Homework 4

```
In [ ]: import numpy as np
   import matplotlib.pyplot as plt
   from sympy import *
```

Q1

If the linear equation have a non-zeros solution, then the determinant of the matrix is zero.

If λ =3 or 1 the determinant of the matrix is zeros.

Q2

No

```
In [ ]: A = np.array([[1,1],[-1,-1]])
B = np.array([[2,2],[-2,-2]])
print(A@A)
print(B@B)
[[0 0]
      [0 0]]
      [[0 0]
      [0 0]]
```

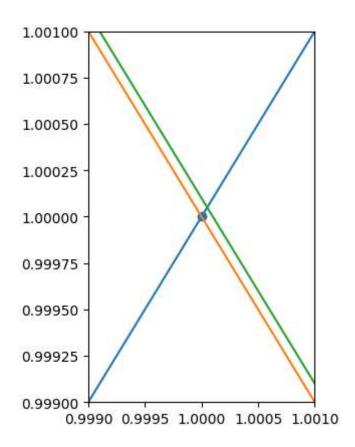
Q3

```
x = np.array([-1.6, 1.2])
print(sum(abs(x)))
print(x.T@x)
print(abs(np.max(x)))
2.8
4.0
1.2
  1.0
                                             1.0
  0.5
                                             0.5
  0.0
                                             0.0
                                            -0.5
 -0.5
-1.0
                                            -1.0
                                                             0
           -1
                      0
                                 1
                                                 -1
  1.0
  0.5
  0.0
 -0.5
-1.0 -
                      0
           -1
```

The matrix isn't ill conditioned

```
In []: x_ = np.linspace(-1,1,100)
    y1= x
    y2 = 2-x
    y3 = 2.0001-x
    fig = plt.figure()
    ax1 = plt.subplot(121)
    ax1.plot(x,y1)
    ax1.plot(x,y2)
    ax1.plot(x,y3)
    ax1.set_xlim(0.999,1.001)
    ax1.set_ylim(0.999,1.001)
    ax1.scatter(1,1)
    A = np.array([[1,-1],[1,1]])
    cond = np.linalg.norm(A,ord=2)
    print('Cond(A)={}'.format(cond))
```

Cond(A)=1.4142135623730954



```
In [ ]: def PLU_Dec(M):
            n = np.size(M,0)
            L = np.eye(n)
            U = np.copy(M)
            p = np.arange(n)
            for i in range(n-1):
                largest_pivot = np.argmax(U[i:n,i])+i
                c = p[i]
                p[i]=p[largest_pivot]
                p[largest_pivot]=c
                C = np.copy(U[i,:])
                U[i,:] = U[largest_pivot,:]
                U[largest_pivot,:] = C
                for j in range(i+1,n):
                    mult = U[j,i]/U[i,i]
                    L[j,i] = mult
                    U[j,:] = U[j,:]-mult*U[i,:]
            P = np.zeros([n,n])
            for i in range(n):
                P[i,p[i]] = 1
            return P,L,U
        def LU_Dec(M):
            n = np.size(M,0)
            L = np.eye(n)
            U = np.copy(M)
            for i in range(n-1):
                for j in range(i+1,n):
                    mult = U[j,i]/U[i,i]
```

```
L[j,i] = mult
            U[j,:] = U[j,:]-mult*U[i,:]
    return L,U
A = np.array([[1,2,3],
              [2,4,5],
              [3,5,6]],dtype='float64')
P,L,U = PLU_Dec(A)
print('P = {}\nL = {}\nU = {}'.format(P,L,U))
A = np.array([[2,1,1,2],
              [2,2,2,3],
              [4,2,4,3],
              [0,0,6,-1]],dtype='float64')
P,L,U = PLU Dec(A)
print('P = {}\nU = {}'.format(P,L,U))
P = [[0. 0. 1.]]
[0. 1. 0.]
 [1. 0. 0.]]
L = [[1.
                                       ]
                 0.
                            0.
 [0.66666667 1.
                        0.
                                   ]
 [0.33333333 0.5
                                   ]]
               5.
                                       ]
U = [[3.
                            6.
             0.66666667 1.
 [0.
                        0.5
                                   ]]
 [0.
             0.
P = [[0. 0. 1. 0.]]
 [0. 1. 0. 0.]
 [0. 0. 0. 1.]
 [1. 0. 0. 0.]]
L = [[ 1.
                                                      ]
                               0.
                                            0.
                                                  ]
 [ 0.5
                           0.
               1.
                                        0.
 [ 0.5
               0.
                           1.
                                        0.
                                                  ]
 [ 0.
                           -0.16666667 1.
                                                  ]]
U = [[ 4.
                   2.
                               4.
                                                      1
                                            3.
 [ 0.
               1.
                           0.
                                        1.5
 [ 0.
               0.
                                       -1.
                           6.
 [ 0.
               0.
                           0.
                                       0.33333333]]
```

```
In [ ]: def LU_Solve(L,U,x,P=None):
    ## PTPAx = Ax = PTb
    if P is not None:
        B = P.T@np.copy(b)
    else:
        B = np.copy(b)
    n = np.size(L,0)
    Y = np.zeros([n,1])
    Y[0] = B[0]
    for i in range(1,n):
        Y[i] = B[i]-L[i,0:i]@Y[0:i]

    X = np.zeros([n,1])
    Y_ = np.copy(Y)
    X[n-1] = Y_[n-1]/U[n-1,n-1]
    for i in range(n-2,-1,-1):
```

```
X[i] = (Y_{i}-U[i,i+1:n]@X[i+1:n])/U[i,i]
    return Y, X
A = np.array([[10,-7,0,1],
              [-3,2.099999,6,2],
              [5 ,-1 ,5,-1],
[2 ,1 ,0,2]],dtype='float64')
b = np.array([ 8 ,5.900001 ,5,1] ,dtype='float64').T
P,L,U = PLU Dec(A)
Y,X = LU_Solve(L,U,b,P)
print('Y={}\nX={}\nPLUX={}'.format(Y,X,P@L@U@X))
L,U = LU_Dec(A)
Y,X = LU Solve(L,U,b)
print('Y={}\nX={}\nLUX={}'.format(Y,X,L@U@X))
Y=[[ 8.
[ 7.4
 [ 1.90000396]
 [-6.18399734]]
X=[[ 1.38582661]
[ 0.66299193]
 [ 0.78330726]
 [-1.21732257]]
PLUX=[[8.
 [5.900001]
 [5.
          ]
 [1.
          ]]
Y=[[8.0000000e+00]
 [8.30000100e+00]
 [2.07500035e+07]
 [5.07999891e+00]]
X = [[-6.03391825e-10]
 [-1.00000000e+00]
 [ 1.00000000e+00]
 [ 1.0000000e+00]]
LUX=[[8.
              1
 [5.900001]
 [5.
          ]
          11
 [1.
```

The aim of the solution is to find a Ax = b, in which the max norm of x converges with the convergence criterion.

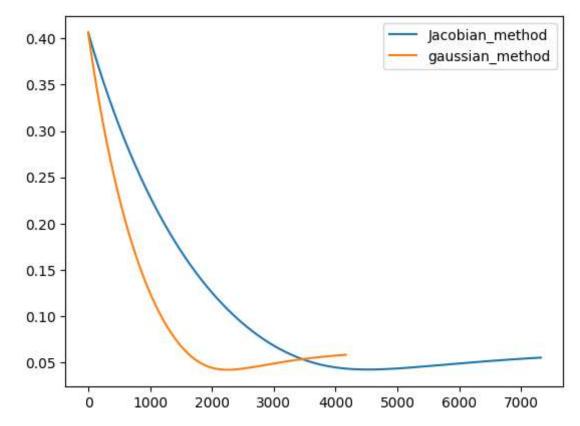
```
In []: epsillon = 1
a = 0.5
n = 100
def get_Matrix(epsillon,a,n):
    h = 1.0/n
    D = np.eye(n)*(-2*epsillon-h)
    L = np.eye(n)*(epsillon)
    L[1:n,:] = L[0:n-1,:]
    L[0,:] = np.zeros(n)
```

```
U = np.eye(n)*(epsillon+h)
                U[:,1:n] = U[:,0:n-1]
                U[:,0] = np.zeros(n)
                A = D+L+U
                B = np.ones(n)*a*h**2
                B[n-1]=(epsillon+h)
                P = np.linspace(0,1,n)
                return A, B, P
A,B,P = get_Matrix(epsillon,a,n)
def Jacobian Iter(A,B,P,delta,max iter=10000):
                n = np.size(A,0)
                X = np.zeros(n)
                X_list = np.zeros([max_iter,n])
                iter = 0
                for i in range(max_iter):
                                for j in range(n):
                                                 c = A[j,0:j]@P[0:j].reshape([j,1])+A[j,j+1:n]@P[j+1:n].reshape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+A[j,j+1:n].perchape([n-j-1])+
                                                X[j] = (B[j]-c)/A[j,j]
                                 err = np.linalg.norm(X-P,ord=np.inf)
                                 relative_err = err/(np.linalg.norm(X,ord=np.inf)+1e-10)
                                 P = np.copy(X)
                                X_{list[i,:]} = np.copy(X)
                                 if err<delta:</pre>
                                                iter = i
                                                 break
                X_list = X_list[0:iter_,:]
                return X, X list
jacobian_Y,jacobian_Y_list = Jacobian_Iter(A,B,P,delta=1e-6,max_iter=10000)
def Gaussian_Iter(A,B,P,delta=1e-6,max_iter=10000):
                n = np.size(A,0)
                X = np.zeros(n)
                X list = np.zeros([max iter,n])
                iter = 0
                for i in range(max_iter):
                                new_P = np.copy(P)
                                for j in range(n):
                                                 c = A[j,0:j]@new_P[0:j].reshape([j,1])+A[j,j+1:n]@new_P[j+1:n].reshape([j,1])+A[j,j+1:n]@new_P[j+1:n].reshape([j,1])+A[j,j+1:n]@new_P[j+1:n].reshape([j,1])+A[j,j+1:n]@new_P[j+1:n].reshape([j,1])+A[j,j+1:n]@new_P[j+1:n].reshape([j,1])+A[j,j+1:n]@new_P[j+1:n].reshape([j,1])+A[j,j+1:n]@new_P[j+1:n].reshape([j,1])+A[j,j+1:n]@new_P[j+1:n].reshape([j,1])+A[j,j+1:n]@new_P[j+1:n].reshape([j,1])+A[j,j+1:n]@new_P[j+1:n].reshape([j,1])+A[j,j+1:n]@new_P[j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].reshape([j,1])+A[j,j+1:n].resha
                                                X[j] = (B[j]-c)/A[j,j]
                                                 new P[j] = X[j]
                                 err = np.linalg.norm(X-P,ord=np.inf)
                                 relative_err = err/(np.linalg.norm(X,ord=np.inf)+1e-10)
                                 P = np \cdot copy(X)
                                X_{list[i,:]} = np.copy(X)
                                 if err<delta:</pre>
                                                iter = i
                                                 break
                X_list = X_list[0:iter_,:]
                return X, X_list
gaussian_Y,gaussian_Y_list = Gaussian_Iter(A,B,P,delta=1e-6,max_iter=10000)
```

```
In []: fig = plt.figure()
    ax = fig.subplots()
    x = np.arange(0,100)*1/n
    analystic_Y= (1-a)/(1-np.exp(-1/epsillon))*(1-np.exp(-x/epsillon))+a*x
    max_iter = np.size(jacobian_Y_list,0)
    loss_y = np.zeros(max_iter)
    for i in range(max_iter):
```

```
loss_y[i] = np.linalg.norm(analystic_Y-jacobian_Y_list[i,:],ord=2)
idx = np.arange(max_iter)
ax.plot(idx,loss_y,label='Jacobian_method')
max_iter = np.size(gaussian_Y_list,0)
loss_y = np.zeros(max_iter)
for i in range(max_iter):
    loss_y[i] = np.linalg.norm(analystic_Y-gaussian_Y_list[i,:],ord=2)
idx = np.arange(max_iter)
ax.plot(idx,loss_y,label='gaussian_method')
ax.legend()
```

Out[]: <matplotlib.legend.Legend at 0x2ddc9d602e0>



```
In []: fig = plt.figure()
## epsillon=0.1
ax = plt.subplot(3,1,1)
x = np.arange(0,100)*1/n
epsillon = 0.1
analystic_Y= (1-a)/(1-np.exp(-1/epsillon))*(1-np.exp(-x/epsillon))+a*x
ax.plot(x,analystic_Y)
ax.plot(x,gaussian_Y1,'--')
## epsillon = 0.01
```

```
ax = plt.subplot(3,1,2)
epsillon = 0.01
analystic_Y= (1-a)/(1-np.exp(-1/epsillon))*(1-np.exp(-x/epsillon))+a*x
ax.plot(x,analystic_Y)
ax.plot(x,gaussian_Y2,'--')
## epsillon = 0.001
ax = plt.subplot(3,1,3)
epsillon = 0.001
analystic_Y= (1-a)/(1-np.exp(-1/epsillon))*(1-np.exp(-x/epsillon))+a*x
ax.plot(x,analystic_Y)
ax.plot(x,gaussian_Y3,'--')
```

Out[]: [<matplotlib.lines.Line2D at 0x2ddc7f62700>]

