

McGILL UNIVERSITY

DEPARTMENT OF ELECTRICAL & COMPUTER ENGINEERING

ECSE 457 - FINAL REPORT

**Research & Development of a Real-Time Object
Tracking System**

Authors:

Benjamin BROWN
benjamin.brown2@mail.mcgill.ca
260450182

Taylor DOTSIKAS
taylor.dotsikas@mail.mcgill.ca
260457719

Supervisors:

Warren GROSS, PROF.

Arash ARDAKANI

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Abstract

This project aimed to research object tracking algorithms, and implement the algorithm best suited to meet project requirements in both software and hardware. The software implementation used static input videos and the hardware implementation used a real-time input video stream. Phase 1 of the project focused on algorithm research and software implementation, and is documented in [15]. Phase 2 of the project focused on hardware implementation, and this report encompasses all of Phase 2 from the design process to the final results. It was found that hardware implementation of the algorithm provides a much faster way of tracking an object when compared to the equivalent software implementation. It was also found that using a background subtraction, delta frame based algorithm to determine object position is not the best choice for real-time situations, due to the inability to cope with lighting and background variations. It was also found that while a Kalman filter does greatly improve object tracking results, it comes at a heavy cost in terms of hardware utilization.

Acknowledgments

We would like to thank Professor Warren Gross and Arash Ardakani for overseeing this project and providing advice, insight, and direction over the course of the year. We would also like to thank Analog Devices for providing us with the Altera DE2 board needed for hardware implementation, and in particular Leah Magaldi for being our main point of contact. Finally, the example TV decoder Verilog code provided by Terasic was a major asset for hardware implementation, we would like to thank them for making it open and available to developers like ourselves.

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1 Abbreviations & Notation

FPGA - Field Programmable Gate Array

VGA - Video Graphics Array

HDL - Hardware Description Language

2 Introduction

The ability to determine the position of an object in a scene is a highly relevant challenge for industries like surveillance and robotics. This task represents a challenge from an engineering perspective due to the real-time timing constraints placed on these systems, and the large amount of data that must be processed when working with video data. In order to overcome these challenges, custom hardware implementation has become an increasingly popular solution as demonstrated in [1], [2], and [3].

TODO: This section needs more work.

3 Background

This section contains the prerequisite information regarding video tracking needed to understand the system architecture and design. For background information regarding the basics of video processing, Kalman filtering, fixed-point representation, and optical flow, please see [15].

TODO: Is just referencing this background information enough? Or do we need to repeat the Kalman equations, delta frame generation, and other topics again?

3.1 Moving Average Filter

A moving average filter replaces the current input data sample with a mean of some number of past input data samples. The number is referred to as the moving average filter length, N . This type of filter is particularly useful when data samples are arriving in a time series (i.e. at constant time intervals) as it will remove outliers, creating a smoother trend in data samples. Given that p_i is the current input data sample, the filtered results, p'_i , is given using the following equation,

$$p'_i = \frac{\sum_{k=i}^{i+N} p_{k-N}}{N} \quad (1)$$

3.2 Saturation Filter

A saturation filter, in the context of this project, refers to either flooring a grayscale pixel intensity to a minimum value or ceiling the grayscale pixel intensity to a maximum value, if it is below or above a threshold. The pseudo-code for the saturation filter is given below.

```
function [output] = saturation_filter(input, min, max, thresh)
    if (input > thresh)
        output = max;
    else
        output = min;
```

```

    end
end

```

This makes the data set binary, in the sense that all values are one of two values, and makes data processing significantly easier. In the context of object tracking, values above the threshold indicate an object present in the scene, and values below indicate no object present.

3.3 Determining Position

The algorithm implemented in software, and presented in [15], for determining the (x, y) position of an object in the delta frame used a raster scan technique to determine the leftmost, rightmost, top, and bottom pixels. By intersecting two lines formed between these points, the center of the object can be estimated. It was found that despite the success of this algorithm in software, it would not be conducive to hardware implementation. This is mainly due to the fact that in the software implementation, data arrived in discrete frames from MATLAB's `VideoReader` class at constant time steps. In hardware implementation, data is handled at the pixel level, with a constant stream of pixels being placed in and extracted from a FIFO frame buffer storage module (to be discussed later). Thus using an algorithm that requires an entire frame requires more memory and extra logic. A new algorithm was developed to determine the local of the object.

This algorithm assumes that pixel data arrives in binary format; either a string of all zeros representing no object present or a string of all ones representing an object present. Note that this data format is achieved by using the saturation filter just discussed. The algorithm tests if the pixel is an object pixel or not, and if it is the x and y coordinates of the pixel are each added to a rolling summation (x_{sum} and y_{sum}). A counter, n , is also incremented. At the end of the frame, the rolling summations are divided by the counter, the position is outputted, and the three values are cleared. This algorithm is essentially just taking an average of the (x, y) coordinates of the object pixels as the center of the object.

$$x = \frac{x_{sum}}{n} \quad (2)$$

$$y = \frac{y_{sum}}{n} \quad (3)$$

Note that this algorithm achieves best results when the object is highly symmetric.

3.4 Video Pipeline

The phrase *video pipeline* refers to a series of image processing modules that exist between the video input device (e.g. the camera) and the video output device (e.g. the display). The pipeline can be implemented in software or hardware, but for the scope of this report the pipeline will refer to hardware implementation, and modules will be referred to in the Verilog sense. Modules in the video pipeline generally consist of decoding and encoding the video data into various formats, and performing video processing (e.g. applying algorithms of interest) in the middle. Implementing a video pipeline is directly coupled with implementing an object tracking algorithm in hardware, as there are no libraries or classes to convert and store the incoming video data when working at this low of a level.

3.5 The VGA Interface

The ADV7123 High-Speed Video DAC converts three 10-bit RGB digital signals to the corresponding analog signals needed to transmit video data using the VGA interface. To give the user flexibility with timing and resolution, it does not generate the digital synchronization signals. These two signals are horizontal and

vertical synchronization (sync). To properly generate these signals and thus drive the VGA interface, the VGA timing specifications must be understood.

A VGA display generates images using a raster scan technique. The display is a grid of pixels, and each pixel is displayed individually starting at the top left corner, moving left to right in rows, and ending in the bottom right corner. This operation occurs at a speed known as the refresh rate. The refresh rate is fast enough such that the user perceives an entire image display instantaneously, despite the fact that the image is being displayed discretely per pixel. Due to the motion just described, VGA timing specifications are described in terms of horizontal (i.e. the rows) and vertical (i.e. the number of rows) parameters. VGA resolution is described in terms of the number of visible pixels in a single row, and the total number of visible rows in the display. It is important to note that the number of visible pixels and rows is less than the total number of pixels and rows, as there are blanking periods due to timing constraints. The final term that should be defined is the pixel clock, f_{pixel} . The pixel clock is used to count each pixel in the grid, and is much faster than the refresh rate. In fact, the refresh rate of a VGA display can be related to the pixel clock and VGA resolution,

$$f_{refresh} = \frac{f_{pixel}}{H_{pixels} \times V_{lines}} \quad (4)$$

Figure 1 shows the timing specifications, and has four notable regions: front porch, back porch, sync, and active video [16]. As the figure shows, the front porch, back porch, and sync regions form what is known

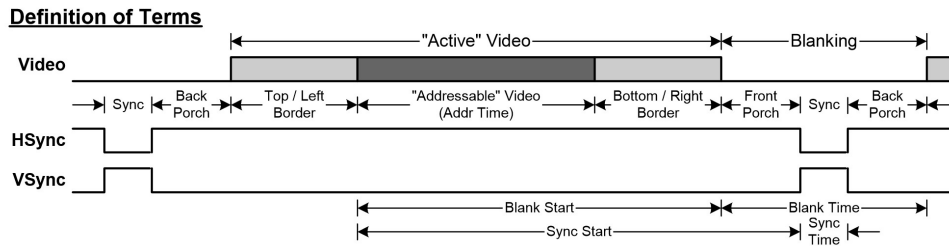


Figure 1: VGA Timing Signals [16]

as the blanking period, since no valid video data is displayed during this period. The horizontal sync signal must be pulsed (high or low depending on polarity) prior to the valid data is displayed on each row, for a specific amount of time depending on resolution. Similarly, the vertical sync signal must be pulsed prior to the entire frame is displayed. For the typical 640×480 VGA resolution, the timing constraints shown in Table 1.

Note: Some specifications, such as [16], add two more regions denoted left and right borders. However, since they are functionally irrelevant for this application, they have been lumped into the front and back porches respectively for this discussion.

Table 1: 640×480 VGA Timing Specifications [16]

Region	Time	Pixels
Hor. Front Porch	.636 us	16
Hor. Sync Period	3.813 us	96
Hor. Back Porch	1.906 us	48
Hor. Visible	25.422 us	640
Ver. Front Porch	0.317 ms	10
Ver. Sync Period	0.064 ms	2
Ver. Back Porch	1.048 ms	33
Ver. Visible	15.523 ms	480

4 Requirements

Due to the fact that the scope of the project did not change between Phases 1 and 2, the requirements listed in this section are largely based off of the Phase 1 report [15]. Some additional requirements were added at the beginning of Phase 2 when hardware implementation became clearer.

4.1 Time Frame

The project shall be completed by December 2015. The project will be broken into two phases. Phase 1 of the project will be January - April 2015, and Phase 2 of the project will be September - December 2015.

4.2 Algorithm

An algorithm shall be chosen such that the system is capable of tracking an object that is in motion, in real-time. The algorithm will be simple enough such that it can be implemented in hardware within the project's duration and within the scope of the author's skills. To ensure this, the chosen algorithm will contain basic operations in order to ease hardware implementation. The algorithm will be chosen during Phase 1 of the project.

4.3 Software

Software implementation of the algorithm shall be performed during Phase 1 of the project to act as a proof of concept for the algorithm, and to understand the strengths and weaknesses of the algorithm. Software implementation shall be performed in MATLAB. Upon completion, the software will be able to track an object in motion with a clear and visible cursor, for an input video that satisfies all assumptions and prerequisite conditions the algorithm may have.

4.3.1 Floating-Point

The first iteration of the software implementation shall be in floating-point to provide a high-level structure of the algorithm. The software shall not have any dependencies on MATLAB (i.e. use built in functions) with the exception of simple helper functions such as `eye`, `zeros`, `round`, and `fix`. MATLAB's `VideoReader` and `VideoWriter` class may be used for I/O.

4.3.2 Fixed-Point

The second iteration of the software implementation shall be in fixed-point to make hardware implementation easier and more efficient. No values in the fixed-point software shall have fractional portions (e.g. reals, floats, or doubles) with the exception of constants declared directly before conversion and preparing data for MATLAB's `VideoWriter` class as output.

4.4 Hardware

Hardware implementation of the algorithm shall be performed during Phase 2 of the project. The completed hardware implementation of the algorithm shall be the final system. The system will contain embedded hardware which the algorithm has been implemented on. The system will include a video camera providing a live feed of the chosen scene. Upon completion, the system will be able to track an object in motion in real-time and identify the object in motion on a display with a clear and visible cursor. All components of the system shall be as low priced as possible.

4.4.1 Platform

The system hardware shall be an Altera DE2 breakout board. This breakout board contains video input/output peripherals, off-chip memory, and an FPGA (Altera Cyclone IV) that contains more than enough resources for the system.

4.4.2 Method

The system hardware shall be implemented using Verilog HDL. The sample designs included with the Altera DE2 breakout board include Verilog modules for video input decoding. In order to utilize these resources and avoid mixing languages, Verilog is the logical choice.

5 Design

5.1 Generating VGA Output

Make sure mention the design references for this.

5.2 Simple Video Pipeline

This is where you should cite [13] and talk about the video input wrapper module.

5.3 Storing the Base Frame

This is where you should cite [14] and talk about the SRAM.

5.4 RGB to Grayscale Conversion

Simple discussion here.

5.5 Modified Video Pipeline

Talk about moving the RGB conversion further upstream prior to the SDRAM such that we can work in RGB.

5.5.1 Delta Frame Generation

Discuss the poor man's absolute value implemented and the saturation filter.

5.5.2 Moving Average Filter

Show the two pictures of the before/after moving average filter here.

5.6 Measuring Object Position

Make sure to mention it's using the VGA module counters.

5.7 Kalman Filter

Lots of meat here.

6 Testing & Verification

TODO: This section.

7 Impact on Society

Since the scope and application of the project did not change between Phases 1 and 2, discussion of the project's impact on society is entirely based off of the Phase 1 report [15].

Our dedicated hardware object tracking system will have some impacts in society, but no profound negative impacts. The most notable impact will be its automating effect in the applications it is integrated with. Its introduction to another system will fulfill a task that could have been previously accomplished by a human or even by software previously embedded within the system. The consumer would have to purchase our hardware to implement our system. Our choice of hardware, the Altera DE2 Board has an upfront cost of around \$500. This can be a substantial price for the consumer but the performance benefits outweigh that of embedded software, and in the case of surveillance for example, the cost of running our system will be less than paying a worker to watch a video feed for objects. The introduction of new technology that makes a manned job obsolete is often viewed as a negative impact on society, but this sort of progress is inevitable.

In the design process so far, we have not made any sort of environmental impact as we have just been developing software. Our system is going to be a combination of previously manufactured parts which were presumably made with the environment in mind. There are so many electronics already in use in society today, the addition of our system's environmental impact can not be quantified. Our product does not produce waste nor does it need constant physical additions. It requires a small constant supply of electricity. When it is no longer needed it can be properly recycled in the same way that computers and other electronics are. Our system does not pose any health or safety risks.

As mentioned earlier, the value of our device comes from the additional performance gained by a secondary system using it. It is possible that the secondary system has some sort of malicious intent but our product cannot be responsible for any negative consequences the secondary system produces.

8 Allocation of Work

9 Conclusion

The final hardware implementation was a success, and the system can display a red dot that shows the (x, y) measurement of the object in the frame as well as green dot showing the improved (x', y') position the Kalman filter produces. Due to the underlying principles of the delta frame generation, the system is an object tracking algorithm and not a motion tracking algorithm. However, the Kalman filter theoretical model assumes constant velocity. This is why when the object stops moving, the green dot slowly converges to the red one. This shows the system favoring the measurements over the model to reduce error, and demonstrates the Kalman filter functioning properly. When the object is moving with constant velocity, the Kalman filter provides a smoother, improved output.

9.1 Future Work

9.1.1 Underlying Algorithm

The system could be significantly improved by using an object measurement algorithm that has fewer requirements and assumptions regarding the environment. To produce a valid delta frame, the scene background must have constant lighting and no objects in it. In a real-time situation, we found it very difficult to ensure these requirements were met.

9.1.2 Motion Tracking

The delta frame generation algorithm detects objects in the scene regardless of whether or not they are in motion. However the Kalman filter assume a constant velocity model, which means that when it receives measurements they might break this model. In order to ensure the scene reflects the model, the underlying algorithm just mentioned could be modified to be a motion tracking algorithm instead of an object tracking algorithm. We have discussed and attempted to implement this by simply constantly grabbing a new base frame. The idea is that if the base frame is updated very frequently, only objects in motion will be measured.

9.1.3 Feature or Color Detection

Using specific color or feature detection could be a major system improvement, and would be more useful for applications like surveillance. The simple averaging algorithm implemented for object measurement assumes highly symmetric objects and is perhaps the simplest possible way of locating an object in a scene. This improvement would require more complicated video processing, and using color would require major changes to the system architecture, which was designed for grayscale.

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Appendices