



Lab 03. Solve linear system and matrix decompositions

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2IA

1 Code

1.1 The elimination of Gauss-Jordan to Solve a system S of linear

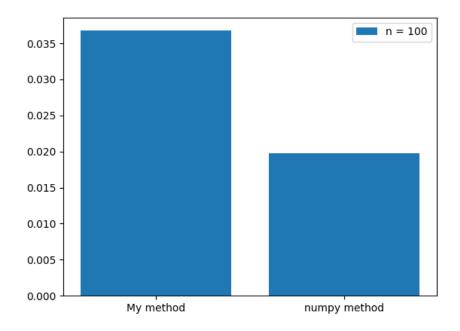
Use the elimination of Gauss-Jordan to Solve a system S of linear equations A x = b defined by the matrix A and the vector b.

 ${\rm code\ file: Solve_a_system_by_the_elimination_of_Gauss.py}$

```
import numpy as np
   from time import time
   import matplotlib.pyplot as plt
   np.random.seed(42)
   def for_Elimination(A,b):
       n = A.shape[0]
6
       arr = np.zeros((n, n+1))
       arr[:,:n] = A.copy()
       arr[:,n] = b.copy()
       for i in range(n-1):
10
            p_ind = i + np.argmax(np.abs(arr[i:,i]))
11
            arr[i,i:], arr[p_ind,i:] = arr[p_ind,i:].copy(),arr[i,i:].copy
           pivot = arr[i,i]
13
           if pivot == 0 : print("le systeme n admet pas de solution
               \hookrightarrow unique ")
            for k in range(i+1, n):
15
                m = arr[k,i]/pivot
16
                arr[k,i:] -= m*arr[i, i:]
17
       return arr[:,:n], arr[:,n]
18
19
   def back_Substitution(A,b):
20
       n = A.shape[0]
21
       X = np.empty(n)
22
       X[-1] = b[-1]/A[-1,-1]
23
       for i in range(-2, -(n+1), -1):
24
            X[i] = (b[i] - np.dot(A[i,i+1:], X[i+1:]))/A[i,i]
25
       return X
27
28
   def Solve(A,b):
29
30
        Solve a system S of linear equations A x = b defined \setminus
31
            by the matrix A and the vector b
32
       a, bb = for_Elimination(A,b)
34
```

```
return back_Substitution(a, bb)
35
   if __name__ == "__main__":
36
       n = 100
37
       A = np.random.randn(n,n)
38
39
       b = np.random.randn(n)
40
41
       names = ["My method", "numpy method"]
42
43
44
       t1 =time(); X = Solve(A,b);t.append(time()-t1)
45
       t1 =time(); x = np.linalg.solve(A,b);t.append(time()-t1)
46
47
       plt.bar(names,t, label = "n = "+str(n))
48
       plt.legend()
49
       plt.show()
50
51
       print(list(zip(names,t)))
52
       print(f"the percentage of X == x is : {np.sum(np.around(X,9) == np.}
53
           \hookrightarrow around(x,9))*100/n} % ")
```

1.1.1 Screenshots



```
[Running] python -u "c:\Users\lenovo\Desktop\ENSI@S CODING\PYTHON code\Numerical Analysis and Optimization\LAB3\Solve_a_system_by_the_elimination_of_Gauss.py"
[('My method', 0.03676295280456543), ('numpy method', 0.019762516021728516)]
the percentage of X == x is : 100.0 %

[Done] exited with code=0 in 27.923 seconds
```

1.2 Use the elimination of Gauss-Jordan to find the inverse of a matrix A

code file: Find_the_inverse_of_a_matrix_by_elimination_of_Gauss.py

```
from Solve_a_system_by_the_elimination_of_Gauss import Solve
  import numpy as np
  def Inverse(A):
3
       n = A.shape[0]
       inv_A = np.empty((n,n))
       b = np.array([0]*n)
       for i in range(n):
           b[i] = 1
           inv_A[:,i] = Solve(A,b)
           b[i] = 0
10
11
       return inv_A
12
   if __name__ == "__main__" :
13
       A = np.array([[25,5,1]],
14
                       [64,8,1],
15
                       [144,12,1]])
16
17
       invA = Inverse(A)
18
       print("A@invA = \n", np.around(A@invA))
19
```

1.2.1 Screenshots

```
[Running] python -u "c:\Users\lenovo\Desktop\ENSI@S CODING\PYTHON code\Numerical
Analysis and Optimization\LAB3\Find_the_inverse_of_a_matrix_by_elimination_of_Gauss.py"
A@invA =
  [[ 1. -0. 0.]
  [ 0. 1. 0.]
  [-0. -0. 1.]]

[Done] exited with code=0 in 2.391 seconds
```

1.3 Python implementation for the LU decomposition.

code file: Implementation_for_the_LU_decomposition.py

```
import numpy as np
  import time as t
  import scipy.linalg
  np.random.seed(42)
  # Define the matrix
  n = 50
6
  A = np.random.randn(n,n)
  # -----| implementation for the LU decomposition|-----
  def LU(A, ameliorer = False) :
      n = len(A)
10
      U = A.copy()
11
      L = np.eye(n)
12
       for i in range(n-1):
13
14
           pivot = U[i,i]
15
           if pivot == 0 : print("pivot = 0") ; return L,U
16
17
           for k in range(i+1, n):
18
               L[k,i] = U[k,i]/pivot
19
```

```
U[k] = U[k] - L[k,i]*U[i]
20
        return L,U
21
22
   def PLU(A):
23
       n = A.shape[0]
24
       U = A.copy()
25
       L = np.eye(n, dtype= np.double)
26
       P = np.eye(n, dtype= np.double)
27
28
       for i in range(n):
            for k in range(i,n):
30
                 if ~np.isclose(U[i,i], 0) :
31
                     break
32
                 U[[k,k+1]] = U[[k+1, k]]
                 P[[k,k+1]] = P[[k+1, k]]
34
35
            for k in range(i+1, n):
36
                 L[k,i] = U[k,i]/U[i,i]
37
                 U[k] = U[k] - L[k,i]*U[i]
38
        return P,L,U
39
40
41
42
   if __name__ == "__main__":
43
       # -----
44
       t1 = t.time(); L,U = LU(A); t1 = t.time() - t1
45
        t2 = t.time(); p, l, u = scipy.linalg.lu(A); t2 = t.time() - t2
46
        t3 = t.time(); p, pL, pU = PLU(A); t3 = t.time() - t3
47
       # -----
       print("my LU time : ",np.around(t1,10))
print("my PLU time : ",np.around(t3,10))
49
50
51
        print("numpy lu time : ",np.around(t2,10))
52
53
        print(f"A is a matrix of shape {n}x{n}")
54
        print(f"L == 1 : {np.sum(np.around(1) == np.around(L))*100/len(1)}
55

    **2}%")

        print(f"U == u : {np.sum(np.around(u) == np.around(U))*100/len(1)
56
           \hookrightarrow **2}%")
        print(f"A == L*U : {np.sum(np.around(A) == np.around(L@U))*100/len(
57
           \hookrightarrow 1)**2}%")
        print(f"A == 1*u : {np.sum(np.around(A) == np.around(1@u))*100/len(
58
           \hookrightarrow 1)**2}%")
        print(f"L@U == 1*u : {np.sum(np.around(L@U) == np.around(1@u))*100/
59
           \hookrightarrow len(1)**2}%")
60
        print(f"pL == 1 : {np.sum(np.around(1) == np.around(pL))*100/len(1)}
61
           \hookrightarrow **2}%")
        print(f"pU == u : {np.sum(np.around(u) == np.around(pU))*100/len(1)}
           \hookrightarrow **2}%")
        print(f"A == pL*pU : {np.sum(np.around(A) == np.around(pL@pU))*100/
63
           \hookrightarrow len(1)**2}%")
        print(f"pL@pU == 1*u : {np.sum(np.around(pL@pU) == np.around(1@u))
64
           \hookrightarrow *100/len(1)**2}%")
```

```
[Running] python -u "c:\Users\lenovo\Desktop\ENSI@S CODING\PYTHON code\Numerical
Analysis and Optimization\LAB3\Implementation_for_the_LU_decomposition.py"
my LU time : 0.0071957111
my PLU time : 0.0099756718
numpy lu time : 0.0112290382
A is a matrix of shape 50x50
L == 1 : 63.96%
U == u : 53.12%
A == L*U : 100.0%
A == 1*u : 27.84%
L@U == 1*u : 27.84%
pL == 1 : 63.96%
pU == u : 53.12%
A == pL*pU : 100.0%
pL@pU == 1*u : 27.84\%
[Done] exited with code=0 in 2.184 seconds
```

1.3.1 Screenshots

1.4 Use the LU decompositions to solve the system A x = b

code file: Solve_a_system_by_LU_decomposition.py

```
from Implementation_for_the_LU_decomposition import LU, PLU
   import numpy as np
2
   def Solve_Lz_b(L,b):
       n = L.shape[0]
       Z = np.empty(n)
6
       Z[0] = b[0]/L[0,0]
       for i in range(1,n):
           Z[i] = (b[i] - np.dot(L[i,:i], Z[:i]))/L[i,i]
       return Z
10
   def Solve_Ux_z(U,z):
12
       n = U.shape[0]
13
       X = np.empty(n)
14
       X[-1] = z[-1]/U[-1,-1]
       for i in range(-2, -(n+1), -1):
16
           X[i] = (z[i] - np.dot(U[i,i+1:], X[i+1:]))/U[i,i]
17
       return X
18
   def Solve(L,U,b):
20
       z = Solve_Lz_b(L,b)
21
       x = Solve_Ux_z(U,z)
22
       return x
23
   if __name__ == "__main__":
24
       A = np.array([[2,4,-2],
25
                       [4,9,-3],
26
                       [-2, -3, 7]])
27
       b = np.array([2,8,10])
28
29
       L,U = LU(A)
       Xlu = Solve(L,U,b)
31
       print("x by LU = ",Xlu)
32
33
       P,L,U = PLU(A)
34
       Xlu = Solve(L,U,np.dot(P,b))
35
       print("x by PLU = ",Xlu)
36
```

1.4.1 Screenshots

```
[Running] python -u "c:\Users\lenovo\Desktop\ENSI@S CODING\PYTHON code\Numerical
Analysis and Optimization\LAB3\Solve_a_system_by_LU_decomposition.py"
x by LU = [-1. 2. 2.]
x by PLU = [-1. 2. 2.]
[Done] exited with code=0 in 0.727 seconds
```

1.5 Python implementation for the Choleski's decomposition

code file: Choleski_decomposition.py

```
import numpy as np
2
  def Choleski_decomposition(A):
       L = np.zeros(A.shape, dtype= np.double)
       n = A.shape[0]
5
       for k in range(n):
6
           L[k,k] = np.sqrt(A[k,k] - np.dot(L[k,:k], L[k,:k]))
           for i in range(k+1, n):
9
               L[i,k] = (A[i,k] - np.dot(L[i,:k], L[:k,k]))/L[k,k]
10
11
       return L
12
  if __name__ == "__main__":
13
14
       A = np.array([[2, -1, 0], [-1, 2, -1], [0, -1, 2]], dtype= np.double)
15
16
17
       L = Choleski_decomposition(A)
18
       l = np.linalg.cholesky(A)
       print("Choleski_decomposition L = \n", L)
20
       print("numpy choleski l = \n",1)
```

1.5.1 Screenshots

```
[Running] python -u "c:\Users\lenovo\Desktop\ENSI@S CODING\PYTHON code\Numerical
Analysis and Optimization\LAB3\Choleski_decomposition.py"
Choleski_decomposition L =
[[ 1.41421356 0. 0. ]
[-0.70710678 1.22474487 0. ]
[ 0. -0.81649658 1.15470054]]
numpy choleski l =
[[ 1.41421356 0. 0. ]
[-0.70710678 1.22474487 0. ]
[ 0. -0.81649658 1.15470054]]

[Done] exited with code=0 in 0.489 seconds
```

1.6 Use the LU decomposition to find the inverse matrix of A

```
from Solve_a_system_by_LU_decomposition import Solve_Lz_b as for_sub from Solve_a_system_by_LU_decomposition import Solve_Ux_z as back_sub from Implementation_for_the_LU_decomposition import LU, PLU
```

```
import numpy as np
5
   def LU_Inverse(L,U, P = None):
       n = L.shape[0]
       Inv = np.empty((n,n), dtype=np.double)
9
       x = np.empty(n, dtype=np.double)
10
       b = np.zeros(n, dtype=np.double)
11
       for i in range(n):
12
           b[i] = 1
13
           \# Solve the equation L * X = b for X using forward substitution
           if P is not None:
15
                x = for_sub(L, np.dot(P,b))
16
           else :
17
                x = for_sub(L, b)
           # Solve the equation L^T * Y = X for Y using back substitution
19
           x = back_sub(U, x)
20
           # Set the i-th column of the inverse matrix
21
           Inv[:,i] = x
22
           b[i] = 0
23
       return Inv
24
25
   if __name__ == "__main__":
26
       A = np.array([[2, -1, 0], [-1, 2, -1], [0, -1, 2]], dtype= np.double)
27
28
       L, U = LU(A)
29
30
       LU_InvA = LU_Inverse(L,U)
31
       P,L,U = PLU(A)
32
       PLU_InvA = LU_Inverse(L,U, P)
33
       numpy_invA = np.linalg.inv(A)
34
       print(f"Inverse of A by LU decomposition is \n {LU_InvA}")
35
       print(f"Inverse of A by PLU decomposition is \n {PLU_InvA}")
36
       print(f"Inverse of A by numpy is \n {numpy_invA}")
```

1.6.1 Screenshots

```
[Running] python -u "c:\Users\lenovo\Desktop\ENSI@S CODING\PYTHON code\Numerical Analysis and Optimization\LAB3\LU_Inverse.py"

Inverse of A by LU decomposition is

[[0.75 0.5 0.25]
[0.5 1. 0.5 ]
[0.25 0.5 0.75]]

Inverse of A by PLU decomposition is

[[0.75 0.5 0.25]
[0.5 1. 0.5 ]
[0.25 0.5 0.75]]

Inverse of A by numpy is

[[0.75 0.5 0.25]
[0.5 1. 0.5 ]
[0.75 0.5 0.25]
[0.5 1. 0.5 ]
[0.25 0.5 0.75]]

[Done] exited with code=0 in 0.712 seconds
```

1.7 Use the Choleski's decomposition to find the inverse matrix of A

code file : Inverse_by_Choleski_decomposition.py

```
from Choleski_decomposition import Choleski_decomposition
  from Solve_a_system_by_LU_decomposition import Solve_Lz_b as for_sub
  from Solve_a_system_by_LU_decomposition import Solve_Ux_z as back_sub
   import numpy as np
   def Inverse_by_choleski(L):
6
       n = L.shape[0]
       Inv = np.empty((n,n), dtype=np.double)
       x = np.empty(n, dtype=np.double)
9
       b = np.zeros(n, dtype=np.double)
10
       for i in range(n):
11
           b[i] = 1
12
           # Solve the equation L * X = b for X using forward substitution
13
           x = for_sub(L, b)
           # Solve the equation L^T * Y = X for Y using back substitution
15
           x = back_sub(L.T, x)
16
           # Set the i-th column of the inverse matrix
17
           Inv[:,i] = x
           b[i] = 0
19
       return Inv
20
21
       return Inv
22
   if __name__ == "__main__":
23
       A = np.array([[2, -1, 0], [-1, 2, -1], [0, -1, 2]], dtype= np.double)
24
25
       L = Choleski_decomposition(A)
26
       Choleski_InvA = Inverse_by_choleski(L)
27
       numpy_invA = np.linalg.inv(A)
28
       print(f"Inverse of A by choleski decomposition is \n {Choleski_InvA
29
          \hookrightarrow }")
       print(f"Inverse of A by numpy is \n {numpy_invA}")
```

1.7.1 Screenshots

```
[Running] python -u "c:\Users\lenovo\Desktop\ENSI@S CODING\PYTHON code\Numerical Analysis and Optimization\LAB3\Inverse_by_Choleski_decomposition.py"

Inverse of A by choleski decomposition is

[[0.75 0.5 0.25]
[0.5 1. 0.5 ]
[0.25 0.5 0.75]]

Inverse of A by numpy is

[[0.75 0.5 0.25]
[0.5 1. 0.5 ]
[0.25 0.5 0.75]]

[Done] exited with code=0 in 0.73 seconds
```

1.8 Comparing these solving methods

code file : comparing_these_solving_methods.py

```
from Find_the_inverse_of_a_matrix_by_elimination_of_Gauss import

Inverse as Gauss_Inv
from Choleski_decomposition import Choleski_decomposition
from Inverse_by_Choleski_decomposition import Inverse_by_choleski as

Choleski_Inv
```

```
from Solve_a_system_by_LU_decomposition import Solve as LU_solve
   from Solve_a_system_by_the_elimination_of_Gauss import Solve as
      → Gauss_solve
  from Implementation_for_the_LU_decomposition import LU, PLU
  from LU_Inverse import LU_Inverse as LU_Inv
  from time import time
   import matplotlib.pyplot as plt
   import numpy as np
10
  np.random.seed = 42
11
  n = 10
  names = ["LU_inv", "PLU_inv", "numpy_inv", "Gauss_inv",
13
           "Choleski_inv", "LU_solve",
14
           "PLU_solve", "Gauss_solve", "numpy_solve"]
15
  times = []
16
17
   \# A = np.array([[2, -1, 0], [-1, 2, -1], [0, -1, 2]], dtype= np.double)
18
   A = np.random.randn(n,n)
19
  b = np.random.randn(n)
  # -----
21
  t = time()
22
  l,u = LU(A)
23
  LU_Inv(1,u)
24
   times.append(time() - t)
25
   # -----
26
  t = time()
27
  p,l,u = PLU(A)
  LU_Inv(1,u,p)
  times.append(time()-t)
30
31
  t = time()
32
  np.linalg.inv(A)
33
  times.append(time() - t)
34
  t = time()
36
  Gauss_Inv(A)
37
  times.append(time() - t)
38
   # -----
   t = time()
40
  1 = Choleski_decomposition(A)
41
   Choleski_Inv(1)
42
  times.append(time() - t)
  # ----
44
  t = time()
45
  1,u = LU(A)
46
  LU_solve(1,u,b)
   times.append(time() - t)
48
   # ----
49
  t = time()
  p,l,u = PLU(A)
  LU_solve(1,u, np.dot(p,b))
52
  times.append(time() - t)
53
54
  t = time()
55
  Gauss_solve(A,b)
56
  times.append(time() - t)
57
  t = time()
60 np.linalg.solve(A,b)
```

```
times.append(time() - t)

# ------

plt.bar(names[:-4], times[:-4], alpha = 0.6 )

plt.bar(names[-4:], times[-4:], alpha = 0.6 )

plt.title("runing time of these method")

plt.xlabel("methods names")

plt.ylabel("time (s)")

plt.show()
```

1.8.1 Screenshots

