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Multiple Objects Tracking in Video Sequences - Survey

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Abstract- Multiple object tracking is important in research area and in computer vision task. There are many virtual applications like video surveillance, traffic monitoring, robot navigation, security and object recognition. Multiple object tracking can be achieved by detecting objects in individual video frames and then linking across frames. There are several methods proposed for the multiple object tracking. In this paper, we discuss various multiple object tracking methods.

Keywords - Multiple object tracking, Observation model, Dynamic model.

1. INTRODUCTION

Videos are represented as scene, shot and frame. A video shot can be defined as the video frames that consist of continuous action. The frames in video shots are captured by a camera in single operation. The complete video sequence is formed by joining two or more video frames which is input to the video tracking. The main goal of multiple object tracking is the process of segmenting the features of an object from video frames and tracks its motion and shape. Object tracking requires location and shape of the objects in video frames. So, object detection and representation (classification) are the preceding steps of object tracking in computer vision applications [1]. The basic black diagram of object detection and tracking is shown in below figure 1.

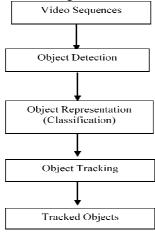


Fig: 1 Basic Block Diagram of Object Detection and Tracking.

The basic steps for tracking an object [2] are described below:

a) Object Detection: Object Detection is a process to identify objects of interest in the video sequence. Object detection can be done by various techniques such as temporal differencing, frame differencing, Optical flow and Background subtraction [3].

b) Object Representation: Object Representation involves various methods such as Shape-based representation, Motion-based representation, colourbased representation and texture-based representation [4] where object can be represented as vehicles, birds and other moving objects.

c) Object Tracking: Tracking is the process of locating moving objects over a time in video sequences. Object tracking can be done by different methods such as [5] Point based tracking, Kernel tracking and Silhouette based tracking.

2. MULTIPLE OBJECT TRACKING

In general, Multiple Object Tracking (MOT) classification is difficult. Existing Visual tracking system has provided a MOT classification based on several factors. MOT mainly classified based on factors like initialization, way of data processing (or learning), mathematical methods and based on state estimation model [6] will be discussed in later sections.

2.1 Based on initialization method

Based on this method MOT classified into two sets those are Detection Based Tracking (DBT) and Detection Free Tracking (DFT).

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Detection Based Tracking (DBT): In this, first of all, objects are localized in each frame. A specific type of object detection (background subtraction) [7] is applied to each frame to obtain object hypotheses. DBT made on a pre- trained object detector which produces object hypotheses and object hypotheses treated as observations. Hence object tracking is used to link detection hypotheses into trajectories. DBT is more popular than DFT [8] why because new objects are discovered and disappearing objects are terminated.

Detection Free Tracking (DFT): It processes observations sequentially. It does not require pretrained object detector to provide object hypotheses. DFT require manual initialization of fixed number of objects. Those fixed objects localized sequentially [9] (i.e. in subsequent frames).

2.2 Based on data processing

Based on processing the data, MOT can be grouped into online tracking and offline tracking.

Online tracking: It uses observations up to the current time instant for estimation process. It is a sequential process that why it is also called as sequential tracking. It gives appropriate results when video stream is obtained sequentially. So, this approach is used in practise.

Offline tracking: It contains both past and future observations for estimation. Observations of all frames are required in advance to estimate the final output. It cannot handle the all frames at a time due its complexity [10]. So, it cannot be used in practise.

2.3 Based on state estimation model

Generally, MOT can be formulated as a multivariable estimation problem. Dynamic and Observation models are used to estimate the states of objects.

2.3.1 Dynamic model

This model specifies the states transition across frames. This model again classified into probabilistic optimization and [11] deterministic optimization. This model is also called as mathematical model.

2.3.1.1 Probabilistic approach

This approach represents object states such as position, velocity (motion) and size with probabilistic distribution. The main aim of probabilistic approach

is to estimate the probability distribution of objects status by using existing observations. Then objects states estimated with this probabilistic distribution. Different types of probabilistic distribution methods are used in MOT [12], such as Kalman filter, particle filter and multiple hypothesis filters.

Kalman filter: It is a set of mathematical equations that provides an efficient recursive state estimation process [12]. It calculates estimations of past, present and also future States. It has feedback control to update the states. Kalman filter algorithm involves two steps, prediction and correction (update step).

The first step uses previous state to predict the current state. The second step uses the current measurements like location of objects to update the state. Kalman filter provides optimal solutions when assumptions of state variable are normally distributed (Gaussian).

Particle filter: Particle filter generates all the models like initial state, noise covariance, number of particles etc for one variable before moving to the next variable. This algorithm uses contours, color features and texture mapping [13]. The drawback of Kalman filter is overwhelmed by using particle filtering.

Multiple hypothesis filters: It is widely used method for solving data association problem in MOT. It is an iterative algorithm. Several frames have been observed for better tracking. Iterations begin with a set of existing track hypothesis. For each hypothesis, a prediction of object's position in the succeeding frame is made. The predictions are compared by calculating a distance measure [14]. It gives optimal solutions in MOT and handles occlusions.

2.3.1.2 Deterministic approach

In this approach, first observations from each frame in the image sequence are obtained. Final estimation of tracking results is obtained by defining the similarities among observations. Various kinds of deterministic optimization have been applied to MOT such as [15] Bipartite graph matching, Dynamic programming, Min- cost Max-flow Network flow, Conditional random field and Maximum weight independent set.

2.3.2 *Observation model*

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It measures similarities between object States and observations. It includes appearance model, motion model, interaction model, exclusion model and occlusion models [16].

3 COMPARATIVE STUDY OF MULTIPLE OBJECT TRACKING METHODS

According to survey [7] Table 1 describes comparative study between DBT and DFT.

| S.No | Parameter | DBT | DFT |
|------|-------------------|---|---|
| 1 | Initialization | Automatic and Imperfect | Manual and perfect |
| 2 | Number of objects | Variable | Number of objects must be fixed here. |
| 3 | Applications | In most of the cases, it is used for a particular type of objects only | It used for any type objects |
| 4 | Advantages | It has the ability to handle the varying number of objects. | No need of object detectors |
| 5 | Disadvantage s | Performance of this approach depends on object detection methods | It requires manual initialization |

Table 2 describes comparative study between Online and Offline Tracking.

| S.No | Parameter | Online tracking | Offline tracking |
|------|------------------------------|---|--|
| 1 | Number of inputs required | Up to time observations | All observations |
| 2 | Procedures (methodology) | Existing trajectories and current observations | It link observations into trajectories |

| 3 | Advantages | It is used in online tasks | It provides theoretical optimal solution |
|---|---------------|--|--|
| 4 | Disadvantages | It suffer from shortage of observation | Delay exists in producing final outputs. |

Table 3 [16, 17, 18] describes Comparative study of various kinds of Probabilistic Approach.

| Probabilistic | Time | Acour | Advantages | Disadvantages |
|-----------------|--------|-------|--------------|----------------|
| 110040110000 | | | Auvantages | Disadvantages |
| Tracking | Comple | acy | | |
| | xity | | | |
| Kalman Eller | | | T4 | :4 . : |
| Kalman Filter | meatum | | | it gives poor |
| | | ate | optimal | performance |
| | | | solution. it | when objects |
| | | | is used to | assumptions |
| | | | track points | not normally |
| | | | in noisy | distributed (|
| | | | images | Gaussian) |
| | | | | |
| Particle Filter | high | high | in complex | due to its |
| | | | background | complexity, it |
| | | | and | is not used in |
| | | | occlusion, | real time |
| | | | it gives | applications |
| | | | good | |
| | | | results | |
| | | | | |
| Multiple | low | low | it has the | in both time |
| Hypothesis | | | capability | and memory, it |
| Filter | | | to deal with | is |
| | | | new object | computationall |
| | | | entries and | y exponential |
| | | | exit | J |
| | | | existing | |
| | | | object | |
| | | | 00jeet | |
| | | | | |

4 CONCLUSION

This paper presents a review of multiple object tracking. The proposed multiple object tracking methods have been explained and compared. Among all those methods, probabilistic approaches are generally used for multiple object tracking. Especially Kalman filter is widely used due to its

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accuracy and optimal solutions of objects tracking. Advantages and disadvantages of each technique also described in this paper.

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