

Note: Basic instruction to follow before or while performing practicals Step

1: Install latest Version of python.

```
C:\Users\User>python --version
Python 3.12.3
C:\Users\User>pip --version
pip 24.0 from C:\Program Files\Python312\Lib\site-packages\pip (python 3.12)
```

Step 2: Set the environment variable path.

Step 3: Install packages as per required for particular practical.

i. Install numpy (cmd : pip install numpy)

ii. Install matplotlib (cmd : pip install matplotlib)

```
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iii.Install opency (cmd : pip install opency-python)

iv. Install tensorflow (cmd: pip install tensorflow)

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For checking Version of tensorflow

```
Python 3.12 (64-bit)
>>>
>>> import tensorflow as tf
>>> print(tf.__version__)
2.16.1
>>>
```

v. Install scikit-learn (cmd : pip install scikit-learn)

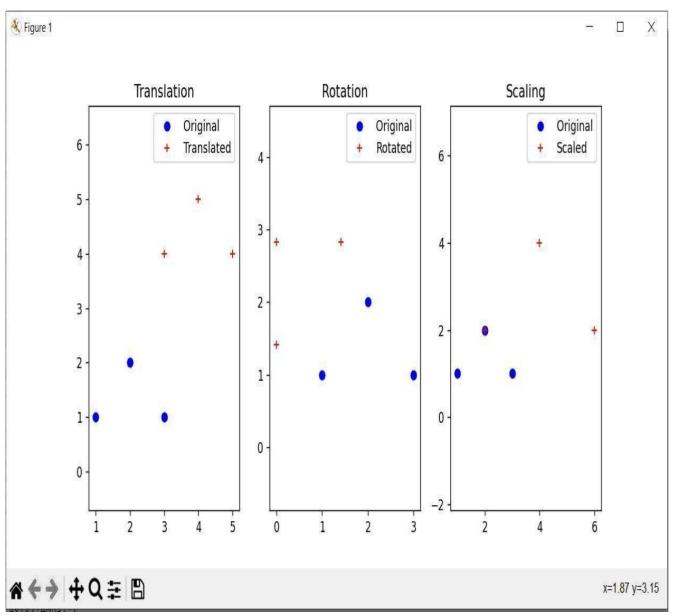
```
C:\Users\User>
C:\Users\User>pip install scikit-learn
Defaulting to user installation because normal site-packages is not writeable
Collecting scikit-learn
Downloading scikit_learn-1.5.0-cp312-cp312-win_amd64.whl.metadata (11 kB)
Requirement already satisfied: numpy>=1.19.5 in c:\users\users\user\uperapratatar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperapratar\uperap
```

4

Practical 1 Perform geometric transformation using python

```
import numpy as np import
matplotlib.pyplot as plt
import matplotlib.transforms as transforms
# Original points points = np.array([[1,
1], [2, 2], [3, 1]])
# Translation translation matrix = np.array([[1, 0, 2], [0, 1, 3], [0, 0, 1]]) # Translation by (2, 3)
translated points = np.dot(translation matrix, np.hstack([points, np.ones((points.shape[0],
1))]).T).T[:, :2] # Rotation theta = np.pi / 4 # Rotation angle (45 degrees) rotation matrix =
np.array([[np.cos(theta), -np.sin(theta), 0], [np.sin(theta), np.cos(theta), 0], [0, 0,
1]]) rotated points = np.dot(rotation matrix, np.hstack([points, np.ones((points.shape[0],
1))]).T).T[:, :2]
# Scaling scaling matrix = np.array([[2, 0, 0], [0, 2, 0], [0, 0, 1]]) # Scaling by a factor of 2
scaled points = np.dot(scaling matrix, np.hstack([points, np.ones((points.shape[0], 1))]).T).T[:, :2]
# Plotting plt.figure(figsize=(10, 5)) plt.subplot(1, 3, 1) plt.title('Translation')
plt.plot(points[:,
                              points[:,
                                                     'bo'.
                                                              label='Original')
                      0],
                                            1],
plt.plot(translated points[:, 0], translated points[:, 1], 'r+', label='Translated')
                                                  3, 2) plt.title('Rotation')
plt.axis('equal') plt.legend() plt.subplot(1,
plt.plot(points[:,
                      0],
                              points[:,
                                            1],
                                                     'bo',
                                                              label='Original')
plt.plot(rotated points[:, 0], rotated points[:, 1], 'r+', label='Rotated')
plt.axis('equal')
                  plt.legend() plt.subplot(1,
                                                   3,
                                                        3) plt.title('Scaling')
plt.plot(points[:, 0], points[:, 1], 'bo', label='Original') plt.plot(scaled points[:,
0], scaled points[:, 1], 'r+', label='Scaled') plt.axis('equal') plt.legend()
plt.show()
Output:
```

Code:



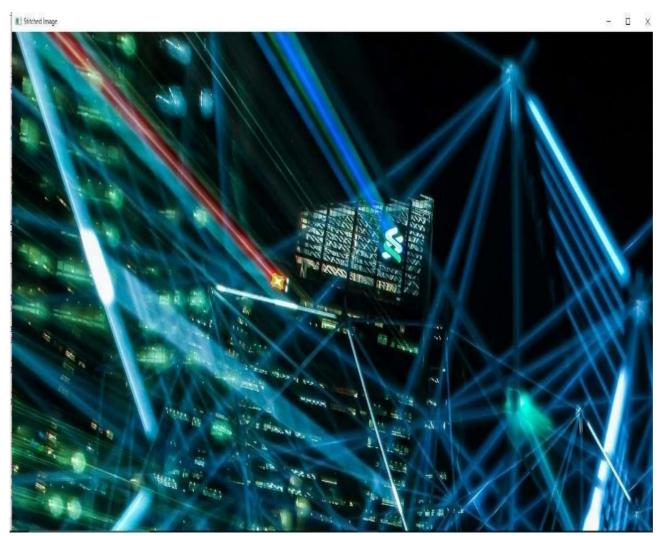
Practical No.2
Perform Image Stitching.

import cv2 import numpy as np # Load images image2 = cv2.imread("D:\Abhishek CV\ComputerVisionPractical\imageFolder\pex1.jpg") image1 = cv2.imread("D:\Abhishek CV\ComputerVisionPractical\imageFolder\pex.jpg") print("Image 1 shape:", image1.shape) print("Image 2 shape:", image2.shape) # Convert images to grayscale gray1 = cv2.cvtColor(image1, cv2.COLOR_BGR2GRAY) gray2 = cv2.cvtColor(image2, cv2.COLOR_BGR2GRAY)

```
# Detect keypoints and compute descriptors sift =
cv2.SIFT create() keypoints1, descriptors1 =
sift.detectAndCompute(gray1, None) keypoints2, descriptors2 =
sift.detectAndCompute(gray2, None)
# Match descriptors between the two images matcher
         cv2.BFMatcher()
                                matches
matcher.match(descriptors1, descriptors2)
# Sort matches by distance matches = sorted(matches,
key=lambda x: x.distance)
# Extract matched keypoints 1 = np.float32([keypoints1[match.queryIdx].pt for match
in matches]).reshape(-1, 1, 2) print("Number of points in points1:", len(points1)) points2 =
np.float32([keypoints2[match.trainIdx].pt for match in matches]).reshape(-1, 1, 2)
print("Number of points in points2:", len(points2))
      Find
               homography
                               matrix
                                           homography,
cv2.findHomography(points1, points2, cv2.RANSAC)
# Warp image1 to align with image2 height, width = gray2.shape
stitched image = cv2.warpPerspective(image1, homography, (width, height))
# Combine the stitched image with image2 stitched image[0:image2.shape[0],
0:image2.shape[1]] = image2
     Display
                         stitched
                 the
                                     image
cv2.imshow('Stitched
                                    Image',
stitched image)
                            cv2.waitKey(0)
cv2.destroyAllWindows()
```

Output:

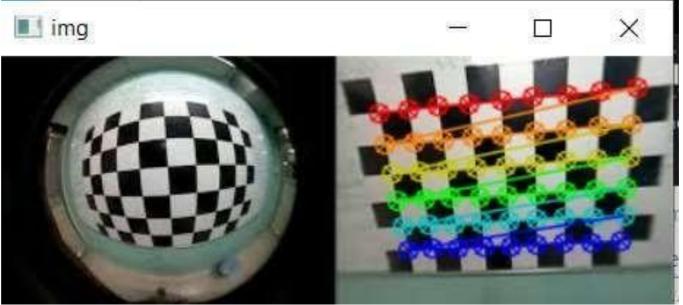
Code:



Practical No. 3 Camera Calibration

import numpy as np import cv2 import glob # Define the number of corners in the chessboard num_corners_x = 9 num_corners_y = 6 # Prepare object points, like (0,0,0), (1,0,0), (2,0,0), (6,5,0) objp = np.zeros((num_corners_x * num_corners_y, 3), np.float32) objp[:, :2] = np.mgrid[0:num_corners_x, 0:num_corners_y].T.reshape(-1, 2) # Arrays to store object points and image points from all the images objpoints = [] # 3d point in real world space image image plane.

```
Load
             images
                       images
                                      glob.glob("D:\Abhishek
CV\calibration images/*.jpg")
# Loop through images and find chessboard corners for
fname in images:
  img = cv2.imread(fname) gray = cv2.cvtColor(img,
  cv2.COLOR BGR2GRAY)
  # Find the chessboard corners ret, corners = cv2.findChessboardCorners(gray,
  (num corners x, num corners y), None)
  # If found, add object points, image points (after refining them) if
  ret:
    objpoints.append(objp) corners2 = cv2.cornerSubPix(gray,
    corners, (11, 11), (-1, -1),
criteria=(cv2.TERM CRITERIA EPS + cv2.TERM CRITERIA MAX ITER, 30, 0.001))
    imgpoints.append(corners2) # Draw and display the corners img =
    cv2.drawChessboardCorners(img, (num corners x, num corners y), corners2, ret)
    cv2.imshow('img', img) cv2.waitKey(5000)
cv2.destroyAllWindows() #
Perform camera calibration
ret, mtx, dist, rvecs, tvecs = cv2.calibrateCamera(objpoints, imgpoints, gray.shape[::-1], None, None)
# Save calibration results np.savez('calibration.npz', mtx=mtx, dist=dist,
rvecs=rvecs, tvecs=tvecs)
    Print
            calibration
                        results
print("Camera
                      matrix:")
             print("\nDistortion
print(mtx)
coefficients:") print(dist) Output:
```



Code:

Practical No.4

Face detection Code:

```
import cv2
```

Load the pre-trained Haar Cascade face detector

face cascade = cv2.CascadeClassifier(cv2.data.haarcascades +

'haarcascade frontalface default.xml')

Load the image = cv2.imread("D:\Abhishek

CV\ComputerVisionPractical\imageFolder\IMG1.jpg")

Convert the image to grayscale (face detection works on grayscale images) gray

= cv2.cvtColor(image, cv2.COLOR BGR2GRAY)

Detect faces in the image faces = face_cascade.detectMultiScale(gray, scaleFactor=1.1, minNeighbors=5, minSize=(30, 30))

Draw rectangles around the faces for

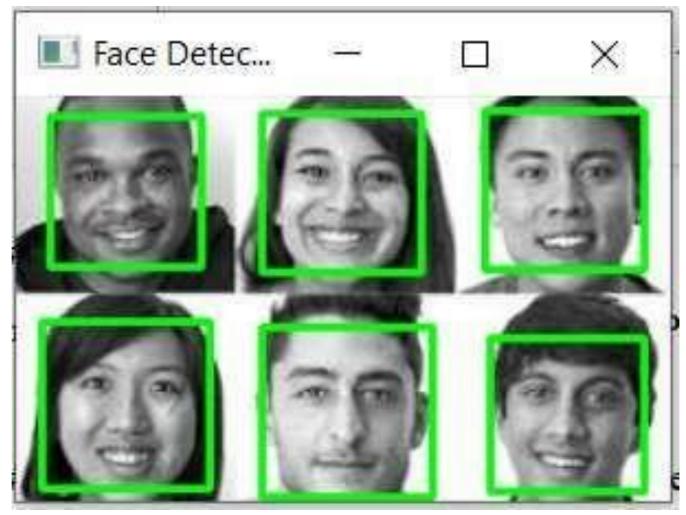
(x, y, w, h) in faces:

cv2.rectangle(image, (x, y), (x+w, y+h), (0, 255, 0), 2)

Display the result cv2.imshow('Face Detection',

image) cv2.waitKey(0)

cv2.destroyAllWindows() Output:



Practical No.5
Object Detection

Code: import numpy as np import os import tensorflow as tf import cv2

```
# Load the pre-trained model

MODEL_NAME = 'ssd_mobilenet_v2_coco_2018_03_29'

PATH_TO_CKPT = os.path.join(MODEL_NAME, 'frozen_inference_graph.pb')

NUM_CLASSES = 90

detection_graph = tf.Graph() with
detection_graph.as_default():
```

od_graph_def = tf.compat.v1.GraphDef() with tf.io.gfile.GFile(PATH TO CKPT, 'rb') as fid:

od_graph_def.ParseFromString(serialized_graph)
tf.import graph def(od graph def, name=")

Load label map

serialized graph

```
#PATH TO LABELS = os.path.join('data', 'mscoco label map.pbtxt')
PATH TO LABELS = "D:\Abhishek
CV\ComputerVisionPractical\imageFolder\mscoco label map (2).pbtxt"
category index = {} with open(PATH TO LABELS, 'r') as f: lines =
f.readlines() for line in lines:
    if 'id:' in line:
       id index = int(line.strip().split(':')[1])
    if 'display name:' in line:
       name = line.strip().split(':')[1].strip().strip('''') category index[id index]
       = {'name': name}
# Function to perform object detection def
detect objects(image):
  with detection graph.as default():
    with tf.compat.v1.Session(graph=detection graph) as sess:
       # Expand dimensions since the model expects images to have shape: [1, None, None,
       3] image_expanded = np.expand dims(image, axis=0) image tensor =
       detection graph.get tensor by name('image tensor:0')
       # Each box represents a part of the image where a particular object was detected.
       boxes = detection graph.get tensor by name('detection boxes:0')
       # Each score represents the level of confidence for each of the objects. #
       The score is shown on the result image, together with the class label.
       scores = detection graph.get tensor by name('detection scores:0')
       classes = detection graph.get tensor by name('detection classes:0')
       num detections =
       detection graph.get tensor by name('num detections:0') # Actual
       detection.
       (boxes, scores, classes, num detections) = sess.run([boxes, scores, classes,
num detections], feed dict={image tensor: image expanded}) # Visualization of
the results of a detection.
       for i in range(len(scores[0])):
         if scores[0][i] > 0.5: # Adjust confidence threshold as needed
            class id
                               int(classes[0][i])
                                                    class name
                                                     score
            category index[class id]['name']
            float(scores[0][i])
            ymin, xmin, ymax, xmax = boxes[0][i]
            (left, right, top, bottom) = (xmin * image.shape[1], xmax * image.shape[1], ymin *
image.shape[0], ymax * image.shape[0]) cv2.rectangle(image, (int(left), int(top)), (int(right),
            int(bottom)), (0, 255, 0), 2) cv2.putText(image, '{}: {:.2f}'.format(class name,
            score), (int(left), int(top - 5)),
cv2.FONT HERSHEY SIMPLEX, 0.5, (0, 255, 0), 2) return
  image
    Object
              detection
                                                                   cv2.imread("D:\Abhishek
                          on
                                an
                                      image
                                               input image
CV\ComputerVisionPractical\imageFolder\IMG.jpg")
                                                                 output image
detect objects(input image) cv2.imshow('Object Detection', output image) cv2.waitKey(0)
cv2.destroyAllWindows() Output:
```



Practical No.6

Pedestrian detection

Code:

import cv2

- # Load the pre-trained pedestrian detector pedestrian_cascade = cv2.CascadeClassifier(cv2.data.haarcascades + 'haarcascade_fullbody.xml')
- # Load the input image image = cv2.imread("D:\Abhishek CV\ComputerVisionPractical\imageFolder\pedestrainimg.jpg")
- # Convert the image to grayscale gray = cv2.cvtColor(image, cv2.COLOR BGR2GRAY)
- # Detect pedestrians in the image pedestrians = pedestrian_cascade.detectMultiScale(gray, scaleFactor=1.1, minNeighbors=1, minSize=(5, 5))
- # Draw rectangles around the detected pedestrians for (x, y, w, h) in pedestrians:
 - cv2.rectangle(image, (x, y), (x+w, y+h), (0, 255, 0), 5)
- # Display the image with pedestrian detections
- cv2.imshow('Pedestrian Detection', image)
- cv2.waitKey(0) cv2.destroyAllWindows()

Output:



Practical No.7 Feature

extraction using RANSAC Code:

```
import numpy as np from sklearn.linear model import
RANSACRegressor import matplotlib.pyplot as plt #
Generate some noisy data points np.random.seed(0) x
= np.random.uniform(0, 10, 100) y = 2 * x + 1 +
np.random.normal(0, 1, 100)
# Add outliers outliers index = np.random.choice(100, 20,
replace=False)
                  y[outliers index]
                                                10
np.random.normal(0, 1, 20)
# Stack the points for RANSAC data
= np.vstack((x, y)).T
# Initialize RANSAC model (using sklearn) ransac
= RANSACRegressor()
# Fit the model ransac.fit(data[:, 0].reshape(-
1, 1), data[:, 1])
# Extract the inliers and outliers inlier mask
= ransac.inlier_mask_
outlier mask = np.logical not(inlier mask)
```

```
# Extract the line parameters line slope =
ransac.estimator .coef [0] line intercept =
ransac.estimator .intercept
# Plot the data points plt.scatter(data[inlier mask][:, 0], data[inlier mask][:, 1],
c='b', label='Inliers') plt.scatter(data[outlier mask][:, 0], data[outlier mask][:, 1],
c='r', label='Outliers')
# Plot the fitted line plt.plot(x, line slope * x + line intercept, color='g',
label='RANSAC line') plt.xlabel('X') plt.ylabel('Y') plt.legend()
plt.grid(True)
                   K Figure 1
plt.show()
Output:
                                    Inliers
                                    Outliers
                                    RANSAC line
                         20
                         10
                                                                                             10
                          → + Q 至 □
                                                                                            x=6.33 y=-3.8
                                                            No.8
                                           Practical
```

Colorization of an image Code:

Convert grayscale image to BGR (3-channel) for colorization color_image

= cv2.cvtColor(gray_image, cv2.COLOR_GRAY2BGR)

Define a basic colorization lookup table color_lookup_table = np.zeros((256, 1, 3), dtype=np.uint8) for i in range(256):

color lookup table[i, 0, 0] = i # Blue channel color lookup table[i, 0, 1] = 127 # Green channel color lookup table[i, 0, 2] = 255 - i # Red channel # Apply thecolorization lookup table to the grayscale colorized image = cv2.LUT(color image, color lookup table) # Display the original grayscale image and the colorized image cv2.imshow('Grayscale Image', gray image) cv2.imshow('Colorized Image', colorized image) cv2.waitKey(0) cv2.destroyAllWindows() Output:





Practical No.9 Image

matting and composting Image composting:

Code:

import cv2 import numpy as np def estimate_alpha(image, trimap):

Placeholder function, replace with your matting algorithm implementation

This example simply sets alpha values based on trimap (e.g., foreground = 1, background = 0, unknown = interpolated) alpha = np.zeros_like(trimap, dtype=np.float32) alpha[trimap == 255] = 1.0 # Foreground alpha[trimap == 0] = 0.0 # Background alpha[(trimap > 0) & (trimap < 255)] = 0.5 # Interpolated return alpha

def image_matting(image, trimap): #
Convert image and trimap to float32
image = image.astype(np.float32) / 255.0
trimap = trimap.astype(np.float32) / 255.0

Estimate alpha matte using a matting algorithm
Replace this with your desired matting algorithm alpha = estimate_alpha(image, trimap)

Clip alpha values to [0, 1] alpha
= np.clip(alpha, 0, 1)

```
return alpha
def composit foreground background(foreground, background, alpha):
  # Resize background to match the foreground size background
  cv2.resize(background, (foreground.shape[1], foreground.shape[0]))
  # Convert alpha to 3 channels alpha =
  np.stack((alpha, alpha, alpha), axis=2)
  # Composite foreground and background using alpha matte
  composited image = alpha * foreground + (1 - alpha) * background
  return composited_image
# Example usage if __name
  == " main ":
  # Read foreground, background, and trimap images foreground = cv2.imread("D:\Abhishek
  CV\ComputerVisionPractical\imageFolder\model.jpg") background = cv2.imread("D:\Abhishek
  CV\ComputerVisionPractical\imageFolder\model.jpg") trimap = cv2.imread("D:\Abhishek
  CV\ComputerVisionPractical\imageFolder\model.jpg", cv2.IMREAD GRAYSCALE)
     Perform image
                     matting alpha
  image matting(foreground, trimap)
  #
             Perform
                               compositing
                                                    composited image
  composit foreground background(foreground, background, alpha)
  # Display result cv2.imshow("Composited Image",
composited image)
                                    cv2.waitKey(0)
cv2.destroyAllWindows() Output:
```



OR Image Matting

display alpha matte cv2.imshow("Alpha

cv2.waitKey(0)

alpha)

cv2.destroyAllWindows() Output:

Code:

Matte",

```
import cv2 import numpy as np def
estimate alpha(image, trimap):
                      Convert
                                                         to
                                                                          float
                                                                                                  image
        image.astype(np.float32) / 255.0
        # Normalize trimap to [0, 1] trimap = trimap.astype(np.float32) / 255.0 # Compute alpha
        matte using Closed-Form matting foreground = np.where(trimap > 0.95, 1.0, 0.0) \#
        Foreground mask alpha = np.where(trimap > 0.05, 1.0, 0.0) # Alpha initialization for in
        range(5): # Iterative refinement alpha = (image[:, :, 0] - image[:, :, 2] * alpha) / (1e-12 + 1e-12) / (1e-12 + 1e-12)
        foreground + (1.0 - trimap) * alpha = np.clip(alpha, 0, 1)
        return alpha
# Example usage
    if _name____= "_main_":
      #
                          Read
                                                        image
                                                                                                                    trimap
                                                                                                                                                                                                             cv2.imread("D:\Abhishek
                                                                                          and
                                                                                                                                                        image
       CV\ComputerVisionPractical\imageFolder\model.jpg")
         trimap = cv2.imread("D:\Abhishek CV\ComputerVisionPractical\imageFolder\model.jpg",
cv2.IMREAD GRAYSCALE)
       # Estimate alpha matte alpha =
estimate alpha(image, trimap) # Save or
```

