```
exercise1 (Score: 17.0 / 17.0)

1. Test cell (Score: 1.0 / 1.0)

2. Task (Score: 5.0 / 5.0)

3. Test cell (Score: 1.0 / 1.0)

4. Task (Score: 2.0 / 2.0)

5. Task (Score: 5.0 / 5.0)

6. Test cell (Score: 1.0 / 1.0)

7. Task (Score: 2.0 / 2.0)
```

Lab 3

- 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(stduent_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-3 (https://yuanyuyuan.github.io/itcm/lab-3.html), 有任何問題歡迎找助教詢問。
- 4. Deadline: 10/30(Wed.)

```
In [1]:
```

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

Exercise 1

Let $g(x) = \ln(4 + x - x^2)$ and α is a fixed point of g(x) i.e. $\alpha = g(\alpha)$.

- ### Part A. Implement your fixed-point algorithm and solve it with initial guess $x_0 = 2$ within tolerance 10^{-10} , and answer the questions of error behavior analysis below.
- ### Part B. Redo Part A. by applying Aitken's acceleration.

Import libraries

```
In [2]:
```

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

Implement the target function $g(x) = \ln(4 + x - x^2)$

```
In [3]:
```

(Ton

```
In [4]:
```

```
cell-c0f08330aec65e17
assert round(g(0), 4) == 1.3863
### BEGIN HIDDEN TESTS
import random
x = random.random()
assert q(x) == np.log(4 + x - x**2), 'Failed on x = f' % x
### END HIDDEN TESTS
```

Run built-in fixed-point method

(https://docs.scipy.org/doc/scipy/reference/generated/scipy.optimize.fixed_point.html#rf(1) with Python SciPy, and use this accurate value as the fixed point α

In [5]:

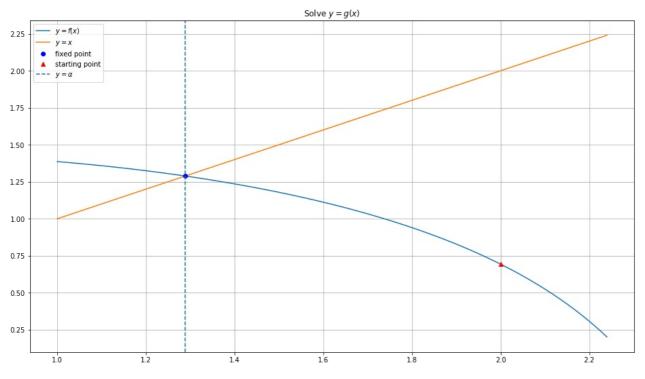
```
from scipy import optimize
alpha = optimize.fixed_point(g, x0=2, xtol=1e-12)
print('The fixed point is', alpha)
```

The fixed point is 1.2886779668238684

Visualization

In [6]:

```
x range = np.arange(1, 2.25, 0.01)
plt.figure(figsize=(16, 9))
plt.title(r'Solve $y=g(x)$')
plt.plot(x_range, g(x_range), label=r'$y=f(x)$')
plt.plot(x_range, x_range, label=r'$y=x$')
plt.plot(alpha, g(alpha), 'bo', label='fixed point')
plt.plot(2.0, g(2.0), 'r^', label='starting point')
plt.axvline(x=alpha, linestyle='--', label=r'$y=\alpha$')
plt.gca().legend()
plt.grid()
plt.show()
```



Part A.

1. Find the fixed point of g(x) using your fixed-point iteration to within tolerance 10^{-10} with initial guess $x_0 = 2$.

1-1. Implement the fixed point method

In [7]:

```
def fixed_point(
    func,
    x_0,
    tolerance=1e-7,
   max_iterations=5,
    '''Find the fixed point of the given function func
    Parameters
    ____.
    func : function
       The target function.
    x 0 : float
        Initial guess point for a solution func(x)=x.
    tolerance: float
       One of the termination conditions. Error tolerance.
    max_iterations : (positive) integer
        One of the termination conditions. The amount of iterations allowed.
   Returns
    _____
    solution : float
        Approximation of the root.
    history: dict
        Return history of the solving process
       history: {'x_n': list}
    # ==== 請實做程式 =====
    x n = x 0
    num iterations = 0
    history = {'x_n':[]}
    while True:
        f_of_x_n = func(x_n)
        num iterations += 1
        error = abs(f_of_x_n - x_n)
        history['x_n'].append(x_n)
        if error < tolerance:</pre>
            print('Found solution after', num iterations, 'iterations.')
            return x_n, history
        if num iterations < max iterations:</pre>
           x_n = f_of_x_n
        else:
           print('Terminate since reached the max iterations.')
           return x_n, history
```

1-2. Find the root

```
In [8]:
```

Found solution after 29 iterations.

In [9]:

```
cell-2d72f68109ee500c (Top)

print('My estimation is', solution)
### BEGIN HIDDEN TESTS
assert np.round(solution, 9) == np.round(alpha, 9), 'Wrong answer!'
### END HIDDEN TESTS
```

My estimation is 1.2886779668876651

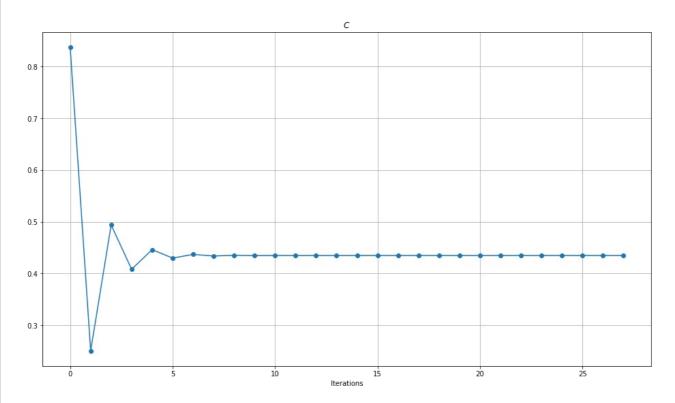
(Top)

2. Estimate graphically the asymptotic error constant C

$$\lim_{n\to\infty} \frac{|x_{n+1} - \alpha|}{|x_n - \alpha|} = C$$

```
In [10]:
```

```
(Top)
1.1.1
Hint:
    1. Prepare the sequences: x_n (from the history of algorithm)
    2. Compute the error of sequence: e_n
    3. Compute the sequence: e_{n+1}/e_{n}
    4. Plot the curve
    5. Fill in the name of x,y axes
    6. Show the plot
# ===== 請實做程式 =====
x_n = history['x_n']
e_n = abs(alpha - x_n)
plt.figure(figsize = (16,9))
plt.plot(e_n[1:] / e_n[:-1], 'o-')
plt.title(r'$C$')
plt.xlabel('Iterations')
plt.grid()
plt.show()
```



Part B.

(Top)

- 1. Accelerate the convergence of the sequence $\{x_n\}$ obtained in *Part A.* using Aitken's Δ^2 method, yielding sequence $\{\hat{x}_n\}$.
- 1-1. Introduce Aitken's acceleration into the original method.

```
In [11]:
def aitken(
    func,
    x Θ,
    tolerance=1e-7,
    max iterations=5,
):
    '''Approximate solution of f(x)=0 on interval [a,b] by the secant method.
    Parameters
     func : function
        The target function.
    x 0 : float
        Initial guess point for a solution f(x)=x.
     tolerance: float
        One of the termination conditions. Error tolerance.
    max_iterations : (positive) integer
        One of the termination conditions. The amount of iterations allowed.
    Returns
     solution : float
        Approximation of the root.
    history: dict
        Return history of the solving process
        history: {'x_n': list}
    # ==== 請實做程式 =====
    x_n = x_0
    num\_iterations = 0
    history = {'x_n':[]}
    while True:
        x 1 = func(x 0)
        x 2 = func(x 1)
        x_n = x_2 - ((x_2 - x_1)^{**2} / ((x_2 - x_1) - (x_1 - x_0)))
        num_iterations += 1
        error = abs(x_n - x_0)
        history['x_n'].append(x_n)
         if error < tolerance:</pre>
             print('Found solution after', num_iterations, 'iterations.')
             return x_n, history
         if num iterations < max iterations:</pre>
            x_0 = x_n
```

1-2. Find the root

else:

==========

return x n, history

In [12]:

```
solution, history hat = aitken(
   # ===== 請實做程式 =====
   x 0 = 2.0,
   tolerance=1e-10,
   max iterations=100
   )
```

print('Terminate since reached the max iterations.')

Found solution after 5 iterations.

cell-5c862e35ba0aa7d9

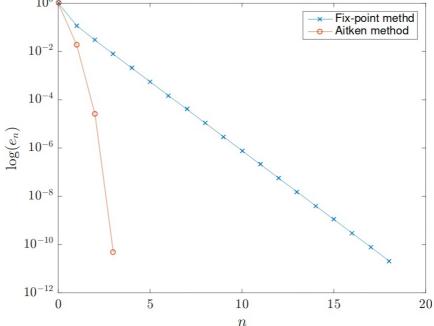
(Top)

```
print('My estimation is', solution)
### BEGIN HIDDEN TESTS
assert np.round(solution, 9) == np.round(alpha, 9), 'Wrong answer!'
### END HIDDEN TESTS
```

My estimation is 1.2886779668238684

(Top)

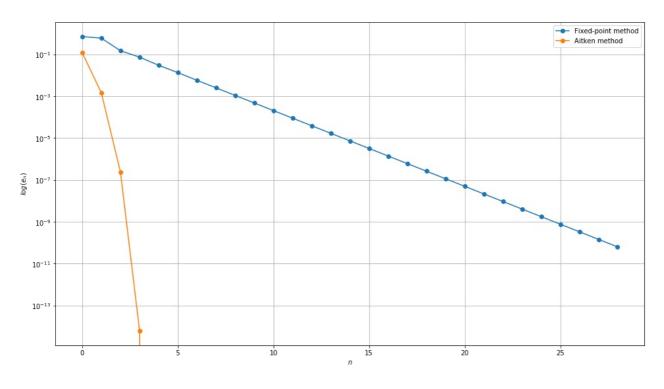
2. Plot the error curves of each algorithm w.r.t iterations n in log scale to compare the convergence rates. You may see a figure like the one in our lecture.



Ref. Page15 of cmath2019 note1 aitken.pdf (note1 aitken.pdf (note1 note1

```
In [14]:
```

```
(Top)
1.1.1
Hint:
    1. Prepare the sequences: x_n, x_n_hat(from the history of each algorithm)
    2. Compute the error of sequences: e_n, e_n_hat
    3. Plot the curves of e n, e n hat respectively
    4. Change scale into log
    5. Fill in the name of x,y axes
    6. Enable legend(show curve names)
    7. Show the plot
# ===== 請實做程式 =====
x_n = history['x_n']
x_n_{hat} = history_{hat}['x_n']
e_n = abs(alpha - x_n)
e_n_{at} = abs(alpha - x_n_{hat})
plt.figure(figsize = (16,9))
plt.plot(e_n, 'o-', label = 'Fixed-point method')
plt.plot(e_n_hat, 'o-', label = 'Aitken method')
plt.yscale('log',basey=10)
plt.xlabel('$n$')
plt.ylabel('$log(e_n)$')
plt.gca().legend()
plt.grid()
plt.show()
# =========
```



In []: