```
exercise1 (Score: 18.0 / 20.0)

1. Task (Score: 4.0 / 4.0)

2. Test cell (Score: 2.0 / 2.0)

3. Test cell (Score: 4.0 / 4.0)

4. Test cell (Score: 2.0 / 2.0)
```

Task (Score: 4.0 / 4.0)
 Task (Score: 2.0 / 4.0)

7. Comment

## Lab 4

- 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-4 (https://yuanyuyuan.github.io/itcm/lab-4.html), 有任何問題歡迎找助教詢問。
- 4. Deadline: 11/20(Wed.)

## In [1]:

```
name = "林以翎"
student_id = "B06201024"
```

# **Exercise 1. Finite Difference**

## Part 0.

Import necessary libraries. Note that diags library from scipy is used to construct the differentiation matrix below.

#### In [2]:

```
import numpy as np
import matplotlib.pyplot as plt
from scipy.sparse import diags
```

#### Part 1.

Given a function u(x) which we want to find its derivative with numerical methods.

Consider a uniform grid partitioning x into  $\{x_1, x_2, ..., x_n\}$  with grid size  $\Delta x = x_{j+1} - x_j, j \in \{1, 2, ..., n\}$ , and a set of corresponding data values  $U = \{U_1, U_2, ..., U_n\}$ , where

$$U_{j+k} = u(x_j + k\Delta x) = u(x_{j+k}), j \in \{1, 2, ..., n\}.$$

We want to use one-sided finite-difference formula

$$\alpha_1 U_j + \alpha_2 U_{j+1} + \alpha_3 U_{j+2}$$

to approximate the derivative of u at all the points  $x_{j}, j \in \{1, 2, ..., n\}$ , that is

$$u'(x_j) \approx W_j \triangleq \alpha_1 U_j + \alpha_2 U_{j+1} + \alpha_3 U_{j+2}.$$

(Top)

#### **Part 1.1**

Find the coefficients  $\alpha_j$  for j=1,2,3 which make the stencil above accurate for as high degree polynomials as possible.

Write down your derivation in detail with Markdown/LaTeX.

$$\begin{split} U_{j} &= u(x_{j}) \\ U_{j+1} &= u(x_{j} + 1 \cdot \Delta x) = u(x) + \Delta x u'(x) + \frac{(\Delta x)^{2}}{2} u''(x) + O(u^{3}) \\ U_{j+2} &= u(x_{j} + 2 \cdot \Delta x) = u(x) + 2\Delta x u'(x) + \frac{4(\Delta x)^{2}}{2} u''(x) + O(u^{3}) \end{split}$$

$$\implies \begin{cases} \alpha_1 + \alpha_2 + \alpha_3 = 0 \\ \Delta x \alpha_2 + 2\Delta x \alpha_3 = 1 \\ \frac{(\Delta x)^2}{2} \alpha_2 + \frac{4(\Delta x)^2}{2} \alpha_3 = 0 \end{cases}$$

$$\implies (\alpha_1,\alpha_2,\alpha_3) = \left(\frac{-3}{2\Delta x},\frac{2}{\Delta x},\frac{-1}{2\Delta x}\right)$$

## Part 1.2

Fill in the tuple variable alpha of length 3 with your answer above. (Suppose  $\Delta x = 1$ )

#### In [3]:

#### In [4]:

```
cell-e7c9469885bebc80 (Top)

print('My alpha =', alpha)
### BEGIN HIDDEN TESTS

assert alpha == [-1.5, 2, -0.5] or alpha == (-1.5, 2, -0.5)
### END HIDDEN TESTS
```

```
My alpha = [-1.5, 2, -0.5]
```

#### Part 2.

Suppose we use the finite-difference formula above to approximate and assume the problem is periodic, i.e. take  $U_0 = U_n$ ,  $U_1 = U_{n+1}$ , and so on.

Find the differentiation matrix D so that the numerical differentiation problem can be represented as a matrix-vector multiplication  $W \triangleq DU$ , where  $D \in \mathbb{R}^{n \times n}$ ,  $U \in \mathbb{R}^{n}$ , and  $W \in \mathbb{R}^{n}$ .

#### **Part 2.1**

Complete the following function to construct the desired differentiation matrix under the **periodic boundary condition** with given number of partition n, coefficients of 3-point finite-difference formula  $\alpha$ , and mesh size  $\Delta x$ .

```
In [5]:
```

```
def construct_differentiation_matrix(n, alpha, delta_x):
    ''' Construct
    Parameters
    n : int
        number of partition
    alpha: tuple of length 3
       alpha = (\alpha 1, \alpha 2, \alpha 3)
    delta\_x : float
        mesh size
    Returns
    D : scipy.sparse.diags
    # ===== 請實做程式 =====
    D = [
        alpha[0] * np.ones(n),
        alpha[1] * np.ones(n-1),
        alpha[2] * np.ones(n-2),
        alpha[2] * np.ones(2),
        alpha[1] * np.ones(1)
    diagnoal matrix = diags(D, offsets=[0, 1, 2, -(n-2), -(n-1)])
    diagnoal_matrix /= delta_x
    return diagnoal matrix
    return D
```

#### Part 2.2

Print and check your implementation.

```
In [6]:
           cell-2ca00ba5ff115302
print("For n = 8 and mesh size 1, D in dense form is")
sparse_D = construct_differentiation_matrix(8, alpha, 1)
dense_D = sparse_D.toarray()
print(dense D)
### BEGIN HIDDEN TESTS
answer = np.array([
     [-1.5, 2., -0.5, 0., 0.,
                                     0.,
                                            0.,
                                     0.,
                                           0.,
                                                  0.],
     [ 0.,
           -1.5, 2., -0.5, 0.,
    [ 0.,
            0., -1.5, 2., -0.5, 0., 0., -1.5, 2.,
                                     0.,
                                            0.,
                                                  0.],
     [ 0.,
                                    -0.5,
                                           0.,
                                                  0.
                  0.,
                         0., -1.5,
                                    2.,
    [ 0.,
             0.,
                                          -0.5, 0.],
                  0.,
                              0., -1.5, 2., -0.5],
0., 0., -1.5, 2.],
             0.,
                         0.,
     [ 0.,
                         0.,
     [-0.5, 0.,
                   0.,
                                      Θ.,
           -0.5, 0.,
                         0.,
                               Θ.,
                                           Θ.,
     [ 2.,
                                                 -1.5]
])
assert np.linalg.norm(dense_D - answer) < 1e-7</pre>
### END HIDDEN TESTS
```

```
For n = 8 and mesh size 1, D in dense form is
[[-1.5 \ 2. \ -0.5 \ 0. \ 0. \ 0. \ 0.
                                  0.]
      -1.5 2.
               -0.5 0.
                                    0.]
 [ 0.
                          0.
                               0.
           -1.5 2.
                    -0.5 0.
 [ 0.
       0.
                               0.
                                    0.]
               -1.5 2. -0.5 0.
            0.
 [ 0.
       0.
                                    0.1
                 0. -1.5 2. -0.5 0.]
 [ 0.
       0.
            0.
       0.
            0.
                     0. -1.5 2. -0.5]
 [ 0.
                0.
 [-0.5 0.
            0.
                0.
                     0.
                          0. -1.5 2.]
 [ 2.
     -0.5
           0.
                 0.
                     0.
                          0.
                               0. -1.5]]
```

## Part 3.

Take  $u(x)=e^{\sin x}$  on the domain  $[-\pi,\pi]$ . Find the finite difference approximation W for  $\{u^{'}(x_{j})\}_{j=1}^{n}$  for various values of  $n=2^{k}$ , k=3,4,...,10, and analyze the errors.

#### **Part 3.1**

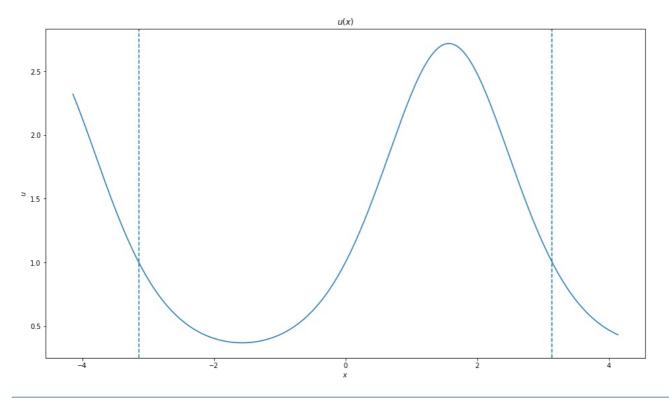
Define the functinos u and u'(x).

## In [7]:

Plot and check the functions

cell-f97d6fb0842a6055 (Top)

```
x_range = np.linspace(-np.pi-1, np.pi+1, 2**8)
plt.figure(figsize=(16, 9))
plt.plot(x_range, u(x_range))
plt.avvline(x=np.pi, linestyle='--')
plt.avvline(x=-np.pi, linestyle='--')
plt.ylabel(r'$u$')
plt.vlabel(r'$u$')
plt.title(r'$u(x)$')
plt.show()
### BEGIN HIDDEN TESTS
assert u(1) == np.exp(np.sin(1))
assert d_u(1) == np.cos(1) * np.exp(np.sin(1))
assert d_u(0) == np.cos(0) * np.exp(np.sin(0))
### END HIDDEN TESTS
```



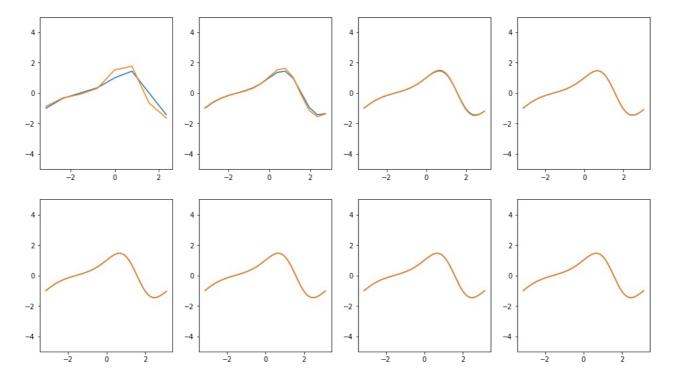
(Top)

## Part 3.2

Plot the  $u^{'}$  and W together for each point  $x_{j}, j \in \{1, 2, ..., n\}$  with  $n = 2^{k}, k \in \{3, 4, ..., 10\}$ . Note that there're total 8 figures to be plotted. And you need to compute the error, display them in the plots, and store them into the list variable error\_list for further analysis below.

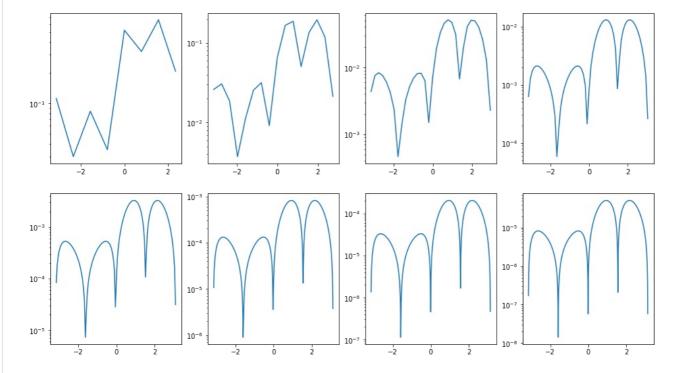
(Top

```
error_list = []
fig, axes = plt.subplots(2, 4, figsize=(16,9))
for idx, ax in enumerate(axes.flatten()):
    '''Hints:
    For each case in this for loop, you may follow the steps below
        1. Use idx to set k and n.
        2. Prepare n partition points of the domain.
       3. Construct D.
       4. Find u', U, and W.
       5. Compute the error between u' and W.
        6. Append the error into error_list.
        7. Use ax to plot u', W with proper labels, title
       8. Enable legend to show the labels of curves.
       9. To make the plots more readable, set a consistent range of y-axis e.g. ax.set\_ylim([-3, 3])
    # ==== 請實做程式 =====
    k = idx + 3
    n = 2**k
    x = np.linspace(-np.pi,np.pi,n,endpoint=False)
    D = construct differentiation matrix(n, alpha, 2*np.pi/n)
    U = u(x)
    u_diff = d_u(x)
    W = D @ U
    error = np.abs(u_diff - W)
    ax.set_ylim([-5,5])
    ax.plot(x,u_diff)
    ax.plot(x,W)
    error_list.append(error)
    # =========
```



Plot the error list with respect to k = 3, 4, ..., 10 in log scale to show the error behavior.

(Top)



(Top)

## Part 3.3

From the figure above, what rates of convergence do you observe as  $\Delta x \rightarrow 0$ ?

## **Comments:**

For what r?

The rate of convergence  $\approx (\Delta x)^r$  for some r > 0 as  $\Delta x \to 0$ .