

AUTOMATED ANALYSIS OF SURVEILLANCE VIDEOS

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Automated Analysis of Surveillance Videos

by

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Certificate

This is to certify that the thesis titled “**Automated Analysis of Surveillance Videos**” being submitted by Ayesha Choudhary to the Indian Institute of Technology, Delhi for the award of the degree of Doctor of Philosophy is a record of bona fide research work carried out by her under our supervision. In our opinion, the thesis has reached the standards fulfilling the requirements of the regulations relating to the degree.

The work presented in this thesis has not been submitted elsewhere, either in part or full, for the award of any degree or diploma.

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Abstract

Automated analysis of surveillance videos is an important area of research. This is mostly due to the growing need for surveillance of wide areas, such as airports, military bases and public places. As the areas under surveillance are growing wider and more complex, the difficulties of manual analysis of surveillance videos are increasing. At the same time, with the increase in size and complexity of the areas under surveillance, automated analysis of surveillance videos has become a challenging area of research. This has led to an increase in the various aspects of automated analysis of surveillance videos that need to be addressed.

In this thesis, we address some of the aspects of automated analysis of surveillance videos, such as event discovery, summarization and mining in single camera surveillance systems and real-time distributed composite event recognition in a calibrated pan-tilt camera network. We also address the problem of distributed self-calibration of a pan-tilt camera network. In the following paragraphs, we briefly outline our contributions.

We first propose a framework for multi-perspective summarization and mining of video data using unsupervised and incremental learning. We propose a tupular representation of the analytics data of a video that takes into consideration the fact that video data consists of information in multiple attributes or components and in multiple layers of abstraction. We introduce the concept of component based clustering to summarize the video data in the various attributes, chosen by the user. This allows the user to discover and learn the patterns in each of the attribute of the video data independently. We propose an incremental clustering algorithm that can be used with any type of data, symbolic or numeric, and is independent of pre-defining the number of clusters and the cluster radii. We apply our incremental clustering algorithm to cluster the data in each of the attribute spaces on-the-fly. This allows us to deal with the large amounts of video data easily as well as get a summary of the past as the video is analyzed. We also extend our incremental clustering algorithm for semi-supervised incremental clustering to deal with situations when labeled data is present *a priori*. We also extend

our incremental clustering algorithm for temporal incremental clustering to be able to find temporal patterns in the video data. This is essential for discovering events that are temporally related as well as for discovering events that are unusual on the basis of their time of occurrence. We develop a cluster algebra that defines operations on clusters from various attribute spaces and give a rough set [1] based analysis that shows that the algebraic operations on the component clusters leads to forming clusters in the higher dimensional spaces that are close approximation of the clusters formed by directly clustering in the higher dimensional spaces. We also define a *usualness* measure of a cluster that defines how usual a cluster is and use it to discover unusual patterns as well as learn usual ones. Our cluster algebra together with the *usualness* measure gives the user a flexible and powerful tool to explore the space of all clusters and their various combinations for multi-perspective summarization and mining of the video data patterns based on the requirement of the system.

Next, we propose an event discovery and anomaly detection framework for learning the usual and detecting the unusual events in the area under observation, given a sufficient amount of training data. We develop a generic segmentation procedure to segment the objects in the form of space-time patches. Our segmentation scheme uses video epitomes [2] to be able to segment a moving object in the form of space-time patches, even when the video has low resolution and lost frames. We use probabilistic Latent Semantic Analysis (pLSA) [3], to discover and learn a pre-defined number of event classes from the training data. We extend pLSA to detect unusual events in novel video clips of that area. We further extend pLSA to *time pLSA* to learn the temporal distribution of the discovered event classes. We show that this framework directly extends to a static multi-camera scenario for learning the correlations among events related in time.

If the area under surveillance is wide, a single camera becomes inadequate to observe the entire area and capture all the events that occur at the various locations of this area. Therefore, multi-camera networks are required to observe and analyze events that occur across space and time in a wide area. Pan-tilt camera networks are better suited for wide area surveillance, as compared to static camera networks as a fewer number of cameras are capable of observing wide areas for surveillance. However, a multi-camera network requires to be calibrated with respect to a global coordinate system for real-time surveillance and activity analysis. In this thesis, we propose a method for distributed self-calibration of a pan-tilt camera network using multi-layered belief propagation. We also propose a distributed self-calibration algorithm for a large static

camera network using multi-layered belief propagation that requires automatic computation of corresponding points between at most three views at a time. We also develop an algorithm for automatic computation of corresponding points among multiple views of the same scene. We use this algorithm to discover the set of cameras that have a common view of the area. Since the cameras can pan and tilt, it is essential to discover the cameras that view a common region every time a camera pans and/or tilts. We show that by applying multi-layered belief propagation, we are able to obtain globally consistent and accurate camera parameters for each camera in the pan-tilt camera network in a distributed manner.

For a calibrated pan-tilt camera network, we propose a real-time distributed framework for composite event recognition. A composite event in a pan-tilt camera network comprises of events that are observed by smaller sub-networks of cameras, either simultaneously or sequentially in different locations across space and time. Distributed recognition of composite events requires distributed multi-camera, multi-object tracking and distributed recognition of events that comprise the composite event. We propose a multi-layered belief propagation based approach for reaching a consensus, within and across sub-networks, on the global identities of the objects that move around in the pan-tilt camera network. This allows us to perform multi-camera, multi-object tracking in a pan-tilt camera network in a distributed manner. For distributed recognition of an event by multiple cameras simultaneously viewing the event, we propose the use of belief propagation to reach a consensus on the event recognized by each camera independently. We propose a hidden Markov model (HMM) [4] based approach for composite event recognition as they are best suited for recognition of a temporal sequence of observations when the state of the system is dynamic and unknown. We also present an application of our framework, where an event implies interaction between a pair of objects. In this case, the composite event is composed of various different interactions between pairs of objects as they move in the complete area under observation of the pan-tilt camera network. We also propose a novel pLSA [3] based algorithm for discovery and learning of pair-wise interactions as well as to detect unusual pair-wise interactions. Our online, real-time composite event recognition framework is capable of recognizing usual and detecting unusual events and composite events in a distributed manner in a pan-tilt camera network.

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