

# REPORT

## PROBLEM 1.1:

*Write a MATLAB function `[pts] = extract_corners(img, THRESH)` that returns the locations of all corner points in an image, `img`, using the Harris corner detection algorithm. Use `THRESH` as the threshold on the response value of the detector. `pts` is a vector containing the locations of all corner points determined using non-maximal suppression. You may use the MATLAB function `imgradientxy` to compute the image derivatives.*

Implementation:

- Returns edge points from the image.
- `THRESH` is the 'N' number of edge points required.
- `img` is the image path.
- Implementation is showed commented in the program.

To run:

`pts1 = extract_corners('Data/mall1.jpg', 30)`



Note:

- Results for different images has been stored in **Output** folder with **P\_1.1 <img> .jpg**
- Uncomment the end code to get the plots.

## PROBLEM 1.2:

Write a MATLAB function `[mpts1,mpts2] = match_corners(img1,img2,pts1,pts2,WSIZE)` that takes as input two images, `img1` and `img2`, along with their detected corner locations, `pts1` and `pts2` (found using the `extract_corners` function), and a window size, `WSIZE`. The output of the function will be two vectors, `mpts1` and `mpts2`, containing the location of corresponding corners in the two images. Specifically, `mpts1` and `mpts2` contain the location of the corresponding corners in `img1` and `img2` found by computing the normalized cross correlation. The normalized cross correlation is defined as

$$NCC = \frac{(f - \bar{f})(g - \bar{g})}{\sqrt{\sum (f - \bar{f})^2 \sum (g - \bar{g})^2}}$$

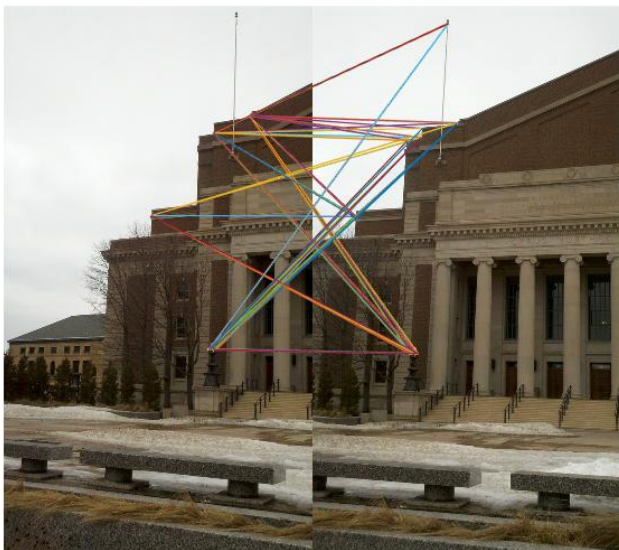
where  $f$  is a corner point in the first image,  $g$  is a corner point in the second image, and  $\bar{f}$  and  $\bar{g}$  are the mean intensity values within a window of size `WSIZE` centered at  $f$  and  $g$  in their respective images. The NCC is computed for every pair of corner points where the larger the value of the correlation the more likely that the corner points are matches.

### Implementation:

- Returns matched points from the image1 and image2.
- `pts1` and `pts2` are points returned for `img1` and `img2` from `extract_corners` function.
- `img1` and `img2` are the image paths.
- `WSIZE` is the window size [converted to nearest odd number size]
- Implementation is showed commented in the program.

To run:

```
[mpts11,mpts12] = match_corners('Data/mall1.jpg','Data/mall2.jpg',pts1,pts2,5);
```



Note:

- Uncomment the end code to get the plots.

## PROBLEM 1.3:

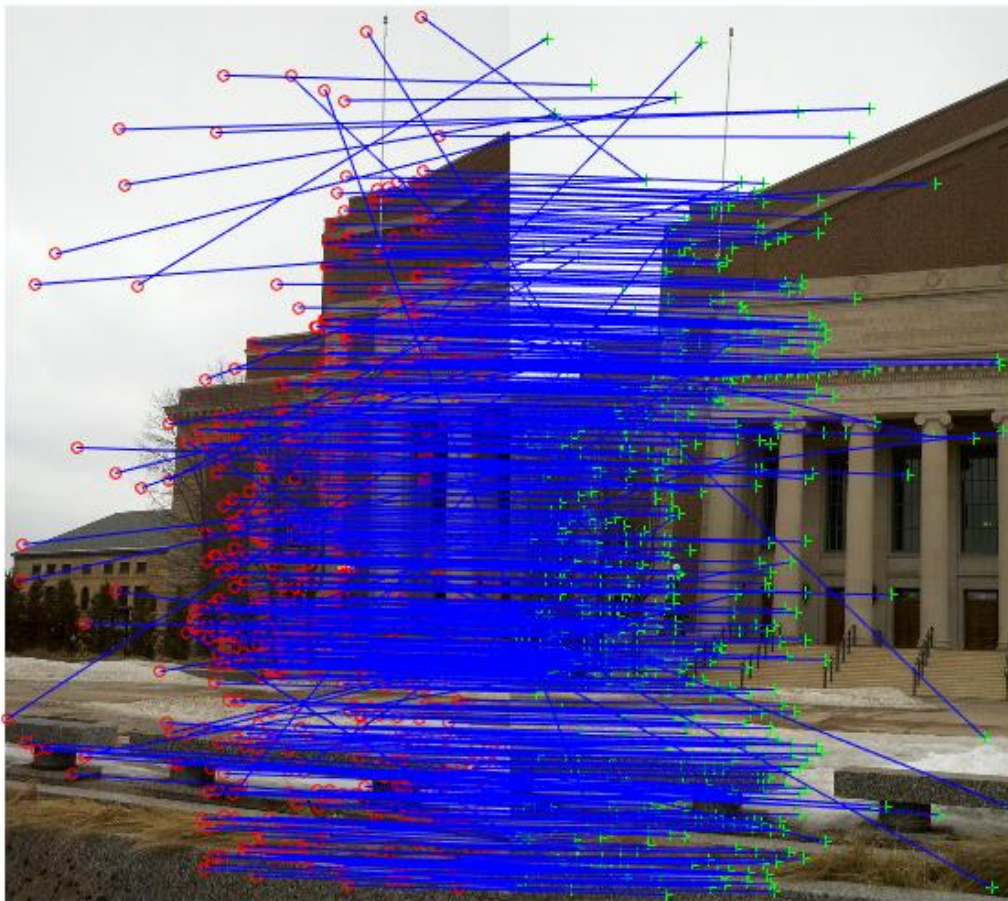
Use the `vl_sift` function from the SIFT toolbox available at [www.vlfeat.org](http://www.vlfeat.org) to detect the SIFT features in an input image. Match the obtained SIFT features from two images using the `vl_ubcmatch` function. In the PDF report, submit two images marking the location of the detected features and the matches found. You can show the matches using arrows in the two images placed side by side, or mark the position of the features using text labels on the image. Use the input images 'mall1.jpg' and 'mall2.jpg' for finding the features.

### Implementation:

- Install VLFeat from [www.vlfeat.org](http://www.vlfeat.org)
- `Img1` and `img2` are the image paths.

### To run:

- `sift_comparision('Data/mall1.jpg','Data/mall2.jpg');`



## PROBLEM 2.1:

Write a MATLAB function  $H = \text{compute\_homography}(mpts1, mpts2)$  that takes as input two  $N \times 3$  vectors of homogeneous image coordinates, and returns as output the  $3 \times 3$  homography matrix such that  $mpts2 = H \cdot mpts1$  using least squares fitting. Note that the number of input points,  $N$ , must be greater than or equal to four in order to estimate the homography.

Implementation:

- Returns Homography matrix such that  $T(mpts2) = H * T(mpts1)$
- $mpts1$  and  $mpts2$  are matched points generated by 2 images from `match_corners` function.
- Implementation is showed commented in the program.

To run:

```
 $H = \text{compute\_homography}(mpts1, mpts2);$ 
```

Proof:

```
>> a = [randi(10, 4, 2), ones(4, 1)]
```

a =

```

9   8   1
5   6   1
4   4   1
7   2   1
```

```
>> b = [randi(10, 4, 2), ones(4, 1)]
```

b =

```

6   3   1
3   5   1
1   7   1
8   4   1
```

```
>> H = compute_homography(a, b)
```

H =

```

-3.8323 -1.0000 17.8944
 0.6273 -1.3478 -7.1615
-0.2298 -0.3789  1.0000
```

```
>> X = H * transpose(a)
```

X =

```
-24.5963 -7.2671 -1.4348 -10.9317
-12.2981 -12.1118 -10.0435 -5.4658
-4.0994 -2.4224 -1.4348 -1.3665
```

```
>> convertToHomogenous(transpose(X))
```

```
ans =
```

```
6.0000 3.0000 1.0000
3.0000 5.0000 1.0000
1.0000 7.0000 1.0000
8.0000 4.0000 1.0000
```

```
>> b
```

```
b =
```

```
6 3 1
3 5 1
1 7 1
8 4 1
```

Here **b** and **ans** are same.

## PROBLEM 2.2:

Write a MATLAB function  $H = \text{compute\_homography\_ransac}(mpts1, mpts2, THRESH)$  that uses the RANSAC algorithm to reject the outlier matches in  $mpts1$  and  $mpts2$ , and then uses only the inliers to find  $H$ . Since  $H$  can be computed using four points, at each iteration the of algorithm you must sample four matches and 2 use these corresponding pairs to determine the candidate  $H^*$ . Use the symmetrical reprojection error as the distance function to determine the inliers. The symmetrical reprojection error is defined as

$$d = ||mpts2 - H*mpts1|| + ||mpts1 - H^{-1}*mpts2||.$$

Since  $mpts1$  and  $mpts2$  are homogeneous coordinates, you have to normalize  $H*mpts1$  and  $H^{-1}*mpts2$  before computing the norm. To decide if a point is an inlier, use  $THRESH$  as a threshold for the error. If  $THRESH$  is 0, then use  $2.5\sigma$  as the threshold where  $\sigma$  is the standard deviation of the error. Try out different values of  $THRESH$  to see which ones give reasonable performance. Use only the inliers found to compute the return value using the least squares compute homography function. Additional information can be found in [1]. In particular, refer to steps 1-3 of Algorithm 4.6 in section 4.8 'Automatic computation of a homography'.

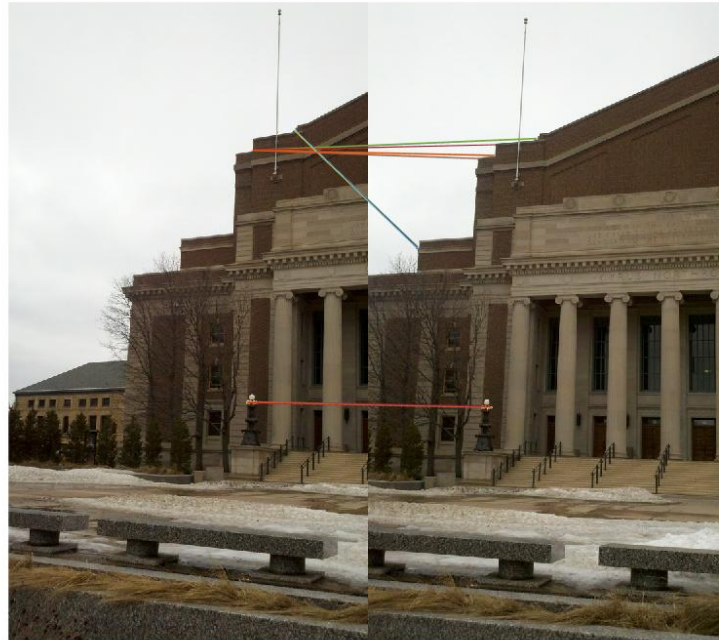
Implementation:

- Returns Homography matrix such that  $T(mpts2) = H * T(mpts1)$  based on RANSAC generated inliers.
- $Mpts1$  and  $mpts2$  are matched points generated by 2 images from `match_corners` function.

- THRESH is the value compared with d.

To run:

```
H = compute_homography_ransac(mpts1,mpts2,200,'Data/mall1.jpg','Data/mall2.jpg');
```



## PROBLEM 3.1:

*After estimating the homography between two images, they can be stitched together using Laplacian pyramid blending. Write a MATLAB function `blend = blend_images(img1,img2,mask1,mask2)` that takes as input two images and their corresponding binary masks and blends them together into a single output image using five pyramid levels. For additional information, see section 3.5.5 in [2] (available online).*

**Implementation:**

- Returns gaussian series of blended images.
- `img1` and `img2` are path to respective images.
- Only single mask has been used. [As said by Prof.]
- Actual implementation explained in the code.
- Mask with respect to `img1`.

To run:

```
mask = [ones(979,276)];
mask = [mask1,zeros(979,276)];
blend_images('Data/mall1.jpg','Data/mall2.jpg',mask);
```



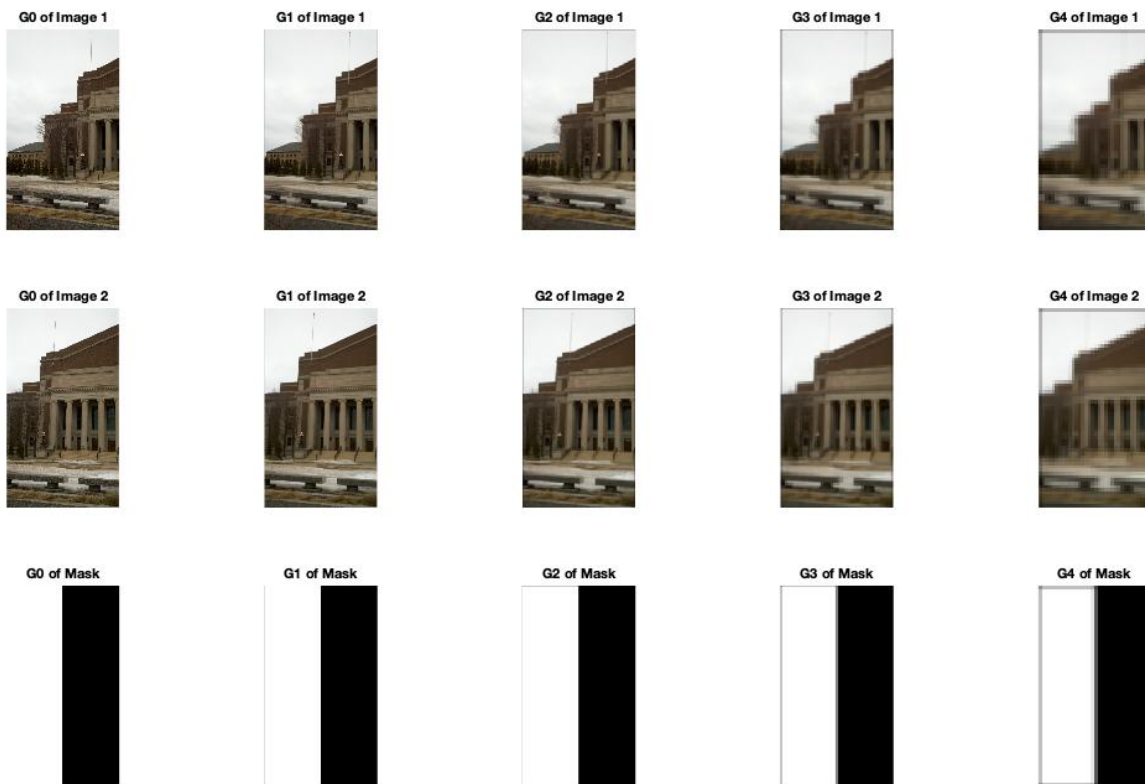


Figure 1 Gaussian

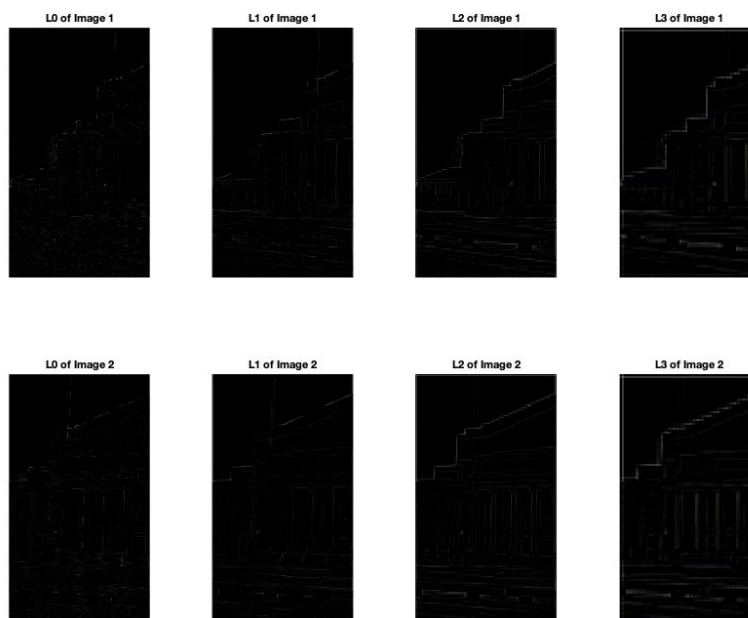


Figure 2 Laplacians



Figure 3 Blended Images

Note:

- $\text{mask2} = 1 - \text{mask}$  [Mask for second image]
- 5 levels of Gaussian pyramids are constructed.
- Uncomment the required code to generate plots.

## PROBLEM 4.1:

*After estimating the homography between two images, they can be stitched together using Laplacian pyramid blending. Write a MATLAB function `blend = blend_images(img1, img2, mask1, mask2)` that takes as input two images and their corresponding binary masks and blends them together into a single output image using five pyramid levels. For additional information, see section 3.5.5 in [2] (available online).*

[Partially done]