

# Assgn1

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## 1 Computer Vision Assignment 1

The Assignment comprises of various tasks

1. Camera Calibration(DLT)
2. Camera Calibration(Ransac)
3. Create a wireframe using Calibration Matrix(Ransac)
4. Radial Distortion
5. Zhang's algorithm
6. Create a wireframe using Calibration Matrix(Zhang)

In [5]: *# Important Imports*

```
import os
import cv2
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
```

In [6]: *# Helper functions*

```
def display_images(img_list, shape,fig_size=(8,8),is_gray=None):
    """
        Display multiple images using matplotlib
        @param img_list=> mxn matrix of images to be displayed
        @param shape=> mnx shape
        @param is_gray=> mnx matrix, is the i,j th the image grayscaled

        return None
    """

    if is_gray is None:
        is_gray = np.zeros(shape)

    m,n = shape
    fig = plt.figure(figsize=fig_size)
```

```

for i in range(m):
    for j in range(n):
        ax = fig.add_subplot(m,n,i*n + j+1)
        if is_gray[i,j] == 1:
            ax.imshow(img_list[i][j],cmap='gray')
        else:
            ax.imshow(img_list[i][j])
        ax.axis('off')

```

## 1.1 Task1: Camera Calibration using DLT

DLT is one of the robust methods for calibration

Using DLT we calculate the interinsic(K) and extrinsic camera matrix( $M=[R|T]$ ) also known as Derivation

1. Define P
2. Get the 6 points
3. Restructure P to M
4. Formulate least square error equation
5. Caclulate SVD for M
6. Prove minimum egien value is the solution for P
7. Calualte P
8. Calculate t from P
9. QR decomposition for K and R
10. Make sure R is positive

Code:

1. Vectorize p = p12
2. Take 6 Points
3. Create M
3. Calulate SVD Get least eigen vector
5. Calulate t, R, K

Result:

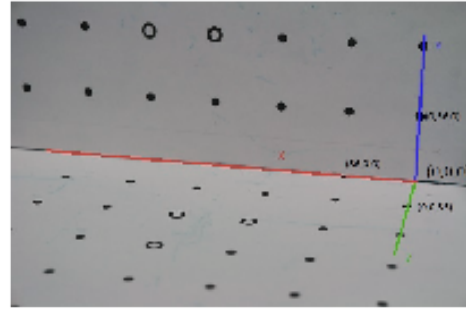
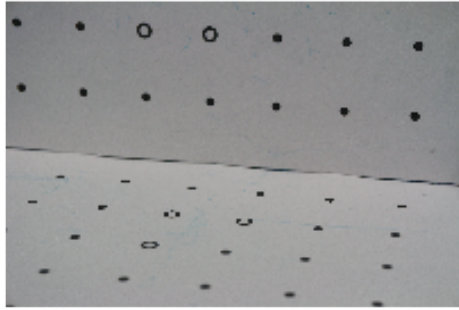
Ref: [DLT](#)

### 1.1.1 Get our 6 points of choice

```

In [28]: calb_im = cv2.imread('./images/IMG_5455.JPG')
         ref_im = cv2.imread('./measurements.jpg')
         display_images([[calb_im,ref_im]],(1,2))

```



Instead of using Harris detector, as there are only 6 points I am manually finding their x,y positions in the image

```
In [29]: # v,u image coordinates
img_coord = [[4826.01,2167.89],
             [3982.35,2091.31],
             [4738.43,2443.27],
             [4881.82,1374.34],
             [3878.07,2378.09],
             [4047.53,1322.20]]

img_coord = np.array(img_coord).astype('float32')

# z,y,x original coordinates
real_coord = [[0 ,0 , 0],
              [36, 0, 0],
              [0 , 0,36],
              [0 ,36, 0],
              [36,0 ,36],
              [36,36,0]]

real_coord = np.array(real_coord).astype('float32')
```

## 2 DLT function

```
In [45]: def dlt(img_coord,real_coord,verbose=False,img=None):
        """
        Compute Camera calibration matrix using 6 2d and 3d points
        @param => img_coord: 2d matrix with image coordinates, shape=6,2
        @param => real_coord: 3d matrix with image coordinates, shape=6,3

        return:
            ret: Error due to real world data
            P => Projection matrix
            K => Intrinsic matrix, upper triangle matrix
            R => Rotation Matrix, 3x3 Matrix Orthogonal Matrix
```

```

        t => Translation Vector, 3x1 Translation Matrix
    """
#    Convert to float
img_coord = np.array(img_coord).astype('float32')
real_coord = np.array(real_coord).astype('float32')

#    Check for errors in input
if img_coord.shape != (6,2):
    print("2D coordinates of shape 6,2")

if real_coord.shape != (6,3):
    print("3D coordinates of shape 6,3")

### Now using these coords we can define M
M = np.zeros((12,12))
for i in range(6):
    X,Y,Z = real_coord[i,:]
    u,v = img_coord[i,:]
    M[2*i,:] = [0,0,0,0,-X,-Y,-Z,-1,u*X,u*Y,u*Z,u]
    M[2*i+1,:] = [-X,-Y,-Z,-1,0,0,0,0,v*X,v*Y,v*Z,v]
if verbose:
    print("M Matrix")
    print(M.astype('int'))

### Calculate the svd for M and return the minimum eigen vector as p
u,s,vh = np.linalg.svd(M)
ind = np.argmin(s) # Generally 11
Pt = vh[ind,:]
P = Pt.reshape(3,4)

#    Caculate mse loss between predicted and real 2d points
img_pred = P.dot(np.concatenate((real_coord.T,np.ones((1,6))),axis=0))
img_pred = img_pred[[1,0],:]/img_pred[2,:]
mse = np.mean((img_coord-img_pred.T)**2)
if verbose:

    print("Projection Matrix:",P)
    print("Predicted 2d Points from 3d coordinates\n",img_pred.T)
    print("Original 2d Coords:\n",img_coord)
    print("MSE:",mse)

    if img is not None:
        for i in range(6):
            cv2.circle(img,(int(img_pred[0,i]),int(img_pred[1,i])) ,6,(0,255,0),im

        plt.imshow(calb_im)
        plt.show()

```

```

# Calculate K,R,T
hom = P[:,0:3]
pt = P[:,3]
hom_inv = np.linalg.inv(hom)
t = hom_inv.dot(pt)
R,K = np.linalg.qr(hom)
K = K/K[2,2]

# Calculate error due to using the lowest eigen value
x = M.dot(Pt)
error = x.dot(x)
if verbose:
    print("Intrinsic Matrix:",K)
    print("Rotation Matrix:",R)
    print("Translation Matrix:",t)

    print("Error",error)

return P,K,R,t,error,mse

```

```
In [31]: P,K,R,t,error,mse = dlt(img_coord,real_coord,verbose=True,img=calb_im)
```

M Matrix

```

[[ 0 0 0 0 0 0 0 0 -1 0 0
  0 4826]
 [ 0 0 0 -1 0 0 0 0 0 0 0
  0 2167]
 [ 0 0 0 0 -36 0 0 -1 143364 0
  0 3982]
 [ -36 0 0 -1 0 0 0 0 75287 0
  0 2091]
 [ 0 0 0 0 0 0 -36 -1 0 0
 170583 4738]
 [ 0 0 -36 -1 0 0 0 0 0 0
 87957 2443]
 [ 0 0 0 0 0 -36 0 -1 0 175745
 0 4881]
 [ 0 -36 0 -1 0 0 0 0 0 49476
 0 1374]
 [ 0 0 0 0 -36 0 -36 -1 139610 0
 139610 3878]
 [ -36 0 -36 -1 0 0 0 0 85611 0
 85611 2378]
 [ 0 0 0 0 -36 -36 0 -1 145711 145711
 0 4047]
 [ -36 -36 0 -1 0 0 0 0 47599 47599
 0 1322]]

```

Projection Matrix:  $\begin{bmatrix} -1.90155101e-04 & -4.03793551e-03 & 1.23751908e-03 & 4.08962030e-01 \\ -4.12878831e-03 & 6.35296236e-04 & -9.54132077e-04 & 9.12531528e-01 \\ 7.68103385e-08 & 7.05781722e-08 & -1.03584856e-07 & 1.89071780e-04 \end{bmatrix}$

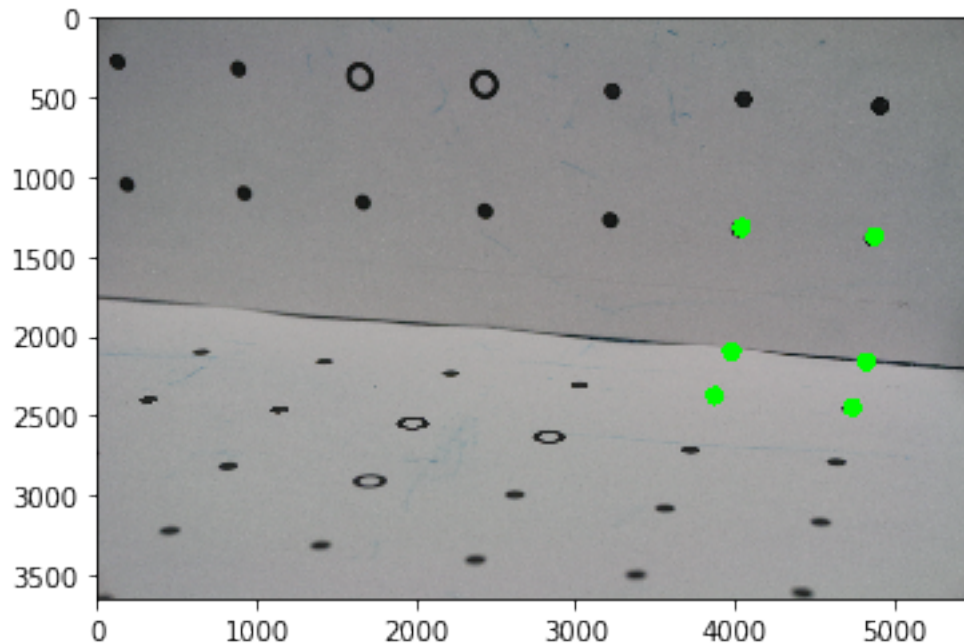
Predicted 2d Points from 3d coordinates

$\begin{bmatrix} 4826.37614506 & 2162.99878282 \\ 3982.00211548 & 2096.136546 \\ 4738.15615403 & 2446.88706571 \\ 4881.73649086 & 1375.67341292 \\ 3878.34006367 & 2374.52621375 \\ 4047.61217591 & 1320.88547236 \end{bmatrix}$

Original 2d Coords:

$\begin{bmatrix} 4826.01 & 2167.89 \\ 3982.35 & 2091.31 \\ 4738.43 & 2443.27 \\ 4881.82 & 1374.34 \\ 3878.07 & 2378.09 \\ 4047.53 & 1322.2 \end{bmatrix}$

MSE: 6.410424697741436



Intrinsic Matrix:  $\begin{bmatrix} -4.33363428e+04 & 4.70620049e+03 & -9.39654476e+03 \\ -0.00000000e+00 & 4.25994874e+04 & -1.34219408e+04 \\ -0.00000000e+00 & -0.00000000e+00 & 1.00000000e+00 \end{bmatrix}$

$\begin{bmatrix} -0.00000000e+00 & 4.25994874e+04 & -1.34219408e+04 \\ -0.00000000e+00 & -0.00000000e+00 & 1.00000000e+00 \end{bmatrix}$

Rotaion Matrix:  $\begin{bmatrix} -4.60071415e-02 & 9.98941111e-01 & 2.02589200e-05 \\ -9.98941111e-01 & -4.60071419e-02 & 1.76705600e-05 \\ 1.85839038e-05 & -1.94244961e-05 & 1.00000000e+00 \end{bmatrix}$

$\begin{bmatrix} -4.60071415e-02 & 9.98941111e-01 & 2.02589200e-05 \\ -9.98941111e-01 & -4.60071419e-02 & 1.76705600e-05 \end{bmatrix}$

$\begin{bmatrix} 1.85839038e-05 & -1.94244961e-05 & 1.00000000e+00 \end{bmatrix}$

Translation Matrix: [ 173.85393538 -795.46539864 -2238.36285262]  
Error 2.756832372035938e-06

## 2.1 Task2: Camera Calibration using Ransac

What is the algorithm

Ranac use DLT to give a more robust calibration

Code:

1. Take n Points
2. Take random 6 points from n
3. Calculate DCT and error
4. Store the best 6 case
5. Compute error on all points

Ref: [RANSAC](#)

```
In [65]: # Let us define our N points. Took way too long to calculate then thought
# 2d points
u = [4.916500e+03,4.068500e+03,3.244500e+03,2.428500e+03,1.652500e+03,8.925000e+02,1.40
v = [5.645000e+02,5.165000e+02,4.605000e+02,4.205000e+02,3.725000e+02,3.325000e+02,2.76
u = np.array(u)
v = np.array(v)

# 3d points
X = []
X.extend(range(0,216+1,36))
X.extend(range(0,216+1,36))
X.extend(range(0,180+1,36))
X.extend(range(0,180+1,36))
X.extend(range(0,144+1,36))
X.extend(range(0,144+1,36))
X = np.array(X)
Y = np.vstack([72*np.ones((7,1)), 36*np.ones((7,1)), np.zeros((22,1))]).reshape(-1)
Z = np.vstack([np.zeros((14, 1)), 36*np.ones((6,1)), 72*np.ones((6,1)), 108*np.ones((5,
```

### 2.1.1 Run Ransac

```
In [66]: # Now from the 36 points take 6 random points to compute DLT, 200 iterations
```

```
n_iter = 200
batch_size = X.shape[0]
best_values = {}
least_mse = 1e10
for i in range(n_iter):
    mini_batch = np.random.choice(batch_size, 6)

    img_coord = np.array([u[mini_batch],v[mini_batch]]).T
```

```

real_coord = np.array([X[mini_batch],Y[mini_batch], Z[mini_batch]]).T
try:
    P,K,R,t,error,mse = dlt(img_coord,real_coord)
except Exception as l:
#         print("Encountered Singular Matrix:",l)
    pass
if error < least_mse:
    least_mse = error
    best_values['P'] = P
    best_values['R'] = R
    best_values['t'] = t
    best_values['K'] = K
    best_values['mse'] = mse
least_mse,best_values

```

```

Out[66]: (2.7066817362216085e-53,
{'K': array([[ 5.23388045e+00, -7.37961443e+22,  2.94093749e+00],
 [ 0.00000000e+00,  2.11577502e+27,  6.24155808e-04],
 [ 0.00000000e+00,  0.00000000e+00,  1.00000000e+00]]),
'P': array([[ -4.51276333e-29,  6.68214034e-17,  4.47204003e-28,
  2.69331135e-25],
 [ -2.47332984e-27,  9.11943733e-14, -1.39839574e-27,
  6.01207876e-25],
 [ 8.62816615e-32, -1.00000000e+00, -2.46519033e-31,
  1.25881285e-28]]),
'R': array([[ -1.82426634e-02, -6.36286093e-07,  9.99833589e-01],
 [ -9.99833588e-01, -3.48732086e-05, -1.82426634e-02],
 [ 3.48790129e-05, -9.99999999e-01, -1.93302405e-15]]),
'mse': 161.456879888349,
't': array([ -5.52086890e+02, -3.08249810e-28,  5.46544214e+02]))}

```

## 2.1.2 Calculate error on all elements and the real and predicted points

In [67]: `P = best_values['P']`

```

ind_list = np.arange(batch_size)
img_coord = np.array([u[ind_list],v[ind_list]]).T
real_coord = np.array([X[ind_list],Y[ind_list], Z[ind_list]]).T

img_pred = P.dot(np.concatenate((real_coord.T,np.ones((1,batch_size))),axis=0))
img_pred = img_pred[[1,0],:]/img_pred[2,:]
mse = np.mean((img_coord-img_pred.T)**2)
print("MSE:",mse)

calb_im = cv2.imread('./images/IMG_5455.JPG')
for i in range(batch_size):
#     print(real_coord[i,:])
#     print(img_coord[i,:])

```



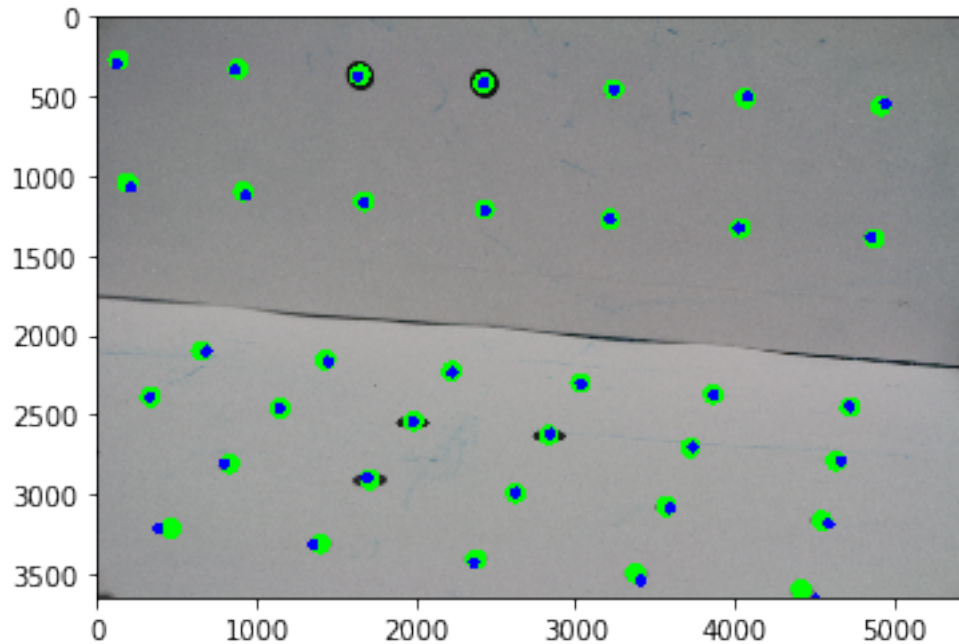
```

#     print(img_pred[:,i])
cv2.circle(calb_im,(int(img_coord[i,0]),int(img_coord[i,1])  ),6,(0,255,0),120)
cv2.circle(calb_im,(int(img_pred[0,i]),int(img_pred[1,i])  ),6,(0,0,255),60)

plt.imshow(calb_im)
plt.show()

```

MSE: 513.2822463657253

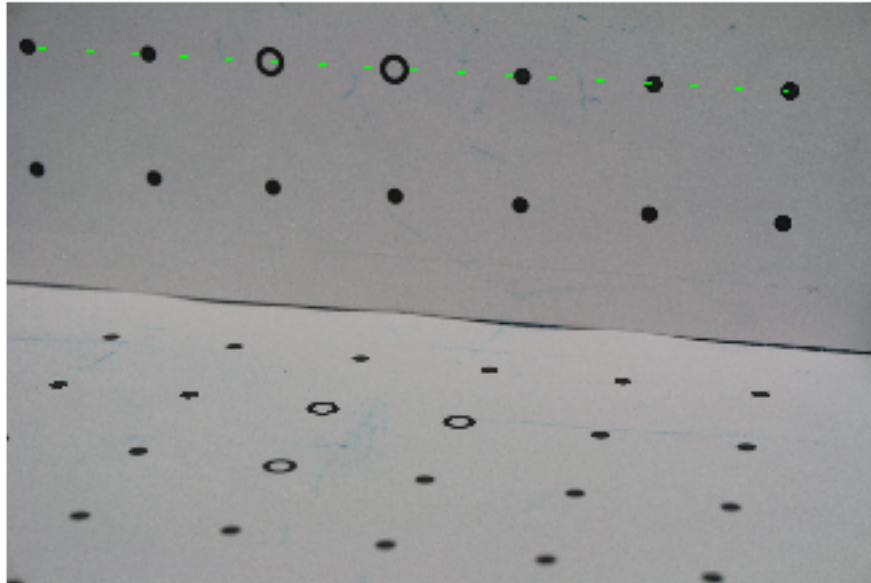


## 2.2 Correcting Radial Distorsion

```

In [37]: # Create a line on image to see distortion
calb_im = cv2.imread('./images/IMG_5455.JPG')
x1 = 4907
y1 = 561
x2 = 138
y2 = 287
new_im = cv2.line(calb_im, (x1, y1), (x2, y2), (0,255,0), 2)
plt.imshow(new_im)
plt.axis('off')
plt.show()

```



We can notice there is very less distortion in the image. Manually calculating it will be very hard

## 2.3 Create a wireframes from 3d points

We can calculate all the 3d points as all dots are separated by 36mm  
 Once we have the 3d points we use P(Projection Matrix) to get all the 2d points in the image  
 Finally we use cv2 to draw the wireframe

```
In [68]: calb_im = cv2.imread('./images/IMG_5455.JPG')
```

```
def draw_lines(start,end):
    X1,Y1,Z1 = start
    X2,Y2,Z2 = end
    v1,u1,c1 = P.dot([X1,Y1,Z1,1])
    v2,u2,c2 = P.dot([X2,Y2,Z2,1])
    cv2.line(calb_im, (int(u1/c1), int(v1/c1)), (int(u2/c2), int(v2/c2)), (0,255,0), 2)
    cv2.circle(calb_im,(int(u1/c1), int(v1/c1)),6,(0,0,255),60)
    cv2.circle(calb_im,(int(u2/c2), int(v2/c2)),6,(0,0,255),60)

# Top 2 horizontal lines
for j in range(1,3):
    for i in range(6):
        X1 = i*36
        X2 = (i+1)*36
```

```

        Y = j*36
        Z = 0
        draw_lines((X1,Y,Z),(X2,Y,Z))

# Top vertical lines
for i in range(7):
    X = i*36
    Y1 = 36
    Y2 = 72
    Z = 0
    draw_lines((X,Y1,Z),(X,Y2,Z))

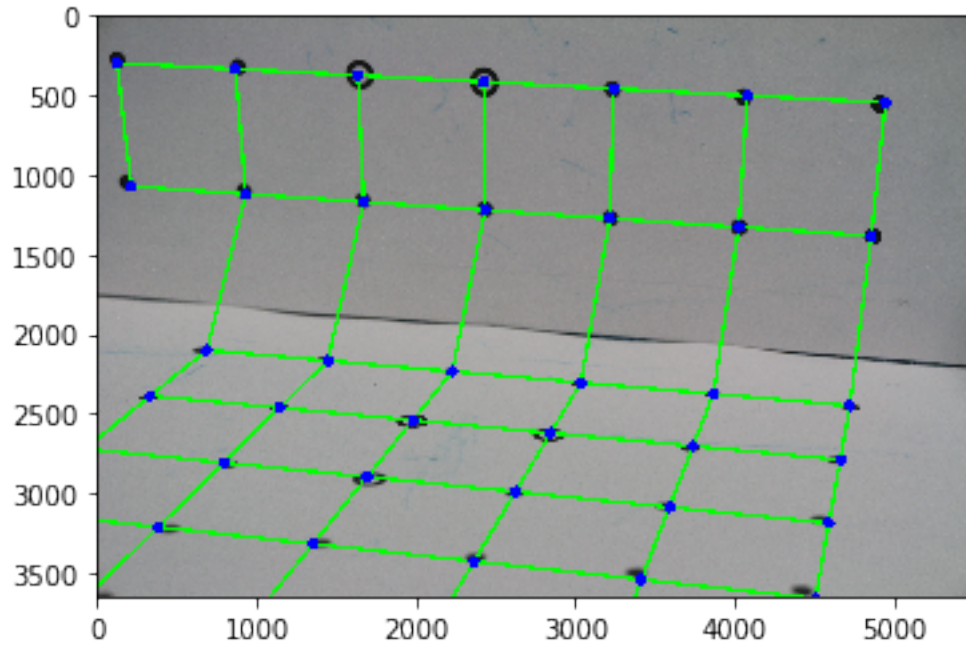
# Out of plane lines
for i in range(6):
    X = i*36
    Y1 = 36
    Y2 = 0
    Z1 = 0
    Z2 = 36
    draw_lines((X,Y1,Z1),(X,Y2,Z2))

# Bottom horizontal lines
for k in range(1,5):
    for i in range(5):
        X1 = i*36
        X2 = (i+1)*36
        Y = 0
        Z = k*36
        draw_lines((X1,Y,Z),(X2,Y,Z))

# Bottom vertical lines
for i in range(6):
    for k in range(1,5):
        X = i*36
        Y = 0
        Z1 = k*36
        Z2 = (k+1)*36
        draw_lines((X,Y,Z1),(X,Y,Z2))

plt.imshow(calb_im)
plt.show()

```



### 3 Zhangs Algorithm for Calibration

```
In [69]: # Store all images in a list, Total 15 images. Storing as 3x5 matrix for better display
zhang_img_list = []
row_list = []
for i, filename in enumerate(os.listdir('./images/')):
    # Skip the first image
    if i==0:
        continue
    # Read Image
    zhang_img = cv2.imread(os.path.join('./images/', filename))
    row_list.append(zhang_img)
    if len(row_list) == 5:
        zhang_img_list.append(row_list)
        row_list = []
```

3.0.1 Find and display all the corners detected by the cv2.

3.0.2 If we found corner points run calibration using these points

Code inspired from the tutorial

Step1: Initialization

```
In [40]: zhang_img_list_vis = []
        _3d_points=[]
```

```

_2d_points=[]

# Define 3rd world points if all corners are detected
x,y=np.meshgrid(range(8),range(6))
world_points=np.hstack((x.reshape(48,1),y.reshape(48,1),np.zeros((48,1)))).astype(np.float32)
print(world_points.shape)

```

(48, 3)

Step 2 Loop over all images

```

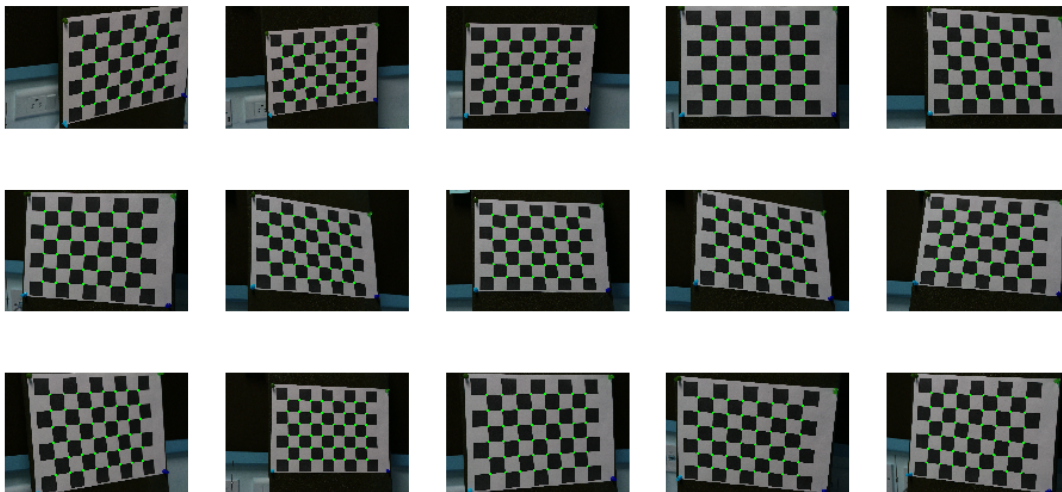
In [58]: for row in zhang_img_list:
          row_list_vis = []
          for im in row:
              # print(im.shape)
              ret, corners = cv2.findChessboardCorners(im, (8,6))

              if ret: #add points only if checkerboard was correctly detected:
                  _2d_points.append(corners) #append current 2D points
                  _3d_points.append(world_points) #3D points are always the same

              im_vis=im.copy()
              for corner in corners:
                  cv2.circle(im_vis,(corner[0,0],corner[0,1]),6,(0,255,0),60)
              row_list_vis.append(im_vis)
          zhang_img_list_vis.append(row_list_vis)

display_images(zhang_img_list_vis,shape=(3,5),fig_size=(16,8))

```



### Step3: Run Opencv built-in calibration module

```
In [59]: # Requires all 2d points and their corresponding 3d points and size of the image
ret, mtx, dist, rvecs, tvecs = cv2.calibrateCamera(_3d_points, _2d_points, (im.shape[1],
im.shape[0]), None, None)

print ("Ret(Error):",ret)
print ("Mtx(Intrinsic Matrix):\n",mtx," -----> [",mtx.shape,"]")
print ("Distortion:\n",dist," -----> [",dist.shape,"]")
print ("All rotation vecs:\n",rvecs," -----> [",rvecs.shape,"]")
print ("Translation vecs:\n",tvecs," -----> [",tvecs.shape,"]")

Ret(Error): 2.4810222497615757
Mtx(Intrinsic Matrix):
[[1.36634816e+04 0.00000000e+00 3.33651275e+03]
 [0.00000000e+00 1.36813888e+04 1.49657985e+03]
 [0.00000000e+00 0.00000000e+00 1.00000000e+00]] -----> [ (3, 3) ]
Distortion:
[[ 9.51409607e-02  1.01150408e+01 -1.52797290e-02  2.87204086e-02
 -1.60701382e+02]] -----> [ (1, 5) ]
All rotation vecs:
[array([[ 0.34957222],
       [-0.56388263],
       [-0.06027045]]), array([[ 0.22457259],
       [-0.40589059],
       [-0.05699634]]), array([[ 0.24427709],
       [-0.32261877],
       [ 0.02472153]]), array([[ -0.0131805 ],
       [-0.05189049],
       [-0.00310176]]), array([[ -0.19221837],
       [-0.38502272],
       [-0.00381435]]), array([[ -0.00723452],
       [ 0.22652406],
       [ 0.03532837]]), array([[ -0.37472804],
       [-0.48684336],
       [ 0.02851478]]), array([[ -0.32566728],
       [-0.2223195 ],
       [-0.01742815]]), array([[ -0.44150121],
       [-0.42569902],
       [ 0.05239111]]), array([[ -0.38084201],
       [ 0.00764579],
       [ 0.06908189]]), array([[ 0.01633787],
       [-0.61839111],
       [-0.07755476]]), array([[ -0.11332449],
       [-0.29806761],
       [ 0.0048852 ]]), array([[ -0.00906775],
       [-0.41619773],
       [-0.03572027]]), array([[ 0.31114199],
       [ 0.06633199],
       [ 0.07053903]]), array([[ 0.0045647 ],
       [ 0.00000000],
       [ 0.00000000]])]
```

```

    [0.30778694],
    [0.0654393 ]]]) -----> [ (3, 1) ]
Translation vecs:
[array([[-2.1671936 ],
        [-1.3173103 ],
        [29.36436163]])], array([[-4.69367327],
        [-0.88961653],
        [38.73630118]])], array([[-5.65033434],
        [-1.20810313],
        [35.01643278]])], array([[-5.14434731],
        [-2.05116345],
        [30.87691032]])], array([[-3.59093251],
        [-2.03069512],
        [31.96548236]])], array([[-5.07784942],
        [-1.96189183],
        [32.36121508]])], array([[-5.17937693],
        [-2.03793857],
        [34.74924872]])], array([[-5.15413122],
        [-1.85738242],
        [35.99344774]])], array([[-4.50526253],
        [-2.27297481],
        [33.34097639]])], array([[-4.2599015 ],
        [-2.25819832],
        [35.09462368]])], array([[-4.15548871],
        [-1.73547376],
        [28.36760298]])], array([[-4.18731971],
        [-1.93088379],
        [38.61305498]])], array([[-4.53126195],
        [-1.76821816],
        [29.02888018]])], array([[-5.31369287],
        [-1.68869488],
        [30.52756066]])], array([[-4.81207108],
        [-2.12403368],
        [32.6976201 ]]]) -----> [ (3, 1) ]

```

### 3.0.3 Draw wireframes on all images using cv2.draw chessboard function

```

In [85]: # Project 3d points of all cameras to 2d using rotation matrix and camera matrix
pred2d_list = []
h,w,c = zhang_img_list_vis[0][0].shape
for i in range(len(_3d_points)):
    pred2d, _ = cv2.projectPoints(_3d_points[i], rvecs[i], tvecs[i], mtx, dist)
    pred2d_list.append(pred2d)

    cv2.drawChessboardCorners(zhang_img_list_vis[i//5][i%5], (8,6), pred2d, 5)

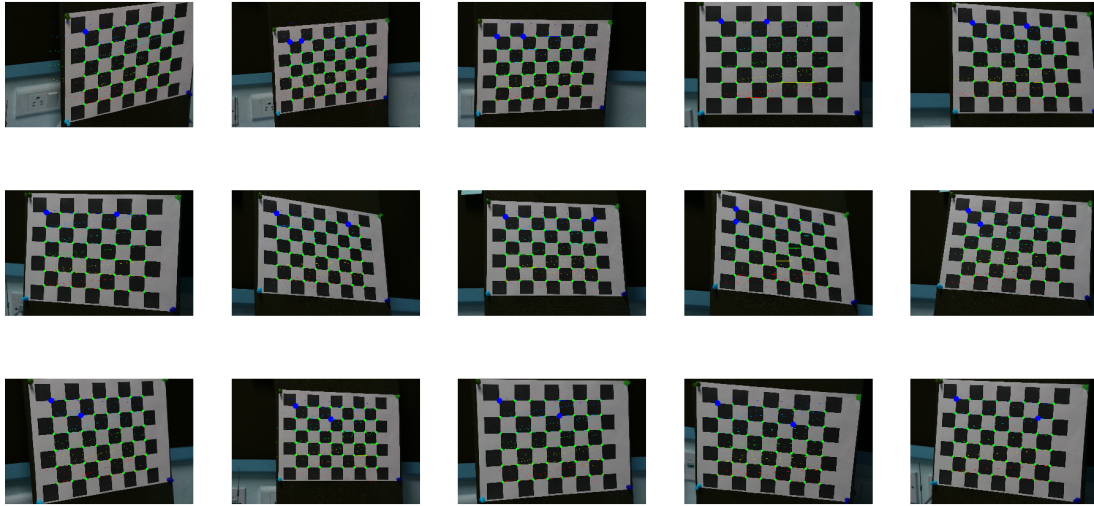
#     Draw center for the next question

```

```

cv2.circle(zhang_img_list_vis[i//5][i%5],(pred2d[0,0,0],pred2d[0,0,1]),20,(0,0,255))
display_images(zhang_img_list_vis,shape=(3,5),fig_size=(32,16))

```



### 3.1 Where is the origin

For Zhang Algorithm we can see the blue dot on the chess board in the images above.

For DLT we can notice the green dot on the paper in Figure 2, (use measurement.jpg as reference)

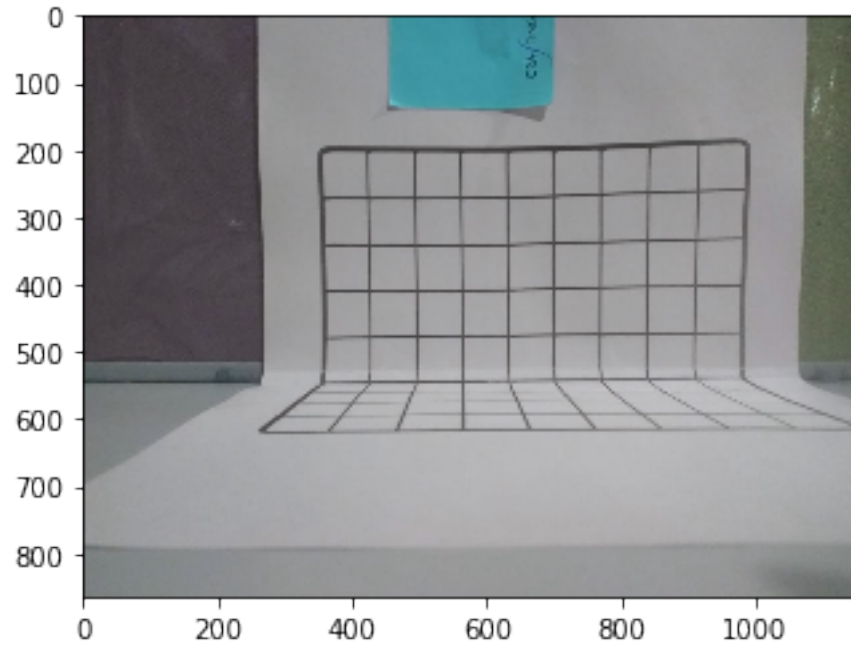
## 4 Test on your own camera'

```

In [41]: calb_im = cv2.imread('./phone.jpeg')
plt.imshow(calb_im)
plt.show()

```





```
In [43]: # v,u image coordinates
img_coord = [[432,477],
             [500,410],
             [477,589],
             [861,553],
             [426,273],
             [497,341]]

img_coord = np.array(img_coord).astype('float32')

# z,y,x original coordinates
real_coord = [[0 ,0 , 0],
              [1, 1, 0],
              [1 , -1,3],
              [6 ,-1, 1],
              [0,3 ,0],
              [1,2,0]]

real_coord = np.array(real_coord).astype('float32')
P,K,R,t,error,mse = dlt(img_coord,real_coord,verbose=True,img=calb_im)
```

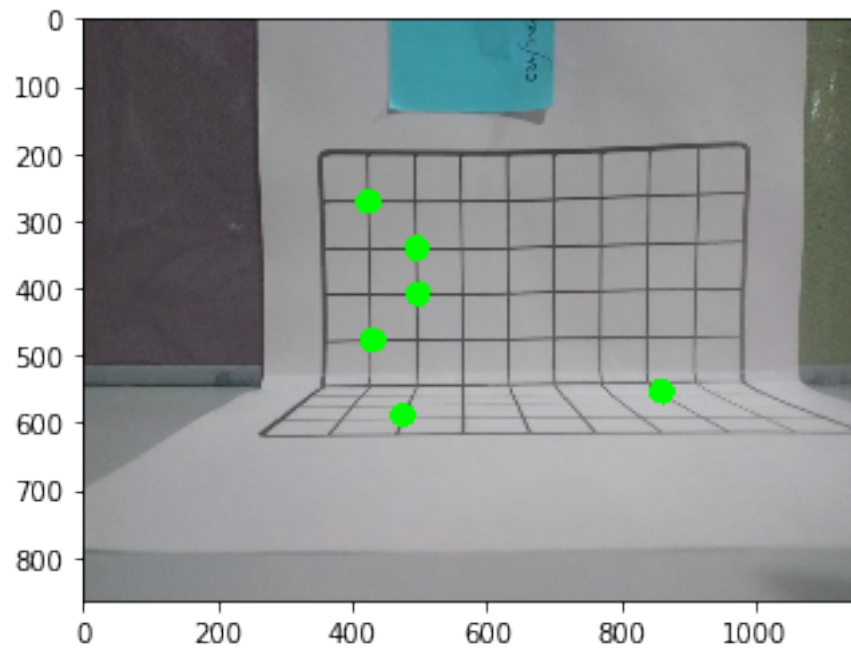
M Matrix

```
[[ 0  0  0  0  0  0  0  0 -1  0  0  0 432]
 [ 0  0  0 -1  0  0  0  0  0  0  0  0 477]
 [ 0  0  0  0 -1 -1  0 -1 500 500  0 500]
 [ -1 -1  0 -1  0  0  0  0 410 410  0 410]
 [ 0  0  0  0 -1  1 -3 -1 477 -477 1431 477]
 [ -1  1 -3 -1  0  0  0  0 589 -589 1767 589]
```

```

[  0  0  0  0 -6  1 -1 -1 5166 -861 861 861]
[ -6  1 -1 -1  0  0  0  0 3318 -553 553 553]
[  0  0  0  0  0 -3  0 -1  0 1278  0 426]
[  0 -3  0 -1  0  0  0  0  0  819  0 273]
[  0  0  0  0 -1 -2  0 -1 497 994  0 497]
[ -1 -2  0 -1  0  0  0  0 341 682  0 341]]
Projection Matrix: [[-2.14066966e-03  1.05188944e-01  4.57631342e-02 -7.30111625e-01]
[-1.08747647e-01  4.77669685e-03  6.92183957e-02 -6.61158866e-01]
[-5.70709158e-06  3.61247616e-06  1.16795582e-04 -1.52980902e-03]]
Predicted 2d Points from 3d coordinates
[[432.18392302 477.25671225]
[499.46340994 409.33602792]
[476.99845298 588.98560503]
[861.0036506  553.03528228]
[425.8333578  272.91148475]
[497.51849443 341.47577343]]
Original 2d Coords:
[[432. 477.]
[500. 410.]
[477. 589.]
[861. 553.]
[426. 273.]
[497. 341.]]
MSE: 0.11339882115423942

```



```

Intrinsic Matrix: [[ 973.38544262 -61.26564413 -627.38468094]
 [ 0.          940.32597597 397.26973428]
 [ 0.           0.           1.          ]]
Rotaion Matrix: [[-1.96809319e-02  9.99806311e-01 -3.19881738e-05]
 [-9.99806310e-01 -1.96809336e-02 -5.18504595e-05]
 [-5.24699737e-05  3.09615127e-05  9.99999998e-01]]
Translation Matrix: [ -2.36120996 -1.25728225 -13.17466588]
Error 3.185376838159553e-06

```

## 4.1 RANSAC

## 4.2 Zhangs Method

```

In [86]: # Store all images in a list, Total 9 images. Stroing as 3x3 matrix for better displayin
        zhang_img_list = []
        row_list = []
        for i,filename in enumerate(os.listdir('./phone_images_zhang/')):
            # Read Image
            zhang_img = cv2.imread(os.path.join('./phone_images_zhang/',filename))
            row_list.append(zhang_img)
            if len(row_list) == 5:
                zhang_img_list.append(row_list)
                row_list = []

In [88]: zhang_img_list_vis = []
        _3d_points=[]
        _2d_points=[]

        # Define 3rd world points if all corners are detected
        x,y=np.meshgrid(range(8),range(5))
        world_points=np.hstack((x.reshape(40,1),y.reshape(40,1),np.zeros((40,1)))) .astype(np.fl
        print(world_points.shape)

```

(40, 3)

```

In [92]: for row in zhang_img_list:
        row_list_vis = []
        for im in row:
            # print(im.shape)
            ret, corners = cv2.findChessboardCorners(im, (8,5))
            print("Found Corners:",ret)
            if ret: #add points only if checkerboard was correctly detected:
                _2d_points.append(corners) #append current 2D points
                _3d_points.append(world_points) #3D points are always the same

            im_vis=im.copy()
            for corner in corners:

```

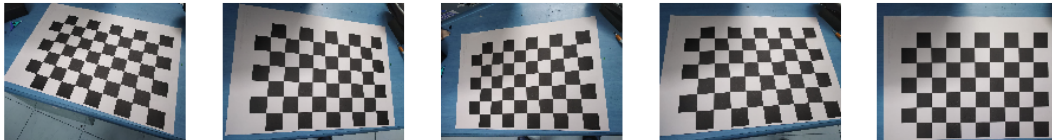
```

        cv2.circle(im_vis,(corner[0,0],corner[0,1]),6,(0,255,0),int(0.03*im.sha
    else:
        im_vis = np.zeros((240,240))
        row_list_vis.append(im_vis)
        zhang_img_list_vis.append(row_list_vis)

display_images(zhang_img_list,shape=(1,5),fig_size=(16,8))

```

Found Corners: False  
 Found Corners: False  
 Found Corners: False  
 Found Corners: False  
 Found Corners: False



**Not calculating Zhang**