# **Assignment 3**

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1

Data points (-1,1), (0,0), (1,1).

a

Since there are 3 data points,  $\phi_1(t) = 1$ ,  $\phi_2(t) = x$ ,  $\phi_3(t) = t^2$ ,  $\theta = [a, b, c]^\top$ .

$$\Phi_{i,j} = \phi_j(i) \tag{1}$$

$$\Phi = \begin{bmatrix} 1 & -1 & 1 \\ 1 & 0 & 0 \\ 1 & 1 & 1 \end{bmatrix} \tag{2}$$

$$\Phi\theta = y \Leftrightarrow \begin{bmatrix} 1 & -1 & 1 \\ 1 & 0 & 0 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 0 \\ 1 \end{bmatrix} \Leftrightarrow \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$
(3)

Thus,  $p(t) = \phi(t)^{\top} \theta = t^2$ .

b

As for Lagrange basis,

$$\phi_i(t) = \prod_{0 \le j \le n \land i \ne j} \frac{t - t_j}{t_i - t_j} \tag{4}$$

$$\phi_1(t) = \frac{t-0}{-1-0} \frac{t-1}{-1-1} = \frac{t(t-1)}{2} \tag{5}$$

$$\phi_2(t) = \frac{t+1}{0+1} \frac{t-1}{0-1} = -t^2 + 1 \tag{6}$$

$$\phi_3(t) = \frac{t+1}{1+1} \frac{t-0}{1-0} = \frac{t(t+1)}{2} \tag{7}$$

$$p(t) = \phi(t)^{\top} y = t^2 \tag{8}$$

C

As for Newton basis,  $\theta = [a, b, c]^{\top}$ ,

$$\phi_i(t) = \prod_{0 \le j \le n \land i \ne j} (t - t_j) \tag{9}$$

$$\phi_1(t) = 1 \tag{10}$$

$$\phi_2(t) = t + 1 \tag{11}$$

$$\phi_3(t) = t(t+1) \tag{12}$$

$$\theta_1 = y_1 = 1 \tag{13}$$

$$\theta_{i+1} = \frac{y_{i+1} - p_i(t_{i+1})}{\phi_{i+1}(t_{i+1})} \tag{14}$$

Thus,  $\theta = [1, -1, 1]^{\top}, p(t) = \phi(t)^{\top}\theta = t^2$ .

d

 $p(t) = t^2$  is always obtained.

e

As for linear interpolating,

$$p(t) = \begin{cases} -t, -1 \le t < 0 \\ t, 0 \le t \le 1 \end{cases}$$
 (15)

f

As for cubic natural spline interpolation,

$$p(t) = \begin{cases} p_1(t) = \alpha_1 + \alpha_2 t + \alpha_3 t^2 + \alpha_4 t^3, -1 \le t < 0 \\ p_2(t) = \beta_1 + \beta_2 t + \beta_3 t^2 + \beta_4 t^3, 0 \le t \le 1 \end{cases}$$
(16)

Interpolate the data at the endpoints,

$$p_1(-1) = \alpha_1 - \alpha_2 + \alpha_3 - \alpha_4 = 1 \tag{17}$$

$$p_1(0) = \alpha_1 = 0 \tag{18}$$

$$p_2(0) = \beta_1 = 0 \tag{19}$$

$$p_2(1) = \beta_1 + \beta_2 + \beta_3 + \beta_4 = 1 \tag{20}$$

The first derivative of the interpolating function is continuous at t = 0,

$$\frac{\partial p_1(t)}{\partial t}\Big|_{t=0} = \alpha_2 = \beta_2 = \frac{\partial p_2(t)}{\partial t}\Big|_{t=0}$$
(21)

The second derivative of the interpolating function is continuous at t = 0,

$$\frac{\partial^2 p_1(t)}{\partial t^2}\Big|_{t=0} = 2\alpha_3 = 2\beta_3 = \frac{\partial^2 p_2(t)}{\partial t^2}\Big|_{t=0}$$
 (22)

A natural spline has second derivative equal to zero at the endpoints,

$$\frac{\partial^2 p_1(t)}{\partial t^2}\Big|_{t=-1} = 2\alpha_3 - 6\alpha_4 = 0 \tag{23}$$

$$\frac{\partial^2 p_2(t)}{\partial t^2}_{|t=1} = 2\beta_3 + 6\beta_4 = 0 \tag{24}$$

From (17) to (24),

$$[\alpha_1, \alpha_2, \alpha_3, \alpha_4, \beta_1, \beta_2, \beta_3, \beta_4]^\top = \left[0, 0, \frac{3}{2}, \frac{1}{2}, 0, 0, \frac{3}{2}, -\frac{1}{2}\right]^\top \tag{25}$$

Thus,

$$p(t) = \begin{cases} p_1(t) = \frac{1}{2}t^3 + \frac{3}{2}t^2, -1 \le t < 0 \\ p_2(t) = -\frac{1}{2}t^3 + \frac{3}{2}t^2, 0 \le t \le 1 \end{cases}$$
 (26)

Notice that [-1, 1] is a symmetric interval.

$$P_1(x)P_2(x)w(x) = (3x^3 - x)w(x)$$
(27)

is odd function when w(x) is even.

That is, when w(x) is even,

$$\int_{-1}^{1} P_1(x) P_2(x) w(x) dx = 0$$
 (28)

3

$$T_k(x) = \cos(k\arccos(t)), t \in [-1, 1]$$
(29)

a

$$\int_{-1}^{1} \frac{T_i(t)T_j(t)}{\sqrt{1-t^2}} dt = \int_{-\pi}^{0} \frac{T_i(\cos x)T_j(\cos x)}{\sqrt{1-\cos^2 x}} d\cos x$$
 (30)

$$= \int_{-\pi}^{0} \frac{\cos ix \cos jx}{-\sin x} (-\sin x dx) \tag{31}$$

$$= \int_{-\pi}^{0} \cos ix \cos jx dx \tag{32}$$

When  $i \neq j$ ,

$$\int_{-\pi}^{0} \cos ix \cos jx dx = \int_{-\pi}^{0} \frac{1}{2} [\cos (i+j)t + \cos (i-j)t] dx = 0$$
 (33)

When i = j,

$$\int_{-\pi}^{0} \cos ix \cos jx dx = \int_{-\pi}^{0} \cos^{2} ix dx \tag{34}$$

$$= \int_{-\pi}^{0} \frac{1 + \cos 2ix}{2} \mathrm{d}x \tag{35}$$

$$=\frac{\pi}{2}\tag{36}$$

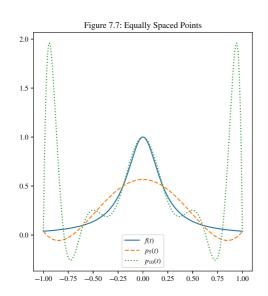
Thus,

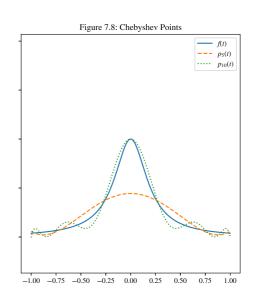
$$\int_{-1}^{1} \frac{T_i(t)T_j(t)}{\sqrt{1-t^2}} dt = \frac{\pi}{2} \delta_{i,j}$$
(37)

#### **Core Code**

```
def chebyshev_nodes(n, a, b):
    return 0.5 * (a + b) + 0.5 * (b - a) * np.cos(np.pi * (2 *
np.arange(1, n + 1) - 1) / (2 * n))
# Lagrange polynomial interpolation
def polynomial_interpolation(points, x_values):
    x_points, y_points = points
    polynomial = np.zeros_like(x_values)
    n = len(x_points)
    for i in range(n):
        term = y_points[i]
        for j in range(n):
            if i != j:
                term *= (x_values - x_points[j]) / (x_points[i] -
x_points[j])
        polynomial += term
    return polynomial
```

### Result





5

a

$$M(f) = (b-a)f\left(\frac{a+b}{2}\right) = 1 \cdot \left(\frac{1}{2}\right)^3 = \frac{1}{8}$$
 (38)

$$T(f) = \frac{b-a}{2}(f(a)+f(b)) = \frac{1}{2}(0+1) = \frac{1}{2}$$
(39)

b

$$E(f) = \frac{T(f) - M(f)}{3} = \frac{1}{8} \tag{40}$$

$$-2E(f) = -\frac{1}{4} \tag{41}$$

C

$$S(f) = \frac{2M(f) + T(f)}{3} = \frac{1}{4} \tag{42}$$

d

Yes, since when I(f) = S(f),  $\frac{\partial^3 R(f)}{\partial x^3} = 0$ , which means the Simpson rule is of degree 3.

e

$$\int_0^1 x^3 \mathrm{d}x = A_0 f(x_0) + A_1 f(x_1) \tag{43}$$

There are 4 parameters.

$$\int_0^1 1 = 1 = A_0 + A_1 \tag{44}$$

$$\int_0^1 x = \frac{1}{2} = A_0 x_0 + A_1 x_1 \tag{45}$$

$$\int_0^1 x^2 = \frac{1}{3} = A_0 x_0^2 + A_1 x_1^2 \tag{46}$$

$$\int_0^1 x^3 = \frac{1}{4} = A_0 x_0^3 + A_1 x_1^3 \tag{47}$$

The solution is  $[A_0,A_1,x_0,x_1]^ op=[rac12,rac12,rac{3-\sqrt3}6,rac{3+\sqrt3}6]^ op$ 

$$G(f) = A_0 f(x_0) + A_1 f(x_1) = \frac{1}{4}$$
(48)

The result is exact since G(f) is of degree 4-1=3.

f

$$M_2(f) = (0.5 - 0)f(0.25) + (1 - 0.5)f(0.75) = 0.21875$$
 (49)

$$f(x) = L_n(x) = \sum_{0 \le k \le n} f(x_k) l_k(x)$$
 (50)

$$I(f) = \int_a^b L_n(x) \mathrm{d}x = \int_a^b \sum_{0 \le k \le n} f(x_k) l_k(x) \mathrm{d}x \tag{51}$$

$$= \sum_{0 \le k \le n} f(x_k) \int_a^b l_k(x) \mathrm{d}x \tag{52}$$

$$=\sum_{0\leq k\leq n}^{-1}w_kf(x_k)\tag{53}$$

7

$$e = C \cdot h^n \Leftrightarrow \log e = \log C + n \log h \tag{54}$$

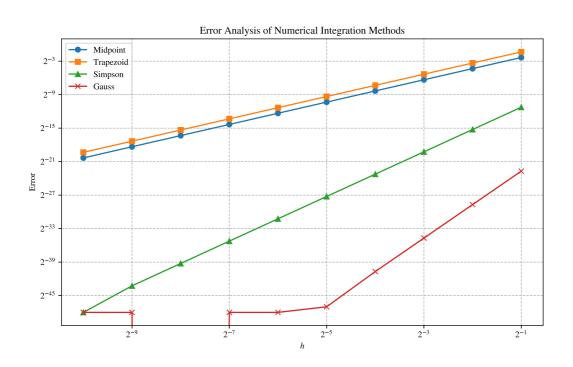
# Output

#### # (C, order)

- 0.7923224542232744 1.9992933732410973
- 1.587137649897742 1.9995965300241643
- 0.007343895022866783 4.041227723532598
- 9.372066198851523e-06 5.99563201077984

That is, the dominant term in error of composite midpoint or Trapezoid rules is  $O(h^2)$ , the dominant term in error of the composite Simpson rule is  $O(h^4)$  and the dominant term in error of the composite 3-points Gauss rules is  $O(h^6)$ .

#### Result



#### 8.12

Average them,

$$\frac{\frac{f(x+h)-f(x)}{h} + \frac{f(x)-f(x-h)}{h}}{2} = \frac{f(x+h) - f(x-h)}{2h}$$
 (55)

By Taylor series,

$$f(x+h) = f(x) + f'(x)h + \frac{1}{2}f''(x)h^2 + O(h^3)$$
(56)

$$f(x-h) = f(x) - f'(x)h + \frac{1}{2}f''(x)h^2 + O(h^3)$$
(57)

Thus,

$$\frac{f(x+h) - f(x-h)}{2h} = f'(x) + O(h^2)$$
 (58)

, which is second-order accurate.

## 8.13

$$f(x+h) = f(x) + f'(x)h + \frac{1}{2}f''(x)h^2 + O(h^3)$$
(59)

$$f(x+2h) = f(x) + 2f'(x)h + 2f''(x)h^{2} + O(h^{3})$$
(60)

Consider the linear composition af(x) + bf(x+h) + cf(x+2h),

$$\frac{b}{2} + 2c = 0 (61)$$

Suppose b=4, c=-1, then a=-b-c=-3 and b+2c=2.

That is,

$$f'(x) = \frac{-3f(x) + 4f(x+h) - f(x+2h)}{2h} + O(h^2)$$
 (62)

, which is second-order accurate.