1、

$$S = \frac{X \times X_0}{N} = \frac{X - 0.12}{0.12}$$

$$P(X) = f_0 + S + L f_0 + \frac{S(S-1)}{2!} + L^2 f_0 \quad (degree 2)$$

$$= 0.79168 - 0.01834 \times \frac{X + 0.12}{0.12} - 0.01129 \times \frac{(X + 0.12)}{0.12} + \frac{(X - 0.24)}{0.12} \times \frac{2!}{0.12} = 0.791510$$

$$P(0.231) = 0.79168 - 0.01834 \times \frac{0.11!}{0.12} - 0.01129 \times \frac{0.11!}{0.12} \times \frac{-0.009}{0.12} = 0.791510$$

$$|V(0)| = \int_{0}^{\infty} \int_{0}$$

Effor of part (a) =
$$a^3 f \times \frac{5(S-1)(S-2)}{3!} = 0.00034 \times \frac{0.11)}{0.12} \cdot \frac{-0.009}{0.12} \cdot \frac{-0.129}{6}$$

From of part (b) = $a^4 f = \frac{5(S-1)(S-2)}{4!} = 0.00038 \times \frac{0.111}{0.12} \times \frac{(-0.009)(-0.129)}{0.12} \times \frac{(-0.249)}{4!}$

=-0,00000245

$$0 = 0.39 \Rightarrow 0.00 = 0.005 \times \frac{31}{0.000} \times \frac{0.00}{0.000} = 0.000 152$$

 $0 = 0.54 \Rightarrow 0.00 = 0.00105 \times \frac{31}{0.0000} \times \frac{0.00}{0.0000} = 0.0001002$

the error start from No=0,24 is smaller, so chaose No start from 0.19 is better.

Use the below equation to solve all the S0 ~ Sn (h0 ~ hn-1 is easy to calculate)

```
Matrix for Computing Cubic Spline Coefficients
\begin{bmatrix} h_0 & 2(h_0 + h_1) & h_1 & & & & \\ h_1 & 2(h_1 + h_2) & h_2 & & & & \\ & h_2 & 2(h_2 + h_3) & h_3 & & & \\ & & \ddots & \ddots & \ddots & \\ & & & h_{n-2} & 2(h_{n-2} + h_{n-1}) & h_{n-1} \end{bmatrix} \begin{bmatrix} S_0 \\ S_1 \\ S_2 \\ S_3 \\ \vdots \\ S_{n-1} \\ S_n \end{bmatrix}
= 6 \begin{bmatrix} f[x_1, x_2] - f[x_0, x_1] \\ f[x_2, x_3] - f[x_1, x_2] \\ f[x_3, x_4] - f[x_2, x_3] \\ \vdots \\ f[x_{n-1}, x_n] - f[x_{n-2}, x_{n-1}] \end{bmatrix} \qquad h_i = x_{i+1} - x_i \\ S_i = g''_i(x_i), \quad i = 1, \dots, n-1 \\ \vdots \\ Numerical Methods © Wen-Chieh Lin 15
```

Since the problem says the slope at the ends are 0, we should slightly modify the matrix.

According to the image below, when A=B=0, $2h_0S_0 + h_0S_1 = 6(f[x_0, x_1] - 0)$. We can use this to modify the matrices.

```
f'(x_0) = A \text{ and } f'(x_n) = B,
Left: i = 0 	 2h_0 S_0 + h_0 S_1 = 6(f[x_0, x_1] - A)
Right: i = n 	 h_{n-1} S_{n-1} + 2h_{n-1} S_n = 6(B - f[x_{n-1}, x_n])
```

I construct the big matrix A and the corresponding matrix b and solve all the S0~Sn first. The first six lines of codes is the modification of matrix because of 0 slope at the ends.

```
A[0,0] = 2*h[0]

A[0,1] = h[0]

b[0] = 6 * ((y_nodes[1] - y_nodes[0]) / h[0] - 0)

A[-1,-2] = h[-1]

A[-1,-1] = 2*h[-1]

b[-1] = 6 * (0 - (y_nodes[-1] - y_nodes[-2]) / h[-1])

for i in range(1, n-1):

A[i, i-1] = h[i-1]

A[i, i] = 2*(h[i-1] + h[i])

A[i, i+1] = h[i]

b[i] = 6*((y_nodes[i+1] - y_nodes[i]) / h[i] - (y_nodes[i] - y_nodes[i-1]) / h[i-1])

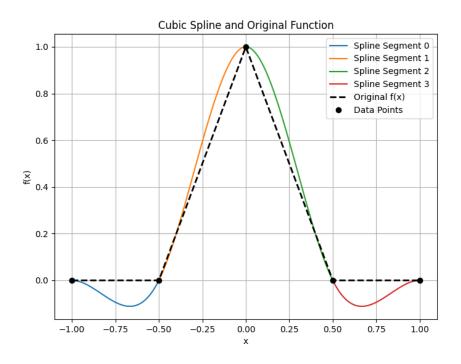
S = no.linale.solve(A, b)
```

Once we have all the S value, we can use it to solve all the coefficients for each segment

Solve the coefficients in $f_i(x) = ai(x - xi)^3 + bi(x-xi)^2 + ci(x-xi) + di for each segment.$

```
def compute_segment_coeffs(x_nodes, y_nodes, 5):
    n = len(x_nodes)
    a, b, c, d = [], [], [],
    for i in range(n - 1):
        h_i = x_nodes[i+1] - x_nodes[i]
        a.append((S[i+1] - S[i]) / (6 * h_i))
        b.append(S[i] / 2)
        c.append((y_nodes[i+1] - y_nodes[i]) / h_i - (2 * h_i * S[i] + h_i * S[i+1]) / 6)
        d.append(y_nodes[i])
    return a, b, c, d
```

Plot the spline curve together with f(x)



3.

Use the equation below to compute the polynomial for each segment

$$\mathbf{P}(u) = \begin{bmatrix} u^3 & u^2 & u & 1 \end{bmatrix} \begin{bmatrix} 2 & -2 & 1 & 1 \\ -3 & 3 & -2 & -1 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ -3 & 3 & 0 & 0 \\ 0 & 0 & -3 & 3 \end{bmatrix} \begin{bmatrix} p_0 \\ p_1 \\ p_2 \\ p_3 \end{bmatrix}$$

Construct x, y and the matrices m1, m2 for the above equation

```
x = np.array([10, 50, 75, 90, 105, 150, 180, 190, 160, 130])
y = np.array([10, 15, 60, 100, 140, 200, 140, 120, 100, 80])

m1 = np.array([
      [2, -2, 1, 1],
      [0, 0, 1, 0],
      [1, 0, 0, 0]
])

m2 = np.array([
      [1, 0, 0, 0],
      [0, 0, 0, 1],
      [-3, 3, 0, 0],
      [0, 0, -3, 3]
])
```

Use the first four points (p0~p3) to compute the first segment and we will reuse point p3 for the second segment. Therefore, I add 3 to the index for each loop. I call the function Bezier_curve for each segment to compute the coefficient using the matrices m1, m2. u_vals contains 100 values evenly between 0 and 1 for plotting each curve.

```
while idx + 3 < len(x):
    ctrl_x = x[idx:idx+4]
    ctrl_y = y[idx:idx+4]

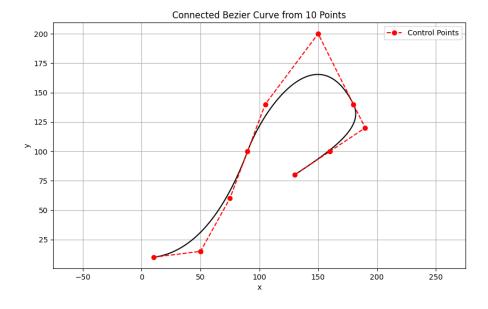
bx, by = bezier_curve(ctrl_x, ctrl_y, m1, m2, u_vals)
    plt.plot(bx, by, 'black')

idx += 3</pre>
```

```
def bezier_curve(ctrl_x, ctrl_y, m1, m2, u_vals):
    M = m1 @ m2
    coeff_x = M @ ctrl_x.T
    coeff_y = M @ ctrl_y.T

curve_x = np.polyval(coeff_x, u_vals)
    curve_y = np.polyval(coeff_y, u_vals)
    print(f'coeff_x: {coeff_x}, coeff_y: {coeff_y}')
    return curve_x, curve_y
```

(a) Draw the graph determined by the ten points.



(b)

Since the slope between point 2 and point 3 is the same as the slope between point 3 and point 4. i.e. the points [p2, p3, p4] are colinear. Also, the points [p5, p6, p7] have the same property. Therefore, the curve is C1 continuity, it smoothly connected at points 3 and 6

We know that the besier curve for each segment is compare like this. $B(0) = \begin{bmatrix} c + u^3 \\ 3 u + u u^2 \\ 3 u + u^2 \end{bmatrix} \begin{bmatrix} p_{1+1} \\ p_{1+2} \\ p_{1+2} \\ p_{1+3} \end{bmatrix} = p_1 \cdot (1 + u)^3 + p_{1+1} \cdot 3 u(1 + u)^2 + 3 u^2(1 + u) + u^3$ for each segment use for the first segment use don't have to modify the equation for the second segment $U = t - 1 \text{ for } t \in [1, 2]$ $U = t - 1 \text{ for } t \in [1, 2]$ $U = t - 1 \text{ for } t \in [1, 2]$ $U = t - 1 \text{ for } t \in [1, 2]$ $U = t - 1 \text{ for } t \in [1, 2]$

The modified equation
$$B_2(t) = B(u=t-t) = \begin{bmatrix} (2-t)^3 \\ 3(t+1)(2-t)^3 \\ (t-1)^3 \end{bmatrix} \begin{bmatrix} p_i \\ p_{i+1} \\ p_{i+2} \\ p_{i+3} \end{bmatrix}$$
for $t \in [1,2]$

Similarly, for the third segment

$$B_{3}(t) = B(u=t-z) = \begin{bmatrix} (3-t)^{3} \\ 3(t-1)^{3}(3-t) \end{bmatrix} \begin{bmatrix} p_{1}^{2} \\ p_{2}^{2} \\ p_{3}^{2} \\ p_{2}^{2} \\ p_{3}^{2} \\ p_{3}^{2} \\ p_{2}^{2} \\ p_{3}^{2} \\ p_{3}^{2} \\ p_{4}^{2} \\ p_{3}^{2} \\ p_{4}^{2} \\ p_{4}^{2} \\ p_{5}^{2} \\ p_{5}^{2}$$

4. Use the equation below to compute each segment of the B-spline curve

$$B_{i}(u) = \mathbf{u}^{T} \mathbf{M} \mathbf{p} = \frac{1}{6} \begin{bmatrix} u^{3} & u^{2} & u & 1 \\ 0 & 1 & 1 \\ 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} -1 & 3 & -3 & 1 \\ 3 & -6 & 3 & 0 \\ -3 & 0 & 3 & 0 \\ 1 & 4 & 1 & 0 \end{bmatrix} \begin{bmatrix} p_{i-1} \\ p_{i} \\ p_{i+1} \\ p_{i+2} \end{bmatrix}$$

(c)

The implementation of the code is similar to the Bezier Curve I implemented above. We need to change the matrix m and the way we compute the coefficient.

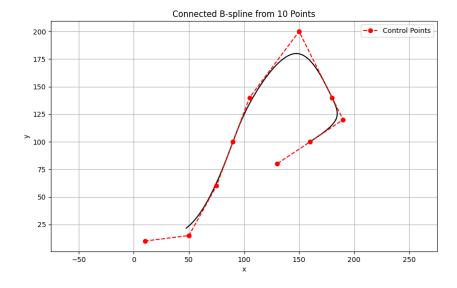
```
def B_spline(ctrl_x, ctrl_y, m, u_vals):
    coeff_x = (m @ ctrl_x.T) / 6
    coeff_y = (m @ ctrl_y.T) / 6
    coeff_y = (m @ ctrl_y.T) / 6
    curve_x = np.polyval(coeff_x, u_vals)
    curve_y = np.polyval(coeff_y, u_vals)
    print(f'coeff_x: {coeff_y}, coeff_y: {coeff_y}')
    return curve_x, curve_y
```

Also, we need to change the loop we have implemented before since B_spline will share more points between each segments.

```
idx = 1
while idx + 2 < len(x):
    ctrl_x = x[idx-1:idx+3]
    ctrl_y = y[idx-1:idx+3]

bx, by = B_spline(ctrl_x, ctrl_y, m, u_vals)
    plt.plot(bx, by, 'black')
    idx += 1</pre>
```

(a)



(b)

Since B-spline is C2-continuity, the graph smoothly connected at points 3 and 6.

(c)

It's the same as what I've done in 3(x)

If we want to move us [Oil] to us[1:2]

Just change all u in the equation to unit 11 shift the equation right by 1

5.(a)

Use Least-square solution to determine X $A^TA \cdot X = A^Tb$

(b)

Solve the above equation to find the least-squares solution

The result: Least squares solution: [1.59609218 -0.70238136 0.22066603]

The equation: z = 1.59609218 x - 0.70238136 y + 0.22066603

(c)

Find the sum of the least squares of the deviations of the points from the plane

```
Z fit = A \otimes X
Z residuals = B - Z fit
Z_residuals_sq = Z_residuals**2
Z_residuals_sq_sum = np.sum(Z_residuals_sq)
print(f"Z residuals squared sum: {Z_residuals_sq_sum}")
```

The result: Z residuals squared sum: 0.3193951297516514

$$| (bs)(x) |$$

$$(bs)(x) = (1 - \frac{x^2}{2!} + \frac{x^4}{4!} - (1 -$$

$$=) (OS^{2}(X)) = \frac{1-3X^{2}}{1+3X^{2}}$$

2,
$$Sin(x^4-x) = -xt6x^3+x^4-\frac{1}{120}x^5-...$$

$$= \frac{Q_0 + Q_1 x + Q_2 x^2 + Q_3 x^3}{| f b_1 x + b_2 x^2 + b_3 x^3}$$

$$\begin{array}{lll} x^{0}: Q_{0} = 0 & Q_{0} = 1 \\ x^{1}: Q_{1} = -1 & Q_{1} = -1 \\ x^{2}: Q_{2} = b_{1} & b_{1} = \frac{102}{2153} & -X + \frac{102}{129180} X^{2} + \frac{2134}{2153} X^{3} \\ x^{4}: Q = -b_{3} + b_{1} + 1 & b_{2} = \frac{102}{93060} & =) & 1 - \frac{102}{2153} X + \frac{14393}{43060} X^{2} + \frac{2134}{2153} X^{3} \\ x^{5}: Q = \frac{102}{6} b_{3} + b_{1} + \frac{1}{120} b_{1} & Q_{2} = \frac{102}{2153} \\ X^{6}: Q = \frac{102}{6} b_{3} + b_{2} + \frac{1}{120} b_{1} & Q_{3} = \frac{-21644}{129180} \end{array}$$

$$\begin{array}{c} Q_{3} = \frac{-21644}{129180} & Q_{3} = \frac{-21$$

$$3, \chi e^{\chi} = \chi t \chi^2 + \pm \chi^2 + \frac{\chi^6}{b} + \frac{\chi^6}{120} - \dots = \frac{(b_0 + a_1 \chi + a_2 \chi^2 + a_3 \chi^2)}{(1 + b_1 \chi + b_2 \chi^2 + b_3 \chi^2)}$$

7.

(a)

$$x(u) = a_x u^3 + b_x u^2 + c_x u + d_x$$
$$y(u) = a_y u^3 + b_y u^2 + c_y u + d_y$$

$$\begin{bmatrix} u^3 & u^2 & u & 1 \end{bmatrix} \begin{bmatrix} 2 & -2 & 1 & 1 \\ -3 & 3 & -2 & -1 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_i \\ x_{i+1} \\ x'_i \\ x'_{i+1} \end{bmatrix}$$

It's similar to p3 and p4. First, construct the matrix and the differentiation function we need.

```
m1 = np.array([
        [2, -2, 1, 1],
        [-3, 3, -2, -1],
        [0, 0, 1, 0],
        [1, 0, 0, 0]

])
x = np.array([1, 2, 3])

def f(x):
    return x*np.exp(-x)
    def diff(x):
    return (1-x)*np.exp(-x)
```

Use the control point [p[0], p[1], p'[0], p'[1]] to get the coefficient of the first segment of the Hermite curve. u_v als is used to plot the curve. We can find the second segment using the same way. Note that x'(1), x'(2) and x'(3) should plug in 1 since we let the parametric equation of x = u + 1 and x = u + 2

```
def hermite_interpolation(ctrl_x, ctrl_y, m1, u_vals):
    coeff_x = m1 @ ctrl_x.T
    coeff_y = m1 @ ctrl_y.T

    curve_x = np.polyval(coeff_x, u_vals)
    curve_y = np.polyval(coeff_y, u_vals)
    print(f'coeff_x: {coeff_x}, coeff_y: {coeff_y}')
    return curve_x, curve_y ,coeff_x, coeff_y
```

```
x_input = np.array([1, 2, 1, 1])
y_input = np.array([f(1), f(2), diff(1), diff(2)])
bx, by, cx, cy= hermite_interpolation(x_input, y_input, m1, u_vals)
plt.plot(bx, by, 'black')
```

Find the coefficient of two segments

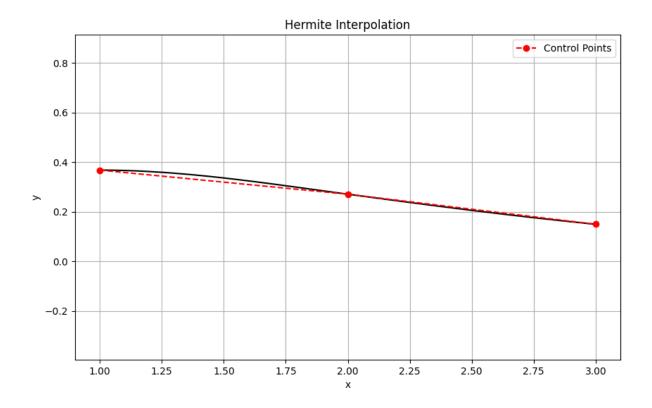
Eg: the first segment is

x(u) = u + 1 (As we expected)

$$y(u) = 0.059 * u^3 - 0.1563 * u^2 + 0.3679$$

The second segment:

The Hermite curve:



(b) To find the approximate f(1.5) using Hermite curve, we plug in u = 0.5 to the polynomial of first segment. Since we know that x(0.5) = 1.5 for first segment. The approximate value is close to the true value.

The approximate f(1.5) = 0.33619191422691047Real value : f(1.5) = 0.33469524022264474