exercise2 (Score: 13.0 / 13.0)

- 1. Test cell (Score: 2.0 / 2.0)
- 2. Test cell (Score: 2.0 / 2.0)
- 3. Coding free-response (Score: 2.0 / 2.0)
- 4. Written response (Score: 2.0 / 2.0)
- 5. Test cell (Score: 1.0 / 1.0)
- 6. Coding free-response (Score: 2.0 / 2.0)
- 7. Written response (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 2

It is known that when interpolating a function f(x) with a polynomial p_{m+1} of degree m that using x_j for j=0,1,...,m as interpolation points the error has the form

$$|f(x) - p_{m+1}(x)| = \frac{\left| f^{(m+1)}(\xi_x) \right|}{(m+1)!} \left| \prod_{k=0}^m (x - x_k) \right|,$$

where $\xi_x \in [x_0, x_m]$.

Therefore, the polynomial $\omega_m(t) := \prod_{k=0}^m (t-x_k)$ influences the size of the interpolation error.

1. Put m+1 *distinct equidistant points* in the interval [-1,1], and plot $\omega_m(t)$ for m=5,10,15,20.

Part 0. Import libraries.

import matplotlib.pyplot as plt
import numpy as np

Part 1. Define $\omega_m(t)$ function.

```
In [3]:
```

In [4]:

```
mega

# Test
print('w_5(0.5) =', omega_m(0.5, np.linspace(-1, 1, 6)))

### BEGIN HIDDEN TESTS
from random import random

rd_number = random()
x = np.linspace(-1, 1, 11)

m = len(x)
product = 1

for i in range(m):
    product *= (rd_number - x[i])

assert omega_m(rd_number, np.linspace(-1, 1, 11)) == product, 'omega_m is wrong!'
### END HIDDEN TESTS
```

 $w_5(0.5) = 0.017325000000000007$

Part 2. Define the equidistant points function.

For example, if m = 4, then m + 1 distinct equidistant points in the interval [-1, 1] should be [-1, -0.5, 0, 0.5, 1].

So the results of equidistant_points(4) will be [-1. -0.5 0. 0.5 1.].

In [5]:

```
In [6]:
```

```
points

# Test
m = 4
print("Equidistant points:", equidistant_points(m))

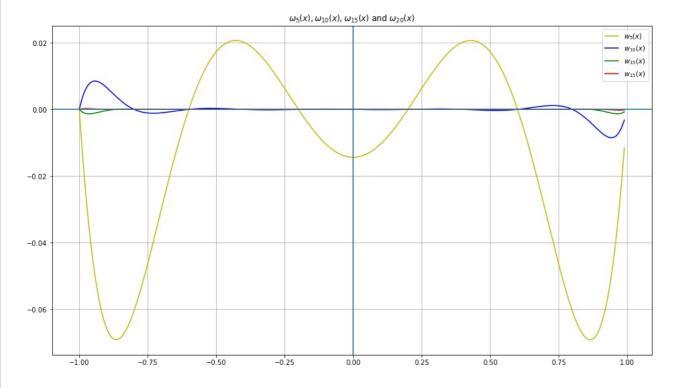
### BEGIN HIDDEN TESTS
m = 10
assert np.mean(np.array(equidistant_points(m)) - np.linspace(-1, 1, m+1)) < 1e-7, 'equidistant_points is wrong!'
### END HIDDEN TESTS</pre>
```

Equidistant points: [-1. -0.5 0. 0.5 1.]

Part 3. plot $\omega_m(t)$ for m = 5, 10, 15, 20.

Please refer parts of plotting in " lagrange.ipynb ".

In [7]:



Part 4. What's your observation of the above figure?

The higher the degree, the smaller the error.\ However, the error near the end points is larger than others.

2. Redo " Problem 1. " using *zeros of the Chebyshev polynomial (Chebyshev nodes)* as the interpolation points.

Part 1. Define Chebyshev nodes.

Please refer the part of Chebyshev nodes in " lagrange.ipynb ".

```
In [8]:
```

In [9]:

```
chebv_nodes

# Test
m = 5
print("Chebyshev nodes:", chebv_nodes(m))

### BEGIN HIDDEN TESTS
m = 10
assert np.mean(np.array(chebv_nodes(m)) - np.cos(np.linspace(0, np.pi, m+1))) < 1e-7, 'chebv_nodes is wrong!'
### END HIDDEN TESTS</pre>
```

Chebyshev nodes: [-1. -0.80901699 -0.30901699 0.30901699 0.80901699 1.]

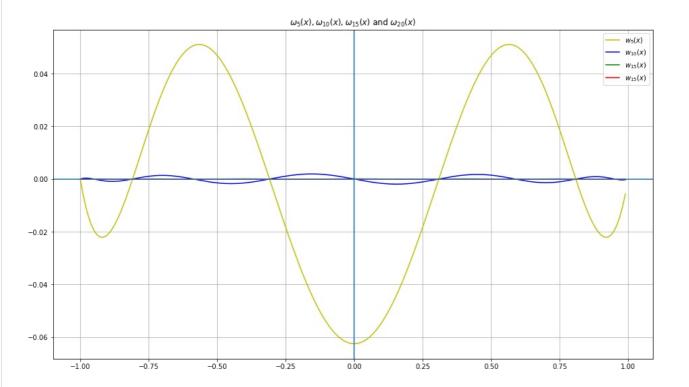
Part 2. plot $\omega(t)$ for m = 5, 10, 15, 20.

Please refer parts of plotting in " lagrange.ipynb ".

```
In [10]:
```

plt.legend(loc='upper right')

ax.grid(True)
ax.axhline(y=0)
ax.axvline(x=0)
plt.show()



Part 3. What's your observation of the above figure?

(Top)

The higher the degree, the smaller the error.\ After using the Chebyshev nodes, the error near the end points is smaller, compared to using the equidistant points. The error also distributes more average.

```
In [ ]:
```