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ICV21_Assignment #1

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Problem 1. 2D Fourier Transform of Images

```
import numpy as np
import matplotlib.pyplot as plt
import cv2
```

1. load images img1, and img2 with 'imread'

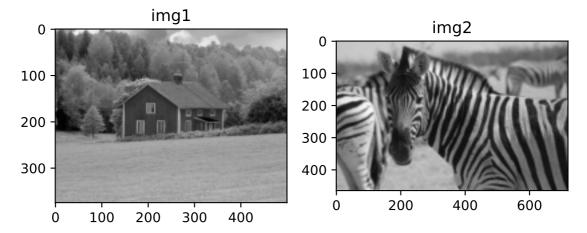
```
img1 = cv2.imread('data/img1.jpg', cv2.IMREAD_GRAYSCALE)
img2 = cv2.imread('data/img2.jpg', cv2.IMREAD_GRAYSCALE)
```

1. Display the images

```
fig, ax = plt.subplots(1, 2)

ax[0].imshow(img1, cmap='gray')
ax[1].imshow(img2, cmap='gray')
ax[0].set_title("img1")
ax[1].set_title("img2")

fig.tight_layout()
plt.show()
```



1. Next resize the img2 to be the same size as the img1

```
in [4]: [img2 = cv2.resize(img2, dsize=(img1.shape[1],img1.shape[0]))
```

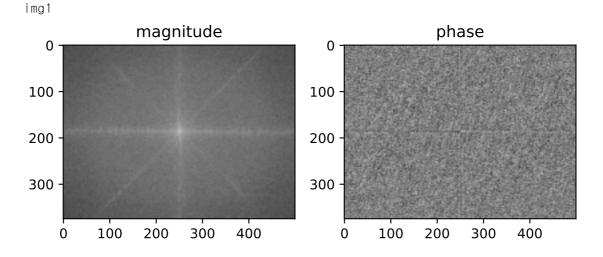
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1. Perform 2D DFT on img1, and img2

```
img1_fft = np.fft.fftshift(np.fft.fft2(img1))
img2_fft = np.fft.fftshift(np.fft.fft2(img2))
```

1. Display magnitude and phase of eache image

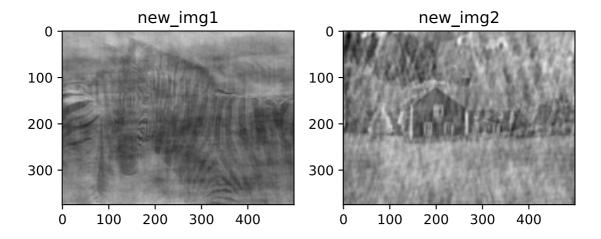
```
img1_mag = np.abs(img1_fft)
img1_phase = np.angle(img1_fft)
fig, ax = plt.subplots(1, 2)
ax[0].imshow(20*np.log(img1_mag), cmap='gray')
ax[1].imshow(img1_phase, cmap='gray')
ax[0].set_title("magnitude")
ax[1].set_title("phase")
fig.tight_layout()
print("img1")
plt.show()
img2_mag = np.abs(img2_fft)
img2_phase = np.angle(img2_fft)
fig, ax = plt.subplots(1, 2)
ax[0].imshow(20*np.log(img2_mag), cmap='gray')
ax[1].imshow(img2_phase, cmap='gray')
ax[0].set_title("magnitude")
ax[1].set_title("phase")
fig.tight_layout()
print("img2")
plt.show()
```



img2

magnitude phase

1. Synthesize new images new_img1 and new_img2 by combining img1_mag + img2_phase and img2_mag + img1_phase, respectively



Problem 1. Discussion

By looking new_img1 and new_img2, we can see that both images look similar with the original image which has the same phase.

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It is because phase has the information of the contour of the images.

Problem 2. Perspective Image Transforms

```
import numpy as np
import matplotlib.pyplot as plt
import cv2
```

1. Perform the 2D projective transforms defined by the following 4 matrices for img1

1) Read img1

```
img1 = cv2.imread('data/img1.jpg')
img1 = cv2.cvtColor(img1, cv2.COLOR_BGR2GRAY)
```

2) Given M_transposes

3) function to get projection image by given image and M matrix used equation below

```
[p \ q \ r]^T = \mathbf{M} [x \ y \ I]^T, and u = p/r and v = q/r.
```

```
def get_proj_img(img, M):
    proj_img = np.zeros(img1.shape)
    for i in range(img.shape[0]):
        for j in range(img.shape[1]):
        p, q, r = np.matmul(M, np.array([i,j,1]))
        u = p/r
        v = q/r
        if u <= img.shape[0] and v <= img.shape[1]:</pre>
```

```
proj_img[int(u)][int(v)] = img[i][j]
return proj_img
```

4) get projection images and save as projected{i}.png file

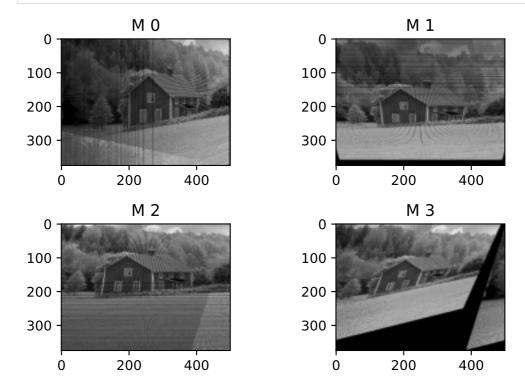
(image fie will be saved in directory projected_imgs/)

```
In [22]:
```

```
fig, ax = plt.subplots(2, 2)

for i, M_T in enumerate(M_T_list):
    proj_img = get_proj_img(img1, np.transpose(M_T))
    ax[i//2, i%2].set_title('M {}'.format(i))
    ax[i//2, i%2].imshow(proj_img, cmap='gray')
    plt.imsave('projected_imgs/projected{}.png'.format(i+1), proj_img,
    cmap='gray')

fig.tight_layout()
plt.show()
```



Problem 2. Discussion

using the equation, projection image seems to be well made

Problem 3. Camera Calibration

```
import numpy as np
import matplotlib.pyplot as plt
import cv2
```

3.a) Show that the solution p is the last column of V, corresponding to the smallest eigenvalue of A.

$$p = min_p(||Ap||), \; A = UDV^T$$

$$||Ap|| = p^TA^TAp = p^TVD^TU^TUDV^Tp = p^TVD^2V^Tp = p^Tegin{pmatrix} \lambda_1 & 0 & \cdots & 0 \ 0 & \lambda_2 & \cdots & 0 \ dots & dots & \ddots & dots \ 0 & 0 & \cdots & \lambda_n \end{pmatrix}p$$

 $p \ can \ be \ decomposed \ into \ \ p = \Sigma c_i v_i$

 $since \ ||p|| = 1, \quad following \ is \ true \ \Sigma c_i^2 = 1$

$$p^T egin{pmatrix} \lambda_1 & 0 & \cdots & 0 \ 0 & \lambda_2 & \cdots & 0 \ dots & dots & \ddots & dots \ 0 & 0 & \cdots & \lambda_n \end{pmatrix} p = \Sigma c_i^2 \lambda_i^2$$

$$min_p(||Ap||) = min_p(\Sigma c_i^2 \lambda_i^2)$$

In np. linalg.
$$svd(), \ \lambda_i > 0 \ and \ (\lambda_1 > \lambda_2 > \cdots > \lambda_n)$$

So, λ_n is smallest eigenvalue and p minimizing the expression ||Ap|| is as below

$$c_n=1, p=v_n$$

3.b) Given the following correspond points, determine the camera projection matric P using the SVD method.

1) Given (u,v) and (x,y,z)

```
In [14]:
         uv = np.array([[880,214],[43,203],[270,197],[886,347],[745,302],
         [943.128].
                          [476,590], [419,214], [317,335], [783,521], [235,427],
         [665,429],[655,362],[427,333],
                          [412,415], [746,351], [434,415], [525,234], [716,308],
         [602, 187]], dtype=np.float32)
         XYZ = np.array([[312.747,309.14,30.086],[305.796,311.649,30.356],
                              [307.694,312.358,30.418],[310.149,307.186,29.298],
         [311.937,310.105,29.216].
                              [311.202,307.572,30.682],[307.106,306.876,28.66],
                              [309.317,312.49,30.23], [307.435,310.151,29.318],
                              [308.253,306.3,28.881],[306.65,309.301,28.905],
                              [308.069.306.831.29.189].[309.671.308.834.29.029].
                              [308.255,309.955,29.267],[307.546,308.613,28.963],
                              [311.036,309.206,28.913],[307.518,308.175,29.069],
                              [309.95.311.262.29.99].[312.16.310.772.29.08].
         [311.988,312.709,30.514]], dtype=np.float32)
```

2) make A by Given equation below and Given (u,v) and (x,y,z)

```
\begin{split} & m_{11}X_i + m_{12}Y_i + m_{13}Z_i + m_{14} - m_{31}u_iX_i - m_{32}u_iY_i - m_{33}u_iZ_i - m_{34}u_i = 0 \\ & m_{21}X_i + m_{22}Y_i + m_{23}Z_i + m_{24} - m_{31}v_iX_i - m_{32}v_iY_i - m_{33}v_iZ_i - m_{34}v_i = 0 \end{split}
```

3) use SVD method to decompose A, then last column of V is p

```
u, s, vh = np.linalg.svd(A, full_matrices=True)

p = vh[-1]

p /= p[-1]
```

4) reshape vector p into matrix P

```
P = np.reshape(p, (3,4))
print(P)

[[-2.33261286e+00 -1.10020234e-01 3.37490343e-01 7.36686704e+02]
[-2.31041064e-01 -4.79515779e-01 2.08721751e+00 1.53626639e+02]
[-1.26376391e-03 -2.06774463e-03 5.14662103e-04 1.00000000e+00]]
```

3.c) In equation in b), set $m_34 = 1$, get equation Ap = b, then use pseudo inverse to get the least square solution p

1) make b, A with Given equation below p is size 11 (m_34 is 1 so will be added later)

```
m_{11}X_i + m_{12}Y_i + m_{13}Z_i + m_{14} - m_{31}u_iX_i - m_{32}u_iY_i - m_{33}u_iZ_i = u_i + m_{21}X_i + m_{22}Y_i + m_{23}Z_i + m_{24} - m_{31}v_iX_i - m_{32}v_iY_i - m_{33}v_iZ_i = v_i + m_{24}X_i + m_{24}X_i + m_{25}X_i + m_{2
```

```
b = np.zeros((2*len(uv),))

for i in range(len(uv)):
    u_i, v_i = uv[i]
    b[2*i] = u_i
    b[2*i+1] = v_i

A = np.zeros((2*len(uv),11))

for i in range(len(uv)):
```

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```
u_i, v_i = uv[i]
X_i, Y_i, Z_i = XYZ[i]
A[2*i] = np.array([X_i, Y_i, Z_i, 1, 0, 0, 0, 0, -u_i*X_i, -u_i*Y_i,
-u_i*Z_i])
A[2*i+1] = np.array([0, 0, 0, 0, X_i, Y_i, Z_i, 1, -v_i*X_i, -v_i*Y_i, -v_i*Z_i])
```

2) get pseudo inverse of A

```
A_T = np.transpose(A)
A_pseudo_inv = np.matmul(np.linalg.inv(np.matmul(A_T, A)), A_T)
```

3) use equation below to get p

```
\mathbf{p} = \mathbf{A}^{+}\mathbf{b}
```

4) reshape vector p into matrix P

```
(add m_{34} = 1 in matrix P)
```

```
P = np.resize(p, (3,4))

P[-1][-1] = 1

print(P)

[-2.33259421e+00 -1.09988242e-01  3.37390954e-01  7.36674050e+02]
```

```
[[-2.33259421e+00 -1.09988242e-01 3.37390954e-01 7.36674050e+02]
[-2.31047155e-01 -4.79506733e-01 2.08717179e+00 1.53627131e+02]
[-1.26378942e-03 -2.06771124e-03 5.14584938e-04 1.00000000e+00]]
```

Problem 3. Discussion

If we have world coordinates and projection coordinates, we can get projection matrix.