Video based human activity recognition for surveillance application

## Introduction:

Human activity recognition attempt automatically analysis and recognize human- activities of subject with the environmental conditions using acquired information from the various types of sensors.

Human activity has an inherent hierarchical structure that indicates the differentlevels of it. Gestures, actions, interaction, group-activities, which can be considered as a four levels categorization. First: for the bottom level, there is an “gesture” that are simple motion of a part of body. for instance, “raising an arm and moving a leg”. After action primitive level, the action/activity comes as the second level, for example: strolling, waving, walking. after that, “interaction” are human activities that involves at least two persons or objects, as an example, somebody utilize laptop, two persons checking hands.it is human to objects and human to human interaction. finally, group activities are that activities played by group composed of individuals or objects. Such as group of people playing basketball, fighting by two groups people.

Over the years, with the development of depth sensors, new opportunities arise to improve and advance this field. Recent advancement in the depth sensor (i.e., Microsoft Kinect) has made it feasible to capture the depth images/videos in real-time as well as colour images. Moreover, depth images have appropriate resolution (i.e., 640×480) and accuracy with respect to specific subject distance. As compare to the RGB images, depth images have several advantages for the problem of activity recognition. For example, depth images show appropriate geometry and shape information, which can be more discriminative than the RGB images in many problems such as segmentation, object detection and activity recognition. Depth images also provide additional body shape and structure information, which have been successfully applied to recover skeleton joints from a single depth image. Furthermore, depth images are insensitive to illumination changes. In this context, it seems very natural and important to use depth images/videos for computer vision problems. In the recent researches, it is shown that the approaches based on traditional RGB sequences could not perform well. On the other hand, depth and colour images have much different properties. The feature extraction method based on brightness, gradient and optical flow works very well for the RGB images but may not perform well for the depth images. It is therefore important to design such feature extraction method which is based on specific characteristics of depth sequences.

Briefly, the recognition of human activity involves input information from sensors, preprocessing, detection and segmentation of objects, feature extraction and representation, and classification

Application and challenge:

Human activity recognition has a hot scientific topic in computer vision community.it is involved in the development of many applications such as human computer interaction (HCI), virtual realities (VR), video surveillance for outdoor and indoor activities, home monitoring, video game, medical environment, patient monitoring, abnormal/normal activity identification, health care, and elderly pe0ople/children monitoring in specific area. HAR becomes popular and importantdomain in artificial intelligence and computer vision field.

In the resent years, recognizing human activities from a series of video frame or still image is a challenging task due to problems, such as:

**human behaviour**: Some humans can perform multiple tasks at the same time, which makes the process of recognition more difficult.

**Intra-class variability**: A given activity, can be performed differently by different individuals.

**inter-class similarity**: Classes that are fundamentally different, may show similar characteristics. Example of such activity may be, skipping and running has the similar pose, which becomes challenging when pose based algorithms are used for activity recognition

**illumination changes**: There may be dynamic variations in factors such as brightness, contrast and other parameters in a video, or these parameters may be affected due to the factors such as change in environmental condition like change in weather

**shadow effect**: The shadow of the object or human may create silhouettes, which may lead to false detection and tracking of the activity.

**Occlusion**: One body part of the person may block the view of other part of the body (which may be significant in recognizing the activity), which typically occurs due to the viewpoint. Subject of interest whose action is to be recognized may be occluded by another object in the video. This makes it difficult to identify the activity

**camera jitters**: The quality of video may be degraded due to low resolution, or poor quality of the recording device.

**noise image or frames in video**: The pixels related to people involved in the activity may be similar to the pixels in the background, leading to difficulty in segmenting out the humans or object of interest from the background.

**background clutter**: moving background object or human that it may lead to the wrong activity being identified, as they may be considered as the actual objects or humans involved in activity recognition.

Related work (review of literature):

Most of the proposed techniques of video based human activity recognition was divided four categories: information resource, pro-processing, feature extraction and representation, and classification.

(L. Minh Dang a, 2020) A comprehensive survey of state-of-the-art methods, along with their pros and cons for vision-based HAR and sensor-based HAR has been provided. introduction a classification of HAR methodologies and show advantages and weaknesses for methods in each category. (Michalis Vrigkas1, 2015) in the paper, provide a detailed review of recent and state-of-the-art research advances in the field of human activity classification. And discuss their advantages and limitations of human activity methodologies. We provide a comprehensive analysis of the existing, publicly available human activity classification datasets and examine the requirements for an ideal human activity recognition dataset. (Shugang Zhang, 2017) this review highlights the advances and limited of state-of-the-art activity recognition approaches, especially for the activity representations and classification methods. And representative and available datasets are introduced. (Allah Bux Sargano 1, 2017) this review paper presents a comprehensive survey of both hand-crafted and learning-based action representations, offering comparison, analysis, and discussions on these approaches. (Agrawal1, 2017)this paper, we present the state-of-the-art which demonstrates the overall progress of suspicious activity recognition from the surveillance videos in the last decade. We include a brief introduction of the suspicious human activity recognition with its issues and challenge. (Chhavi Dhiman a, 2019)this paper aims to summarize various existing abnormal human activity recognition (Ab-HAR) hand-crafted and deep approaches with the variation of the type of information available such as two-dimensional or three-dimensional data. (Ashwin Geet D’Sa, 2019)through this paper, approaches involved in activity recognition, advantage and challenges involved in activity recognition are discussed. (Djamila Romaissa Beddiar1, 2020) this paper attempts to review and summarize the progress of HAR systems from the computer vision perspective. Most computer vision applications, challenges are discussed (Thakkar1, 2020)in this survey, we discussed various ML and DL techniques for HAR for the years 2011-2019.the paper discusses the characteristics of public datasets used for HAR.it also presents a survey of various action recognition techniques along with the HAR applications. The advantages disadvantages, and challenges of action representation, dimensionality reduction and action analysis methods are also provided.

(Ahmad Jalal1, 2015) In this paper, a video-based novel approach for human activity recognition is presented using robust hybrid features and embedded Hidden Markov Models, In the proposed HAR framework, depth maps are analyzed by temporal motion identification method to segment human silhouettes from noisy background and compute depth silhouette area for each activity to track human movements in a scene. Several representative features, including invariant, depth sequential silhouettes and Spatial-Temporal body joints features were fused together to explore gradient orientation change, intensity differentiation, temporal variation and local motion of specific body parts. (Zayed, 2015) In this paper, a system for human activity recognition is proposed. We have considered the task of obtaining a descriptive labelling of the activities being performed through labelling human sub-activities. The activities we consider happen over a long period, and comprise several sub-activities performed in a sequence. The proposed activity descriptor makes the activity recognition problem viewed as a sequence classification problem. The proposed system employs Hidden Markov Models (HMMs) to recognize human activities. (Ahmad Jalal a, 2016) In this paper, propose novel multi-fused features for on-line human activity recognition (HAR) system that recognizes human activities from continuous sequences of depth map. The proposed on-line HAR system segments human depth silhouettes using temporal human motion information as well as it obtains human skeleton joints using Spatial-temporal human body information. Then, it extracts the Spatial-temporal multi-fused features that concatenate four skeleton joint features and one body shape feature. (Youssef Hbali1, 2017) the authors propose a new skeleton-based approach to describe the Spatial-temporal aspects of a human activity sequence, using the Minkowski and cosine distances between the 3D joints. (Boufama, 2017) This paper proposes a new method using trajectories, as mid-level features, for human activity recognition. (Ahmad Jalal, 2019) In this paper, we have made an effort to allow machines understand the behaviour in outer environment by proposing a novel methodology to recognize human interactions. (MUHAMMAD HAMEED SIDDIQI 1, 2019) The proposed method is an extension of the max-relevance and min-redundancy method. The ability of this method is to combine the strengths of different extraction techniques. we exploit the curve transform for feature extraction, and linear discriminant analysis for reduction of feature space. we use the hidden Markov model for activity recognition based on our proposed method of feature selection. Finally, by using the benchmark datasets such as KTH and Weizmann datasets compare the proposed scheme with state-of-the-art. Simulation results show that the proposed scheme is not only more accurate for some datasets, but outperforms competing method by weighted average accuracy 98%.

In classification. (D.K. Vishwakarma ⇑, 2015) The aim of this paper is to present a new approach for human activity recognition in a video sequence by exploiting the key poses of the human silhouettes, and constructing a new classification model. The proposed hybrid classification model is a combination of SVM and 1-NN model and termed as ‘SVM–NN’. (Uddin, 2016) A novel approach is proposed here for depth video based human activity recognition, using joint-based Spatial-temporal features of depth body shapes and hidden Markov models. From depth video, different body parts of human activities are first segmented using a trained random forest. The activity features are then further enhanced using generalized discriminant analysis to classify them non-linearly in order to convert them to more robust features. (S. U. Parka, 2016)In this paper, we propose a new HAR system via Recurrent Neural Network (RNN) which is one of deep learning algorithms. We utilize joint angles from multiple body joints changing in time which are represented a Spatial-temporal feature matrix (i.e., multiple body joint angles in time). With these derived features, we train and test our RNN for HAR. Our test results show that the proposed RNN-based HAR is able to recognize twelve human activities reliably and outperforms the HMM- and DBN-based HAR. We have achieved the average recognition accuracy of 99.55% for the activities. The results are 7.06% more accurate than that of the HMM-based HAR and 2.01% more accurate than that of the DBN-based HAR. (Mohanad Babiker1, 2017)A robust neural network was built based on the human activities features database, which was extracted from the frame sequences. multi-layer feed forward perception network used to classify the activities model in the dataset. The classification results show a high performance in all of the stages of training, testing and validation. (Rajat Khurana, 2018) One of the most pioneering technique for Human Activity Recognition is based upon deep learning and this paper focuses on various approaches based on that. Convolution Neural Network and Recurrent Neural Networks are mostly used in deep learning architectures. (Sumaira Ghazal1, 2019)Supervised machine learning was implemented for recognizing four activity classes including sit, stand, walk and fall. Performance of five techniques including K-nearest neighbours (KNNs), support vector machine, Naive-Bayes, linear discriminant and feed-forward back- propagation neural network was compared to find the best classifier for the proposed method. All techniques performed well with best results obtained through the KNN classifier.

Finally, (Vishwakarma, 2019) categorized these datasets into two part, the first part consists two-dimensional (2D-RGB) datasets and the second part has three-dimensional (3D-RGB) datasets. presented the state-of-the-art algorithms that give the highest accuracy on these datasets

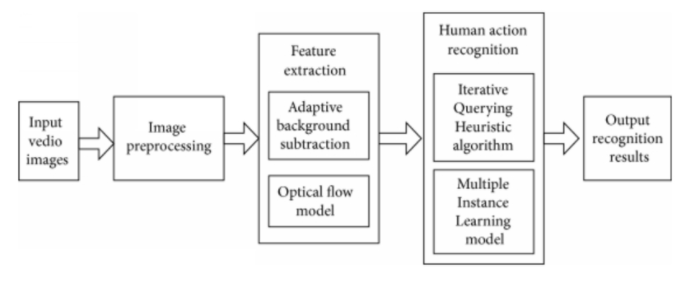
Motivation:

Currently, based on limited intelligent human activity recognition system which can robustly and efficiently recognition each class of human activity. Based on the challenging and necessary of computer vision domain, HAR should be made more attention in survey and development.

Objectives & Scope of the Propose Work:

The aim of this research is to gain an insight into the rapidly developing of Human Activity Recognition research.in order to get more accurate recognition in HAR,we focusing on image/frame preprocessing,segmentation,feature extra and representation,and classification.an excellent feature extra and representation phase is of great significance to classify and understanding accurately. In HAR, the following are the objective of this work:

1. image/frame of video input from different sensors.
2. Image/frame preprocessing,it include image transform ,nosing remove,object segmentation from foreground and background.
3. Feature extract and representation based on preprocessed image.
4. Finally, we classify features from several classificatory method. The different behaviour/action are achieved in system.



HAR system flowchart

Propose approach:

The recognition of human activity (HAR) involves input information from sensors/visual, pre-processing, detection and segmentation of objects, feature extraction and representation, and classification steps.

Sensor technology has achieved exceptional developments in multiple perspectives. The fast development of smart-phone, wearable devices, and CCTV systems has motivated researchers to improve HAR systems under practical situations. Now days, the depth sensor has made it feasible to capture the depth images/videos in real-time as well as colour images. The several data source such as RGB, RGB D, and various wearable devices signal from different sensor or visual used in human activity recognition systems.

Background subtraction, feature detecting and tracking, feature extraction is an important step. Skin-colour based, shape based, pixel values based, 3D models based, motion based, anisotropic-diffusion based method that various methods is obtained for detecting. Template-based methods such as features tracking, contours based tracking, or optimal estimation, particle filters, Cam shift can be sued to track the action.in (Kimia, 2016)a commercial depth-sensor-based camera, is utilized in this work to acquire the RGB as well as the depth images of different human activities. The depth silhouette is then extracted from each depth image after a background subtraction. After obtaining a depth silhouette, a corresponding skeleton-body model that provides 15 joint positions is obtained through the OpenNI library, therefore, from each depth silhouette, 15 body joints are obtained for the calculation of the spatial-temporal features. Therefore, proper pre-processing operating that include foreground/background frame segmentation for action feature detection, feature tracking, is huge magnitude for feature extraction and representation.

Finally, the features of action/activity are performed by classification methods. (Djamila Romaissa Beddiar1, 2020) For instance, Support Vector machine (SVM), Naïve Bayesian classifier, algorithm of K-Nearest neighbour, K-mean, Mean shift clustering, Machines finite state, Hidden Markov Model, Dynamic time warping, Neural networks (CNN, DNN, RNN), and hybrid method to classify the feature of Human activity.

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