

# Object Tracking System Using Mean Shift Algorithm and Implementation on FPGA

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**Abstract-** Various attempts have been made to implement human vision in real-time computer vision. In most cases, however, this capability has been found to be too complex to implement on a practical systems. In short, human vision capability looks like a simple system but in reality it is difficult to model. One of the difficulties is the computational complexity required to process large amounts of data in real-time, and consequently, the design of real-time object tracking using vision remains a challenging problem. However, in this thesis, we propose and implement hardware architecture for real-time object tracking system. We have chosen Mean-Shift algorithm for implementation. We will implement firstly this algorithm on MATLAB and simulated in different environments to verify the performance. After approving its robustness we implemented the algorithm on hardware.

**Keywords** -object tracking; Mean-Shift Algorithm ;FPGA;

**INTRODUCTION:** Computer vision can be defined as a methodology that provides vision capabilities to a machine. Various attempts are being made in the computer vision industry to model human vision and to apply the same techniques to machine vision. In most of these cases researchers have had difficulty with the complexity of implementing such a capability on practical systems. These implementation problems are due to various constraints, including the lack of needed computational capability to process large amounts of data in a real-time, object motion, noise, complex background and the inability to model human vision exactly. Real-time object tracking remains a challenging problem in the computer vision industry and hence there is a huge interest on the market to make technical equipment “smart” and “self learning”[12].

## Research Objective

The objectives of the current dissertation research are

1. Verification of tracking results by software implementation of Mean-Shift Algorithm on MATLAB.
  2. To propose a novel hardware Architecture for Mean-Shift algorithm. The algorithm should be easy for hardware implementation.
- To confirm experimentally soundness of the proposed hardware Architecture implemented through Hardware Description Language.

## Mean-Shift Algorithm

We assume in the sequel the support of two modules which should provide (a) detection and localization in the initial frame of the objects to track targets [5] and (b) periodic analysis of each object to account for possible updates of the target models due to significant changes in color.

Color Representation

Target Model :- Let  $\{x_i^*\}_{i=1\dots n}$  be the pixel locations of the target model, centered at zero. We define a function  $b:R^2 \rightarrow \{1\dots m\}$  which associates to the pixel at location  $x_i^*$  the index  $b(x_i^*)$  of the histogram bin corresponding to the color of that pixel. The probability of the color  $u$  in the target model is derived by employing a convex and monotonic decreasing kernel profile  $k$  which assigns a smaller weight to the locations that are farther from the center of the target. The weighting increases the robustness of the estimation, since the peripheral pixels are the least reliable, being often affected by occlusions (clutter) or background. The radius of the kernel profile is taken equal to one, by assuming that the generic coordinates  $x$  and  $y$  are normalized with  $h_x$  and  $h_y$  respectively. Hence, we can write [2]

$$\hat{q}_u = C \sum_{i=1}^n k \left( \|x_i\|^2 \right) \delta [b(x_i^*) - u] \quad \text{-----(1)}$$

Where  $\delta$  is Kronecker delta function.

$$\delta(a) = \begin{cases} 1 & \text{if } a = 0 \\ 0 & \text{otherwise} \end{cases} \quad \text{------(2)}$$

The normalization constant  $C$  is derived by imposing the

condition  $\sum_{u=1}^m \hat{q}_u = 1$ , from where

$$C = \frac{1}{\sum_{i=1}^n k \left( \|x_i^*\|^2 \right)}, \quad \text{------(3)}$$

Since the summation of delta functions for  $u=1\dots m$  is equal to one.

Target Candidates: Let  $\{x_i\}_{i=1\dots n_h}$  be the pixel locations of the target candidate, centered at  $y$  in the current frame. Using the same kernel profile  $k$ , but with radius  $h$ , the probability of the color  $u$  in the target candidate is given by

$$\hat{p}_u(y) = C_h \sum_{i=1}^{n_h} k \left( \left\| \frac{y-x_i}{h} \right\|^2 \right) \delta[b(x_i) - u] \quad \text{---(4)}$$

where  $C_h$  is the normalization constant. The radius of the kernel profile determines the number of pixels (i.e., the scale) of the target candidate. By imposing the condition

that  $\sum_{u=1}^m \hat{p}_u = 1$  we obtain

$$C_h = \frac{1}{\sum_{i=1}^{n_h} k \left\{ \left\| \frac{y-x_i}{h} \right\|^2 \right\}}, \quad \text{-----(5)}$$

Note that  $C_h$  does not depend on  $y$ , since the pixel location  $x_i$  is organized in a regular lattice,  $y$  being one of the lattice nodes. Therefore,  $C_h$  can be precalculated for a given kernel and different values of  $h$ . [12]

Kernel

Flat Kernel

$$k(x) = \begin{cases} 1 & \text{if } \|x\| \leq 1 \\ 0 & \text{otherwise} \end{cases} \quad \text{-----(6)}$$

Gaussian Kernel

$$k(x) = \exp(-\|x\|^2) \quad \text{-----(7)}$$

Epanechnikov Kernel

$$k(x) = \begin{cases} \frac{1}{2} c_d^{-1} (d+2)(1-x) & \text{if } x \leq 1 \\ 0 & \text{otherwise} \end{cases} \quad \text{---(8)}$$

Where  $d$  is the number of dimension and  $C_d$  is the area of a unit circle in  $d$  dimension. [5]

**MATLAB Implementation:**

We have taken several different videos with different resolutions and different environmental condition. We have applied the algorithm and results are shown below. We have compared every video taking two bin sizes 16 and 8 and also calculated false detection rate and frame rate

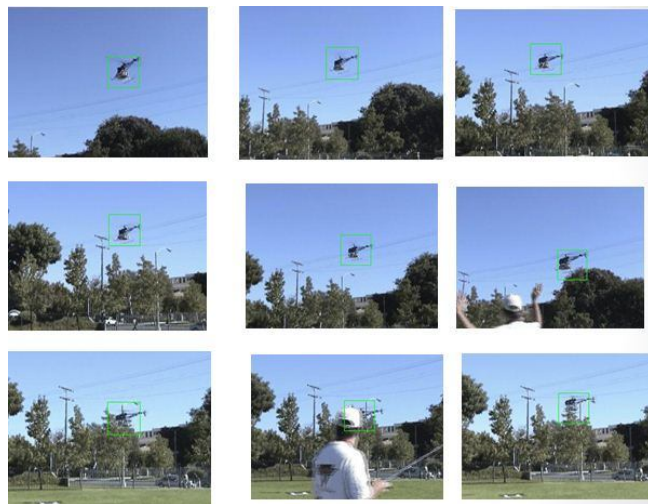


Figure 1 Helicopter is completely track



Figure 2 Car is completely track

**HDL Implementation**

**Top Level Architecture for Object Tracking**

The Object Tracking deals with tracking of objects which are of interest. The algorithm followed for implementation of tracking is Mean-shift. Mean-shift uses Histogram and Bhattacharyya coefficient for tracking the position of object in subsequent frames of videos by taking the fact into consideration that object will not displace much in continuous frames.

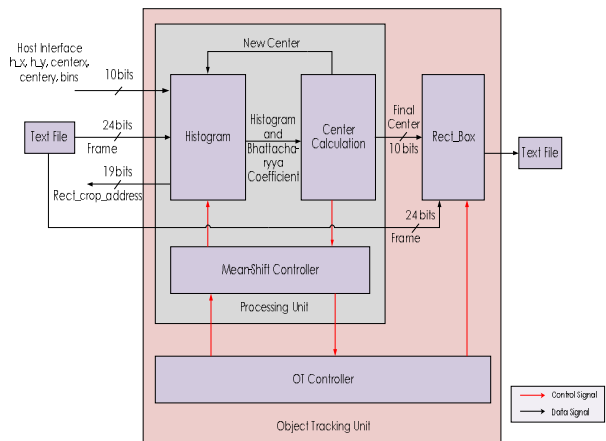


Figure 3 Mean-shift algorithms for Object Tracking [5]

The above is the top level architecture which we have proposed for the hardware implementation of Mean-shift algorithm for Object Tracking. To provide a RGB image as an input to this architecture we have converted an image into a text file representing R,G and B values of each pixel in hexadecimal format[5]. Each pixel is read by histogram block as a frame. A frame is a 24-bit wide R,G,B values, 8-bits each, of a pixel representing intensity values of a pixel

of a colored image. Host interfaces are inputs given by user in order to specify the object to be tracked in an image

**Histogram**

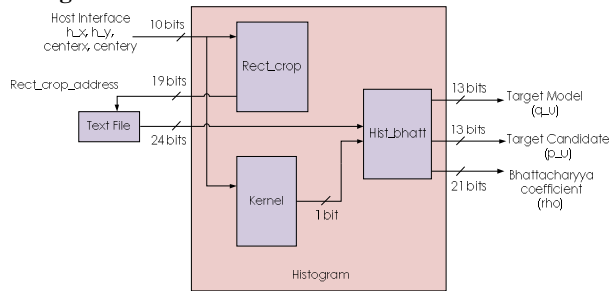


Figure 4 Histogram [5]

Sub modules of Histogram module are – Rect\_crop, Kernel, Hist\_bhaff. Rect\_crop sub module is responsible for generating address of pixels of object of interest using (5.1). As we have explained early that for processing of algorithm we do not require complete image, rather we just want only object of interest. So we need to crop our image and extract object of interest. For that we have designed an equation which calculates addresses of only those pixels related to crop image. [5]

Where  $x_1, x_2, y_1$  and  $y_2$  are corner pixel location of cropped image. When Rect\_crop calculates address of the desired pixel at the same time kernel sub module calculates value of the kernel for the same pixel. For our implementation we have taken Flat Kernel (3.6). Flat Kernel is either '0' or '1' so only 1-bit is required. In Hist\_bhaff sub module histogram for the desired object and Bhattacharyya Coefficient are calculated. Histogram is stored in histogram buffer. Size of the histogram buffer depends on the number of bins provided via host interface input. In our case we have taken it 8 bins, so the size of the histogram buffer would be  $bins * bins * bins = 512$  locations. When pixel intensity value and kernel value reaches to the Hist\_bhaff sub module, address for the histogram buffer can be calculated by pixel intensity value. In order to calculate histogram kernel generated is added to the location addressed by Hsit\_bhaff sub module in histogram buffer.

**Center Calculation**

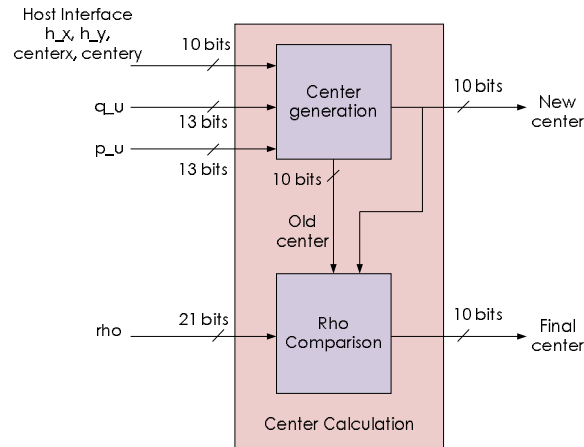


Figure 5 Center Calculation [5]

After the calculation of Target Model and Target Candidate a signal named hist\_done enables Center Generation sub module. This sub module is the heart of the Object Tracking

Architecture because it predicts the next possible center which is stored as new\_center. In Center Generation sub module weight for each pixel of Object of Interest is calculated. Then every weight is multiplied with normalized pixel location in order to predict new\_center.

Rho Comparison is the sub module in which Bhattacharyya coefficient calculated by Histogram module at old\_center and new\_center is compared. Comparison tells us which center is the correct center and it also tells us that shall we go for iteration loop or not. So from the Rho Comparison sub module, exact center is finalized which is stored as final\_center and given to the Rect\_box module to draw the rectangular box around tracked moving object. Then Rect\_box module writes the image pixel values into a output text file.

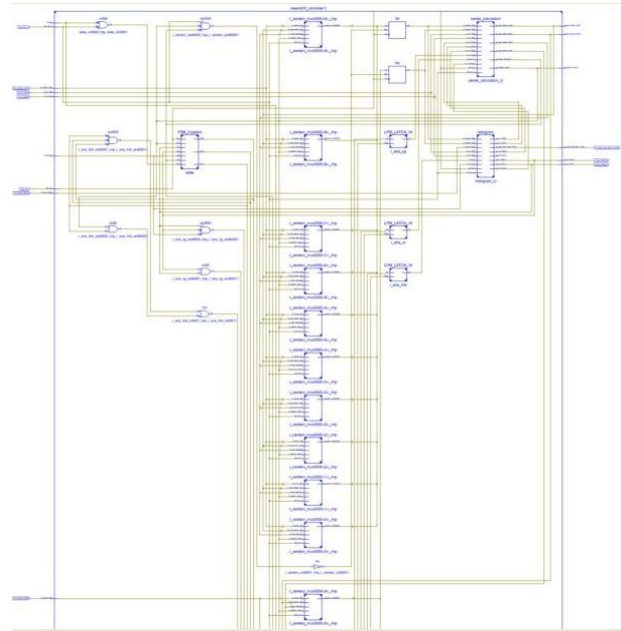


Figure 6 RTL Schematic of Top Module

Resources	Top Module
RAMs	1
Multipliers	6
Adder/ Subtractor	60
Counter	5
Registers	7427
Latches	1041
Comparators	28
Multiplexers	5
Xors	2

Table 1 Synthesis Report for architecture

Resources	Top Module
Slice Registers	20610/28800 (71%)
Slice LUTs	13028/28800 (45%)
IOs	131/480 (27%)
Block RAM	3/60 (5%)

Table 2 Device Utilization Report for architecture

**CONCLUSION**

The objective of our work was to implement hardware design for real-time object tracking. Several algorithms for detection and tracking of moving objects have been proposed worldwide. Most of the researchers implemented software model of all available algorithms but only few of them attempted for their hardware model. Inspiring by wide range of application and excellent performance results presented by other researches we have chosen Mean-Shift algorithm for tracking. In first phase we implemented software model. We verified the robustness, capability of tracking object under partial occlusion, intensity variation and in random environment. In second phase we have implemented hardware architecture for the design and implemented hardware model using HDL

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