden. To overcome this problem, proposed system used combination of deep learning with artificial intelligence model to detect helmet wear in ATM center. These days, the concepts of artificial intelligence is growing in all real world applications, object detection by YOLO, and image processing techniques, makes it possible. YOLO model is used as it can detect helmet as object with high accuracy in the incoming video stream. Then Region of Interest (ROI) is used for detecting whether person is wearing helmet. If a person wearing a helmet is detected, an alert is sent to the ATM centre control centre. The experimental setup demonstrated high accuracy in detecting helmet wear.

KeyWords: Artificial Intelligence, Image processing,

Object detection model, RCNN algorithm, YOLO, Region of Interest

I. INTRODUCTION

With growing pace of economic revolution in country, liquidity of currency flow also increases, only banks can observe people for transaction. Thus smart way of banking is developed for convenience and used everywhere. ATM is the one, which is very simple to use and provides services to most of banking services including deposit, withdrawals, cheque deposits, printing transactions on passbook etc. To use these services, users are provided with security pins for transaction. These ATMs are monitored through surveillance camera round the clock to prevent robbery and fraudulent activities. Surveillance camera is positioned to capture user's face for monitoring purposes. Fraudster may conceal their faces with mask or helmet thus their face cannot be shown to the monitoring camera. When security threats go unnoticed, the surveillance system's goal fails. Widespread use of banking services are accessible through ATM centers, which is available 24x7 facility. Digital transactions are growing in enormous in developing countries like India, has higher demand for smart ATM services. As the number of outlets are

large, it is difficult to manage with security person one each ATM outlet and the maintenance cost will be huge. Thus many banks deployed surveillance camera to monitor the outlets.

ATMs are vulnerable for many types of attacks including man-in-middle attack, data sniffing, spoofing attacks, etc. According to reports, crime and illegal activity are increasing year after year. These problems are arising due to lack of high security measures. According to existing study, 5500 fraud activities

are reported in each year.

Artificial intelligence is backbone for many real world applications utilizing computer vision, like object detection, monitoring services. Artificial intelligence is growing as under image quality improving, haze etc. Helmet detection through pre-trained YOLO model can be used for detection of unusual activities in ATM outlets.

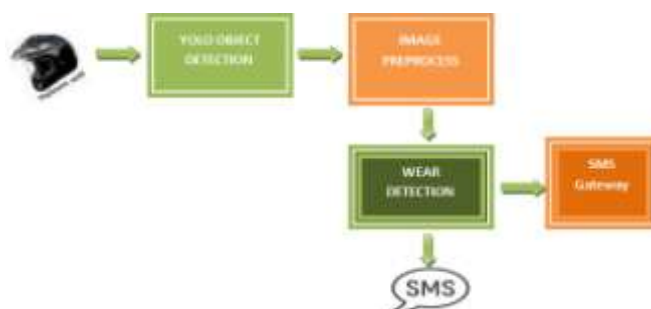


Figure: 1 Helmet Wear Detection

The primary goal of proposed work is to detect safety helmet wear by use of surveillance camera in ATM outlets. This is possible by combining machine vision and deep learning models. The model employed is YOLO pre-trained object detection, which can detect more than 3000 classes of objects. This pre-trained model is through Region Based Convolutional Neural Networks (R-CNN). Live stream of surveillance is accomplished as input for the system. When a person is found wearing helmet and SMS alert message is given to the control center or monitoring team. The main motive of work is to avoid unusual activity or robbery in ATM outlets.

The work implementation is divided into three major phases, the first one consists of detecting helmet object using YOLO model and segment detected object for further processing. In the next phase of work, segmented image is identified whether the user in ATM outlet wearing helmet, by finding the skin color inside the detected object. In the final phase, if any person is detected, then an SMS alert is given to the monitoring team.

In the following chapters, existing work on helmet detection is discussed in detail. In chapter 3, implementation of helmet detection is proposed with its methodology and algorithms. In chapter 4, the experimental set up and results are explained. In chapter 5, this helmet detection is given conclusion and further enhancements are discussed.

II. RELATED WORK

This chapter provides a short discussion of research works handled by authors on Safety Helmet detection and its usage for ATM security. In automated monitoring, machine vision and deep learning are important techniques and found to be growing with its usage. Thus more research on this field makes the problem achievable.

Automated detection of biker riders on roads without helmet was proposed in [1]. Image processing techniques are handled, then applied feature extraction to segmented object with person region proposal. Histogram of Oriented Gradients (HOG), Local Binary pattern (LBP) and scale-invariant feature transform (SIFT) are the three feature extraction techniques used, in this the author experimented that HOG gives better results. Person with bike is classified for detection and remaining are omitted. Person with bike is applied SVM (Support vector machine) classifier to identify person wearing helmet and it is omitted, person not wearing helmet is identified and given output.

Motorcyclist without helmet was proposed in [2], the author used wavelet classification for vehicle classification, this classifies only motorcycles. For helmet detection, author used circular Hough transform (CHT) and HOG feature extractions for detection and multi layer perceptron for classification. The author handles motorcyclist without helmet in two stages, the first stage for identifying the vehicles and classifying only bikes.

Region of proposal was arrived and segmented and applied Random forest (RF) classifier, it given better performance on 50*200 imagepixels.

Crime scene detection in ATM outlets through video surveillance was reviewed by authors in [3]. There were numerous research on this study was prefaced, of which abnormal event detection this is analyzed by activity variations on the input. Some of the study elaborated with skin detection, occulted face detection for robbery threats and illegal objects like weapon detection were studied. Moving objects detection through different image feature extraction algorithms such as HOG, MOG, Distancetransforms, werestudied. Facedetection handlingwithAdaBoost, Viola-Jonesalgorithmwerereviewed.

SafetyHelmetdetectionthroughmodifiedHoughtransformwasstudiedbyauthors[4].Onthedetectionfront, thevideoorimage capturedinfrontofATMmachineisdetectedacircularobject kind of object on the images through Hough transform technique, asthecircularshapedobjectsaremostlyhelmet sare used to occult the face of frauds. Hough transform gives better results on pattern identification on specifically shapes. However, the detection of helmet takes large computations, thus in our work, we proposed with YOLO detection model for accurate detection of Helmet objects.

Automatic detection of safety helmets by deep learning algorithm was proposed in [5]. The deep learning algorithm CNN Convolutional neural network was proposed with architecture VGGNET16 and ALEXNET. The dataset was used on work, which used 1880 still images, out of which, 1000 are wearing helmet and 880 still images person without helmet. RCNN algorithm used for training these images and trained model is generated. The pre-trained YOLO model is considered in RCNN. This can identify the Helmet accurately on the still images.

Safety Helmet detection on construction sites are

identified through YOLO v3 model was proposed in [6]. This approach used Gaussian fuzzy augmentation process to the unbalance data to improve the accuracy of Helmet detection. Unbalanced data with sample images of different sizes were considered for the study, thus pre-processing and feature extractions are applied before detection. However, this system can be applied for helmet detection for construction site, whether employees wearing helmet or not can be identified. It is inferred from above researches that, there are highly suitable models available for Helmet detection. The method of implementation to be improved for considering the accuracy of detection and scope of detection. Many of the existing studied were proposed for safety helmet detection on traffic surveillance. However, there are criminal activities at ATM outlets wearing Helmet or mask was unattended. Thus our proposed study, considered this as major problem and given solution with YOLO and deep learning algorithms.

III. PROPOSED WORK

Helmet wear detection on ATM outlet is challenging as the live stream of video input is considered for detection. The detection system should be speed and quick enough to find the fraudulent. This is achieved by implementing machine vision concept in Python Language. Implementation used Python 3.6 as programming with mandatory libraries namely OpenCV, requests etc. Python application is designed with Tkinter for user interface design. Through this interface, the number to which alert to be sent has been given by user input. When the wear detection is identified an alert sent to the user.

Web camera captures input stream of video, is taken as image frames for processing. Image pre-processing steps are carried to find the Safety Helmet wear detection by feature extraction and skin portion identification.

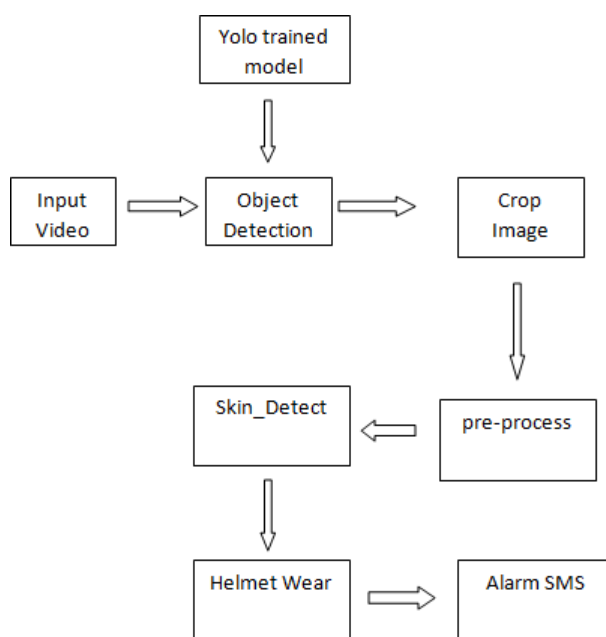


Figure 2: Architecture of proposed work

The above architecture represents proposed system's architecture with all working modules of the system namely camera input, object detection, image pre-processing, skin portion detection and helmet wear modules.

Helmet wear detection follows below mentioned modules

- a. Helmet Detection through YOLO
- b. Localizing Object
- c. Image pre-processing
- d. Skin Identification
- e. Wear detection
- f. SMS alert system

DATA INPUT

There is no specific dataset for implementation as the detection is considered for real world surveillance camera data as input. Thus in our implementation, web camera input has been taken for the detection as we used the pre-trained model.

HELMET DETECTION THROUGH YOLO

The motive is to find the object whether cat, bike, human, helmet, mobile phone from the image frames. This detection paved way for localizing and detection of multiple objects in the scene images. Image frame from web camera input is taken for object detection, the class of object identified in this work is Helmet through YOLO, pre-trained model using algorithm Region based Convolutional Neural Networks (R-CNN). Once the object is detected, it is localized and segmented. Selective search method is used to search given object in the region of proposal is done. Once the object detected a region proposal is converted to fixed image by resizing it. The model search for where there is large chance of presence of an object. This model can search for multiple objects in the given image frames and all the objects can be recognized with high accuracy.

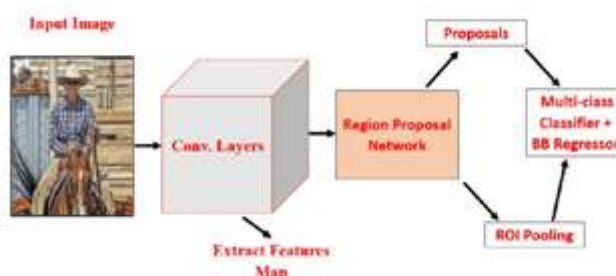


Figure: R-CNN architecture

Proposed work gets input from web camera. This is converted into still images continuously with specified time duration. Classification application and manual feature extractions are eliminated in this model. Object recognition and localizing is done in each grid on the image. The highest confidence detected value on the given region proposal is given as object class. The above architecture represents R-CNN working model. In the given input, safety helmet is detected by running forward propagation in the algorithm. High confidence value of more than 95% is considered for the study.

LOCALIZING OBJECT

Following object identification, localizing object in image is performed, then localized object is grab

cut and stored as fixed image. Image is re-sized with predefined size, foreground and background of image is identified and background subtraction is done. Object localizing is done with Opencv grab cut module, the region of proposal is detected to segment the helmet object.

Background model and foreground model are provided well with bounding boxes. Detected ROI of Background model and foreground model is stored as temporary array.

Numerous iterations to the above input is provided, as the number of iterations increased, detection of foreground and background is done with highest accuracy. Grabcut is used to identify Region of Interest (ROI) and initialized with bounding boxes. The following image shows the segmented image from object localization results.



Figure 6: Localized Object IMAGE PRE_PROCESSING

Localized object image from above module is pre-processed for detecting skin portion. This module must be performed to identify whether person wearing or holding helmet on hands. This module is performed after background subtraction.

Background subtracted resultant image is considered for input in this module. Hue Saturation colour conversion function is used. BGR images are converted to HSV images. This assists in detection of skin

portion in specified region of proposal. With some experiments conducted on development part, the height X width while wearing skin can be identified. This acts as threshold value and fixed for detection of safety helmet. Once the detected skin HXW value is equal or higher the threshold limits, it is considered that person wearing and alert is generated to users as a text sms.

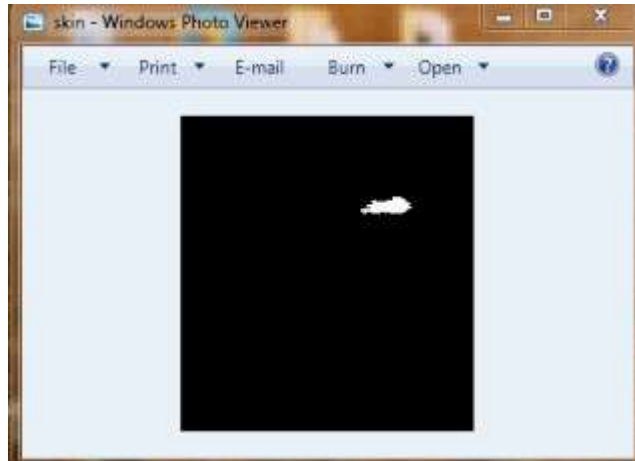


Figure 5: Skin portion detection

Above screen represents output of Hue Saturation converted image, in which white coloured regions show skin portion. The application is integrated to SMS gateway using the 3rd party SMS API (Application programming interface). Based on safety helmet wear detection, this integrated gateway send alert message to user registered to the application. OpenCV library has a function called contour, which can able to detect boundary of image object, this boundary drawn on HSV converted image to identify the skin region of proposal.

Algorithm: Detection of Human Wearing Helmet

- Step 1: Pre-trained model YOLO with weight and class file is imported with class for detection = 'Helmet'
- Step 2: Web Camera video is captured through OpenCV and get still images for processing
- Step 3: Object performed using RCNN if confidence value ≥ 0.92
 Object = 'Helmet'
 else
 no helmet object
- Step 4: Object segmentation on region of proposal is done

- Step 5: Background subtracted image is processed for HSV color value detection
- Step 6: Upper and Lower boundary ranges are identified for skin
- Step 7: Draw a bounding rectangle to measure height and width
 if height & width $>$ thresholds skin detected
- Step 8: Alert message is triggered for helmet wear detection

IV. RESULTS AND DISCUSSIONS

Safety Helmet wear monitoring for ATM security has been proposed as real time monitoring application considering web camera input. The live video stream is taken as still images and on every images, objection detection through R-CNN is proposed. This can detect multiple objects on input frames, the only object proposed for work is Helmet is identified and extracted with region of proposal. This is further processed for wear detection. Experimental results shown good accuracy of detection of safety helmet when a person is wearing. Deployment and maintenance cost of this application is less comparing other system.



The above screenshot depicts the application homepage, where the user can enter their details to store in our application database for sending alert SMS. This interface is designed in Tkinter to ease for people to handle. When the application is

started, camera window opens and starts to monitor continuously.

Following picture represents detection of Safety helmet on application background while executing with input image frames.

```
[UABH:G] terminating some callback
[G:7184844613075256]
[G:9938668189201355]
conf idence:
G:9938668189201355
[G:9817887544631958]
conf idence:
G:9817887544631958
h value
12
y value
31
SKIN DETECTED
Name
MAHA
Mobile No
8610331491
Message
Hi MAHA, Someone wearing Helmet in our ATM 11_02_19
{"return":true,"request_id":"92a52a0c99ec84","message":["Message sent successfuly to number"]}
[G:9946852922439575]
conf idence:
G:9946852922439575
```

Confidence for object detection for finding Safety Helmet is given 90%, whenever the system detects the helmet with confidence 90 or above, the image is cropped for further processing. Similarly, with experimental study, skin region of

proposal height and width are monitored for 10 experiments, to set the threshold for skin detection in application. This confirms person wearing helmet with high accuracy.

Detection	Parameter	Value	Result
Object (Helmet)	Confidence	≥ 0.92	Helmet detected
Object (Helmet)	Confidence	< 0.92	No detection
Skin Detection	Height X Width	≥ 10	Skin detected
Skin Detection	Height X Width	< 10	No detection

Table: Confidence Value for Object and skin detections

Safety helmet detection outperforms when the standard model Helmet is used by users. The detection latency may increase over the time period, as the infrastructure to be well built to handle the live stream of video cautiously. The system latency can be reduced on detection part in further study.

V. CONCLUSIONS

With the growing economy, liquidity increased among people, thus banking services are needed among people for 24x7 and for quick access of money. However, security of ATM outlets are big

threats and need to monitor all time from fraudulent and robbery activities. The proposed work considered ensemble model combining machine vision and image processing for providing ATM security surveillance. The proposed system can be implemented and maintained at low cost, safety helmet wear detection can be accurately identified. YOLO pre-trained R-CNN object detection technique. Experimental results are demonstrated to show the detection and an alert message sent for control centre.

The proposed system can be able to identify only Safety Helmet wear, but the fraudulent robbers can use a mask, this can be extended as future work.

Some more illegal activities and weapon detections such as holding gun or knife can be added on the surveillance system.

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