

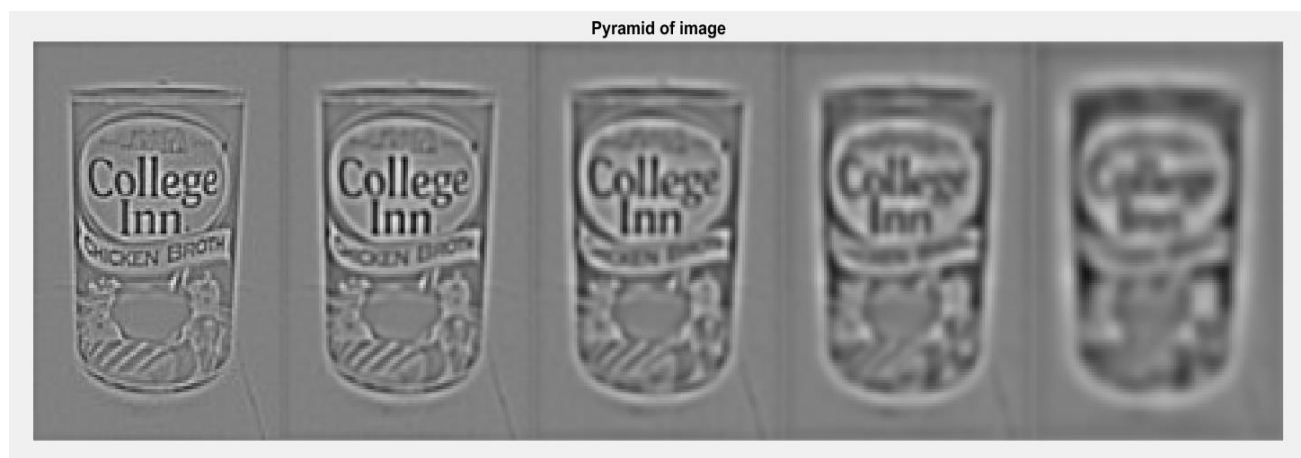
# Computer Vision Home Work 2

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## Question 1.1 Gaussian Pyramid



## Question 1.2 DOG Pyramid



### Question 1.3 Edge Suppression



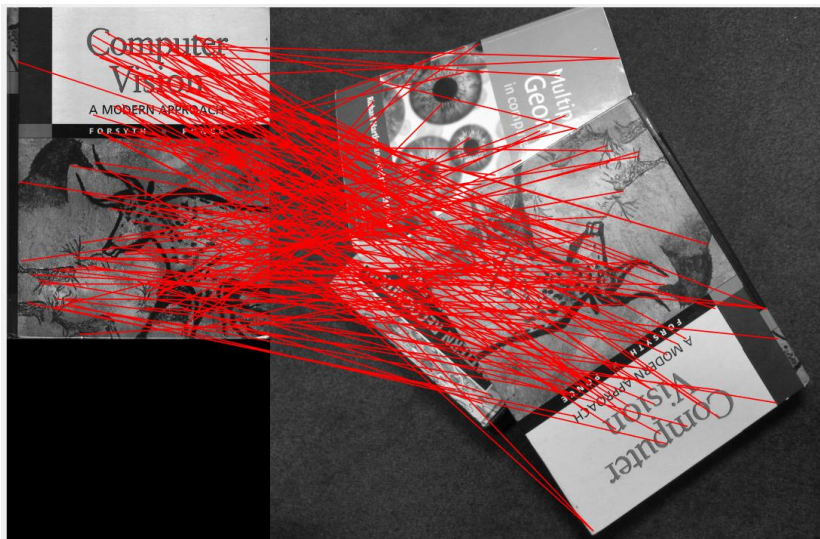
Without Edge Suppression



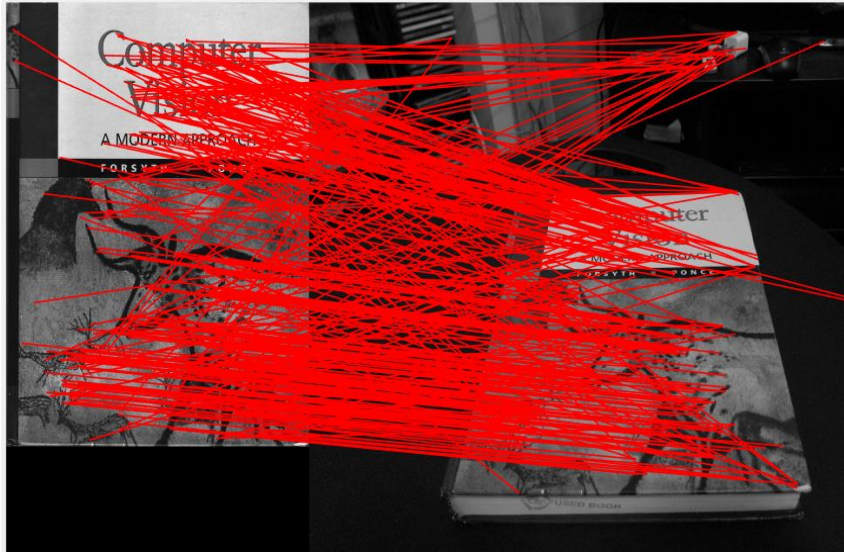
With Edge Suppression

### Question 2.4 Descriptor Matching

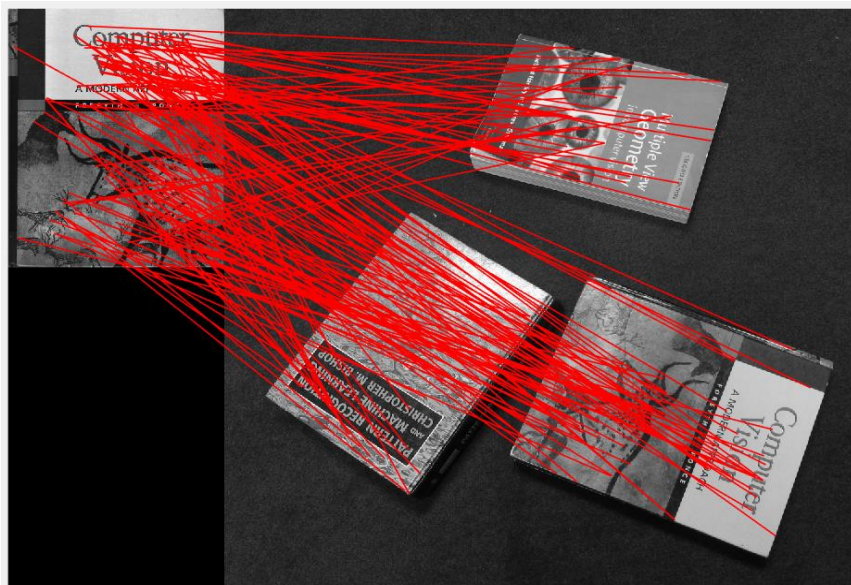
Picture 1: Scanned Image with Pile



**Picture 2: Scanned Image with desk**

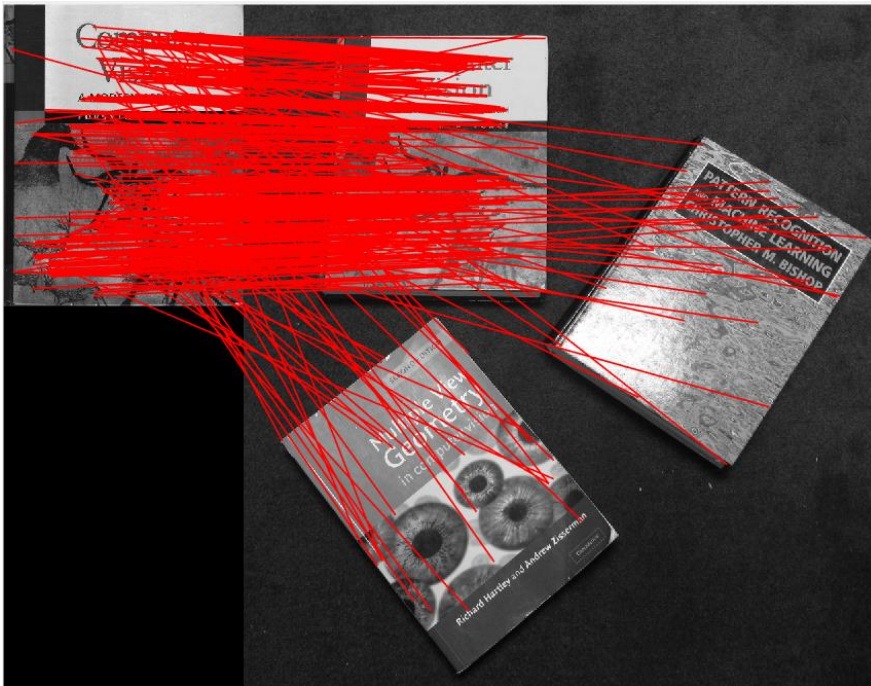


**Picture 3: Scanned Image with Floor\_Rotated**





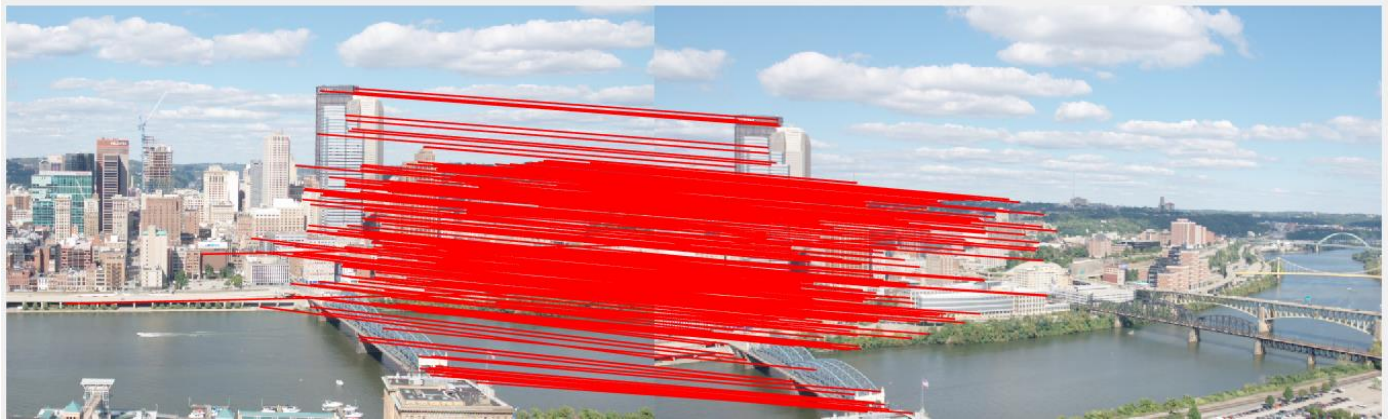
**Picture 4: Scanned Image with Floor**



**Picture 5: Scanned Image with Stand**



**Picture 6: Incline Images L and R with ratio value as 0.5**



The best matches are achieved between the original scanned image and the image lying on the floor which is not rotated as shown in the **picture 4**.

The matches gets degraded when the original scanned image is matched with the rotated image lying on the pile as shown in **picture 1**, because while computing the BRIEF descriptors we take in the location and intensities of nearby pixels centered around a patch and this makes the BRIEF, rotation variant.

### **Question 2.5 BRIEF and Rotations**

**Note:** Since the it was taking a lot of time for computing for 360 rotations, I did it for 4 rotations.

**Image Rotated by 0 deg**



**Image Rotated by 10 deg**



Image Rotated by 20 deg

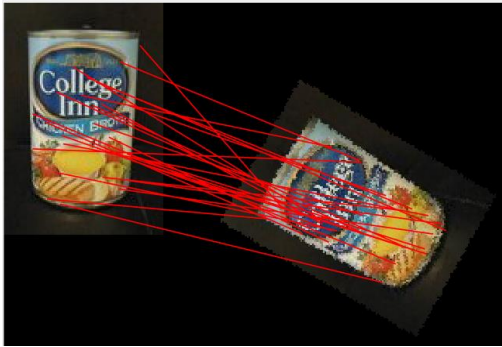
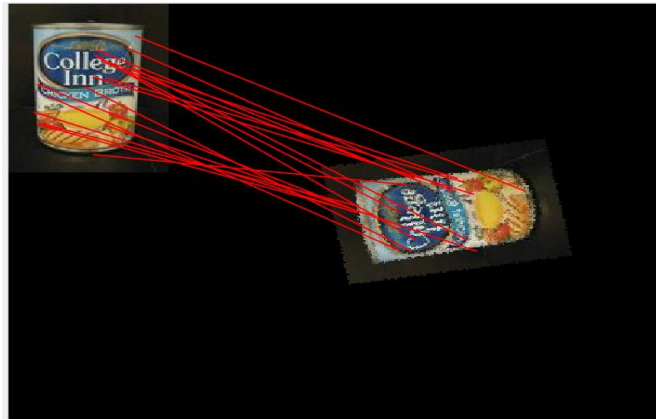


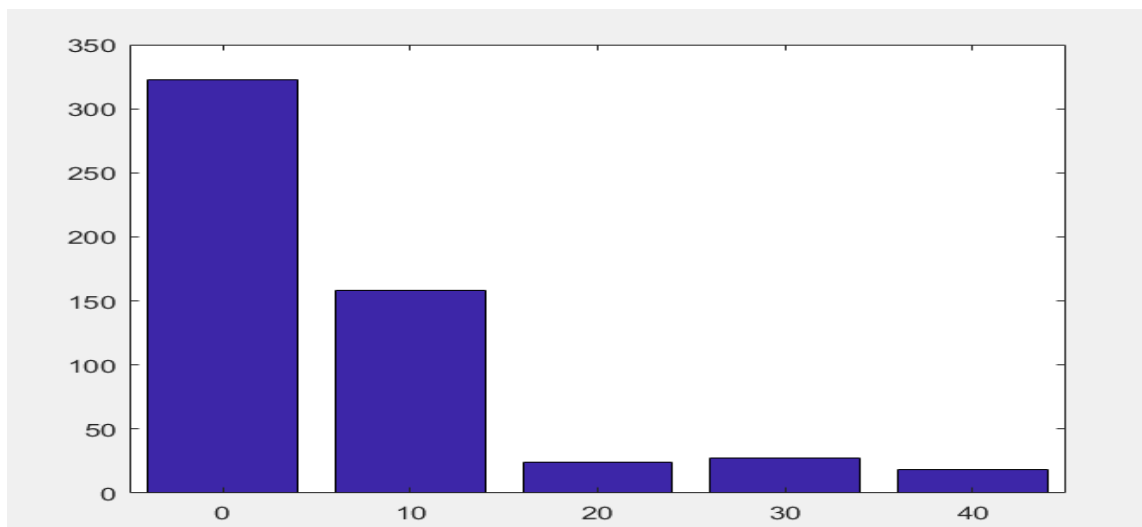
Image Rotated by 30 deg



Image Rotated by 40 deg



Bar Graph:



Bar graph with angles in X direction and number of matches in Y direction

The number of matches degrades drastically, as the images are rotated because the BRIEF uses the intensities of the nearby pixels centered around a patch to generate descriptors. When the image is rotated, the intensities of the pixels are no longer a match with respect to the patch and thus the matches reduces.

### Question 3.0 Planar Homography

a)

Planar Homography

$$\lambda \tilde{x} = H \tilde{u}$$

$$\lambda \begin{bmatrix} x_1 \\ y_1 \\ 1 \end{bmatrix} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix} \begin{bmatrix} u_1 \\ v_1 \\ 1 \end{bmatrix}$$

$$\lambda x = h_{11}u_1 + h_{12}v_1 + h_{13}$$

$$\lambda y = h_{21}u_1 + h_{22}v_1 + h_{23}$$

$$\lambda = h_{31}u_1 + h_{32}v_1 + h_{33}$$

To get  $x$  &  $y$ , we divide it by  $\lambda$

$$\frac{\lambda x_1}{\lambda} = \frac{h_{11}u_1 + h_{12}v_1 + h_{13}}{h_{31}u_1 + h_{32}v_1 + h_{33}}$$

$$\frac{\lambda y_1}{\lambda} = \frac{h_{21}u_1 + h_{22}v_1 + h_{23}}{h_{31}u_1 + h_{32}v_1 + h_{33}}$$

$$\frac{\lambda}{\lambda} = \frac{h_{31}u_1 + h_{32}v_1 + h_{33}}{h_{31}u_1 + h_{32}v_1 + h_{33}}$$

$$x_1(h_{31}u_1 + h_{32}v_1 + h_{33}) = h_{11}u_1 + h_{12}v_1 + h_{13}$$

$$x_1(h_{31}u_1 + h_{32}v_1 + h_{33}) - (h_{11}u_1 + h_{12}v_1 + h_{13}) = 0$$

$$y_1(h_{31}u_1 + h_{32}v_1 + h_{33}) = h_{21}u_1 + h_{22}v_1 + h_{23}$$

$$y_1(h_{31}u_1 + h_{32}v_1 + h_{33}) - (h_{21}u_1 + h_{22}v_1 + h_{23}) = 0$$

Taking coefficients of  $h$



$$\begin{bmatrix} -u_1 & -v_1 & -1 & 0 & 0 & 0 & u_1 x_1 & v_1 x_1 & x_1 \\ 0 & 0 & 0 & -u_1 & -v_1 & -1 & u_1 y_1 & v_1 y_1 & y_1 \\ -u_2 & -v_2 & -1 & 0 & 0 & 0 & u_2 x_2 & v_2 x_2 & x_2 \\ 0 & 0 & 0 & -u_2 & -v_2 & -1 & u_2 y_2 & v_2 y_2 & y_2 \\ - & - & - & - & - & - & - & - & - \\ - & - & - & - & - & - & - & - & - \end{bmatrix} \begin{bmatrix} h_{11} \\ h_{12} \\ h_{13} \\ h_{21} \\ h_{22} \\ h_{23} \\ \vdots \\ h_m \end{bmatrix}$$

$$Ah = 0$$

Here with A matrix has a dimension of  $2N \times 9$  and H is a  $9 \times 1$  Matrix

b) There are in total 9 elements in h. But the degree of freedom is only 8, as the elements of **H matrix** are independent of the scale factor and thus reduces the DOF from 9 to 8.

c) Each new correspondence gives two independent equations (information's). We need 4 pair of points, with 8 independent equations for planar homography.

d)

$$Ah = 0$$

Sum Squared Errors

$$f(h) = \frac{1}{2} (Ah - 0)^T (Ah - 0)$$

$$f(h) = \frac{1}{2} (Ah)^T (Ah)$$

$$f(h) = \frac{1}{2} h^T A^T A h$$

To Minimize h, differentiate w.r.t x

$$\frac{df}{dh} = 0 = \frac{1}{2} (A^T A + (A^T A)^T) h$$

$$A^T A h = 0$$

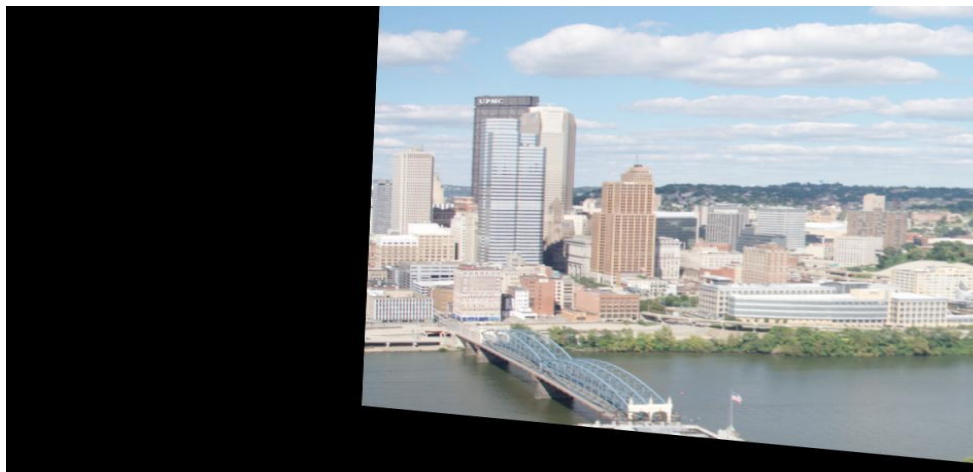
Steps followed are:



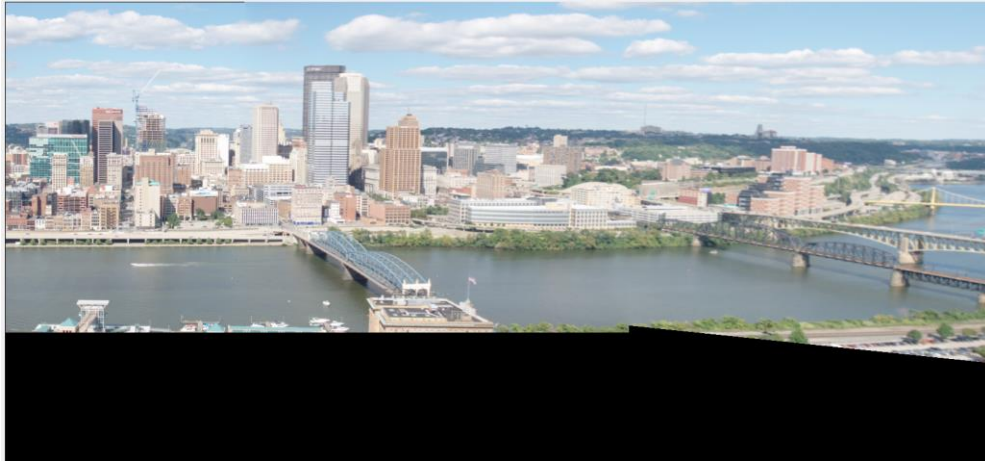
1. Derive  $2N$  independent equations in the form  $Ah=0$ . Here we are using 4 pair of correspondences.
2. The resulting  $A$  matrix would be of size  $8 \times 9$ . In order to perform SVD, we make the  $A$  matrix square by multiplying it by  $A'$ .
3. Then we perform SVD.
4. The last column of  $V$  gives us the eigenvector corresponding to the least eigenvalue, which minimizes the system close to zero.
5. In ideal case  $h$  should be the eigenvector corresponding to zero eigenvalue, but in reality we only find eigenvalue which minimizes system close to zero.

### Question 6.1 Image Stitching

Warping:



Blending:



**Question 6.2 Image Stitching with no clipping with Compute H**



**Question 6.3 Final Panorama with Ransac:**

