

EN3160 IMAGE PROCESSING AND MACHINE VISION

Assignment 2: Fitting and Alignment

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Link to GitHub repository - <https://github.com/Sithum02/EN3160-Assignments>

Question 1

The Code

```
1 # Detect maxima manually
2     coordinates = []
3     (H, W) = img_log.shape
4     k = 1 # Neighborhood size
5     threshold = 0.1 # Threshold for maxima detection
6     for i in range(k, H - k):
7         for j in range(k, W - k):
8             slice_img = img_log[i - k:i + k + 1, j - k:j + k + 1]
9             if np.max(slice_img) >= threshold:
10                 # Get coordinates of the maximum point
11                 x, y = np.unravel_index(slice_img.argmax(), slice_img.shape)
12                 coordinates.append((i + x - k, j + y - k))
13     coordinates = set(coordinates)
14
15 # Plot circles around detected blobs
16 for x, y in coordinates:
17     c = plt.Circle((y, x), sigma * 1.414, color='g', linewidth=1, fill=False)
18     ax.add_patch(c)
19 patches.append(c)
20 ax.plot()
```

Listing 1: Blob detection code.



Figure 1: Blob detection

Radius Range Integers from 1 to 10.

σ Values 0.71, 1.41, 2.12, 2.83, 3.54, 4.24, 4.95, 5.66, 6.36, 7.07.

Largest Circle $\sigma = 7.07$ and Radius = 10.

Question 2

The Code

```

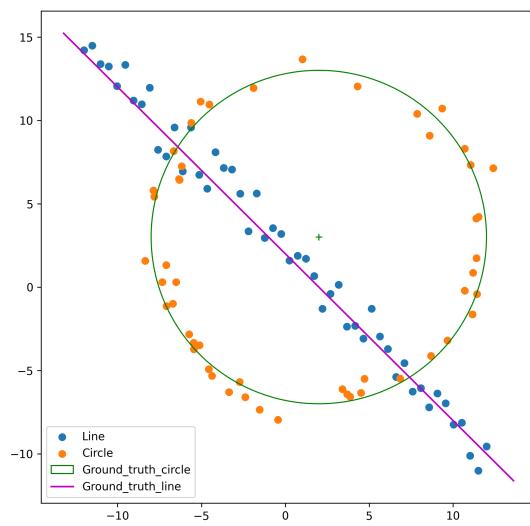
1 def ransac_line_fit(points, max_iter=1000, eps=0.6, min_support=35):
2     best_model = None
3     best_inliers = np.empty((0,2))
4     n_pts = len(points)
5
6     for _ in range(max_iter):
7         idx = np.random.choice(n_pts, 2, replace=False)
8         line_params = fit_line_from_two_pts(points[idx[0]], points[idx[1]])
9         dist = compute_point_line_dist(line_params, points)
10        mask = dist < eps
11        if mask.sum() > len(best_inliers) and mask.sum() >= min_support:
12            best_model = line_params
13            best_inliers = points[mask]
14    return best_model, best_inliers
15
16 def ransac_circle_fit(points, max_iter=1000, eps=1.5, min_support=45):
17     best_model = None
18     best_inliers = np.empty((0,2))
19     n_pts = len(points)
20
21     for _ in range(max_iter):
22         idx = np.random.choice(n_pts, 3, replace=False)
23         try:
24             circ_params = estimate_circle_from_pts(points[idx])
25         except np.linalg.LinAlgError:
26             continue
27         dist = radial_error(circ_params, points)
28         mask = dist < eps
29         if mask.sum() > len(best_inliers) and mask.sum() >= min_support:
30             best_model = circ_params
31             best_inliers = points[mask]
32     return best_model, best_inliers
33
34 # Fit line then circle on residual points
35 line_params, line_inliers = ransac_line_fit(X)
36
37 # Remove line inliers
38 mask = np.ones(len(X), dtype=bool)
39 for pt in line_inliers:
40     mask[np.all(np.isclose(X, pt), axis=1)] = False
41 remaining = X[mask]
42
43 circle_params, circle_inliers = ransac_circle_fit(remaining)

```

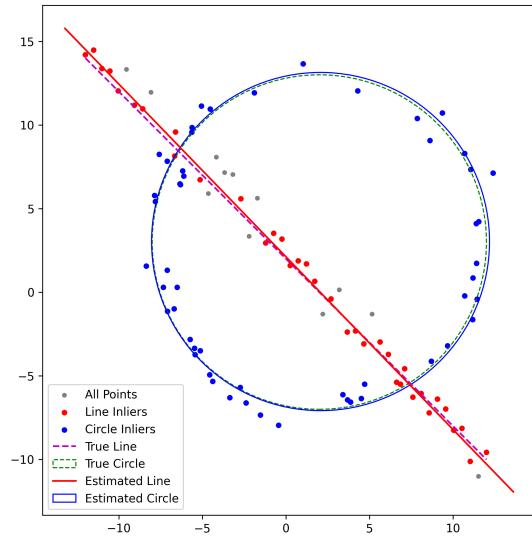
Listing 2: RANSAC code.

RANSAC line fitting parameters are iterations = 1000, threshold = 0.6, and minimum inliers = 35. The parameters of the estimated line are $a = -0.72$, $b = -0.70$, $d = 1.48$. RANSAC circle fitting parameters are iterations = 1000, threshold = 1.5, and minimum inliers = 45. The parameters of the estimated circle are $x = 2.07$, $y = 3.03$, $r = 10.11$.

If we fit the circle first, many line points may be incorrectly classified as circle inliers, leading to an incorrect circle model. As a result, when we subsequently try to fit the line, we are left with a corrupted or reduced set of line points, causing poor line estimation. Therefore, fitting the line first and then the circle gives more robust and accurate results.



(a) Generated Data points



(b) Predictions

Question 3

The Code

```

1 # 4 Points were selected by Mouse Clicks
2
3 # Define source and destination points for homography
4 destination_points = np.array(clicked_points).astype(np.float32)
5 h2, w2 = adding_img.shape[:2]
6 source_points = np.float32([[0, 0], [w2, 0], [0, h2], [w2, h2]])
7
8 # Define the homographic matrix
9 H = cv.getPerspectiveTransform(source_points, destination_points)
10 # Warping the adding image with the homography
11 warped_img = cv.warpPerspective(adding_img, H, (base_img.shape[1], base_img.shape[0]))
12 # Superimposing the warped image with the base image
13 supimp_img = cv.addWeighted(base_img, 1, warped_img, 1, 0)

```

Listing 3: Blob detection code.

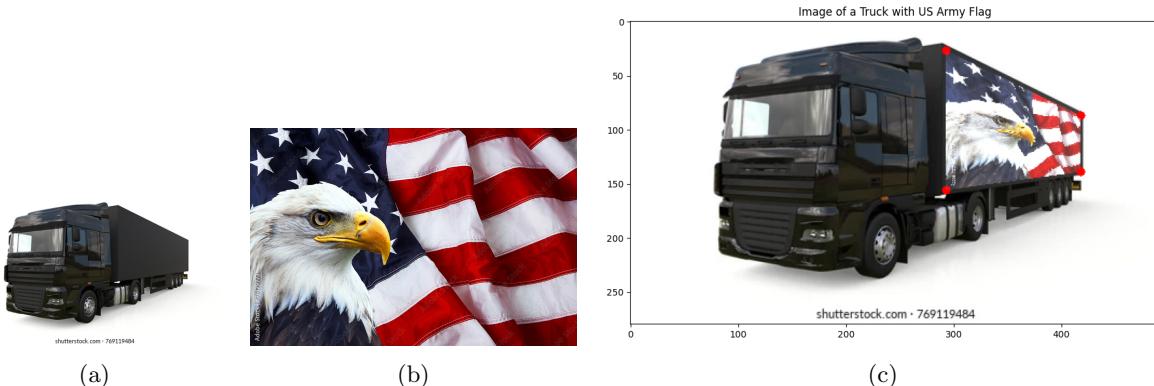


Figure 3: Overlaying the flag image (b) onto image (a) using Homography. The result is (c).



Figure 4: Overlaying the image (b) onto image (a) using Homography. The result is (c).

Question 4

The Code for (a)

```

1 # Convert to grayscale
2 img1_gray = cv2.cvtColor(img1, cv2.COLOR_RGB2GRAY)
3 img2_gray = cv2.cvtColor(img2, cv2.COLOR_RGB2GRAY)
4
5 # Get SIFT descriptors
6 sift = cv2.SIFT_create(nOctaveLayers=3, contrastThreshold=0.09,
7                         edgeThreshold=25, sigma=1)
8 keypoints1, descriptors1 = sift.detectAndCompute(img1_gray, None)
9 keypoints2, descriptors2 = sift.detectAndCompute(img2_gray, None)
10
11 # Match using Brute Force Matcher
12 bf = cv2.BFMatcher()
13 matches = bf.knnMatch(descriptors1, descriptors2, k=2)
14
15 # Lowe's ratio to filter best matches
16 best_matches = []
17 for m, n in matches:
18     if m.distance < 0.75 * n.distance:
19         best_matches.append(m)

```

Listing 4: SIFT feature matching.

The Code for (b)

```

1 # RANSAC parameters
2 num_points = 4
3 thres = 1.0
4 iters = 200

```

```

5 best_homography = None
6 best_inlier_count = 0
7 best_inliers = None
8
9 for i in range(iters):
10     # Get random matches
11     chosen_matches = np.random.choice(good_matches, num_points, replace=False)
12
13     src_points = []
14     dst_points = []
15
16     for match in chosen_matches:
17         src_points.append(np.array(keypoints1[match.queryIdx].pt))
18         dst_points.append(np.array(keypoints2[match.trainIdx].pt))
19
20     src_points = np.array(src_points)
21     dst_points = np.array(dst_points)
22
23     # Estimate homography
24     tform = transform.estimate_transform('projective', src_points, dst_points)
25
26     # Get number of inliers (assuming get_inliers function is defined)
27     inliers = get_inliers(src_full, dst_full, tform, thres)
28
29     # Check if this is the best model so far
30     if len(inliers) > best_inlier_count:
31         best_inlier_count = len(inliers)
32         best_homography = tform
33         best_inliers = inliers

```

Listing 5: RANSAC for homography estimation.

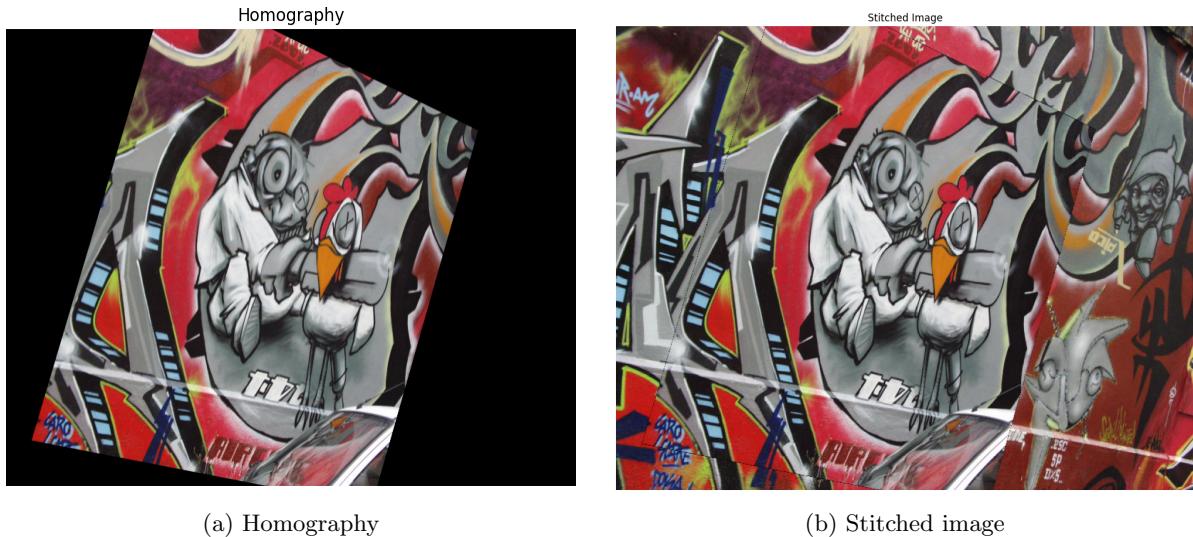


Figure 5: Homography Transformation and stitched image.