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Task 1: Pre-processing

Subtask A

Load calibration images and video file

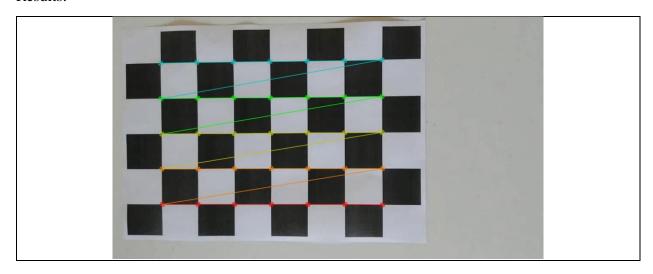
```
def getInputData():
    path = os.path.dirname(os.path.realpath(__file__))
    file_names = [filename for filename in os.listdir(path+'\\Assignment_MV_02_ca
libration\\') if filename.endswith("png")]
    chessImages = [os.path.join(path+'\\Assignment_MV_02_calibration\\', fn) for
fn in file_names]
    vid_name = "Assignment_MV_02_video.mp4"
    video = os.path.join(path, vid_name)
    return chessImages, video
```

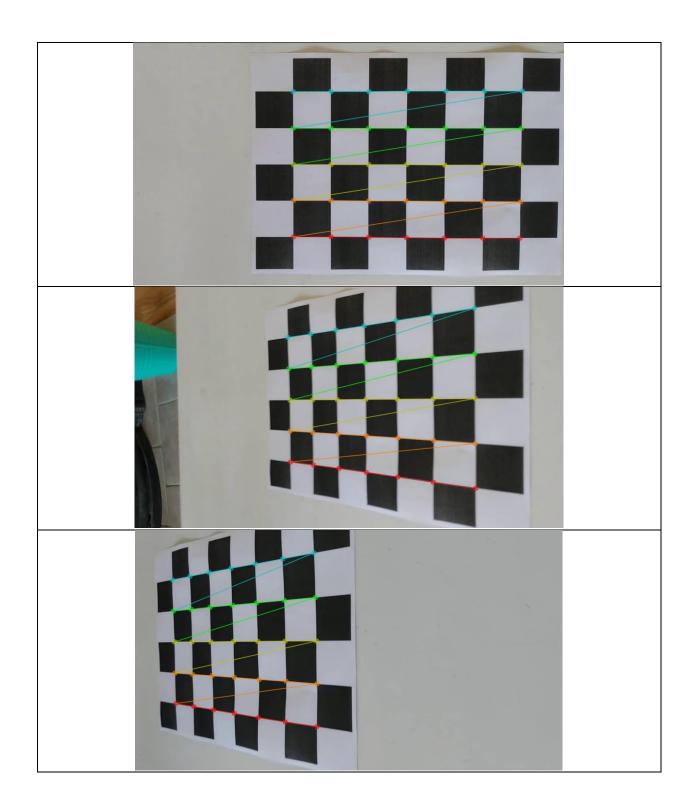
Extract checkerboard corners to subpixel accuracy

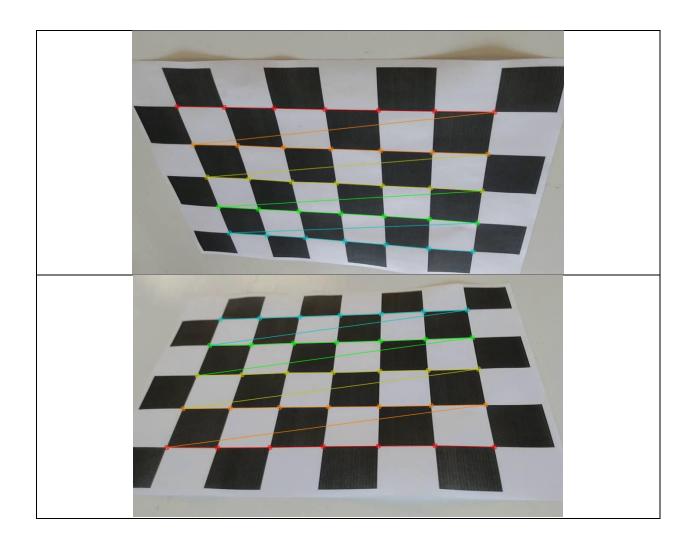
```
ret, corners = cv2.findChessboardCorners(gray, (7,5), None)

if ret == True:
    objpoints.append(objp)
    # get subpixel accurate corner points
    corners2 = cv2.cornerSubPix(gray,corners, (11,11), (-1,-1), criteria)
    imgpoints.append(corners)
    # Draw and display the corners
    cv2.drawChessboardCorners(img, (7,5), corners2, ret)
```

Results:





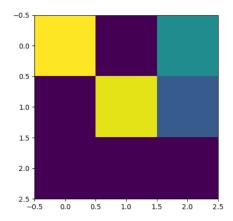


Subtask B

Obtain camera calibration matrix K

ret, mtx_K, dist, rvecs, tvecs = cv2.calibrateCamera(objpoints, imgpoints, gray.shape[::-1], None, None) return mtx_K

Output of matrix K:



```
[[994.93741694 0. 485.13219579]
[ 0. 953.82995345 286.16775074]
[ 0. 0. 1. ]]
```

Subtask C

Identify good features to track for first frame of input video. Refine to subpixel accuracy.

```
def firstFrameFeatures(video):
    # extract 1st frame
    images = extract_frames(video, [1])
    img = images[1].copy()
    new_img = cv2.cvtColor(img, cv2.COLOR_RGB2GRAY)
    corners = cv2.goodFeaturesToTrack(new_img, 200, 0.3, 7)
   criteria = (cv2.TERM_CRITERIA_EPS + cv2.TERM_CRITERIA_MAX_ITER, 100, 0.001)
    # get features with subpixel accuracy
   p0 = cv2.cornerSubPix(new_img,np.float32(corners),(5,5),(-1,-1),criteria)
    # visualize feature points
    for i,points in enumerate(p0):
       x,y = points[0]
        cv2.circle(img,(x,y), 5, (0,255,0), -1)
    cv2.imshow("", img)
    cv2.waitKey()
    cv2.destroyAllWindows()
    saveImg("firstFrameFeatures.png", img)
   return img
```

First frame features visualized:



Subtask D

Track features across whole image sequence and refine coordinates to subpixel accuracy

```
def featureTracking(video):
    stream = cv2.VideoCapture(video)
   while stream.isOpened():
        ret,img= stream.read()
        if ret:
            new img = cv2.cvtColor(img, cv2.COLOR RGB2GRAY)
            p0 = cv2.goodFeaturesToTrack(new_img, 200, 0.3, 7)
            # refine features to subpixel accuracy
            criteria = (cv2.TERM CRITERIA EPS + cv2.TERM CRITERIA MAX ITER, 100,
0.001)
            p0 = cv2.cornerSubPix(new img, np.float32(p0), (5,5), (-1,-
1),criteria)
            break
    # initialise tracks
    index = np.arange(len(p0))
    tracks = {}
    for i in range(len(p0)):
        tracks[index[i]] = {0:p0[i]}
    frame = 0
    frame_img=[]
    while stream.isOpened():
        ret,img= stream.read()
       if not ret:
            break
        frame += 1
        old img = new img
        new_img = cv2.cvtColor(img, cv2.COLOR_RGB2GRAY)
        # calculate optical flow
        if len(p0)>0:
            p1, st, err = cv2.calcOpticalFlowPyrLK(old_img, new_img, p0, None)
            # visualise points
            for i in range(len(st)):
                if st[i]:
                    cv2.circle(img, (p1[i,0,0],p1[i,0,1]), 2, (0,0,255), 2)
                    cv2.line(img, (p0[i,0,0],p0[i,0,1]), (int(p0[i,0,0]+(p1[i][0,
0]-p0[i,0,0])*5),int(p0[i,0,1]+(p1[i][0,1]-
p0[i,0,1])*5)), (0,0,255), 2)
            p0 = p1[st==1].reshape(-1,1,2)
            index = index[st.flatten()==1]
```

Task 2: Fundamental Matrix

Subtask A

Extract and visualize the feature tracks obtained in task 1.



Establish coordinate correspondence between first frame and last frame.

```
def getCorrespondences(tracks, frame1, frame2):
    correspondences = []
    # loop through tracks to get frame correspondences
    for track in tracks:
        if (frame1 in tracks[track]) and (frame2 in tracks[track]):
            x1 = [tracks[track][frame1][0,0],tracks[track][frame1][0,1],1]
            x2 = [tracks[track][frame2][0,0],tracks[track][frame2][0,1],1]
            correspondences.append((np.array(x1), np.array(x2)))
    return correspondences
```

Subtask B

Calculate mean and standard deviations for correspondences coordinates.

```
# calculate mean and standard deviation for each correspondence
mean = np.squeeze(np.mean(np.array([correspondences]), axis=1))
std = np.squeeze(np.std(np.array([correspondences]), axis=1))
```

Normalization homographies:

Subtask C

Select eight feature correspondence at random and build a Kron matrix to calculate the fundamental matrix.

```
# obtain 8 random correspondences
    random_idx = random.sample(range(len(correspondences)), 8)
    random_correspondences = [correspondences[idx] for idx in random_idx]

# normalize the 8 random correspondences
# build matrix using said correspondences
A = np.zeros((0,9))
for x1,x2 in random_correspondences:
    x1n = np.matmul(T,x1)
    x2n = np.matmul(T_,x2)
    ai = np.kron(x1n.T,x2n)
A = np.append(A,[ai],axis=0)
```

Subtask D

Use the 8-point DLT algorithm to calculate the estimated fundamental matrix F-hat. Ensure F-hat is singular and calculate the actual fundamental matrix using the normalization homographies from subtask B with F-hat.

```
U,S,V = np.linalg.svd(A)
F = V[8,:].reshape(3,3).T

U,S,V = np.linalg.svd(F)
# calculate fundamental matrix
F = np.matmul(U,np.matmul(np.diag([S[0],S[1],0]),V))
F = np.matmul(T_.T, np.matmul(F, T))
```

Subtask E

Get remaining unselected feature correspondences. Calculate the value of model equation and variance using the point observation covariance matrix *Cxx*.

```
# get unselected correspondences
new_correspondences = [element for i, element in enumerate(correspondences) if i
not in random idx]
x = np.squeeze(np.array([new_correspondences]))[:,0]
x_ = np.squeeze(np.array([new_correspondences]))[:,1]
# point observation covariance matrix
Cxx = np.array([[1,0,0],
               [0,1,0],
               [0,0,0]])
count_outliers = 0
accumulate_error = 0
inliers = []
‡ loop through unselected correspondences
for x1,x2 in new_correspondences:
   # calculate gi and variance
   value = np.matmul(np.matmul(x2.T, F), x1)
   variance = np.matmul(np.matmul(np.matmul(x2.T, F), Cxx), F.T), x2)
   variance += np.matmul(np.matmul(np.matmul(np.matmul(x1.T, F.T), Cxx), F), x1)
```

Subtask F

Determine for each correspondence if they are outliers by calculating the test statistic with an outlier threshold of 6.635. Sum the test statistics over all inliers.

```
# threshold the test statistics
    if test_stat > 6.635:
        # count outliers
        count_outliers += 1
        # record outliers
        outliers.append((x1,x2))
    else:
        # sum the test statistics for inliers
        accumulate_error += test_stat
        # record inliers
        inliers.append((x1,x2))
```

Subtask G

Repeat the above procedure 10000 times. Select the fundamental matrix and remove all outliers for the selection of eight points which yielded the least number of outliers. Break ties by looking at the sum of the test statistic over the inliers.

```
to record lowest amount of outliers
      if count outliers < best outliers:</pre>
          best error = accumulate error
          best_outliers = count_outliers
          # get fundamental matrix and inliers for lowest amount of outliers
          final Fmat = F
          final_inliers = inliers
          final outliers = outliers
      # break ties for number of outliers
      elif count_outliers == best_outliers:
          # compare sum of test statistics when tied for number of outliers
          if accumulate_error < best_error:</pre>
              best error = accumulate error
              best_outliers = count_outliers
              # get fundamental matrix and inliers
              final Fmat = F
              final_inliers = inliers
              final outliers = outliers
```

Subtask H

Indicate outliers (red) and inliers (blue) for first and last frame:



First Frame



Last Frame

Calculate epipoles:

```
def calculate_epipoles(F):
    U,S,V = np.linalg.svd(F)
    e1 = V[2,:]

    U,S,V = np.linalg.svd(F.T)
    e2 = V[2,:]

    return e1,e2
```

Output coordinates of epipoles:

 $[0.80958472\ 0.58700027\ 0.00180572]$

[-0.80917992 -0.58755817 -0.00180669]

Task 3: Essential Matrix

Subtask A

```
def calculateEssentialMatrix(matrix_K, F):
    # calculate essential mat
    E = np.matmul(np.matmul(matrix_K.T, F), matrix_K)

# decompose E
U,S,V = np.linalg.svd(E)

# ensure positive determinants
if np.linalg.det(U)<0:
    U[:,2] *= -1
if np.linalg.det(V)<0:
    V[:,2] *= -1
return U,V</pre>
```

Subtask B

Determine the four potential combinations of rotation matrices \mathbf{R} and translation vector \mathbf{t} between the first and the last frame. Assume the camera was moving at 50km/h and that the video was taken at 30fps to determine the scale of the baseline \mathbf{t} in meters.

```
getRotate_Translate(U,V):
W = np.array([[0,-1,0],
            [1,0,0],
            [0,0,1]])
Z = np.array([[0,1,0],
            [-1,0,0],
            [0,0,0]])
# assume 50km/h and 30fps
B = (50000/3600) * (30/30)
skew_1 = B * np.matmul(U,np.matmul(Z,U.T))
skew_2 = -(skew_1)
X_{POS} = [skew_1[2,1], skew_1[0,2], skew_1[1,0]]
X_NEG = [skew_2[2,1], skew_2[0,2], skew_2[1,0]]
R1 = np.matmul(U, np.matmul(W, V.T))
R2 = np.matmul(U, np.matmul(W.T, V.T))
t1= np.matmul(np.linalg.inv(R1), X_POS)
t2= np.matmul(np.linalg.inv(R1), X_NEG)
t3= np.matmul(np.linalg.inv(R2), X_POS)
t4= np.matmul(np.linalg.inv(R2), X_NEG)
combo = [(R1,t1), (R1,t2), (R2,t2), (R2,t2)]
return combo
```

```
def get3dpoints(combo, inliers, matrix K):
    print ("length of inliers:",len(inliers))
    # 3d point coords
    x,y,z = [],[],[]
    best solution = len(inliers)+1
    # for each translation and rotation combination
    for c in combo:
        r,t = c[0], c[1]
        count = 0
        # loop through inliers
        for x1,x2 in inliers:
            m = np.matmul(np.linalg.inv(matrix_K),x1)
            m = np.matmul(np.linalg.inv(matrix K),x2)
            # linear equation set up
            LHS = np.array([[np.matmul(m.T, m), np.matmul(-m.T, np.mat-
mul(r.T, m_))],
                            [np.matmul(m.T, np.matmul(r.T, m_)), np.matmul(-
m.T, m_)]])
            RHS = np.array([np.matmul(t.T, m),
                            np.matmul(t.T, np.matmul(r.T, m_))])
            # solve linear equation
            # returns 2 element array containing results from linear equation
            res = np.matmul(np.linalg.inv(LHS), RHS)
            lambd, mue = res[0], res[1]
            # if points are in front of both cameras
            if lambd > 0 and mue > 0:
                count+=1
                # calculate the 3d points for both frame correspondence
                x lambd = lambd * m
                x_mue = t + (mue*np.matmul(r.T, m_))
                # obtain final 3d point by averaging the distance be-
tween xlambda and xmue
                x_{final} = (x_{lambd} + x_{mue}) / 2
                x.append(x_final[0])
                y.append(x_final[1])
                z.append(x_final[2])
tion
        if count < best_solution:</pre>
            best solution = count
            best_x = x
            best_y = y
            best_z = z
            best_t = t
```

Subtask D

Create a 3d plot to show the two camera centers (blue) and all 3d points (red).

```
def plot3d(x,y,z, t):
    import matplotlib.pyplot as plt

fig = plt.figure()
    ax = plt.axes(projection='3d')

# plot 3d points
    ax.scatter3D(x, y, z, c='r')

# plot 1st and 2nd camera center
    ax.scatter3D(0,0,0, c='b')
    ax.scatter3D(t[0],t[1],t[2], c='b')

plt.show()
```

Output:

