

```
In [1]: 1 import numpy as np
        2 import cmath
        3 import math
        4 import matplotlib.pyplot as plt
        5 import csv
        6 import random
        7 random.seed(123)
```

Lab Book 1

```
In [2]: 1 z= 1+1j
        2
        3 print(z.real)
        4 print(z.imag)
```

```
1.0
1.0
```

Lab Book 2

```
In [3]: 1 print(cmath.exp(1j * math.pi))
        2
        3 # check ten random numbers and return the result
        4 for i in range(0,10):
        5     r = random.random()
        6     theta = random.random()
        7     left = r * cmath.exp(1j * theta)
        8     right = r * (math.cos(theta) + 1j * cmath.sin(theta))
        9     print((left - right) == 0)
```

```
(-1+1.2246467991473532e-16j)
True
True
True
True
True
True
True
True
True
True
True
```

Lab Book 3

In [4]:

```

1 a = np.linspace(1,10,10)
2 print(a)
3 a[1:len(a)-1] = 0
4 print(a)

```

```

[ 1.  2.  3.  4.  5.  6.  7.  8.  9. 10.]
[ 1.  0.  0.  0.  0.  0.  0.  0.  0. 10.]

```

Lab Book 4

Numpy will not be able to deal with such computation as it is meaningless and undefined in mathematics and python will return error when executing it. This is not a sensible choice because the dot product between two matrices $A(nm)$ and $B(ij)$ $A @ B$ can only be operated when $m=i$.

Lab Book 5

In [5]:

```

1 def decompositionVector(x, y):
2     v1 = y.dot(x) / np.linalg.norm(y) ** 2 * y
3     v2 = x - v1
4     return v1, v2
5 x = np.array([2.0, -3.0])
6 y = np.array([1.0, 1.0])
7 v1,v2 = decompositionVector(x,y)
8 print(v1,v2)

```

```

[-0.5 -0.5] [ 2.5 -2.5]

```

Lab Book 6

```
In [6]: 1 c = np.linspace(1,100,100)
        2 c = c.reshape(10,10)
        3 csquare = c[0:2,0:2]
        4 print(c)
        5 print(np.linalg.det(csquare))
```

```
[[ 1.  2.  3.  4.  5.  6.  7.  8.  9. 10.]
 [11. 12. 13. 14. 15. 16. 17. 18. 19. 20.]
 [21. 22. 23. 24. 25. 26. 27. 28. 29. 30.]
 [31. 32. 33. 34. 35. 36. 37. 38. 39. 40.]
 [41. 42. 43. 44. 45. 46. 47. 48. 49. 50.]
 [51. 52. 53. 54. 55. 56. 57. 58. 59. 60.]
 [61. 62. 63. 64. 65. 66. 67. 68. 69. 70.]
 [71. 72. 73. 74. 75. 76. 77. 78. 79. 80.]
 [81. 82. 83. 84. 85. 86. 87. 88. 89. 90.]
 [91. 92. 93. 94. 95. 96. 97. 98. 99. 100.]]
-10.000000000000002
```

Lab Book 7a

```
In [7]: 1 def trace(A):
        2     temp = np.ones(A.shape[0])
        3     diagMatrix = np.diag(temp)
        4     return np.sum(diagMatrix * A)
        5 print(trace(csquare))
```

```
13.0
```

Lab Book 7b

```
In [8]: 1 A = c[0:4,0:5]
        2 print(A)
        3 left = trace(A.T @ A)
        4 right = np.sum(A ** 2)
        5 print(left - right)
```

```
[[ 1.  2.  3.  4.  5.]
 [11. 12. 13. 14. 15.]
 [21. 22. 23. 24. 25.]
 [31. 32. 33. 34. 35.]]
0.0
```

Lab Book 8

Lab Book 7c

$$A \in \mathbb{R}^{m,n} \quad m \left\{ \underbrace{\quad}_n \right\} \quad n \left\{ \underbrace{\quad}_m \right\} A^T$$

$$A^T \cdot A = \begin{bmatrix} A_{11} & \dots & A_{m1} \\ \vdots & A_{22} & \vdots \\ A_{1n} & \dots & A_{mn} \end{bmatrix} \cdot \begin{bmatrix} A_{11} & \dots & A_{1n} \\ \vdots & A_{22} & \vdots \\ A_{m1} & \dots & A_{mn} \end{bmatrix}$$

$$\text{identity trace } (A^T A) = \sum_i^m A_{i1}^2 + \sum_i^m A_{i2}^2 + \dots + \sum_i^m A_{in}^2 \\ = \sum_{i=1}^m \sum_{j=1}^n A_{ij}^2$$

```
In [9]: 1 A1 = np.array([[2.0,3.0],[3.0,4.5]])
2 b1 = np.array([3.5, 6.5])
3 x1 = np.linalg.solve(A1,b1)
4 print(x1)
5 A2 = np.array([[1.0,1.0],[1.0,1.0]])
6 b2 = np.array([3.5, 6.5])
7 x2 = np.linalg.solve(A2,b2)
8 print(x2)
```

```
[ 7.50599938e+15 -5.00399959e+15]
```

```
-----
LinAlgError
```

```
Traceback (most recent c
```

```
all last)
```

```
<ipython-input-9-c1e2b0f8d845> in <module>
```

```
5 A2 = np.array([[1.0,1.0],[1.0,1.0]])
```

```
6 b2 = np.array([3.5, 6.5])
```

```
----> 7 x2 = np.linalg.solve(A2,b2)
```

```
8 print(x2)
```

```
<__array_function__ internals> in solve(*args, **kwargs)
```

```
~/opt/anaconda3/lib/python3.7/site-packages/numpy/linalg/linalg.py
```

```
in solve(a, b)
```

```
397 signature = 'DD->D' if isComplexType(t) else 'dd->d'
```

```
398 extobj = get_linalg_error_extobj(_raise_linalgerror_si
ngular)
```

```
--> 399 r = gufunc(a, b, signature=signature, extobj=extobj)
```

```
400
```

```
401 return wrap(r.astype(result_t, copy=False))
```

```
~/opt/anaconda3/lib/python3.7/site-packages/numpy/linalg/linalg.py
```

```
in _raise_linalgerror_singular(err, flag)
```

```
95
```

```
96 def _raise_linalgerror_singular(err, flag):
```

```
----> 97     raise LinAlgError("Singular matrix")
```

```
98
```

```
99 def _raise_linalgerror_nonposdef(err, flag):
```

```
LinAlgError: Singular matrix
```

For some certain singular matrix python will return error of singular matrix and for most cases python will return an extremely small value which also means the given matrix is singular.

Lab Book 9

```

In [10]: 1 def geneHilbert(size):
          2     A = np.zeros([size, size])
          3     for i in range(size):
          4         for j in range(size):
          5             A[i][j] = 1 / (i + j + 1)
          6     return A
          7 A = geneHilbert(15)
          8 b = A[:,0]
          9 x1 = np.linalg.solve(A,b)
         10 x2 = np.linalg.inv(A) @ b
         11 print(x1)
         12 print(x2)
         13 print(x1 - x2)

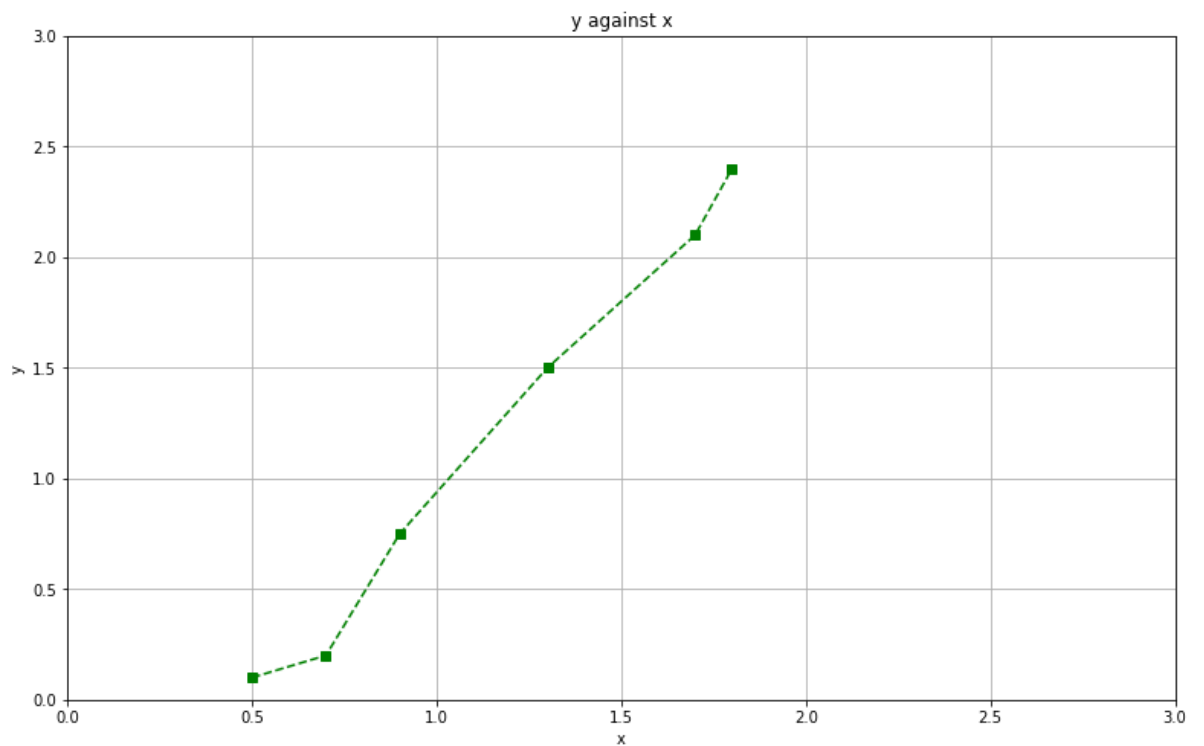
[ 1.  0.  0.  0.  0.  0. -0.  0. -0.  0. -0. -0.  0. -0.  0.]
[-2.16378957e-01  1.47670870e+01 -5.12857056e+01  4.34057617e+01
 1.24613281e+02 -3.44265625e+02  3.32843750e+02 -1.07750000e+02
-5.71875000e+01  6.15000000e+01 -1.80000000e+01  1.00000000e+00
 0.00000000e+00  0.00000000e+00 -3.12500000e-02]
[ 1.21637896e+00 -1.47670870e+01  5.12857056e+01 -4.34057617e+01
-1.24613281e+02  3.44265625e+02 -3.32843750e+02  1.07750000e+02
 5.71875000e+01 -6.15000000e+01  1.80000000e+01 -1.00000000e+00
 0.00000000e+00 -0.00000000e+00  3.12500000e-02]

```

The np.linalg.solve will return an exact value while by calculating $A^{-1}b$ will return float point numbers and those numbers have bigger error with the exact value as the size grow up. This is because when calculating the inverse of A, as the entries in A is too small, there will be error when saving these values. When size getting bigger, the values in Hilbert Matrix become smaller and lead to bigger error.

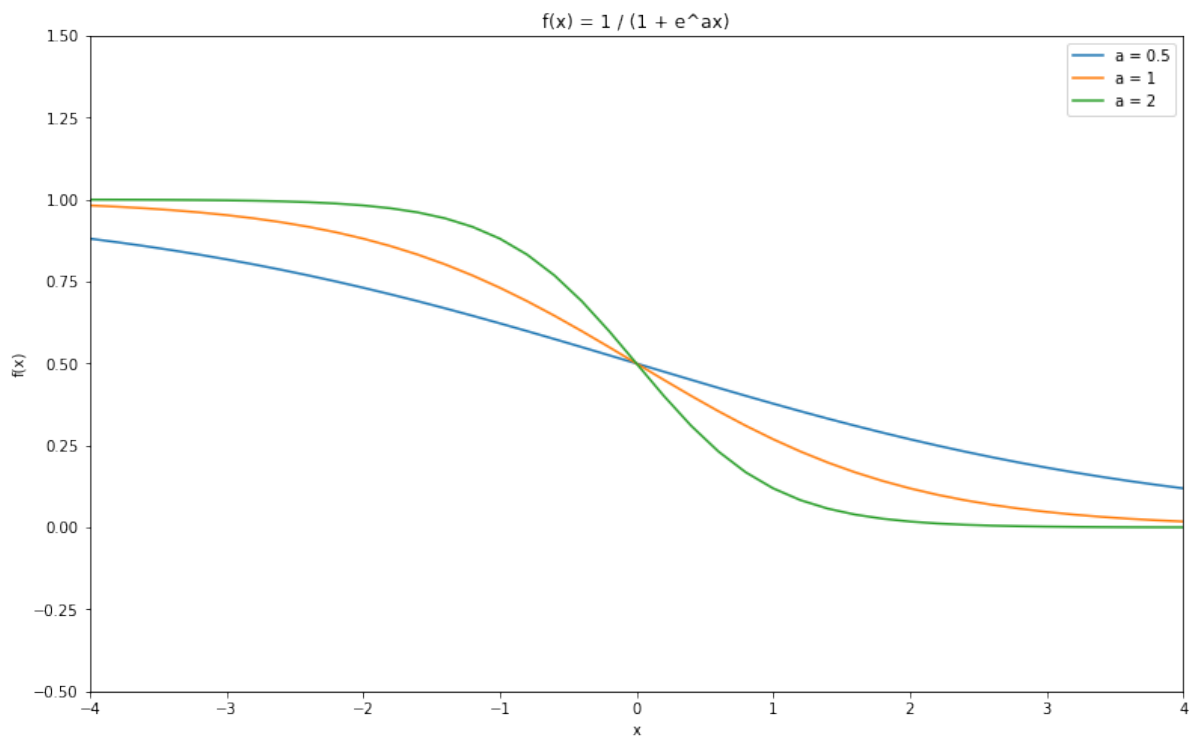
Lab Book 10 & 11

```
In [11]: 1 x = np.array([0.5, 0.7, 0.9, 1.3, 1.7, 1.8])
2 y = np.array([0.1, 0.2, 0.75, 1.5, 2.1, 2.4])
3 plt.figure(figsize = (13,8))
4 plt.clf()
5 plt.plot(x, y, color='g', linestyle='--', linewidth='1.5', mark
6
7 plt.title('y against x')
8 plt.xlabel('x')
9 plt.ylabel('y')
10 plt.grid()
11 plt.xlim(0,3)
12 plt.ylim(0,3)
13
14 plt.show()
```



Lab Book 12

```
In [12]: 1 plt.figure(figsize = (13,8))
2 plt.clf()
3 an = np.array([0.5, 1, 2])
4 x = np.linspace(-4, 4, 41)
5 for a in an:
6     fx = 1 / (1 + np.exp(a * x))
7     plt.plot(x, fx)
8 plt.legend(['a = 0.5', 'a = 1', 'a = 2'])
9 plt.title('f(x) = 1 / (1 + e^ax)')
10 plt.xlabel('x')
11 plt.ylabel('f(x)')
12 plt.xlim(-4, 4)
13 plt.ylim(-0.5, 1.5)
14 plt.show()
```



```
In [ ]: 1
```