

## exercise2 (Score: 18.0 / 21.0)

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## Lab 5

1. 提交作業之前，建議可以先點選上方工具列的**Kernel**，再選擇**Restart & Run All**，檢查一下是否程式跑起來都沒有問題，最後記得儲存。
2. 請先填上下方的姓名(name)及學號(student\_id)再開始作答，例如：

```
name = "我的名字"  
student_id= "B06201000"
```

3. 演算法的實作可以參考[lab-5 \(https://yuanyuyuan.github.io/itcm/lab-5.html\)](https://yuanyuyuan.github.io/itcm/lab-5.html), 有任何問題歡迎找助教詢問。
4. **Deadline: 12/11(Wed.)**

In [1]:

```
name = "黃宇文"  
student_id = "B06201029"
```

## Exercise 2

Suppose that a planet follows an elliptical orbit, which can be represented in a Cartesian coordinate system by the equation of the form

$$\alpha_1y^2 + \alpha_2xy + \alpha_3x + \alpha_4y + \alpha_5 = x^2. \tag{1}$$

Based on the observation of the planet's position:

\$\$\$ \left[ \right.

```
\begin{array}{c}
x \ \backslash \\
y \\
\end{array}
\right ] =
\left [
\begin{array}{c}
1.02 \ 0.95 \ 0.87 \ 0.77 \ 0.67 \ 0.56 \ 0.44 \ 0.30 \ 0.16 \ 0.01 \ 0.39 \ 0.32 \ 0.27 \ 0.22 \ 0.18 \ 0.15 \ 0.13 \ 0.12 \\
& 0.13 \ 0.15
\end{array}
\right ],$$$
```

we want to determine the orbital parameters  $\alpha_i, i = 1, 2, \cdots, 5$ , that solve the linear least squares problem of the form:  $\min_{\alpha_i} \|b - A\alpha\|_2$ , where the vector  $b \in \mathbb{R}^{10}$ ,  $\alpha = [\alpha_1, \alpha_2, \alpha_3, \alpha_4, \alpha_5]^T \in \mathbb{R}^5$  and the matrix  $A \in \mathbb{R}^{10 \times 5}$  can be obtained easily when we substitute the above data to the equation (1).

## Part 0

Import necessary libraries

```
In [2]:
import numpy as np
import matplotlib.pyplot as plt
```

## Part 1

Find the solution of the problem by solving the associated normal equations via Cholesky factorization.

### Part 1.1

Prepare data vector  $x, y$  and store them into 1D arrays: `data_x` , `data_y` .

```
In [3]:


(Top)



```
'''
Hint:
    data_x = ?
    data_y = ?
'''

# ===== 請實做程式 =====
data_x = [1.02 , 0.95 , 0.87 , 0.77 , 0.67 , 0.56 , 0.44 , 0.30 , 0.16 , 0.01]
data_y = [0.39 , 0.32 , 0.27 , 0.22 , 0.18 , 0.15 , 0.13 , 0.12 , 0.13 , 0.15]
# =====
```


```

Check your `data_x` and `data_y`.

In [4]:

cell-3b704739d6fd2990

(Top)

```
print('x =', data_x)
print('y =', data_y)
### BEGIN HIDDEN TESTS
assert np.mean(data_x - np.array([1.02, 0.95, 0.87, 0.77, 0.67, 0.56, 0.44, 0.30, 0.16, 0.01])) < 1e-7
assert np.mean(data_y - np.array([0.39, 0.32, 0.27, 0.22, 0.18, 0.15, 0.13, 0.12, 0.13, 0.15])) < 1e-7
### END HIDDEN TESTS
```

```
x = [1.02, 0.95, 0.87, 0.77, 0.67, 0.56, 0.44, 0.3, 0.16, 0.01]
y = [0.39, 0.32, 0.27, 0.22, 0.18, 0.15, 0.13, 0.12, 0.13, 0.15]
```

## Part 1.2

Construct the matrix  $A$  and the vector  $b$  with the data  $x, y$  and the equation (1).

In [5]:

(Top)

```
def construct_A_and_b(x, y):
    """
    Arguments:
        x : 1D np.array, data x
        y : 1D np.array, data y

    Returns:
        A : 2D np.array
        b : 1D np.array
    """

    # ===== 請實做程式 =====
    arrs = []
    n = len(x)
    b = np.zeros((n,1))

    for i in range(n):
        arrs.append([y[i]**2, (x[i])*(y[i]), x[i], y[i], 1])
        b[i] = x[i]**2

    A = np.ones((n,5))*arrs

    return A, b
# =====
```

Check your  $A$  and  $b$ .

In [6]:

cell-ab0180156b91fc0c

(Top)

```
A, b = construct_A_and_b(data_x, data_y)
print('A:\n', A)
print('b:\n', b)
```

A:

```
[[0.1521 0.3978 1.02   0.39   1.    ]
 [0.1024 0.304   0.95   0.32   1.    ]
 [0.0729 0.2349 0.87   0.27   1.    ]
 [0.0484 0.1694 0.77   0.22   1.    ]
 [0.0324 0.1206 0.67   0.18   1.    ]
 [0.0225 0.084   0.56   0.15   1.    ]
 [0.0169 0.0572 0.44   0.13   1.    ]
 [0.0144 0.036   0.3    0.12   1.    ]
 [0.0169 0.0208 0.16   0.13   1.    ]
 [0.0225 0.0015 0.01   0.15   1.    ]]
```

b:

```
[[1.0404e+00]
 [9.0250e-01]
 [7.5690e-01]
 [5.9290e-01]
 [4.4890e-01]
 [3.1360e-01]
 [1.9360e-01]
 [9.0000e-02]
 [2.5600e-02]
 [1.0000e-04]]
```

## Part 1.3

As the [lecture \(https://ceiba.ntu.edu.tw/course/7a770d/content/cmath2019\\_note4\\_linear\\_system\\_cholesky.pdf\)](https://ceiba.ntu.edu.tw/course/7a770d/content/cmath2019_note4_linear_system_cholesky.pdf) noted, to solve the noraml eqaution via Cholesky factorization we need additional **Forward substitution** and **Backward substitution** besides the **Cholesky factorization**. Please implement and check these three algorithms at below.

**Algorithm 1:** Implement forward substitution to solve

$$Lx = b,$$

where  $L$  is a lower triangular matrix and  $b$  is a column vector.

(Note that you need to implement it by hand, simply using some package functions is not allowed.)

In [7]:

(Top)

```
def forward_substitution(L, b):
    """
    Arguments:
        L : 2D lower triangular np.array
        b : 1D np.array

    Return:
        x : solution to Lx = b
    """

    # ===== 請實做程式 =====
    assert len(L.shape) == 2
    m, n = L.shape
    assert m == n
    assert len(b) == n

    x = np.zeros(n)
    for i in range(n):
        r = sum([L[i,j]*x[j] for j in range(i)])
        x[i] = (b[i]-r) / L[i,i]
    return x
# =====
```

Check your function.

In [8]:

cell-55c3537517a849a7

(Top)

```
L = np.array([
    [1, 0, 0, 0],
    [2, 1, 0, 0],
    [4, 5, 6, 0],
    [1, 2, 3, 4]
])
x = np.array([11, 22, 33, 24])
print('L:\n', L)
print('x:\n', x)
print('My answer:\n', forward_substitution(L, L @ x))
```

```
L:
[[1 0 0 0]
 [2 1 0 0]
 [4 5 6 0]
 [1 2 3 4]]
x:
[11 22 33 24]
My answer:
[11. 22. 33. 24.]
```

**Algorithm 2:** Implement backward substitution to solve

$$Rx = b,$$

where  $R$  is an upper triangular matrix and  $b$  is a column vector.

(Note that you need to implement it by hand, simply using some package functions is not allowed.)

In [9]:

(Top)

```
def backward_substitution(R, b):
    '''
    Arguments:
        R : 2D upper triangular np.array
        b : 1D np.array

    Return:
        x : solution to Rx = b
    '''

    # ===== 請實做程式 =====
    assert len(R.shape) == 2
    m, n = R.shape
    assert m == n
    assert len(b) == n

    x = np.zeros(n)
    for i in reversed(range(n)):
        r = sum([R[i,j]*x[j] for j in range(i, n)])
        x[i] = (b[i]-r) / R[i,i]
    return x
# =====
```

Check your function.

In [10]:

cell-b139cd9ef4098615

(Top)

```
R = np.array([
    [1, 2, 3],
    [0, 4, 5],
    [0, 0, 9]
])
x = np.array([11, 22, 33])
print('R:\n', R)
print('x:\n', x)
print('My answer:\n', backward_substitution(R, R @ x))
```

```
R:
[[1 2 3]
 [0 4 5]
 [0 0 9]]
x:
[11 22 33]
My answer:
[11. 22. 33.]
```

**Algorithm 3:** Implement Cholesky decomposition to decompose a nonsingular [PSD](#) ([https://www.wikiwand.com/en/Definiteness\\_of\\_a\\_matrix](https://www.wikiwand.com/en/Definiteness_of_a_matrix)) matrix  $A$  into

$$A = R^T R,$$

where  $R$  is an upper triangular matrix.

(Note that you need to implement it by hand, simply using some package functions is not allowed.)

In [11]:

(Top)

```
def cholesky_decomposition(A):
    """
    Arguments:
        A : 2D np.array

    Return:
        R : 2D np.array, A = R^T R
    """

    # ===== 請實做程式 =====

    n = len(A)
    R = np.zeros((n,n))

    for i in range(n):
        for k in range(i+1):
            tmp_sum = sum(R[i][j] * R[k][j] for j in range(k))

            if i == k:
                R[i][k] = np.sqrt(A[i][i] - tmp_sum)
            else:
                R[i][k] = (1.0 / R[k][k] * (A[i][k] - tmp_sum))

    return R.T

# =====
```

Check your function.

In [12]:

cell-cc45a402f856cb26

(Top)

```
# Construct a PSD matrix A
_A = np.array([
    [1, 3, 2, 4],
    [4, 2, 1, 7],
    [2, 5, 9, 0],
    [3, 5, 8, 2]
])
A = _A.T @ _A

# Do Cholesky decomposition
R = cholesky_decomposition(A)
print('A:\n', A)
print('R:\n', R)
print('A = R.T @ R:\n', R.T @ R)
```

```
A:
[[ 30  36  48  38]
 [ 36  63  93  36]
 [ 48  93 150  31]
 [ 38  36  31  69]]
R:
[[ 5.47722558  6.57267069  8.76356092  6.93781906]
 [ 0.          4.44971909  7.95555838 -2.15743956]
 [ 0.          0.          3.14787085 -4.01425733]
 [ 0.          0.          0.          0.31282475]]
A = R.T @ R:
[[ 30.  36.  48.  38.]
 [ 36.  63.  93.  36.]
 [ 48.  93. 150.  31.]
 [ 38.  36.  31.  69.]]
```

## Part 1.4

Implement the function `solve_alpha` to find  $\alpha$  from the associated the normal equation.

In [13]:

(Top)

```
def solve_alpha(x, y):
    """
    Arguments:
        x : 1D np.array, data x
        y : 1D np.array, data y

    Returns:
        alpha : 1D np.array

    Hints:
        1. Find matrix A, vector b
        2. Find the associated normal equation
        3. Do Cholesky decomposition
        4. Solve the equation with forward/backward substitution
    """

    # ===== 請實做程式 =====
    A, b = construct_A_and_b(x, y)
    ne = A.T @ A
    R = cholesky_decomposition(ne)

    y = forward_substitution(R.T, A.T @ b)
    alpha = backward_substitution(R, y)

    return alpha
# =====
```

Solve  $\alpha$  !

In [14]:

cell-ada65b7c60848c59

(Top)

```
alpha = solve_alpha(data_x, data_y)
print('alpha:\n', alpha)
### BEGIN HIDDEN TESTS
assert np.mean(alpha - np.array([-2.63562548, 0.14364618, 0.55144696, 3.22294034, -0.43289427])) < 1e-7
### END HIDDEN TESTS
```

```
alpha:
[-2.63562548  0.14364618  0.55144696  3.22294034 -0.43289427]
```

## Part 2

**Perturb the input data slightly by adding to each coordinate of each data point a uniformly distributed random number, and solve the least square problem as before with the perturbed data.**

**Compare the new values for the parameters with those previously computed. What effect does this difference have on the plot of the orbit ?**

### Part 2.1

In order to plot the orbit, we need to transform the equation (1) into a graph  $z = f(x, y, \alpha)$  and then plot the contour at  $z = 0$  by the tool `plt.contour`.

In [15]:

(Top)

```
def ellipse(x, y, alpha):
    """
    Arguments:
        x : 1D np.array, data x
        y : 1D np.array, data y
        alpha : 1D np.array, the coefficients

    Returns:
        z : 1D np.array, z=f(x, y, alpha) from equation (1)
    """
    # ===== 請實做程式 =====
    z = alpha[0] * y**2 + alpha[1] * x*y + alpha[2] * x + alpha[3] * y + alpha[4] - x**2

    return z
# =====
```

Plot the orbit.



In [16]:

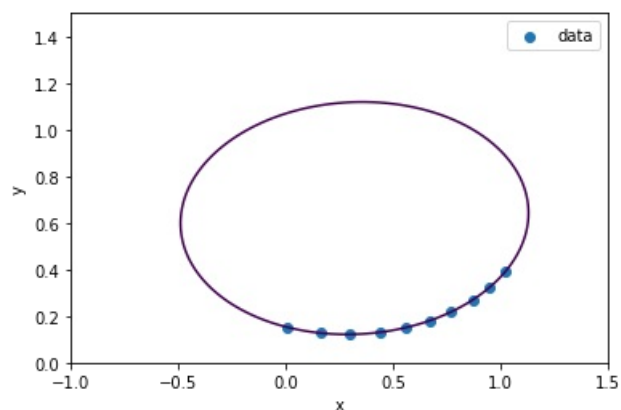
cell-c944b24065f4673f

(Top)

```
# Plot the exact data points (x,y)
plt.scatter(data_x, data_y, label='data')

# Prepare mesh data points (X,Y) to plot the orbit
X, Y = np.meshgrid(
    np.linspace(-1, 1.5, 100),
    np.linspace(0, 1.5, 100)
)
# Plot the level curve at z = 0 only
plt.contour(X, Y, ellipse(X, Y, alpha), [0])

plt.xlabel('x')
plt.ylabel('y')
plt.legend()
plt.show()
```



## Part 2.2

Now perturb the original data with some slight, uniformly random noise and follow the steps as before to find new perturbed\_x , perturbed\_y , perturbed\_alpha .

In [17]:

(Top)

```
'''
Hint:
    perturbed_x = ?
    perturbed_y = ?
    perturbed_alpha = ?
'''

# ===== 請實做程式 =====
p1 = np.random.rand(len(data_x))
p2 = np.random.rand(len(data_y))
perturbed_x = data_x + p1 * 0.005
perturbed_y = data_y + p2 * 0.005
perturbed_alpha = solve_alpha(perturbed_x, perturbed_y)
print('perturbed_alpha:\n', perturbed_alpha)
# =====
```

```
perturbed_alpha:
[-2.97615349  0.21031272  0.55129009  3.32312297 -0.45024145]
```

Overlay the new perturbed orbit on the plot.

In [18]:

cell-7428d2eef3884195

(Top)

```
# Plot the exact data points (x,y)
plt.scatter(data_x, data_y, label='data')

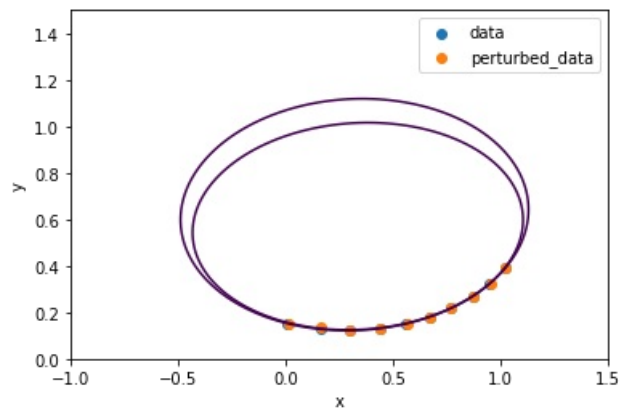
# Plot the perturbed data points
plt.scatter(perturbed_x, perturbed_y, label='perturbed_data')

# Prepare mesh data points (X,Y) to plot the orbits
X, Y = np.meshgrid(
    np.linspace(-1, 1.5, 100),
    np.linspace(0, 1.5, 100)
)

# Plot the level curve at z = 0
plt.contour(X, Y, ellipse(X, Y, alpha), [0])

# Plot the level curve at z = 0 after perturbed
plt.contour(X, Y, ellipse(X, Y, perturbed_alpha), [0])

plt.xlabel('x')
plt.ylabel('y')
plt.legend()
plt.show()
```



(Top)

## Part 2.3

Try some different perturbations and compare the orbits before and after your perturbation. What's your observation?

### Comments:

It's not correct.

No matter how the perturbations change, the graph is still an ellipse.

In [ ]: