Augmented Reality on FPGA

Realtime Object Recognition and Image Processing

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Introduction

- Overlay a digital image on a physical object in realtime.
- In this case, we want to identify a picture frame in captured video, and output video with another image distorted to fit on top of the picture frame.

Example Image



Example Image

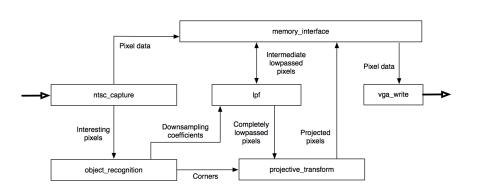


Example Image

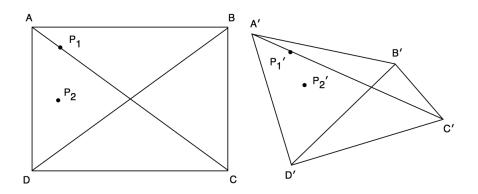


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Top-Level Overview

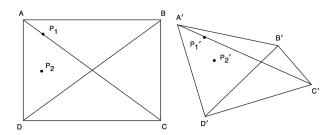


projective_transform: Purpose



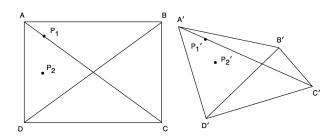
Skew to any arbitrary convex quadrilateral

projective_transform: How the algorithm works



- 1 Calculate the distance of line $\overline{A'D'}$ and assign it to d_{ad} .
- 2 Do the same for $\overline{B_iC_i}$ and assign it to d_{bc} .
- 3 Create two "iterator points," point I_A and I_B initially located at A' and B'.
- 4 Let $o_x = 0$ and $o_y = 0$
- 5 Calculate the distance between the iterator points, assign it to d_i .
- 6 Create a third iterator point, I_C at the location I_A .

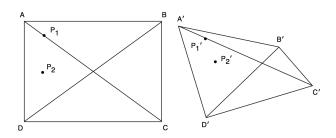
projective_transform: How the algorithm works



- 7 Assign the pixel value of I_C to pixel (o_x, o_y) in the original image.
- 8 Move I_C along line $\overline{I_A I_B}$ by an amount $= \frac{d_i}{width_{original}}$.
- 9 Increment o_x .
- 10 Repeat steps 7–9 until $I_C = I_B$.



projective_transform: How the algorithm works

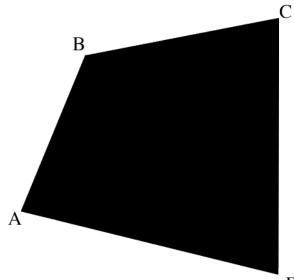


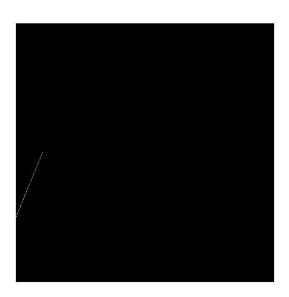
- 11 Move I_A along line $\overline{A_ID_I}$ by an amount $=\frac{d_{ad}}{height_{original}}$.
- 12 Move I_B along line $\overline{B_IC_I}$ by an amount $= \frac{d_{bc}}{height_{original}}$.
- 13 Increment o_y .
- 14 Repeat steps 5–13 until $I_A = DI$ and $I_B = CI$.





Figure: The original image





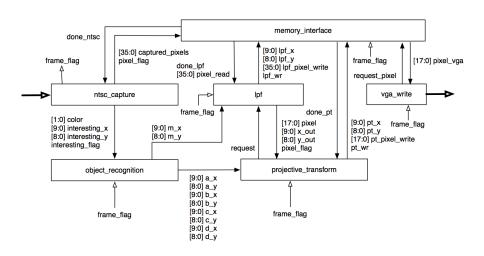




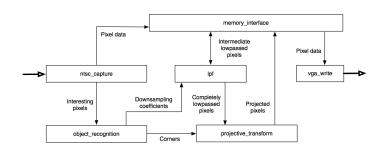
projective_transform: FPGA implementation

- Straightfoward implementation of the above algorithm
- Uses coregen Divider modules for the divisions
- Requires only 2*640*480 + 4*480 multiplications per clock cycle
- Uses an iterative algorithm for finding distances (pipelined at the end of each line of the image)
- Processes pixels "on-the-fly" from LPF
- Negligible memory requirements (a handful of registers)

projective_transform: How it Interfaces



object_recognition



- Mark corners of frame with four differently colored dots.
- Recognition begins in the ntsc_capture module, which detects these colors as it is capturing data and sends the pixel info to the object_recognition module.

object_recognition

- Take linear weighted center of mass for each image
- Sums the (x,y) coordinates for each color as it receives them. (8 running sums, 2 for each color)
- When the frame is done, divide each sum by the number of summed items
- The resulting 4 (x,y) pairs are the corners of the frame
- By looking for pixels in ntsc_capture we significantly reduce the amount of time spent in object_recognition

ullet projective_transform ightarrow aliasing

graphic showing normal signal

ullet projective_transform ightarrow aliasing

graphic aliases

- projective_transform → aliasing
- Aliasing reduces the quality of an image

zoom in on aliased pixels

- projective_transform → aliasing
- Aliasing reduces the quality of an image
- Lowpass filtering prevents aliasing

picture depicting lowpass filter in 2D

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- Information of an image is mostly phase

picture of original picture

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picture of other image

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picture of original phase with other's magnitude

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frequency response of Parks-McClellan filter

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- FIR PM filter reduces mem. acceses to 1.5/pixel



• Given an arbitrary image & skewing coefficients M_x & M_y .

graphic showing the interface between object_recognition and LPF image magnitude fourier plot of image

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- 2 Fetch a filter with cutoff $\frac{\pi}{M_y}$.

magnitude plot of image magnitude plot of filter with cutoff pi/2

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magnitude plot of image magnitude plot of filter with cutoff pi/2 magnitude fourier plot of filtered

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magnitude plot of filtered image magnitude plot of filter with curoff pi/4

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- Filter each column and store in memory.
- Fetch a filter with cutoff $\frac{\pi}{M_x}$.
- Filter each row and output to projective_transform.

magnitude plot of filtered image mangitude plot of filter with cutoff pi/5 magnitude plot of output

- Given an arbitrary image & skewing coefficients M_x & M_y.
- 2 Fetch a filter with cutoff $\frac{\pi}{M_y}$.
- Filter each column and store in memory.
- 4 Fetch a filter with cutoff $\frac{\pi}{M_x}$.
- Filter each row and output to projective_transform.
- Repeat this process every refresh cycle.

magnitude plot of original magnitude plot of filter magnitude plot of output

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 640 · 480 · 24 bits ≈ 0.88MiB

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- Let's store store 18 bits per pixel or 2 per address



memory_interface: operation



system io: ntsc_capture

system io: vga_write

timeline