

# **Computer Vision**

## **Assignment 3**

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### **Image Stitching for Panorama Generation**

#### **1. Dataset**

- Captured/Downloaded images of 5 scenes with 5 images each.
- In a given scene, each image has at least a 20-30% overlap with the previous image.
- The images captured are from stationary scenes with no moving objects.

#### **2. Grayscale and Resizing:**

- All the images are converted to grayscale.
- The images in each scene are resized only if their size is too big while keeping the aspect ratio same.

#### **3. SIFT Feature extraction**

- SIFT and ORB feature extractors available in python-opencv are used to find keypoints and corresponding descriptors in all the images.
- Keypoints and descriptors help in finding correspondence between two images.
- SIFT provides descriptors of size 128.

Now, we describe two ways for image stitching: First by implementing from scratch and second using inbuilt functions.

### **Implementation from Scratch**

#### **4. Feature Matching**

- Between two images, we find the corresponding keypoints using euclidean distances or L2 norm.
- Distance between all keypoints is found using “`scipy.spatial.distance.cdist`”, and then two keypoints from the second image with minimum distance from a given keypoint in the first image are selected.
- We compare the distances of these two keypoints from second image and select only those keypoints where there is a large difference between the difference in the keypoints distance between the keypoint with the minimum distance and the keypoint with the second minimum distance.
- Now we have point correspondences from first image and second image.

## 5. Using RANSAC to find homography and inliers

### - Homography estimation

In order to find homography matrix, we randomly sample four points from the point correspondences obtained using feature matching.

Using,  $Ah = 0$  we find the  $h$  matrix.

$h$  is the last row of  $V$  in the SVD of  $A$ ,  $A=UDV^T$

We reshape  $h$  to obtain  $H$

$H$  is a 3x3 matrix

### - Finding Inliers

We find inliers by transforming points in the source image to points in destination image using  $H$  matrix. The points that lie in a threshold are taken as inliers.

### - Running RANSAC

We run ransac algorithm on the obtained  $H$  matrix and inliers, and prune the results. We run the algorithm till the size of inliers is found to be more than a threshold value( $0.8 \times \text{number of point correspondences}$ ).

## 6. Finding Mapping of each point in destination image to the source image

- After estimating the  $H$  matrix, we find the mapping of all points on the destination image to the source image.

## 7. Image warping

- Finally we create a canvas big enough to hold the stitched image.
- We place the source image as it is on the canvas.
- Next, we use the mapping obtained from the previous step, and place the destination image on the canvas
- The canvas now shows the stitched panoramic image.

## Using inbuilt functions

### 8. Feature Matching

- Between two images, we find the corresponding keypoints using euclidean distances or L2 norm.
- We do this using **BFMatcher** class in cv2, and **knnMatch** function. This works in a similar way to the implementation from scratch described above.
- We find the point correspondences between two images that have the minimum distance from each other.

### 9. Finding homography and inliers

- **Homography estimation**  
Using the point correspondences obtained using feature matching, we find the homography matrix using the function **findHomography**.
- **Finding Inliers**  
The inlier points are given by the same function **findHomography**.

### 10. Finding Mapping of each point in destination image to the source image

- After estimating the **H** matrix, we find the mapping of all points on the destination image to the source image using **WarpPerspective** function.

### 11. Image warping

- A canvas big enough to hold the stitched image is also created by the **WarpPerspective** function on which the destination image is placed.
- We place the source image as it is on the canvas.
- The canvas now shows the stitched panoramic image.

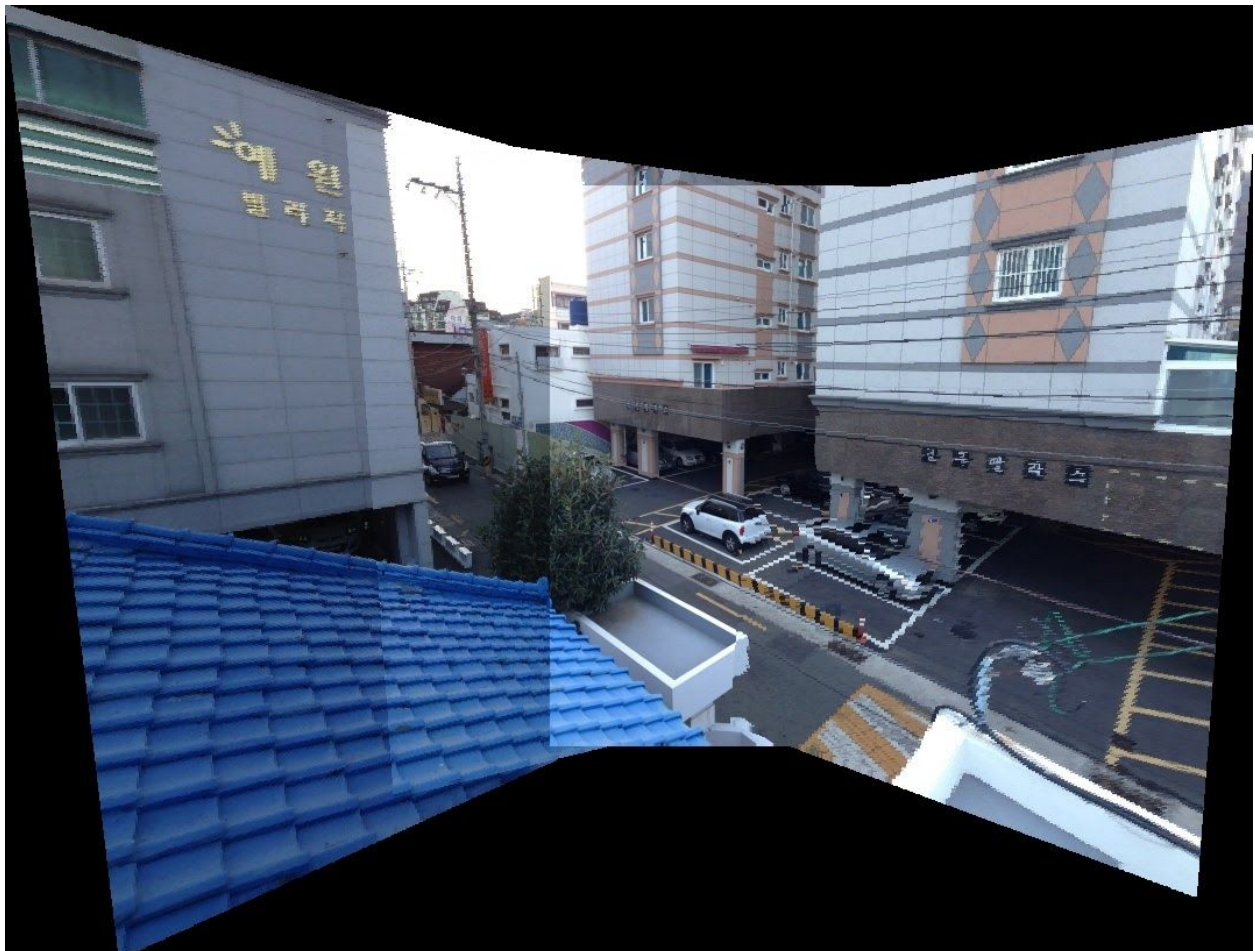
Sample images from the dataset and Panoramas

## Scene 2

Input images



Implementation from scratch



**Inbuilt implementation**



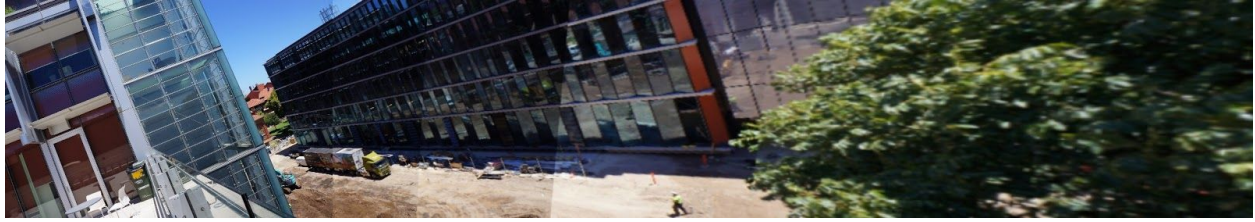
## Scene 1

**Implementation from scratch**



**Inbuilt implementation**



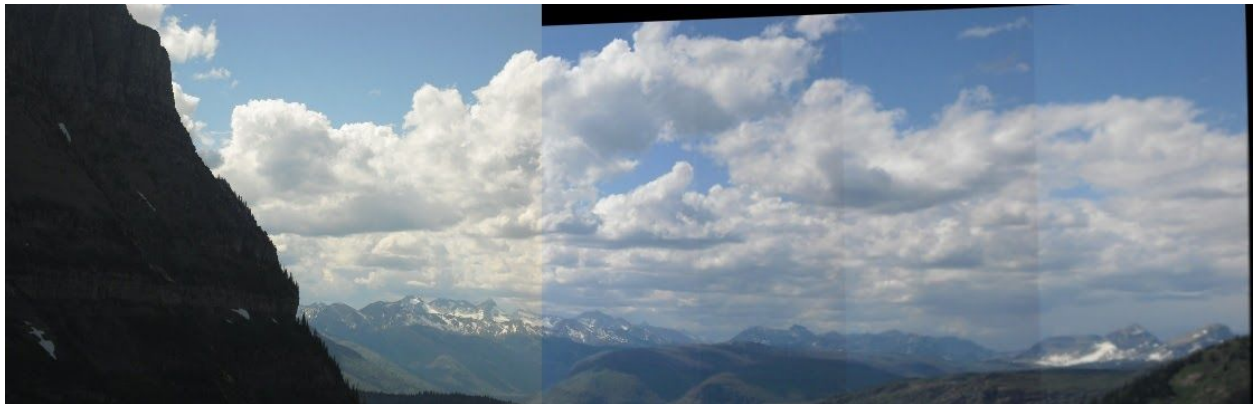


## Scene 3

Implementation from scratch

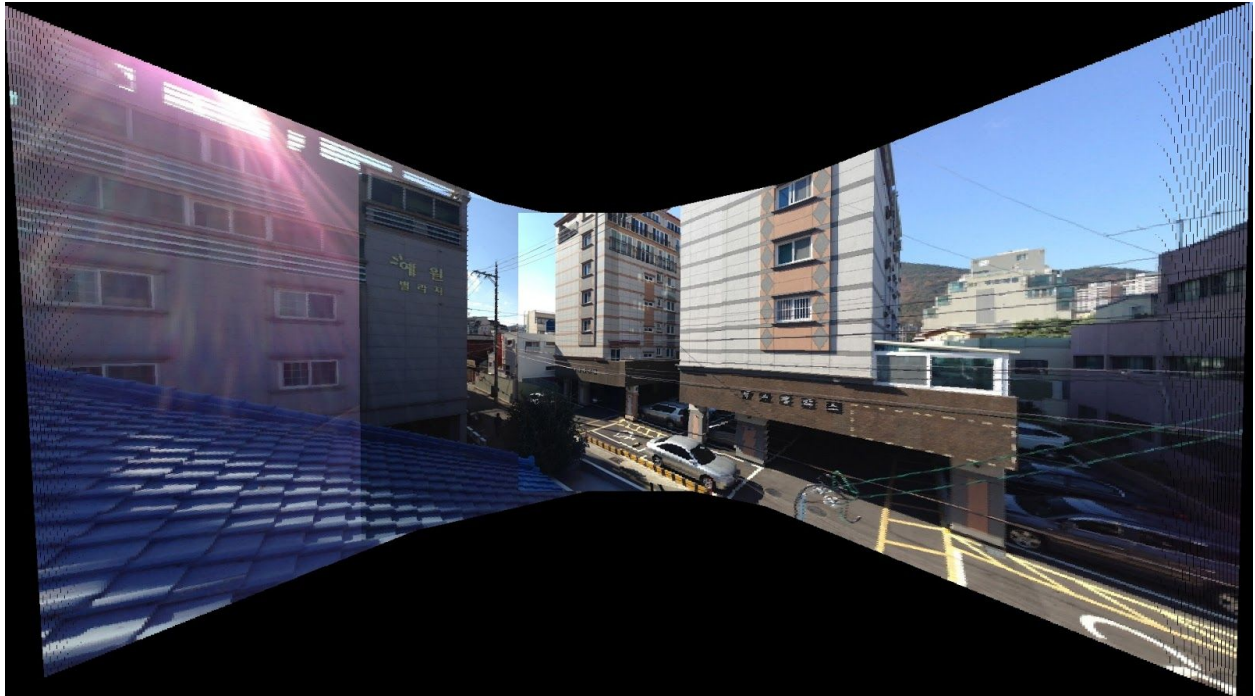


Inbuilt implementation

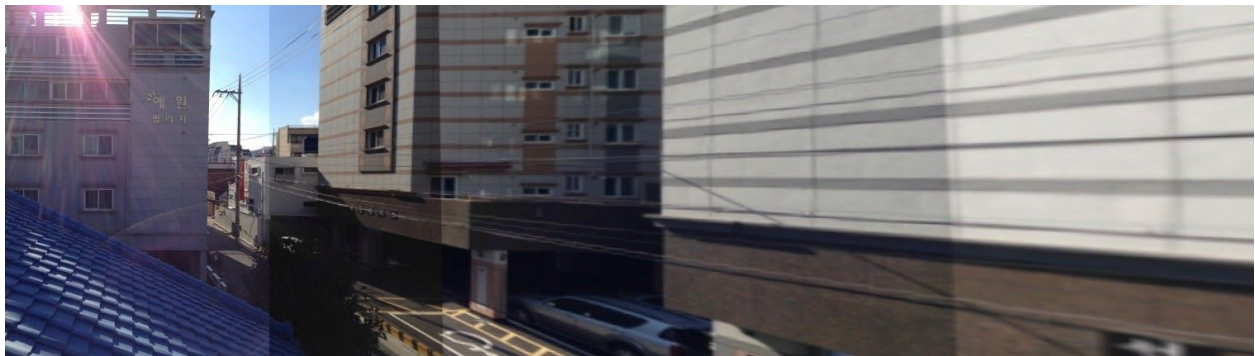


## Scene 4

## Implementation from scratch



## Inbuilt implementation



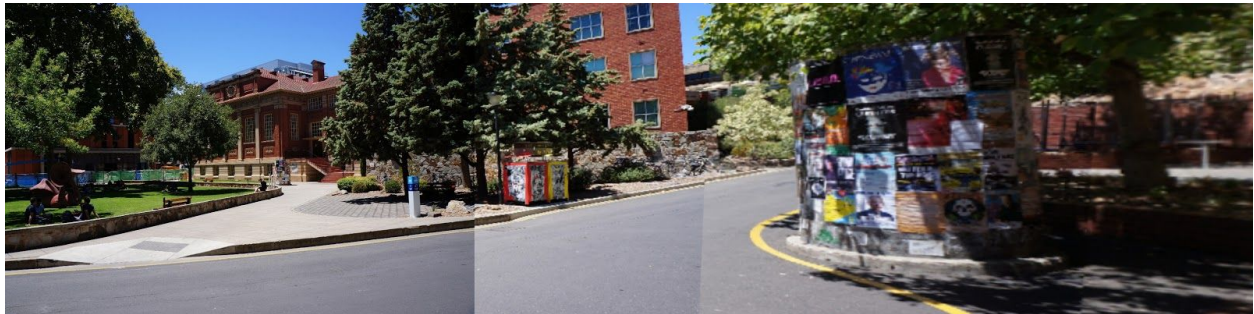
## Scene 5

### Implementation from scratch





### Inbuilt implementation



### References:

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10. <https://inst.eecs.berkeley.edu/~cs194-26/fa18/upload/files/proj6B/cs194-26-aeh/website/>