CSE803 HW6

December 13, 2024

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[1]: import cv2
      import numpy as np
      from PIL import Image
      from matplotlib import pyplot as plt
      import open3d
[19]: def drawlines(img1, img2, lines, pts1, pts2):
          r, c = img1.shape
          img1 = cv2.cvtColor(img1, cv2.COLOR_GRAY2BGR)
          img2 = cv2.cvtColor(img2, cv2.COLOR_GRAY2BGR)
          for r, pt1, pt2 in zip(lines, pts1, pts2):
              color = tuple(np.random.randint(0, 255, 3).tolist())
              x0, y0 = map(int, [0, -r[2]/r[1]])
              x1, y1 = map(int, [c, -(r[2]+r[0]*c)/r[1]])
              img1 = cv2.line(img1, (x0, y0), (x1, y1), color, 1)
              img1 = cv2.circle(img1, tuple(pt1), 5, color, -1)
              img2 = cv2.circle(img2, tuple(pt2), 5, color, -1)
          return img1, img2
      def draw_epipolar(img1, img2, F, pts1, pts2):
          lines1 = cv2.computeCorrespondEpilines(pts2.reshape(-1, 1, 2), 2, F)
          lines1 = lines1.reshape(-1, 3)
          img5, img6 = drawlines(img1, img2, lines1, pts1, pts2)
          lines2 = cv2.computeCorrespondEpilines(pts1.reshape(-1, 1, 2), 1, F)
          lines2 = lines2.reshape(-1, 3)
          img3, img4 = drawlines(img2, img1, lines2, pts2, pts1)
          plt.subplot(121)
          plt.imshow(img5)
          plt.subplot(122)
          plt.imshow(img3)
          plt.show()
      def visualize_pcd(points_3d):
          Visualize the point cloud.
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# create a figure for 3D plotting
         fig = plt.figure(figsize=(10, 8))
         ax = fig.add_subplot(111, projection='3d')
         # plot the 3D points
         ax.scatter(points_3d[0], points_3d[1], points_3d[2], c='r', marker='o', s=5)
         # set labels for the axes
         ax.set xlabel('X')
         ax.set_ylabel('Y')
         ax.set zlabel('Z')
         ax.set_title("3D Point Cloud")
         # show the plot
         plt.show()
     def visualize_point_cloud_with_open3d(points_3d):
         # convert points_3d to Open3D PointCloud format
         points = np.transpose(points_3d) # Nx3 points array
         pcd = open3d.geometry.PointCloud()
         pcd.points = open3d.utility.Vector3dVector(points)
         # visualize the point cloud
         open3d.visualization.draw_geometries([pcd], window_name="3D Point Cloud")
[3]: def fit_projection(X, Y):
         # build matrix A based on the correspondences
         A = \Gamma
         for i in range(X.shape[0]):
             x, y, z = X[i, 0], X[i, 1], X[i, 2]
             u, v = Y[i, 0], Y[i, 1]
             # equation for each correspondence
             A.append([x, y, z, 1, 0, 0, 0, -u * x, -u * y, -u * z])
             A.append([0, 0, 0, 0, x, y, z, 1, -v * x, -v * y, -v * z])
         A = np.array(A)
         # solve Ah = 0 using SVD
         U, S, V = np.linalg.svd(A)
         \# solution h is the smallest singular value of V
         h = V[-1, :]
         # build H matrix & normalization
         h = np.append(h, [1])
         H = h.reshape(3, 4)
```

H = H / H[2, 2]

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return H
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[22]: """
      Q1: Camera Calibration
      points_2d = np.loadtxt("data/pts2d-norm-pic.txt")
      print(points_2d.shape)
      # b = np.append(points_2d, np.ones([points_2d.shape[0], 1], dtype=np.int32),_
       \rightarrow axis=1)
      # print(b.shape)
      points_3d = np.loadtxt("data/pts3d-norm.txt")
      print(points_3d.shape)
      # A = np.append(points_3d, np.ones([points_3d.shape[0], 1], dtype=np.int32),_
       \hookrightarrow axis=1)
      # print(A.shape)
      \# P = A @ np.linalq.inv(A.T @ A) @ A.T
      P = fit_projection(points_3d, points_2d)
      print(P.shape)
      print(f"\nP = \n{P}")
     (20, 2)
     (20, 3)
     (3, 4)
     P =
     [[-2.16139798e-02 -1.99419043e-01 7.24945907e-01 7.02678089e-02]
      [-1.47733446e-02 6.40074852e-02 -2.53073904e-01 -7.28805413e-04]
      [ 2.58411448e-03 -2.38552804e-01 5.51366021e-01 1.00000000e+00]]
 [3]: def normalize_points(points):
          # Normalize points so that the mean is 0 and the average distance to the \Box
       ⇔origin is 1
          mean = np.mean(points, axis=0)
          dist = np.sqrt(np.mean(np.sum((points - mean)**2, axis=1)))
          scale = np.sqrt(2) / dist
          # Create normalization matrix
          T = np.array([[scale, 0, -scale*mean[0]],
                         [0, scale, -scale*mean[1]],
                         [0, 0, 1]])
          # Apply normalization matrix
          ones = np.ones((points.shape[0], 1))
          points_hom = np.hstack([points, ones]) # Convert to homogeneous coordinates
```

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return normalized points[:, :2], T # Return normalized points and the
       ⇔transformation matrix
      def fit fundamental(X, Y):
          # normalize points
          X, T_X = normalize_points(X)
          Y, T_Y = normalize_points(Y)
          # build matrix A based on the correspondences
          A = []
          for i in range(X.shape[0]):
              x, y = X[i, 0], X[i, 1]
             u, v = Y[i, 0], Y[i, 1]
              # equation for each correspondence
              A.append([x * u, x * v, x, y * u, y * v, y, u, v, 1])
          A = np.array(A)
          # solve Af = 0 using SVD
          U, S, V = np.linalg.svd(A)
          # build normalized fundamental matrix
          F_{\text{normalized}} = V[-1, :].reshape(3, 3)
          # enforce the rank-2 constraint by setting the smallest singular value to 0
          U, S, Vt = np.linalg.svd(F_normalized)
          S[2] = 0
          F_rank2 = U @ np.diag(S) @ Vt
          # denormalize the fundamental matrix by applying the inverse of the
       →normalization matrices
          F = T_Y.T @ F_rank2 @ T_X
          # apply scaling normalization
          F /= F[2, 2]
          return F
[55]: """
      Q2: Estimation of the Fundamental Matrix
      temple_npz = np.load("data/temple.npz")
      print(temple_npz)
      pts1 = temple_npz['pts1']
      print(pts1.shape)
```

normalized_points = (T @ points_hom.T).T

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pts2 = temple_npz['pts2']
print(pts2.shape)

img1 = np.asarray(Image.open("data/im1.png").convert('L'))
img2 = np.asarray(Image.open("data/im2.png").convert('L'))
print(img1.shape)
print(img2.shape)

F = fit_fundamental(pts1, pts2)
print(F.shape)
print(f"\nF =\n{F}")
draw_epipolar(img1, img2, F, pts1[:15], pts2[:15])
NpzFile 'data/temple.npz' with keys: K1, K2, pts1, pts2
```

```
(110, 2)

(110, 2)

(640, 480)

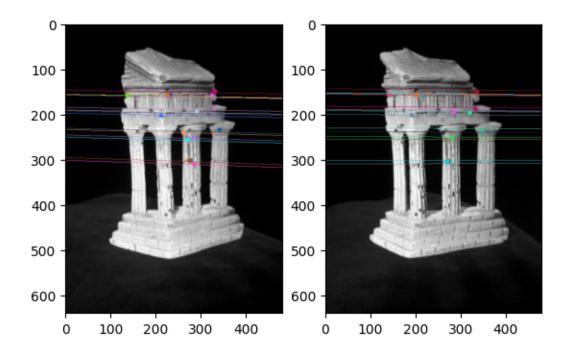
(640, 480)

(3, 3)

F =

[[-1.90213021e-07 6.43692991e-06 -1.74540062e-03]
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[1.01351546e-05 2.35169599e-07 9.27190726e-02] [1.00421527e-04 -9.91702304e-02 1.00000000e+00]]



```
[4]: def extract_essential(F, K1, K2):
          \# E = K2^T * F * K1^{-1}
          return K2.T @ F @ np.linalg.inv(K1)
      def calculate_projection_matrices(K1, K2, R, t):
          # P1 = K1 * [I / O]
          P1 = K1 @ np.hstack((np.eye(3), np.zeros((3, 1))))
          # P2 = K2 * [R / t]
          P2 = K2 @ np.hstack((R, t.reshape(3, 1)))
          return P1, P2
      def triangulate_points(P1, P2, pts1, pts2):
          # convert points to homogeneous coordinates (Nx3) by adding a column of ones
          # pts1_hom = np.hstack((pts1, np.ones((pts1.shape[0], 1))))
          \# pts2\_hom = np.hstack((pts2, np.ones((pts2.shape[0], 1))))
          # triangulate points using cv2.triangulatePoints
          points_4d = cv2.triangulatePoints(P1, P2, pts1.T, pts2.T)
          print(points_4d.shape)
          # convert from homogeneous to non-homogeneous coordinates (divide by the
       ⇔last coordinate)
          epsilon = 1e-10  # a small epsilon value to avoid division by zero
          points_3d = points_4d[:3] / (points_4d[3] + epsilon)
          # points_3d = cv2.convertPointsFromHomogeneous(points_4d.T)
          return points_3d
      def count_positive_depth(points_3d, R, t):
          # check the depth of points (positive depth in both camera views)
          depth_camera_1 = points_3d[2, :] # Z coordinates in the first camera
          depth_camera_2 = (R @ points_3d + t.reshape(3, 1))[2, :] # Z coordinates_
       ⇒in the second camera
          # count how many points have positive depth in both cameras
          positive_depth_count = np.sum((depth_camera_1 > 0) & (depth_camera_2 > 0))
          return positive_depth_count
[20]: """
      Q3: Triangulation
      # 3.1
      temple_npz = np.load("data/temple.npz")
      print(temple_npz)
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K1 = temple_npz['K1']
K2 = temple_npz['K2']
print(K1.shape, K2.shape)
pts1, _ = normalize_points(temple_npz['pts1'])
pts2, _ = normalize_points(temple_npz['pts2'])
print(pts1.shape, pts2.shape)
# 3.2
F = cv2.findFundamentalMat(pts1, pts2)[0]
print(f"\nF = \n{F}")
E = extract_essential(F, K1, K2)
print(f"\nE = \n{E}")
# 3.3
R1, R2, t = cv2.decomposeEssentialMat(E)
poses = [[R1, t], [R1, -t], [R2, t], [R2, -t]]
R, t = poses[1] # this pose has the highest positive depth count (57)
# 3.4
P1, P2 = calculate_projection_matrices(K1, K2, R, t)
print(f"\nP1 = \n{P1}")
print(f"\nP2 = \n{P2}\n")
# 3.5
points_3d = triangulate_points(P1, P2, pts1, pts2)
print(points_3d.shape)
depth_count = count_positive_depth(points_3d, R, t)
print(depth_count)
# 3.6
# visualize_pcd(points_3d)
visualize_point_cloud_with_open3d(points_3d)
<numpy.lib.npyio.NpzFile object at 0x00000158FB3FB400>
(3, 3) (3, 3)
(110, 2) (110, 2)
F =
[[ 1.03799909 -22.70011437 3.49454518]
[ 27.768498
                -1.17112061 -3.96320412]
 [ -6.63163724 -0.05314547 1.
                                        ]]
E =
[[ 1.03799909e+00 -2.26182934e+01 1.05837760e+04]
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[ 2.78689497e+01 -1.17112061e+00 -1.41830870e+04]
  [ 4.71139019e+00 -4.68670778e+00 -1.88219195e+02]]

P1 =
  [[1.5204e+03  0.0000e+00  3.0230e+02  0.0000e+00]
  [0.0000e+00  1.5259e+03  2.4690e+02  0.0000e+00]
  [0.0000e+00  0.0000e+00  1.0000e+00  0.0000e+00]]

P2 =
  [[ 6.18350790e+02 -6.62762912e+02 -1.25753293e+03 -1.42306004e-02]
  [-8.32973510e+02  8.98469160e+02 -9.42464174e+02  2.31875569e-01]
  [ 7.02614732e-01  6.67447327e-01 -2.46671043e-01  9.68699686e-01]]

(4, 110)
  (3, 110)
  57
```