Assignment 1: Camera Calibration

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```
import numpy as np
from IPython.core.display import display, HTML
display(HTML("<style>.container { width:100% !important; }</style>"))
from sympy import Matrix
import random
import cv2 as cv
from math import sqrt
import matplotlib.pyplot as plt
```

```
print(np.__version__)
```

```
1.15.4
```

Camera class

The class represent a camera that has some parameters like K, R, P, t. It also contain functions like :

- get camera parameters: It will calculate camera parameters using DLT.
- get image coordinates: It will map a 3-D real world point to 2-D image point.
- RANSAC: It will calculate camera parameters using DLT.

DLT

It is a Camera Calibration technique that require mapping of 3D real world coordinates to 2D image coordinates. The real world coordinates can be represented as\begin{equation*} $X = [x \ y \ z \ w]$ \end{equation*} while the image coordinates can be represented as \begin{equation*} \bar{x} = [u\ v\ w] \end{equation*} \bar{x} = [u\ v\ w] \end{equation*} \text{ transformation from 3D points to 2D points can be done through a **3 X 3** matrix **P** . Thus combining above \begin{equation} \equiv \begin{equation} \equiv \alpha = \equiv \begin{equation} \equiv \alpha = \equiv \begin{equation} \equiv \equiv \alpha = \equiv \begin{equation} \equiv \alpha = \equiv \begin{equation} \equiv \equiv \equiv \qquation \equiv \equiv \equiv \qquation \equiv \equiv \equiv \qquation \equiv \equiv \equiv \equiv \qquation \equiv \equiv \equiv \qquation \equiv \equ

Where P is the camera matrix and can be represented as\begin{equation*} $P = K[R|t] = KR|Kt \cdot M$ Where K is the intrinsic parameters of camera while R, t are rotational and transational matrices and vectors.

However this can be further solved using\begin{equation*} $KR*(KR)^T = KK^T \cdot \{end\{equation*\}\}$ It can be solved to find K which then can be used to find R and t.

RANSAC

Random sample consensus (RANSAC) is an iterative method to estimate parameters of a mathematical model from a set of observed data that contains outliers, when outliers are to be accorded no influence on the values of the estimates. Therefore, it also can be interpreted as an outlier detection method. I ve set number of iterations to be 100 and threshold to 200. Where 200 is the allowed euclidean distance.

```
class Camera:
    def __init__(self):
       self.P = None
        self.K = self.R = self.C = None
    def change_K(self, K):
        self.K = np.array(K)
        self.KR = np.dot(self.K, self.R)
        self.Kt = np.dot(self.K, self.t)
                  display(self.K, self.R, self.t, self.KR, self.Kt)
        self.P = np.concatenate((self.KR, self.Kt), axis=1)
    def get_camera_parameters(self,
                              real_world_coordinates,
                              image_coordinates,
                              change=True):
        size = len(real_world_coordinates) * 2
        A = np.empty(shape=(size, 12))
        count = 0
        for real_point, image_point in zip(real_world_coordinates,
                                           image_coordinates):
                          print(real_point, image_point)
            A[count] = np.array([
                -real_point[0], -real_point[1], -real_point[2], -1, 0, 0,
0, 0,
                real_point[0] * image_point[0], real_point[1] *
image_point[0],
                real_point[2] * image_point[0], image_point[0]
            ])
            A[count + 1] = np.array([
                0, 0, 0, 0, -real_point[0], -real_point[1], -real_point[2],
-1,
                real_point[0] * image_point[1], real_point[1] *
image_point[1],
                real_point[2] * image_point[1], image_point[1]
            1)
            count += 2
        print(A)
         return A
        S, V, d = np.linalg.svd(A)
        # print(d)
        P = d[-1, :] / d[-1, -1]
        P = np.reshape(P, (3, 4))
        if change:
            self.P = P
```

```
return P
   def get_KRC(self, change=True):
        KR = self.P[np.ix_{([0, 1, 2], [0, 1, 2])}]
                  print("KR = ", KR)
        KRC = self.P[np.ix_{([0, 1, 2], [3])}]
        temp = np.dot(KR, KR.T)
        temp = temp / temp[2][2]
                  print("TEMP = ", temp)
        u0 = temp[2][0]
        v0 = temp[2][1]
        beta = np.sqrt(temp[1][1] - v0 * v0)
        gamma = (temp[1][0] - u0 * v0) / beta
        alpha = np.sqrt(temp[0][0] - gamma**2 - u0**2)
        self.K = np.array([[alpha, gamma, u0], [0, beta, v0], [0, 0, 1]])
        inverse = np.linalg.inv(self.K)
        self.R = np.dot(inverse, KR)
        self.t = np.dot(inverse, KRC)
        return self.K, self.R, self.t
   def get_image_coordinates(self, real_world_coordinates):
        image_coordinates = []
                  print(real_world_coordinates)
        for world_point in real_world_coordinates:
            world_point = np.array(world_point)
                          print(world_point)
            #
                          world_point.append(1)
            #
                          print(world_point)
            world_point = np.append(world_point, 1)
                          print(world_point_)?
            res = np.dot(self.P, world_point.T)
            image_coordinates.append(res[:2] / res[2])
        return np.array(image_coordinates)
   def error(self, X, Y):
        return np.mean(np.sqrt([(x - y)**2 \text{ for } x, y \text{ in } zip(X, Y)]))
   def RANSAC(self,
               real_world_coordinates,
               image_coordinates,
               iterations,
               threshold=200,
               d=1):
        iter = 0
        best fit = None
        best error = 1e+30
        while iter < iterations:
            index = random.sample([i for i in
range(len(image_coordinates))],
                                   6)
            model_real_world_coordinates = [
                real_world_coordinates[i] for i in index
            ]
            model_image_coordinates = [image_coordinates[i] for i in index]
```

```
P = self.get_camera_parameters(model_real_world_coordinates,
                                            model_image_coordinates)
            no_of_inliners = 0
            error = 0.0
            check_real_world_coordinates = [
                real_world_coordinates[i]
                for i in range(len(real_world_coordinates)) if i not in
index
            1
            check_image_coordinates = [
                image_coordinates[i] for i in range(len(image_coordinates))
                if i not in index
            predicted_image_coordinates = self.get_image_coordinates(
                check_real_world_coordinates)
            points_with_allowed_error = [
                self.error(x, y) for x, y in zip(check_image_coordinates,
predicted_image_coordinates)
                          print("fhdskfhadgfdafadslf",
points_with_allowed_error)
            index = []
            for i in range(len(points_with_allowed_error)):
                if points_with_allowed_error[i] < threshold:</pre>
                    index.append(i)
            points_with_allowed_error = [
                i for i in points_with_allowed_error if i < threshold
            ]
            #
                          print(
                              "Hey there delilah whats it like in new york
city I am thousand",
                              points_with_allowed_error)
            if len(points_with_allowed_error) > d:
                model_real_world_coordinates = [
                    real_world_coordinates[i] for i in index
                model_image_coordinates = [image_coordinates[i] for i in
index]
self.get_camera_parameters(model_real_world_coordinates,
                                                model_image_coordinates)
                points_with_allowed_error = [
                    self.error(x, y) for x, y in zip(
                        check_image_coordinates,
predicted_image_coordinates)
                cur_error = np.mean(points_with_allowed_error)
                if cur_error < best_error:</pre>
                    best_fit = P
                    best_error = cur_error
            iter += 1
        self.P = best_fit
        return self.get_KRC()
```

Question 1 (DLT)

```
real_world_coordinates = [[0, 0, 0], [0, 12.3, 0],
                          [14.5, 12.3, 0], [14.5, 0, 0], [0, 0, 14.5],
                           14.5]] #, [14.5,12.3,14.5]]#, [14.5,0,14.5]]
image\_coordinates = [[1302, 1147], [1110, 976], [1411, 863], [1618, 1012],
                     [1324, 812], [1127, 658]] #, [1433, 564]]#, [1645, 704]]
image = [[120.855, 285.603], [904.145, 329.632], [3242.98, 462.119],
         [4070.4, 506.248], [4919.89, 561.41], [198.081, 1046.83],
         [926.21, 1101.99], [1676.4, 1157.15], [2448.66, 1212.31],
         [3220.92, 1267.47], [4026.27, 1333.67], [4875.76, 1388.83],
         [661.435, 2105.93], [1433.69, 2172.12], [2216.98, 2238.31],
         [3033.37, 2315.54], [3860.79, 2381.73], [4721.31, 2458.96],
         [330.468, 2403.8], [1146.85, 2469.99], [3717.37, 2712.7],
         [4655.11, 2800.96], [826.919, 2823.02], [2614.15, 2999.54],
         [3562.92, 3087.8], [4544.79, 3165.02], [462.855, 3231.22],
         [1400.6, 3319.47], [2371.44, 3407.02], [3375.37, 3507.31]]
image = np.array(image)
obj = [[6, 2, 0], [5, 2, 0], [2, 2, 0], [1, 2, 0], [0, 2, 0], [6, 1, 0],
       [5, 1, 0], [4, 1, 0], [3, 1, 0], [2, 1, 0], [1, 1, 0], [0, 1, 0],
       [5, 0, 1], [4, 0, 1], [3, 0, 1], [2, 0, 1], [1, 0, 1], [0, 0, 1],
       [5, 0, 2], [4, 0, 2], [1, 0, 2], [0, 0, 2], [4, 0, 3], [2, 0, 3],
       [1, 0, 3], [0, 0, 3], [4, 0, 4], [3, 0, 4], [2, 0, 4], [1, 0, 4]]
obj = np.array(obj) * 36
```

```
camera = Camera()
A = camera.get_camera_parameters(obj[:-3], image[:-3])
camera.get_KRC()
```

```
display(camera.get_image_coordinates(obj[-3:]), image[-3:])
```

Question 2 (RANSAC)

```
def RMSError(preds, actual):
    return np.sqrt(np.sum((preds - actual)**2, axis=1))
```

```
camera.RANSAC(obj, image, iterations=100)
```

Question 3

Error in DLT and RANSAC

```
predicted_point = camera.get_image_coordinates(obj)
# print(predicted_point.shape, image.shape)
print("Ransac", np.mean(RMSError(image, predicted_point)))
```

```
Ransac 9.875299761233322
```

```
camera = Camera()
A = camera.get_camera_parameters(obj, image)
predicted_point = camera.get_image_coordinates(obj)
# print(predicted_point.shape, image.shape)
print("DLT", np.mean(RMSError(image, predicted_point)))
```

```
DLT 9.05262280856669
```

Question 4 and Question 6

We are given images of checker board. The checkerboard has inner point of shape **8X6**. I've used findChessboardCorners to find corners that are inside. Then I've used cv.cameraCalibrate function It resturns the distortion matrix. Using this distortion matrix and camera matrix we can find new camera matrix. Which in then used to undistort image.

```
from glob import glob
criteria = (cv.TERM_CRITERIA_EPS + cv.TERM_CRITERIA_MAX_ITER, 30, 0.001)
objp = np.zeros((6 * 8, 3), np.float32)
objp[:, :2] = np.mgrid[0:8, 0:6].T.reshape(-1, 2)
objpoints = []
imgpoints = []
img_path = glob('*.JPG')
```

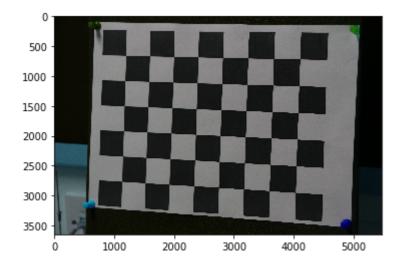
```
img_path.remove("IMG_5455.JPG")
for path in img_path:
    img = cv.imread(path)
    gray = cv.cvtColor(img, cv.COLOR_BGR2GRAY)
    ret, corners = cv.findChessboardCorners(gray, (8, 6), None)
    if ret == True:
        objpoints.append(objp)
        corners2 = cv.cornerSubPix(gray, corners, (11, 11), (-1, -1),
criteria)
        imgpoints.append(corners)
        cv.waitKey(2000)
    cv.destroyAllWindows()
```

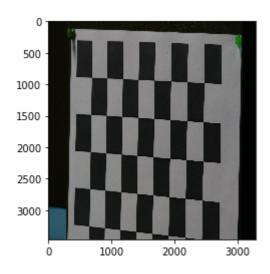
```
ret, mtx, dist, rvecs, tvecs = cv.calibrateCamera(
  objpoints, imgpoints, (gray.shape[1], gray.shape[0]), None, None)
```

```
mtx
```

```
im___ = cv.imread("IMG_5457.JPG")
plt.imshow(im___)
```

<matplotlib.image.AxesImage at 0x7f7dbc47c390>





```
display("ret", ret)
display("mtx", mtx)
display("dist", dist)
display("rvecs", rvecs)
display("tvecs", tvecs)
```

```
'ret'
2.2884151455296062
'mtx'
array([[1.36415094e+04, 0.00000000e+00, 3.31635881e+03],
       [0.00000000e+00, 1.36632517e+04, 1.50037396e+03],
       [0.00000000e+00, 0.0000000e+00, 1.00000000e+00]])
'dist'
array([[ 9.79057908e-02, 9.45876425e+00, -1.53012034e-02,
         2.72096493e-02, -1.48434591e+02]])
'rvecs'
```

```
[array([[ 0.224736 ],
        [-0.40391396],
        [-0.05729597]]), array([[-0.38062218],
        [ 0.00937555],
        [ 0.06939662]]), array([[-0.19331795],
        [-0.38351388],
        [-0.00385092]]), array([[ 0.24480949],
        [-0.32074973],
        [ 0.02437853]]), array([[-0.00934508],
        [-0.41443156],
        [-0.03564632]]), array([[-0.44170342],
        [-0.42401968],
        [ 0.05266029]]), array([[ 0.01632709],
        [-0.61709955],
        [-0.07773616]]), array([[-0.01040574],
        [-0.04719484],
        [-0.00305521]]), array([[0.00513069],
        [0.30996097],
        [0.06536842]]), array([[-0.32562264],
        [-0.22056304],
        [-0.01697744]]), array([[0.3105252],
        [0.0684347],
        [0.07015431]]), array([[-0.37489968],
        [-0.48549731],
        [ 0.02866769]]), array([[ 0.34965336],
        [-0.56196712],
        [-0.06068481]]), array([[-0.11408779],
        [-0.29596135],
        [ 0.00481932]]), array([[-0.00628759],
        [ 0.22827803],
        [ 0.03519236]])]
'tvecs'
[array([[-4.63768612],
        [-0.89986242],
        [38.70403393]]), array([[-4.20789684],
        [-2.26805866],
        [35.06790501]]), array([[-3.54440684],
        [-2.03938703],
        [31.92571791]]), array([[-5.59967776],
        [-1.21713138],
        [34.98571624]]), array([[-4.48832387],
        [-1.77689402],
        [28.99878095]]), array([[-4.4561796],
        [-2.28203658],
        [33.30645999]]), array([[-4.11373552],
        [-1.74328426],
        [28.33627768]]), array([[-5.09954028],
```

```
[-2.0599507],
[30.85011107]]), array([[-4.76286421],
[-2.13319095],
[32.66230913]]), array([[-5.10103684],
[-1.86779325],
[35.9587791 ]]), array([[-5.26909064],
[-1.69789282],
[30.50435301]]), array([[-5.12745147],
[-2.04767454],
[34.70799057]]), array([[-2.12512359],
[-1.32570061],
[29.33561584]]), array([[-4.13203861],
[-1.94212036],
[38.58165265]]), array([[-5.02987713],
[-1.97111128],
[32.32626637]])]
```

Radial Distortion

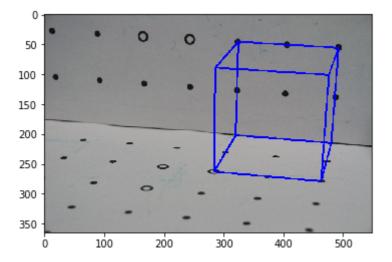
- k1 = 9.79057908e-02
- k2 = 9.45876425e+00
- p1 = -1.53012034e-02,
- p2 = 2.72096493e-02
- k3 = -1.48434591e+02

Question 5

I have drawn image of 3-D cube. I have taken a unit cube with one point at origin. Then used the calculated camera matrix form DLT to find image coordinates. These points are then connected to form cube.

```
measurement = cv.imread("measurements.jpg")
wireframe_iamge = cv.imread("IMG_5455.JPG")
wireframe_iamge = cv.resize(wireframe_iamge, (548, 365))
print(measurement.shape, wireframe_iamge.shape)
predicted_points = camera.get_image_coordinates(obj)
3d_{corners} = np.float32([[0, 0, 0], [1, 0, 0], [1, 0, 1], [0, 0, 1],
                          [0, 1, 0], [1, 1, 0], [1, 1, 1], [0, 1, 1]]) * 72
cube_corners_2d = np.array(camera.get_image_coordinates(_3d_corners)) / 10
cube_corners_2d = np.array(cube_corners_2d, np.int32)
red = (0, 0, 255)
line width = 2
#first draw the base in red
for i in range(4):
    cv.line(wireframe_iamge, tuple(cube_corners_2d[i]),
            tuple(cube_corners_2d[(i + 1) % 4]), red, line_width)
    cv.line(wireframe_iamge, tuple(cube_corners_2d[i]),
            tuple(cube_corners_2d[i + 4]), red, line_width)
```

```
(365, 548, 3) (365, 548, 3)
```

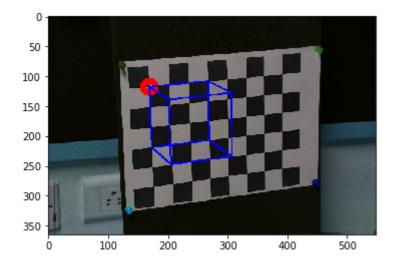


Result

The cube two faces are prefectly aligned with the given surface thus the camera matrix predicted is good. It is also supported by the fact that the rms error is less than 9. Thus the results are satisfying

Question 7

I have drawn image of 3-D cube. I have taken a cube with one point at origin. Then used the calculated camera matrix form DLT to find image coordinates. These points are then connected to form cube.



Results:

The cube formed is quite good in shape thus the camera matrix calculated from cv.CameraCalibrate is good.

Question 8

The image coordinate of origin are (4800, 2163)

```
print(camera.get_image_coordinates([[0, 0, 0]]))
```

```
[[4800.39488964 2158.58816631]]
```

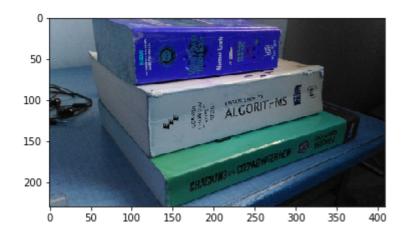
Question 9 and 10

Observations:

Results observed from DLT and RANSAC are satisfactory. Because I've used 33 points to calibrate camera in DLT. The camera used for images are different.

```
img = cv.imread('DLT.jpg')
img = cv.resize(img, (409, 230))
plt.imshow(img)
```

```
<matplotlib.image.AxesImage at 0x7f7dbc2ad630>
```



Calibration using DLT

```
camera = Camera()
A = camera.get_camera_parameters(obj, img)
camera.get_KRC()
```

```
[ 0.79418476],
[ 1. ]]))
```

Error in DLT

```
predicted_point = camera.get_image_coordinates(obj)
# print(predicted_point.shape, image.shape)
print("DLT", np.mean(RMSError(img, predicted_point)))
```

```
DLT 62.276992361320936
```

Calibration using RANSAC

```
camera.RANSAC(obj, img, iterations=100)
```

Error in RANSAC

```
predicted_point = camera.get_image_coordinates(obj)
# print(predicted_point.shape, image.shape)
print("RANSAC", np.mean(RMSError(img, predicted_point)))
```

```
RANSAC 249.2259897021633
```

Challanges

Understanding DLT and how it is solved. Referred to sir's notes and internet for SVD

• RANSAC: The turorial helped in getting a rough idea of what RANSAC is. Later, referred to the wikipedia page to understand more.

• Zhang's Method: Used OpenCV's method's to implement Zhang's method

Learnings

- Learn about the DLT, RANSAC and Zhang's methods for camera calibration
- Learn more about numpy and functions like flatten, etc.
- · Learnt how to plot images using matplotlib