A Review : Object Detection and Object Tracking Methods

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ABSTRACT:
Moving Object Tracking is one of the testing issues in the field of computer vision, surveillance, traffic monitoring and so forth. Distinguishing the objects in the video and tracking its motion to recognize its qualities has been a demanding examination zone in the field of image processing and computer vision. The objective of object tracking is to place a moving object in sequential video frames. The goal of this paper is to review of different moving object detection and object tracking systems. This paper includes various methods for object detection and object tracking. Comparing all the methods researcher can use the better method for their future research.

Keywords: Object detection, Object tracking, Point tracking, Kernel tracking, Silhouette tracking, HSV (hue, saturation, value).

I. INTRODUCTION

Object tracking in video frames is an important topic in the field of computer vision and various research fields. Video surveillance has been a very active research topic in the last few years due to its growing importance in security, military applications etc. The vast amount of data involved makes it infeasible to guarantee alert monitoring by human operations for long periods of time due to boredom and fatigue. As a result, video feeds are usually archived for forensic purposes in the event suspicious activities take place. Videos are actually sequences of images, each of which called a frame. The fundamental in object tracking is object detection, object classification and then finally object tracking. Moving object detection is the first and important step for many video analysis task. Moving object detection aims at extracting moving objects that are important in video sequences. Frequently used methods for moving object detection are background subtraction, temporal frame differencing. Background subtraction is an efficient method for object detection compare to frame differencing. The next step is object tracking. It is an important task within the field of computer vision. It is used not only for video surveillance but also for traffic control, medical imaging, gesture recognition etc. The approaches for object tracking are as follows.

A. Object Detection
Object Detection is to identify objects of interest in the video sequence and to cluster pixels of these objects. Object detection can be done by various techniques such as frame differencing, Optical flow and Background subtraction.

B. Object Classification
Moving regions detected in the video may matches to different objects in real world such as humans, vehicles etc. The approaches to classify the objects are Shape-based classification, Motion-based classification and Color based classification.

C. Object Tracking
The goal of a moving object tracking is to create the path for an object above time by detecting its position in every single frame of the video.
Following are the challenges that should be taken care in object tracking.
• Noise in images.
• Complex object shapes / motion.
• Partial and full object occlusions.
• Scene illumination changes.

II. OBJECT DETECTION
Object detection can be achieved by building a representation of the scene called the background model and then finding deviations from the model for each incoming frame. Any significant change in an image region from the background model signifies a moving object. The pixels constituting the regions undergoing change are marked for further processing. Usually, a connected component algorithm is applied to obtain connected regions corresponding
to the objects. This process is referred to as the background subtraction.[9]. Below figure shows the detection of moving object using reference background.

![Diagram of object detection process](image)

**Fig.1 The flow chart of moving object Detection**

Background subtraction strategy gives exact results for object detection when the reference background is known. It has following approaches:

**A. Mixture of Gaussians**

The background is modeled pixel by pixel using a mixture of Gaussians. Each pixel location is represented by a number (or mixture) of Gaussian functions that sum together to form a probability distribution functions. This moving object detector has also the ability to detect and remove shadows, a great source of false alarms in this kind of visual detectors [1].

**B. Median filtering**

For finding foreground images, Median filtering [5] is the commonly used technique. The background estimation is done by the median at each pixel location of all the frames in the buffer because the value of the median at each pixel stays in the background for more than half of the frames in the buffer. The main advantage of this method is that, the computation speed is high[11].

### III. OBJECT CLASSIFICATION

The Moving regions detected in the video may matches to different objects in real world such as humans, vehicles etc. It is very important to identify the type of a detected object to track it reliably. Object classification has the following approaches:

1. Shape based classification.
2. Motion based classification.

**A. Shape based classification**

Common features used in shape-based classification schemes are the bounding rectangle, area, silhouette and gradient of detected object regions. Different descriptions of shape information of motion regions such as representations of points, box and blob are available for classifying moving objects. Input features to the network is mixture of image-based and scene-based object parameters such as image blob area, apparent aspect ratio of blob bounding box and camera zoom. Classification is performed on each blob at every frame and results are kept in histogram [8].

**B. Motion based classification**

Non-rigid articulated object motion shows a periodic property, so this has been used as a strong cue for moving object classification. Optical flow is also very useful for object classification. Residual flow can be used to analyze rigidity and periodicity of moving entities. It is expected that rigid objects would present little residual flow whereas a non-rigid moving object such as human being had higher average residual flow and even displayed a periodic component [8].

**C. Color based classification**
All video frame formats are based on different color spaces model. The data of different frame can be stored in distinctive color spaces ranging from gray scale, RGB, YCbCr and HSV (hue, saturation, value) color spaces. The data is stored in each frame is the brightness in each spectral band. Color images are denoted as red(R), green (G) and blue (B) layers or RGB. Other distinctive color spaces used in the area of object tracking are YCbCr and HSV[12].

### IV. OBJECT TRACKING

The goal of a moving object tracking is to create the path for an object above time by detecting its position in every single frame of the video. Few object tracking methods are as below:

![Object Tracking Diagram](image)

**Fig 2. Object Tracking Methods**

#### A. Point Tracking approach

Point Tracking can be defined as the correspondence of detected objects represented by points across the frames[9]. There are two methods of correspondence methods namely – deterministic and statistical methods [9]. Point Tracking is a difficult problem particularly in the existence of occlusions, false detections of object. Recognition of points can be done simply by thresholding, at of identification of these points. Point Tracking is capable of dealing with tracking very small objects only.

![Point based tracking](image)

**Fig 3. Point based tracking[9]**

Some approaches of point tracking are as below:

- **Kalman Filter**
  - Kalman filter are based on Optimal Recursive Data Processing Algorithm [9]. Here Gaussian state distribution is assumed. Kalman filtering is composed of two stages, prediction and correction. Prediction of the next state using the current set of observations and update the current set of predicted measurements. The second step is gradually update the predicted values and gives a much better approximation of the next state. Kalman Tracking is capable of dealing with:
    - Kalman filters always give optimal solutions.
    - Another potential approach is to handling noise
    - Tracking is applicable only for single and multiple objects.

![Basic step of Kalman Filter](image)

**Fig 4. Basic step of Kalman Filter**

- **Particle Filter**
  - Particle Filter algorithm can solve object tracking under non-Gaussian or non-linear conditions, its core conception is Monte Carlo integration. In most cases, it can obtain a better tracking result compared to mean-shift[10]. In the method of staring with population of particle, each will assign value to no variables and weight of 1.
At each step, the procedure will be:
1. Generate samples to represent the initial probability
2. Using the prior equation, predict the next state
3. Using the observation, get the weights for the states computed. Predicted states (from step 2) along with the weights collectively represent the state distribution
4. Resample it so as to have the uniformly distributed current state omitting the least-significant representation
5. Continue steps 2 through 4, till all the observations are exhausted.

c. Multiple Hypothesis Tracking (MHT)
The MHT algorithm is based on motion correspondence of several frames together. Better results are obtained if correspondence is established observing several frames rather than using only two frames. The MHT algorithm upholds several suggestions for each object at each time. The final track of object is the most likely set of correspondences over time period of its observation [9]. MHT is an iterative algorithm. Emphasis starts with a set of existing track speculations. Every theories is a group of detach tracks. For every speculation, a prediction of object's movement in the succeeding casing is made. The predictions are then analyzed by computing a separation measure.

B. Kernel Based Tracking Approach
Kernel tracking is typically performed by computing the motion of the object, which is represented by a primitive object region, from one frame to the next. The object motion is generally in the form of parametric motion (translation, conformal, affine, etc.) or the dense flow field computed in subsequent frames[9].

a. Simple Template Matching
Template matching is a brute force technique for analyzing the Region of Interest in the feature. In template matching, a reference image is checked with the frame that is differentiated from the feature. Following could be possible for single question in the feature and covering of item is carried out incompletely. Layout Matching is a procedure for transforming computerized images to discover little parts of an image that matches, or identical model with an image (format) in each one frame. The matching strategy contains the image format for all conceivable positions in the source image and ascertains a numerical file that points out how well the model fits the picture that position. It capable of dealing with:
  • Tracking single image.
  • Partial occlusion of object.
  • Necessity of a physical initialization.

b. Mean Shift Method
Mean-shift tracking [13] tries to find the area of a video frame that is locally most similar to a previously initialized model. The image region to be tracked is represented by a histogram. A gradient ascent procedure is used to move the tracker to the location that maximizes a similarity score between the model and the current image region. In object tracking algorithms target representation is mainly rectangular or elliptical region. It contain target model and target candidate. To characterize the target color histogram is chosen. Target model is generally represented by its probability density function (pdf). Target model is regularized by spatial masking with an asymmetric kernel.

C. Silhouette Based Tracking Approach
Objects having composite shapes for instance, hands, head, and shoulders, are can't be decently characterized by geometric shapes. Silhouette based methodology will give flawless portrayal of state of those objects. The point of the silhouette based tracking is to discover the object area by method for an object model. This model checks the object district in each one casing. Model can be spoken to as color histogram, object edges or shape. We classify silhouette tracking into two classes, specifically, shape matching and form tracking.

a. Contour Tracking
Contour tracking techniques create an unique contour in the foregoing frame to its new position in the present edge, covering of object between the current and next casing. Contour tracking is in the form of state space models[9].

b. Shape Matching
This approach checks for object model in the existing frame [9]. Shape matching performance is similar to template based tracking in kernel approach. Another approach to Shape matching is to find matching silhouettes in two successive frames. Detection based on Silhouette is carried out by background subtraction. Models object are in the form of density functions, silhouette boundary, object edges [9]. Shape matching execution is like template based tracking in kernel approach. An alternate approach to Shape matching is to discover matching silhouettes in two successive frames. Detection based on Silhouette is carried out by background subtraction. Models object are in the form of density functions, silhouette boundary, object edges [9].
IV. COMPARISON BETWEEN VARIOUS TRACKING METHODS

Table 1. Comparison between various tracking methods

<table>
<thead>
<tr>
<th>Sr.No.</th>
<th>Methodology</th>
<th>Advantages</th>
<th>Limitation</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>Kalman Filter[13]</td>
<td>Used to track points in noisy images</td>
<td>State variables are normally distributed (gaussian)</td>
</tr>
<tr>
<td>2</td>
<td>Particle Filter[14]</td>
<td>No Limitation for the complexity of distributions</td>
<td>The most efficient number of particles cannot be calculated</td>
</tr>
<tr>
<td>3</td>
<td>MHT(Multiple Hypotheses Tracking)[13]</td>
<td>Able to deal with entries of new object and exit existing object.</td>
<td>Computationally exponential both in time and memory.</td>
</tr>
<tr>
<td>4</td>
<td>Simple Template Matching[15]</td>
<td>The system is capable of handling entry and exit of object.</td>
<td>It is difficult to track multiple objects at the same time.</td>
</tr>
<tr>
<td>5</td>
<td>Mean Shift Method[16]</td>
<td>Does not assume any predefined shape on data clusters.</td>
<td>Inappropriate window size can cause modes to be merged, or generate additional “shallow” modes</td>
</tr>
<tr>
<td>6</td>
<td>Contour Tracking[15]</td>
<td>The selection of contour edges are not affected by noisy edges or small cross striped textures</td>
<td>The method cannot preserve the contour which has high curvature.</td>
</tr>
<tr>
<td>7</td>
<td>Shape Matching[15]</td>
<td>Less sensitive to appearance variations</td>
<td>objects of arbitrary shapes cannot be detected</td>
</tr>
</tbody>
</table>

V. CONCLUSION

In this paper, an across the board literature survey on different moving object tracking routines has been introduced. Object is followed predominantly on the bases of object detection, object classification and tracking. Near examination is generally accomplished for point tracking, kernel and Silhouette tracking calculations. For example point tracking includes detection in every casing, while kernel based or shape based tracking requires detection when object first shows up in the scene. This survey highlights the features of algorithm for researchers in the area of moving object tracking.

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REFERENCES


[16] wikipedia.org