

Question 1

1. We use the conditions of continuity and smoothness at nodes: $s_i(x_{i+1}) = s_{i+1}(x_{i+1})$, $s'_i(x_{i+1}) = s'_{i+1}(x_{i+1})$

$$s_1(1) = s_2(1)$$

$$\Rightarrow 4 + k_1 + 2 - \frac{1}{6} = 1$$

$$\Rightarrow k_1 = 1 - \frac{35}{6} = -\frac{29}{6}$$

$$s_2(2) = s_3(2)$$

$$\Rightarrow 1 - \frac{4}{3} + k_2 - \frac{1}{6} = 1$$

$$k_2 = 1 + \frac{1}{2} = \frac{3}{2}$$

$$s'_2(2) - s'_3(2) = 0$$

$$-\frac{4}{3} + 2k_2 - \frac{1}{2} = k_3$$

$$-\frac{4}{3} + 3 - \frac{1}{2} = k_3 = -\frac{8}{6} + \frac{18}{6} - \frac{3}{6} = \frac{7}{6}$$

Natural spline? ($s''_1(x_1) = 0$, $s''_{n-1}(x_n) = 0$)

$$s''_1(x) = 4 - x, \quad s''_1(0) \neq 0$$

NOT a natural spline

Parabolically terminated? ($d_1 = d_{n-1} = 0$)

$$d_1 = -\frac{1}{6}$$

NOT parabolically terminated

Not-q-knot? ($d_1 = d_2, d_2 = d_3$)

$$d_1 = -\frac{1}{6} = d_2 = -\frac{1}{6} = d_3 = -\frac{1}{6} \quad \checkmark$$

Yes, not-q-knot holds

Question 2

$$2. \quad x'(t) = 6x + 2cx \cdot t + 3 \cdot 2x \cdot t^2$$

plus in the definition of b_x , $(x, dx$:

$$\begin{aligned} x'(t) = & 3(x_2 - x_1) + 2 \cdot t (3(x_3 - x_2) - 3(x_2 - x_1)) \\ & + 3 \cdot t^2 (x_4 - x_1 - 3(x_2 - x_1) \\ & - (3(x_3 - x_2) - 3(x_2 - x_1))) \end{aligned}$$

$$\begin{aligned} x'(t) = & 3(x_2 - x_1) + 6 \cdot t (x_3 - 2x_2 + x_1) \\ & + 3t^2 (x_4 - x_1 - 3(x_3 - x_2)) \end{aligned}$$

similarly, y is defined the same, we just replace x w/ y :

$$\begin{aligned} y'(t) = & 3(y_2 - y_1) + 6 \cdot t (y_3 - 2y_2 + y_1) \\ & + 3t^2 (y_4 - y_1 - 3(y_3 - y_2)) \end{aligned}$$

clearly,

$$x'(0) = 3(x_2 - x_1), \quad y'(0) = 3(y_2 - y_1)$$

Hence, at $t=0$ (at (x_1, y_1)), the curve has direction $(x_2 - x_1, y_2 - y_1)$

Now we examine $t=1$

$$\begin{aligned}x'(1) &= 3(x_2 - x_1) + 6(x_3 - 2x_2 + x_1) + 3(x_4 - x_1 - 3(x_3 - x_2)) \\&= \underbrace{-3x_1 + 6x_1 - 3x_1}_{+6x_3 - 9x_3} + \underbrace{3x_2 - 12x_2 + 9x_2}_{+3x_4} = 3(x_4 - x_3)\end{aligned}$$

$$\begin{aligned}y'(1) &= 3(y_2 - y_1) + 6(y_3 - 2y_2 + y_1) + 3(y_4 - y_1 - 3(y_3 - y_2)) \\&= 3(y_4 - y_3)\end{aligned}$$

$$\text{thus, } (x'(1), y'(1)) = (3(x_4 - x_3), 3(y_4 - y_3))$$

which is in the direction of $(x_4 - x_3, y_4 - y_3)$

where $t=1$ corresponds to the terminal point (x_4, y_4)

Question 3

3. Define $x \in \mathbb{R}^n$, $y \in \mathbb{R}^n$

a) $\|x\|_2 > 0$ when $x \neq 0$ and $\|x\|_2 = 0$ iff $x = 0$

$$i) \|x\|_2 = \sqrt{\sum_{i=1}^n x_i^2}$$

Assume $\vec{x} \neq \vec{0}$, then there must be an element of x , x_j , s.t. $x_j \neq 0$

$$\text{Hence, } \|x\|_2 = \sqrt{\sum_{i=1}^n x_i^2} > 0$$

$$ii) \|x\|_2 = 0 \Rightarrow x = \vec{0}$$

since $a^2 \geq 0 \quad \forall a$

$$\sqrt{\sum_{i=1}^n x_i^2} = 0 \Rightarrow x_i = 0 \quad \forall i \Rightarrow x = \vec{0}$$

$$x = \vec{0} \Rightarrow \|x\|_2 = 0$$

$$x = \vec{0} \Rightarrow x_i = 0 \quad \forall i \Rightarrow \sqrt{\sum_{i=1}^n x_i^2} = 0 = \|x\|_2$$

$$b) \|Kx\|_2 = |K| \cdot \|x\|_2$$

$$\|Kx\|_2 = \sqrt{\sum_{i=1}^n (Kx_i)^2} = \sqrt{K^2 \sum_{i=1}^n x_i^2}$$

$$= \sqrt{K^2} \sqrt{\sum_{i=1}^n x_i^2} = |K| \cdot \|x\|_2$$

$$c) \|x+y\|_2 \leq \|x\|_2 + \|y\|_2$$

$$\|x+y\|_2^2 = (x+y) \cdot (x+y) = \|x\|_2^2 + 2|x \cdot y| + \|y\|_2^2$$

By Cauchy-Schwarz

$$\leq \|x\|_2^2 + 2\|x\|_2\|y\|_2 + \|y\|_2^2 = (\|x\|_2 + \|y\|_2)^2$$

if we take square root of both sides:

$$\|x+y\|_2 \leq \|x\|_2 + \|y\|_2$$

Question 4

$$4) i) \underline{(A^T)^{-1} = (A^{-1})^T}$$

$$\text{We have that } B^{-1}B = BB^{-1} = I$$

$$\text{and } (BC)^T = C^T B^T$$

$$\text{Also } BC = CB = I \Rightarrow C, B \text{ are inverses} \quad \textcircled{2}$$

$$(A^{-1}A)^T = (AA^{-1})^T = I^T = I$$

$$A^T(A^{-1})^T = (A^{-1})^T A^T = I$$

By ~~2~~ ②, we have shown that $(A^{-1})^T$ is the inverse of A^T

$$\text{In other words, that } (A^{-1})^T = (A^T)^{-1}$$

$$ii) \underline{A\bar{x} = b \Rightarrow A^T A\bar{x} = A^T b}$$

$$A\bar{x} = b \Rightarrow \bar{x} = A^{-1}b$$

$$A^T A\bar{x} = A^T A(A^{-1}b) = A^T b$$

This is the normal equation which we have proved using the unique solution \bar{x}

b) Let $\bar{x} \in \mathbb{R}^n$ s.t. $A^T A \bar{x} = A^T b$

q) $x \in \mathbb{R}^n$, we show $\|A\bar{x} - b\| \leq \|Ax - b\|$

$$\begin{aligned}\|Ax - b\|_2^2 &= \|(A\bar{x} - b) + (Ax - A\bar{x})\|_2^2 \\ &= \|A\bar{x} - b\|_2^2 + \|Ax - A\bar{x}\|_2^2 + \underbrace{\langle A\bar{x} - b, Ax - A\bar{x} \rangle}_{=0} \\ &= \|A\bar{x} - b\|_2^2 + \|Ax - A\bar{x}\|_2^2 = \langle A^T(A\bar{x} - b), x - \bar{x} \rangle\end{aligned}$$

Since the 2-norm is ≥ 0 , $= 0$ b/c $A^T(A\bar{x} - b) = 0$ by \textcircled{a}

$$\|Ax - b\|_2^2 \geq \|A\bar{x} - b\|_2^2$$

$$\Rightarrow \|Ax - b\|_2 \geq \|A\bar{x} - b\|_2$$

Question 5

```
In [ ]: from numpy import cos, sin, pi
from numpy.linalg import norm

# Define data
t = np.arange(0, 1, 1/6, dtype=np.float64)
y = np.array([0., 2., 0., -1., 1., 1.])

### F3 Model

# Define relevant Matrices
A = np.column_stack((np.ones(6, dtype=np.float64), cos(2.*pi*t), sin(2.*pi*t)))

# Solve the normal equations
sol_1 = np.linalg.solve(A.T @ A, A.T @ y).round(3)

# Compute the 2-norm errors
pred_y = A @ sol_1
SSE_1 = norm(y - pred_y, 2).round(3)

print("F3 Model")
print(f"c1: {sol_1[0]}, c2: {sol_1[1]}, c3: {sol_1[2]}")
print(f"SSE: {SSE_1}")
print("")

### F4 Model

# Define relevant Matrices
A = np.column_stack((np.ones(6, dtype=np.float64), cos(2.*pi*t), sin(2.*pi*t), cos(4.*pi*t), sin(4.*pi*t)))

# Solve the normal equations
sol_2 = np.linalg.solve(A.T @ A, A.T @ y).round(3)
```



```
# Compute the 2-norm errors
pred_y = A @ sol_2
SSE_2 = norm(y - pred_y,2).round(3)

print("F4 Model")
print(f"c1: {sol_2[0]}, c2: {sol_2[1]}, c3: {sol_2[2]}, c4: {sol_2[3]}")
print(f"SSE: {SSE_2}")
```

F3 Model
c1: 0.5, c2: 0.667, c3: -0.0
SSE: 2.041

F4 Model
c1: 0.5, c2: 0.667, c3: -0.0, c4: -1.0
SSE: 1.08

The more robust model with a fourth parameter has a significantly better fit to the data. Both models have 0 for c3 (the sin term).

Question 6

$$b) a) S_i = y_i + b_i (X - x_i) + c_i (X - x_i)^2 + d_i (X - x_i)^3$$

for $i = 1, \dots, n-1$

with the following properties

1) S passes through all points:

$$S_i(x_i) = y_i \quad \forall i = 1, \dots, n-1$$

and

$$S_{n-1}(x_n) = y_n$$

2) continuity

$$S_i(x_i) = y_i, \quad S_i(x_{i+1}) = y_{i+1} \quad \forall i = 1, \dots, n-1$$

3) First derivative continuous

$$S'_i(x_{i+1}) = S'_{i+1}(x_{i+1}) \quad \forall i = 1, \dots, n-2$$

4) Second derivative continuous

$$S''_i(x_{i+1}) = S''_{i+1}(x_{i+1}) \quad \forall i = 1, \dots, n-2$$

For a natural spline, we also have

$$S''_1(x_1) = S''_{n-1}(x_n) = 0$$

b) For a not-a-knot spline, we also have

$$d_1 = d_2, \quad d_{n-2} = d_{n-1}$$

By the 0° , 1° , and second order continuity conditions,
this leads to

$$s_1 = s_2, \quad s_{n-2} = s_{n-1}$$

() True

b) True

c) False

Question 7

Solution

$$7. X(t) = x_1 + b_x \cdot t + c_x \cdot t^2 + d_x \cdot t^3$$

$$Y(t) = y_1 + b_y \cdot t + c_y \cdot t^2 + d_y \cdot t^3$$

we clearly see $x_1 = 3$, $y_1 = 2$

$$\text{next, } b_x = 3(x_2 - x_1)$$

$$b_y = 3(y_2 - y_1)$$

$$4 = 3(x_2 - 3)$$

$$-1 = 3(y_2 - 2)$$

$$x_2 = \frac{4}{3} + 3 = \frac{13}{3}$$

$$y_2 = -\frac{1}{3} + 2 = \frac{5}{3}$$

$$\text{Also, } c_x = 3(x_3 - x_2) - b_x$$

$$c_y = 3(y_3 - y_2) - b_y$$

$$-1 = 3x_3 - 3 \cdot \frac{13}{3} - 4$$

$$1 = 3y_3 - 3 \cdot \frac{5}{3} - (-1)$$

$$x_3 = \frac{1}{3}(4 + 13 - 1) = \frac{16}{3}$$

$$y_3 = \frac{1}{3}(1 + 5 - 1) = \frac{5}{3}$$

$$\text{Finally } d_x = x_4 - x_1 - b_x - c_x$$

$$d_y = y_4 - y_1 - b_y - c_y$$

$$x_4 = 2 + 3 + 4 + (-1)$$

$$y_4 = 3 + 2 + (-1) + (1)$$

$$= 8$$

$$y_4 = 5$$

1st end point : (3, 2)

Two control points : $(\frac{13}{3}, \frac{5}{3})$, $(\frac{16}{3}, \frac{5}{3})$

Last end point : (8, 5)

Check Work

In []: `from sympy import symbols, Eq, solve`

`# Define the symbols for the control points`

`P0x, P1x, P2x, P3x, P0y, P1y, P2y, P3y = symbols('P0x P1x P2x P3x P0y P1y P2y P3y')`

```
# Given endpoint P0 and P3
```

```
P0x_value = 3
```

```
P0y_value = 2
```

```
# Equations for x(t) based on the Bézier curve
```

```
eq1 = Eq(-3*P0x_value + 3*P1x, 4) # Coefficient of t
```

```
eq2 = Eq(3*P0x_value - 6*P1x + 3*P2x, -1) # Coefficient of t^2
```

```
eq3 = Eq(-P0x_value + 3*P1x - 3*P2x + P3x, 2) # Coefficient of t^3
```

```
# Equations for y(t) based on the Bézier curve
```

```
eq4 = Eq(-3*P0y_value + 3*P1y, -1) # Coefficient of t
```

```
eq5 = Eq(3*P0y_value - 6*P1y + 3*P2y, 1) # Coefficient of t^2
```

```
eq6 = Eq(-P0y_value + 3*P1y - 3*P2y + P3y, 3) # Coefficient of t^3
```

```
# Solve the equations to find the control points
```

```
solution = solve((eq1, eq2, eq3, eq4, eq5, eq6), (P1x, P2x, P3x, P1y, P2y, P3y))
```

```
print({'P0x': P0x_value, 'P0y': P0y_value}, solution)
```

```
{'P0x': 3, 'P0y': 2} {P1x: 13/3, P1y: 5/3, P2x: 16/3, P2y: 5/3, P3x: 8, P3y: 5}
```

Question 8

8.a) The least squares method gives the value of x that minimizes $\|Ax - b\|_2$:

$$\bar{x} = \min_x \|Ax - b\|_2$$

where $A \in \mathbb{R}^{(m \times n)}$, $b \in \mathbb{R}^m$, $x \in \mathbb{R}^n$

In statistics, this means finding the x that best fits $Ax = b$.

If A is non-singular, we have the exact solution.

Otherwise, x satisfies the normal equation:

$$A^T A x = A^T b$$

b) true

c) true