



# Theoretical Problems

## Numerical analysis 2022

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# Chapter 1 Solving Nonlinear Equations

**Problem 1.1** Consider the bisection method starting with the initial interval  $[1.5, 3.5]$ . In the following questions "the interval" refers to the bisection interval whose width changes across different loops.

- What is the width of the interval at the  $n$ th step?
- What is the maximum possible distance between the root  $r$  and the midpoint of the interval?

**Solution** Note that the interval's width is multiplied by  $\frac{1}{2}$  at each step, and the initial width is 2, hence the width after the  $n$ th step is  $\frac{1}{2^{n-1}}$ .

The maximum distance is not greater than 1 obviously.

Since the loop terminated when  $|f(c)| < \varepsilon$ , we could construct an increasing function  $f$  whose root is  $1.5 + \delta$ , and  $|f(x)| < \varepsilon$  everywhere, hence the bisection loop will terminate at first step, the distance between midpoint and root is  $1 - \delta$ . Let  $\delta \rightarrow 0^+$ , we know the distance could be infinitely close to 1.

**Problem 1.2** In using the bisection algorithm with its initial interval as  $[a_0, b_0]$  with  $a_0 > 0$ , we want to determine the root with its relative error no greater than  $\varepsilon$ . Prove that this goal of accuracy is guaranteed by the following choice of the number of steps,

$$n \geq \frac{\log(b_0 - a_0) - \log \varepsilon - \log a_0}{\log 2} - 1$$

**Solution** Suppose the root is  $r \geq a_0$ . The relative error after the  $n$ th step is

$$\frac{|r - c_n|}{|r|} \quad (1.1)$$

The following inequations hold

$$\frac{|r - c_n|}{|r|} \leq \frac{\frac{1}{2}(b_n - a_n)}{r} \leq \frac{\frac{1}{2}(b_n - a_n)}{a_0} = \frac{b_0 - a_0}{a_0 2^{n+1}} \quad (1.2)$$

Hence when (1.1) holds, we have

$$\begin{aligned} (n+1) \log 2 &\geq \log(b_0 - a_0) - \log \varepsilon - \log a_0 \\ \implies \log 2^{n+1} &\geq \log \left( \frac{b_0 - a_0}{\varepsilon a_0} \right) \\ \implies 2^{n+1} &\geq \frac{b_0 - a_0}{\varepsilon a_0} \implies \frac{b_0 - a_0}{a_0 2^{n+1}} \leq \varepsilon \end{aligned}$$

Hence the conclusion is proved by (1.2).

**Problem 1.3** Perform four iterations of Newton's method for the polynomial equation  $p(x) = 4x^3 - 2x^2 + 3 = 0$  with the starting point  $x_0 = -1$ . Use a hand calculator and organize results of the iterations in a table.

**Solution** Firstly we derivate  $p(x)$

$$p'(x) = 12x^2 - 4x$$

The results are shown as the following table.

$n$	$x_n$	$p(x_n)$	$p'(x_n)$	$x_n - \frac{f(x_n)}{f'(x_n)}$
0	-1	-3	16	-0.8125
1	-0.8125	-0.46582	11.1719	-0.770804
2	-0.770804	-0.0201359	10.2129	-0.768832
3	-0.768832	-3.98011e-05	10.1686	-0.768828
4	-0.768828			

**Problem 1.4** Consider a variation of Newton's method in which only the derivative at  $x_0$  is used,

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_0)} \quad (1.3)$$

Find  $C$  and  $s$  such that

$$e_{n+1} = C e_n^s$$

where  $e_n$  is the error of Newton's method at step  $n$ ,  $s$  is a constant, and  $C$  may depend on  $x_n$ , the given function  $f$  and its derivatives.

**Solution** Assume the root is  $r$ , then  $e_n = x_n - r$ . Let  $g(x) = f(r + x)$ . By (1.3), we derive

$$e_{n+1} = e_n - \frac{g(e_n)}{g'(e_0)} = \left(1 - \frac{g(e_n)}{e_n g'(e_0)}\right) e_n$$

Let  $C(n) = 1 - \frac{g(e_n)}{e_n g'(e_0)}$  and  $s = 1$ , we got  $e_{n+1} = C(n) e_n$ , and

$$\lim_{n \rightarrow \infty} C(n) = 1 - \frac{g'(0)}{g'(e_0)} = 1 - \frac{f'(r)}{f'(x_0)}$$

**Problem 1.5** Within  $(-\frac{\pi}{2}, \frac{\pi}{2})$ , will the iteration  $x_{n+1} = \tan^{-1} x_n$  converge?

**Solution** As we all know that  $0 < \tan^{-1} x < x$  ( $x > 0$ ), so if  $x_0 > 0$ , we derive

$$0 < x_{n+1} = \tan^{-1} x_n < x_n$$

And sequence  $\{x_n\}$  has lower bound 0, so  $\{x_n\}$  is convergent by monotonic sequence theorem.

For  $x_0 < 0$ ,  $\{-x_n\}$  is convergent by the discussion above, hence  $\{x_n\}$  is convergent.

For  $x_0 = 0$ , clearly  $x_n = 0$  ( $\forall n$ ).

**Problem 1.6** Let  $p > 1$ . What is the value of the following continued fraction?

$$x = \frac{1}{p + \frac{1}{p + \frac{1}{p + \dots}}}$$

Prove that the sequence of values converges.

**Solution** We construct a sequence by  $x_1 = \frac{1}{p}$  and  $x_{n+1} = \frac{1}{p+x_n}$  ( $n \geq 1$ ), then  $x = \lim_{n \rightarrow \infty} x_n$  if it exists.

Consider function  $g(x) = \frac{1}{p+x}$ , clearly  $g(x) \in [0, 1]$  for all  $x \in [0, 1]$ . And

$$\lambda = \max_{x \in [0, 1]} |g'(x)| = \max_{x \in [0, 1]} \frac{1}{(x+p)^2} = \frac{1}{p^2} < 1$$

Hence  $g$  is a contraction in  $[0, 1]$ , and consider equation

$$x = g(x) = \frac{1}{p+x}$$

the roots are  $\frac{-p \pm \sqrt{p^2+4}}{2}$ , hence  $g$  has unique fixed-point  $\alpha = \frac{-p + \sqrt{p^2+4}}{2}$  in  $[0, 1]$ .

Recall that  $x_1 = \frac{1}{p} \in [0, 1]$ , and  $x_{n+1} = g(x_n)$ . By Theorem 1.38,  $\{x_n\}$  converges and  $x = \lim_{n \rightarrow \infty} x_n = \alpha$ .

**Problem 1.7** What happens in problem 1.2 if  $a_0 < 0 < b_0$ ? Derive an inequality of the number of steps similar to that in problem 1.2. In this case, is the relative error still an appropriate measure?

**Solution** In this problem we let the absolutely error  $|r - c_n| < \delta$ , we derive

$$|r - c_n| \leq \frac{1}{2} (b_n - a_n) = \frac{b_0 - a_0}{2^{n+1}} \quad (1.4)$$

It is sufficient to let  $\frac{b_0 - a_0}{2^{n+1}} < \delta$ , hence  $n \geq \frac{\log(b_0 - a_0) - \log \delta}{\log 2} - 1$ .

We can't use relative error since  $r$  might be zero.

## Chapter 2 Polynomial Interpolation

**Problem 2.1** For  $f \in \mathcal{C}^2[x_0, x_1]$  and  $x \in (x_0, x_1)$ , linear interpolation of  $f$  at  $x_0$  and  $x_1$  yields

$$f(x) - p_1(f; x) = \frac{f''(\xi(x))}{2}(x - x_0)(x - x_1) \quad (2.1)$$

Consider the case  $f(x) = \frac{1}{x}$ ,  $x_0 = 1$ ,  $x_1 = 2$ .

- Determine  $\xi(x)$  explicitly.
- Extend the domain of  $\xi$  continuously from  $(x_0, x_1)$  to  $[x_0, x_1]$ . Find  $\max \xi(x)$ ,  $\min \xi(x)$  and  $\max f''(\xi(x))$ .

**Solution**

1. The Lagrange's formula yields

$$p_1(f; x) = \frac{(x - 2)}{(1 - 2)} + \frac{1}{2} \times \frac{(x - 1)}{(2 - 1)} = -\frac{1}{2}x + \frac{3}{2}$$

Substitute it to (2.1), with  $f''(x) = 2x^{-3}$ , yield

$$\frac{1}{x} + \frac{1}{2}x - \frac{3}{2} = (x - 1)(x - 2)\xi^{-3}(x)$$

The result follows from it:

$$\xi(x) = \sqrt[3]{2x}$$

2.  $\xi(x)$  is increasing in  $[1, 2]$ , hence

$$\max \xi(x) = \xi(2) = \sqrt[3]{4}, \quad \min \xi(x) = \xi(1) = \sqrt[3]{2}$$

Also

$$f''(\xi(x)) = 2 \left( \sqrt[3]{2x} \right)^{-3} = \frac{1}{x}$$

is decreasing in  $[1, 2]$ , hence

$$\max f''(\xi(x)) = f''(\xi(1)) = 1$$

**Problem 2.2** Let  $\mathbb{P}_m^+$  be the set of all polynomials of degree  $\leq m$  that are non-negative on the real line,

$$\mathbb{P}_m^+ = \{p : p \in \mathbb{P}_m, \forall x \in \mathbb{R}, p(x) \geq 0\}$$

Find  $p \in \mathbb{P}_{2n}^+$  such that  $p(x_i) = f_i$  for  $i = 0, 1, \dots, n$  where  $f_i \geq 0$  and  $x_i$  are distinct points on  $\mathbb{R}$ .

**Solution** Let  $q(x) \in \mathbb{P}_n$  be the unique interpolation polynomial satisfies

$$q(x_i) = \sqrt{f_i}, \quad i = 0, 1, \dots, n$$

Let  $p(x) = q^2(x)$ , then  $p(x) \in \mathbb{P}_{2n}^+$  and

$$p(x_i) = q^2(x_i) = f_i, \quad i = 0, 1, \dots, n$$

Hence  $p(x)$  is what we need. The Lagrange's interpolation formula yields:

$$p(x) = \left( \sum_{i=0}^n \sqrt{f_i} \prod_{j=0, j \neq i}^n \frac{x - x_j}{x_i - x_j} \right)^2$$

**Problem 2.3** Consider  $f(x) = e^x$ .

- Prove by induction that

$$\forall t \in \mathbb{R}, \quad f[t, t+1, \dots, t+n] = \frac{(e-1)^n}{n!} e^t \quad (2.2)$$

- From Corollary 2.22 we know

$$\exists \xi \in (0, n) \text{ s.t. } f[0, 1, \dots, n] = \frac{1}{n!} f^{(n)}(\xi) \quad (2.3)$$

Determine  $\xi$  from the above two equations. Is  $\xi$  located to the left or to the right of the midpoint  $n/2$ .

### Solution

1. The Lagrange's formula yields

$$p(f; x) = \sum_{k=0}^n e^{t+k} \prod_{j=0, j \neq k}^n \frac{x - x_j}{x_k - x_j} = e^t \sum_{k=0}^n e^k \frac{(-1)^{n-k} \prod_{j=0, j \neq k}^n (x - x_j)}{k!(n-k)!}$$

Hence

$$f[t, t+1, \dots, t+n] = e^t \sum_{k=0}^n \frac{(-1)^{n-k} e^k}{k!(n-k)!} = \frac{e^t}{n!} \sum_{k=0}^n \binom{n}{k} (-1)^{n-k} e^k = \frac{(e-1)^n}{n!} e^t$$

2. Let  $t = 0$  in (2.2) and yield

$$f[0, 1, \dots, n] = \frac{(e-1)^n}{n!}$$

Substitute it to (2.3), with  $f^{(n)}(x) = e^x$ , yield

$$\frac{(e-1)^n}{n!} = \frac{e^\xi}{n!}$$

The result follows from it:

$$\xi = n \ln(e-1) > \frac{n}{2}$$

Hence  $\xi$  is located to the right of the midpoint.

**Problem 2.4** Consider  $f(0) = 5$ ,  $f(1) = 3$ ,  $f(3) = 5$ ,  $f(4) = 12$ .

- Use the Newton's formula to obtain  $p_3(f; x)$ ;
- The data suggests that  $f$  has a minimum in  $x \in (1, 3)$ . Find an approximate value for the location  $x_{\min}$  of the minimum.

### Solution

1. The result follows from Newton's interpolation formula:

$$p_3(f; x) = 5 - 2x + x(x-1) + \frac{1}{4}x(x-1)(x-3)$$

Transform it into the canonical form:

$$p_3(f; x) = \frac{1}{4}x^3 - \frac{9}{4}x + 5$$

2. Firstly, calculate the derivative of  $p_3(f; x)$ :

$$p'_3(f; x) = \frac{3}{4}x^2 - \frac{9}{4}$$

The first-order necessary condition  $p'_3(f; x) = 0$  yields that

$$x_{\text{extreme}} = \pm\sqrt{3}$$

In  $x \in (1, 3)$ , the extreme point might be  $x^* = \sqrt{3}$ . The second-order condition shows that

$$p''_3(f; x^*) = \frac{3}{2}x^* = \frac{3\sqrt{3}}{2} > 0$$

Hence  $x^*$  is the minimum, and  $x_{\min} = \sqrt{3} \approx 1.73205$ .

**Problem 2.5** Consider  $f(x) = x^7$ .

- Compute  $f[0, 1, 1, 1, 2, 2]$ .



- We know that this divided difference is expressible in terms of the 5th derivative of  $f$  evaluated at some  $\xi \in (0, 2)$ . Determine  $\xi$ .

**Solution**

1. Solve the Hermite's interpolation with a difference table. The result of Newton's form follows:

$$p(x) = x + 6x(x-1) + 15x(x-1)^2 + 42x(x-1)^3 + 30x(x-1)^3(x-2)$$

Hence

$$f[0, 1, 1, 1, 2, 2] = 30$$

2. The 5th derivative of  $f$  is

$$f^{(5)}(x) = 2520x^2$$

Then  $\frac{f^{(5)}(x)}{5!} = f[0, 1, 1, 1, 2, 2]$  yields

$$\frac{2520}{5!}\xi^2 = 30 \quad \implies \quad \xi = \sqrt{\frac{10}{7}} \approx 1.42857 \in (0, 2)$$

**Problem 2.6**  $f$  is a function on  $[0, 3]$  for which one knows that

$$f(0) = 1, \quad f(1) = 2, \quad f'(1) = -1, \quad f(3) = f'(3) = 0$$

- Estimate  $f(2)$  using Hermite's interpolation.
- Estimate the maximum possible error of the above answer if one knows, in addition, that  $f \in C^5[0, 3]$  and  $|f^{(5)}(x)| \leq M$  on  $[0, 3]$ . Express the answer in terms of  $M$ .

**Solution**

1. The Hermite's interpolation gives that

$$p(x) = 1 + x - 2x(x-1) + \frac{2}{3}x(x-1)^2 - \frac{5}{36}x(x-1)^2(x-3)$$

Hence, estimate  $f(2)$  as

$$f(2) \approx p(2) = \frac{11}{18} \approx 0.611111$$

2. Theorem 2.35 gives that

$$f(x) - p(x) = \frac{f^{(5)}(\xi)}{120}x(x-1)^2(x-3)^2$$

The result follows directly:

$$|f(2) - p(2)| = \left| \frac{f^{(5)}(\xi)}{60} \right| \leq \frac{M}{60}$$

**Problem 2.7** Define forward difference by

$$\Delta f(x) = f(x+h) - f(x), \quad \Delta^{k+1}f(x) = \Delta \Delta^k f(x) = \Delta^k f(x+h) - \Delta^k f(x)$$

and backward difference by

$$\nabla f(x) = f(x) - f(x-h), \quad \nabla^{k+1}f(x) = \nabla \nabla^k f(x) = \nabla^k f(x) - \nabla^k f(x-h)$$

Prove

$$\Delta^k f(x) = k!h^k f[x_0, x_1, \dots, x_k] \tag{2.4}$$

$$\nabla^k f(x) = k!h^k f[x_0, x_{-1}, \dots, x_{-k}] \tag{2.5}$$

where  $x_j = x + jh$ .

**Solution** The Lagrange's interpolation formula yields

$$f[x_0, x_1, \dots, x_k] = \sum_{i=0}^k f(x_i) \frac{1}{\prod_{j=1, j \neq i}^k (x_i - x_j)} = \sum_{i=0}^k \frac{(-1)^{k-i} f(x + ih)}{h^k i! (k-i)!}$$

It yields an equivalent form of (2.4):

$$\Delta^k f(x) = k! h^k f[x_0, x_1, \dots, x_k] = \sum_{i=0}^k \binom{k}{i} (-1)^{k-i} f(x + ih) \quad (2.6)$$

Now prove (2.6) by an induction. For  $k = 1$ , it could be verified directly:

$$\binom{1}{0} (-1)^{1-0} f(x) + \binom{1}{1} (-1)^{1-1} f(x + h) = f(x + h) - f(x) = \Delta f(x)$$

Suppose (2.6) holds for some  $k \geq 1$ , then

$$\begin{aligned} \Delta^{k+1} f(x) &= \Delta \left( \sum_{i=0}^k \binom{k}{i} (-1)^{k-i} f(x + ih) \right) \\ &= \sum_{i=0}^k \binom{k}{i} (-1)^{k-i} f(x + (i+1)h) - \sum_{i=0}^k \binom{k}{i} (-1)^{k-i} f(x + ih) \\ &= f(x + (k+1)h) - (-1)^k f(x) + \sum_{i=1}^k \left( \binom{k}{i-1} (-1)^{k+1-i} f(x + ih) - \binom{k}{i} (-1)^{k-i} f(x + ih) \right) \\ &= f(x + (k+1)h) + (-1)^{k+1} f(x) + \sum_{i=1}^k (-1)^{k+1-i} f(x + ih) \left( \binom{k}{i-1} + \binom{k}{i} \right) \\ &= f(x + (k+1)h) + (-1)^{k+1} f(x) + \sum_{i=1}^k \binom{k+1}{i} (-1)^{k+1-i} f(x + ih) \\ &= \sum_{i=0}^{k+1} \binom{k+1}{i} (-1)^{k+1-i} f(x + ih) \end{aligned}$$

It shows that (2.6) holds for  $(k+1)$ . Hence (2.4) is proved by induction. Now we prove that

$$\Delta^k f(x) = \nabla^k f(x + kh) \quad (2.7)$$

by an induction. For  $k = 1$ , it could be verified directly:

$$\Delta f(x) = f(x + h) - f(x) = \nabla f(x + h)$$

Suppose (2.7) holds for some  $k \geq 1$ , then

$$\begin{aligned} \Delta^{k+1} f(x) &= \Delta \left( \Delta^k f(x) \right) = \Delta \left( \nabla^k f(x + kh) \right) = \nabla^k f(x + (k+1)h) - \nabla^k f(x + kh) \\ &= \nabla \left( \nabla^k f(x + (k+1)h) \right) = \nabla^{k+1} f(x + (k+1)h) \end{aligned}$$

Hence (2.7) is proved by induction. Finally, (2.5) follows immediately from (2.4), (2.7) and Corollary 2.15.

**Problem 2.8** Assume  $f$  is differentiable at  $x_0$ . Prove

$$\frac{\partial}{\partial x_0} f[x_0, x_1, \dots, x_n] = f[x_0, x_0, x_1, \dots, x_n] \quad (2.8)$$

What about the partial derivate with respect to one of the other variables?

**Solution** Firstly, follows from Definition 2.34, we have

$$\frac{\partial}{\partial x_0} f[x_0] = f'(x_0) = f[x_0, x_0]$$

Prove (2.8) by an induction on  $n$ . For  $n = 1$ , verify it directly:

$$\begin{aligned}
\frac{\partial}{\partial x_0} f[x_0, x_1] &= \frac{\partial}{\partial x_0} \left( \frac{f[x_1] - f[x_0]}{x_1 - x_0} \right) \\
&= \frac{-(x_1 - x_0) \frac{\partial}{\partial x_0} f[x_0] + f[x_1] - f[x_0]}{(x_1 - x_0)^2} \\
&= \frac{f[x_0, x_1] - f[x_0, x_0]}{x_1 - x_0} \\
&= f[x_0, x_0, x_1]
\end{aligned}$$

Suppose (2.8) holds for some  $n \geq 1$ , then

$$\begin{aligned}
\frac{\partial}{\partial x_0} f[x_0, x_1, \dots, x_{n+1}] &= \frac{\partial}{\partial x_0} \left( \frac{f[x_1, \dots, x_{n+1}] - f[x_0, \dots, x_n]}{x_{n+1} - x_0} \right) \\
&= \frac{-(x_{n+1} - x_0) \frac{\partial}{\partial x_0} f[x_0, x_1, \dots, x_n] + f[x_1, \dots, x_{n+1}] - f[x_0, \dots, x_n]}{(x_{n+1} - x_0)^2} \\
&= \frac{-(x_{n+1} - x_0) f[x_0, x_0, x_1, \dots, x_n] + f[x_1, \dots, x_{n+1}] - f[x_0, \dots, x_n]}{(x_{n+1} - x_0)^2} \\
&= \frac{-f[x_0, x_0, x_1, \dots, x_n] + f[x_0, x_1, \dots, x_{n+1}]}{x_{n+1} - x_0} \\
&= f[x_0, x_0, x_1, \dots, x_{n+1}]
\end{aligned}$$

It shows that (2.8) holds for  $(n+1)$ , hence proved. Moreover, the order of  $x_0, \dots, x_n$  is not important, hence

$$\frac{\partial}{\partial x_j} f[x_0, x_1, \dots, x_n] = f[x_0, \dots, x_{j-1}, x_j, x_j, x_{j+1}, \dots, x_n], \quad \forall j = 0, \dots, n$$

**Problem 2.9** (A min-max problem) For  $n \in \mathbb{N}^+$ , determine

$$\min_{x \in [a, b]} \max_{a_i \in \mathbb{R}} |a_0 x^n + a_1 x^{n-1} + \dots + a_n| \quad (2.9)$$

where  $a_0 \neq 0$  is fixed and the minimum is taken over all  $a_i \in \mathbb{R}$ ,  $i = 1, 2, \dots, n$ .

**Solution** The map

$$p(x) \mapsto q(x) = \frac{1}{a_0} p \left( a + \frac{b-a}{2}(x+1) \right)$$

yields a bisection relation between polynomials of degree  $n$  defines in  $[a, b]$  with leading coefficient  $a_0$  and polynomials of degree  $n$  defines in  $[0, 1]$  with leading coefficient 1. Chebyshev's Theorem gives that

$$\forall q \in \tilde{\mathbb{P}}_n, \quad \max_{x \in [-1, 1]} \left| \frac{T_n(x)}{2^{n-1}} \right| \leq \max_{x \in [-1, 1]} |q(x)|$$

where  $T_n$  is the Chebyshev's polynomial of order  $n$ . Hence the solution of the min-max problem  $p_{\min}(x)$  satisfies

$$\frac{1}{a_0} p_{\min} \left( a + \frac{b-a}{2}(x+1) \right) = \frac{T_n(x)}{2^{n-1}}$$

The result follows immediately:

$$p_{\min}(x) = \frac{a_0}{2^{n-1}} T_n \left( \frac{2}{b-a}(x-a) - 1 \right)$$

The min value in (2.8) is  $\frac{|a_0|}{2^{n-1}}$ .

**Problem 2.10** (Imitate the proof of Chebyshev's Theorem) Express the Chebyshev polynomial of degree  $n \in \mathbb{N}$  as a polynomial  $T_n$  and change its domain from  $[-1, 1]$  to  $\mathbb{R}$ . For a fixed  $a > 1$ , define  $\mathbb{P}_n^a := \{p \in \mathbb{P}_n : p(a) = 1\}$  and a polynomial  $\hat{p}_n(x) \in \mathbb{P}_n^a$ ,

$$\hat{p}_n(x) := \frac{T_n(x)}{T_n(a)}$$



Prove

$$\forall p \in \mathbb{P}_n^a, \quad \|\hat{p}_n\|_\infty \leq \|p\|_\infty$$

where the max-norm of a function  $f : \mathbb{R} \rightarrow \mathbb{R}$  is defined as  $\|f\|_\infty = \max_{x \in [-1, 1]} |f(x)|$ .

**Solution** First we know that  $\|\hat{p}_n\|_\infty = \frac{1}{|T_n(a)|}$ . And by the property of  $T_n$  we have

$$\hat{p}_n(x'_k) = \frac{(-1)^k}{T_n(a)} \quad \text{for } x'_k = \cos \frac{k}{n} \pi, \quad k = 0, 1, \dots, n$$

Now we prove the conclusion by using reduction to absurdity. Suppose that:

$$\exists p \in \mathbb{P}_n^a, \quad \text{s.t.} \quad \|p\|_\infty < \frac{1}{|T_n(a)|}$$

Let  $q(x) = p(x) - \hat{p}_n(x) \in \mathbb{P}_n$ , then  $q(a) = 0$ . And

$$q(x'_k) = p(x'_k) - \frac{(-1)^k}{T_n(a)}, \quad k = 0, 1, \dots, n$$

We have  $\text{sgn}(q(x'_k)) \neq \text{sgn}(q(x'_{k-1}))$  for  $k = 1, \dots, n$  since  $\|p\|_\infty < \frac{1}{|T_n(a)|}$ . By the continuity of  $q$ ,

$$\exists -1 = x_n < \xi_n < x_{n-1} < \dots < x_1 < \xi_1 < x_0 = 1, \quad \text{s.t.} \quad q(\xi_1) = \dots = q(\xi_n) = 0$$

However,  $q(a) = 0$  and  $a > 1$  shows that  $q$  has at least  $n + 1$  zero points, that contradict to  $q \in \mathbb{P}_n$ .

**Problem 2.11** Prove Lemma 2.48:

$$\forall k = 0, 1, \dots, n, \forall t \in (0, 1), \quad b_{n,k}(t) > 0 \quad (2.10)$$

$$\sum_{k=0}^n b_{n,k}(t) = 1 \quad (2.11)$$

$$\sum_{k=0}^n k b_{n,k}(t) = nt \quad (2.12)$$

$$\sum_{k=0}^n (k - nt)^2 b_{n,k}(t) = nt(1 - t) \quad (2.13)$$

where

$$b_{n,k}(t) = \binom{n}{k} t^k (1 - t)^{n-k}$$

**Solution** (2.10) is clearly since  $t \in (0, 1)$ .

By the Binomial Theorem we have:

$$1 = (t + (1 - t))^n = \sum_{k=0}^n \binom{n}{k} t^k (1 - t)^{n-k} = \sum_{k=0}^n b_{n,k}(t)$$

Hence (2.11) is proved.

Again, by the Binomial Theorem we have:

$$(p + q)^n = \sum_{k=0}^n \binom{n}{k} p^k q^{n-k}$$

Partial derivate with respect to  $p$  to both sides yields:

$$n(p + q)^{n-1} = \sum_{k=0}^n \binom{n}{k} k p^{k-1} q^{n-k}$$

Multiple a  $p$  to both sides, yield

$$np(p + q)^{n-1} = \sum_{k=0}^n \binom{n}{k} k p^k q^{n-k} \quad (2.14)$$

---

Now take  $p = t$  and  $q = 1 - t$ , yield

$$nt = \sum_{k=0}^n \binom{n}{k} kt^k(1-t)^{n-k} = \sum_{k=0}^n kb_{n,k}(t)$$

Hence (2.12) is proved.

Follows from (2.14), partial derivate again with respect to  $p$  to both sides yields:

$$n(p+q)^{n-1} + n(n-1)p(p+q)^{n-2} = \sum_{k=0}^n \binom{n}{k} k^2 p^{k-1} q^{n-k}$$

Multiple a  $p$  to both sides, yield

$$np(p+q)^{n-1} + n(n-1)p^2(p+q)^{n-2} = \sum_{k=0}^n \binom{n}{k} k^2 p^k q^{n-k}$$

Now take  $p = t$  and  $q = 1 - t$ , yield

$$nt + n(n-1)t^2 = \sum_{k=0}^n k^2 b_{n,k}(t)$$

By (2.11), (2.12) and the result above, we got:

$$\begin{aligned} \sum_{k=0}^n (k - nt)^2 b_{n,k}(t) &= \sum_{k=0}^n k^2 b_{n,k}(t) - 2nt \sum_{k=0}^n kb_{n,k}(t) + (nt)^2 \sum_{k=0}^n b_{n,k}(t) \\ &= nt + n(n-1)t^2 - 2(nt)^2 + (nt)^2 = nt - nt^2 = nt(1-t) \end{aligned}$$

Hence (2.13) is proved.

## Chapter 3 Splines

**Problem 3.1** Consider  $s \in \mathbb{S}_3^2$  on  $[0, 2]$ :

$$s(x) = \begin{cases} p(x) & \text{if } x \in [0, 1], \\ (2-x)^3 & \text{if } x \in [1, 2]. \end{cases}$$

Determine  $p \in \mathbb{P}_3$  such that  $s(0) = 0$ . Is  $s(x)$  a natural cubic spline?

**Solution**  $p(x)$  should satisfy the following condition:

$$p(0) = 0, \quad p(1) = 1, \quad p'(1) = -3, \quad p''(1) = 6.$$

Use Hermite interpolation, we got

$$p(x) = 7x^3 - 18x^2 + 12x.$$

$s(x)$  is not a natural cubic spline since  $s''(0) = -36 \neq 0$ .

**Problem 3.2** Given  $f_i = f(x_i)$  of some scalar function at points  $a = x_1 < x_2 < \dots < x_n = b$ , we consider interpolating  $f$  on  $[a, b]$  with a quadratic spline  $s \in \mathbb{S}_2^1$ .

- (a) Why is an additional condition needed to determine  $s$  uniquely?
- (b) Define  $m_i = s'(x_i)$  and  $p_i = s|_{[x_i, x_{i+1}]}$ . Determine  $p_i$  in terms of  $f_i, f_{i+1}$  and  $m_i$  for  $i = 1, 2, \dots, n-1$ .
- (c) Suppose  $m_1 = f'(a)$  is given. Show how  $m_2, m_3, \dots, m_{n-1}$  can be computed.

**Solution** (a) Denote  $p_i = s|_{[x_i, x_{i+1}]} \in \mathbb{P}_2$ , then there're  $3(n-1)$  unknown coefficients in  $p_1, \dots, p_{n-1}$ . First,

$$p_i(x_i) = f_i, \quad p_i(x_{i+1}) = f_{i+1}, \quad i = 1, \dots, n-1$$

gives  $2(n-1)$  equations. And

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

gives  $n-2$  equations. Now there're  $3(n-1)$  unknowns and  $3(n-1) - 1$  equations.

Hence an additional condition is needed.

(b) Suppose that  $p_i(x) = a_i x^2 + b_i x + c_i$ . The conditions give that:

$$\begin{cases} x_i^2 a_i + x_i b_i + c_i = f_i \\ x_{i+1}^2 a_i + x_{i+1} b_i + c_i = f_{i+1} \\ 2x_i a_i + b_i = m_i \end{cases}$$

Solve the linear equation of  $a_i, b_i$  and  $c_i$ , we got

$$\begin{aligned} a_i &= \frac{f_{i+1} - f_i}{(x_{i+1} - x_i)^2} - \frac{m_i}{x_{i+1} - x_i} \\ b_i &= \frac{m_i(x_{i+1} + x_i)}{x_{i+1} - x_i} - \frac{2x_i(f_{i+1} - f_i)}{(x_{i+1} - x_i)^2} \\ c_i &= f_i + \frac{x_i^2(f_{i+1} - f_i)}{(x_{i+1} - x_i)^2} - \frac{m_i x_i x_{i+1}}{x_{i+1} - x_i} \end{aligned}$$

Hence  $p_i$  is determined.

(c) Determine  $p_1$  in terms of  $f_1, f_2$  and  $m_1$ . Let  $m_2 = p'_1(x_2)$ .

Determine  $p_2$  in terms of  $f_2, f_3$  and  $m_2$ . Let  $m_3 = p'_2(x_3)$ .

$\vdots$

Determine  $p_{n-1}$  in terms of  $f_{n-1}, f_n$  and  $m_{n-1}$ .

**Problem 3.3** Let  $s_1(x) = 1 + c(x+1)^3$  where  $x \in [-1, 0]$  and  $c \in \mathbb{R}$ . Determine  $s_2(x)$  on  $[0, 1]$  such that

$$s(x) = \begin{cases} s_1(x) & \text{if } x \in [-1, 0] \\ s_2(x) & \text{if } x \in [0, 1] \end{cases}$$

is a natural cubic spline on  $[-1, 1]$  with knots  $-1, 0, 1$ . How must  $c$  be chosen if one wants  $s(1) = -1$ ?

**Solution** Let  $s_2(x) = \alpha x^3 + \beta x^2 + \gamma x + \theta$ . The following conditions should be satisfied.

$$s_2(0) = s_1(0) = 1 + c, \quad s_2'(0) = s_1'(0) = 3c, \quad s_2''(0) = s_1''(0) = 6c, \quad s_2(1) = s(1) = -1, \quad s_2''(1) = 0.$$

And these conditions give that:

$$\begin{cases} \theta = 1 + c \\ \gamma = 3c \\ 2\beta = 6c \\ \alpha + \beta + \gamma + \theta = -1 \\ 6\alpha + 2\beta = 0 \end{cases}.$$

Solve the linear system, and we got that  $c = -\frac{1}{3}$ .

**Problem 3.4** Consider  $f(x) = \cos\left(\frac{\pi}{2}x\right)$  with  $x \in [-1, 1]$ .

- Determine the natural cubic spline interpolant to  $f$  on knots  $-1, 0, 1$ .
- As discussed in the class, natural cubic splines have the minimal total bending energy. Verify this by tanking  $g(x)$  be (i) the quadratic polynomial that interpolates  $f$  at  $-1, 0, 1$ , and (ii)  $f(x)$ .

**Solution** (a) The natural cubic spline interpolant to  $f$  on knots  $-1, 0, 1$  is

$$s(x) = \begin{cases} -\frac{1}{2}x^3 - \frac{3}{2}x^2 + 1 & \text{if } x \in [-1, 0], \\ \frac{1}{2}x^3 - \frac{3}{2}x^2 + 1 & \text{if } x \in [0, 1]. \end{cases}$$

(b) The bending energy of  $s$  is

$$\int_{-1}^1 [s''(x)]^2 dx = \int_{-1}^0 (-3x - 3)^2 dx + \int_0^1 (3x - 3)^2 dx = 6.$$

The quadratic polynomial that interpolates  $f$  at  $-1, 0, 1$  is

$$p(x) = -x^2 + 1.$$

And its bending energy is

$$\int_{-1}^1 [p''(x)]^2 dx = \int_{-1}^1 4 dx = 8 > 6.$$

The bending energy of  $f$  is

$$\int_{-1}^1 [f''(x)]^2 dx = \int_{-1}^1 \left[ -\frac{\pi^2}{4} \cos\left(\frac{\pi}{2}x\right) \right]^2 dx = \frac{\pi^4}{16} \approx 6.0881 > 6.$$

**Problem 3.5** The quadratic B-spline  $B_i^2(x)$ .

- Derive the same explicit expression of  $B_i^2(x)$  as that in the notes from the recursive definition of B-splines and the hat function.
- Verify that  $\frac{d}{dx} B_i^2(x)$  is continuous at  $t_i$  and  $t_{i+1}$ .
- Show that only one  $x^* \in (t_{i-1}, t_{i+1})$  satisfies  $\frac{d}{dx} B_i^2(x^*) = 0$ . Express  $x^*$  in terms of the knots within the interval of support.
- Consequently, show  $B_i^2(x) \in [0, 1]$ .
- Plot  $B_i^2(x)$  for  $t_i = i$ .

## Solution

(a) See that

$$B_i^1(x) = \begin{cases} \frac{x-t_{i-1}}{t_i-t_{i-1}} & x \in (t_{i-1}, t_i], \\ \frac{t_{i+1}-x}{t_{i+1}-t_i} & x \in (t_i, t_{i+1}], \\ 0 & \text{otherwise.} \end{cases} \quad B_{i+1}^1(x) = \begin{cases} \frac{x-t_i}{t_{i+1}-t_i} & x \in (t_i, t_{i+1}], \\ \frac{t_{i+2}-x}{t_{i+2}-t_{i+1}} & x \in (t_{i+1}, t_{i+2}], \\ 0 & \text{otherwise.} \end{cases}$$

And by the recursive definition we have

$$B_i^2(x) = \frac{x-t_{i-1}}{t_{i+1}-t_{i-1}} B_i^1(x) + \frac{t_{i+2}-x}{t_{i+2}-t_i} B_{i+1}^1(x)$$

For  $x \in (t_{i-1}, t_i]$ ,

$$B_i^2(x) = \frac{x-t_{i-1}}{t_{i+1}-t_{i-1}} \cdot \frac{x-t_{i-1}}{t_i-t_{i-1}} + \frac{t_{i+2}-x}{t_{i+2}-t_i} \cdot 0 = \frac{(x-t_{i-1})^2}{(t_{i+1}-t_{i-1})(t_i-t_{i-1})}.$$

For  $x \in (t_i, t_{i+1}]$ ,

$$B_i^2(x) = \frac{(x-t_{i-1})(t_{i+1}-x)}{(t_{i+1}-t_{i-1})(t_{i+1}-t_i)} + \frac{(t_{i+2}-x)(x-t_i)}{(t_{i+2}-t_i)(t_{i+1}-t_i)}.$$

For  $x \in (t_{i+1}, t_{i+2}]$ ,

$$B_i^2(x) = \frac{x-t_{i-1}}{t_{i+1}-t_{i-1}} \cdot 0 + \frac{t_{i+2}-x}{t_{i+2}-t_i} \cdot \frac{t_{i+2}-x}{t_{i+2}-t_{i+1}} = \frac{(t_{i+2}-x)^2}{(t_{i+2}-t_i)(t_{i+2}-t_{i+1})}.$$

Hence we derived

$$B_i^2(x) = \begin{cases} \frac{(x-t_{i-1})^2}{(t_{i+1}-t_{i-1})(t_i-t_{i-1})} & x \in (t_{i-1}, t_i], \\ \frac{(x-t_{i-1})(t_{i+1}-x)}{(t_{i+1}-t_{i-1})(t_{i+1}-t_i)} + \frac{(t_{i+2}-x)(x-t_i)}{(t_{i+2}-t_i)(t_{i+1}-t_i)} & x \in (t_i, t_{i+1}], \\ \frac{(t_{i+2}-x)^2}{(t_{i+2}-t_i)(t_{i+2}-t_{i+1})} & x \in (t_{i+1}, t_{i+2}], \\ 0 & \text{otherwise.} \end{cases} \quad (3.1)$$

(b) Follows from (3.1), we derived

$$\frac{d}{dx} B_i^2(x) = \begin{cases} p_1(x) = \frac{2(x-t_{i-1})}{(t_{i+1}-t_{i-1})(t_i-t_{i-1})} & x \in (t_{i-1}, t_i], \\ p_2(x) = \frac{t_{i+1}+t_{i-1}-2x}{(t_{i+1}-t_{i-1})(t_{i+1}-t_i)} + \frac{t_{i+2}+t_i-2x}{(t_{i+2}-t_i)(t_{i+1}-t_i)} & x \in (t_i, t_{i+1}], \\ p_3(x) = \frac{2(x-t_{i+2})}{(t_{i+2}-t_i)(t_{i+2}-t_{i+1})} & x \in (t_{i+1}, t_{i+2}], \\ 0 & \text{otherwise.} \end{cases} \quad (3.2)$$

We have

$$\begin{aligned} p_1(t_i) &= \frac{2(t_i-t_{i-1})}{(t_{i+1}-t_{i-1})(t_i-t_{i-1})} = \frac{2}{t_{i+1}-t_{i-1}} \\ p_2(t_i) &= \frac{t_{i+1}+t_{i-1}-2t_i}{(t_{i+1}-t_{i-1})(t_{i+1}-t_i)} + \frac{t_{i+2}+t_i-2t_i}{(t_{i+2}-t_i)(t_{i+1}-t_i)} \\ &= \frac{t_{i-1}-t_i}{(t_{i+1}-t_{i-1})(t_{i+1}-t_i)} + \frac{1}{t_{i+1}-t_{i-1}} + \frac{1}{t_{i+1}-t_i} \\ &= \frac{t_{i-1}-t_i+t_{i+1}-t_{i-1}}{(t_{i+1}-t_{i-1})(t_{i+1}-t_i)} + \frac{1}{t_{i+1}-t_{i-1}} \\ &= \frac{2}{t_{i+1}-t_{i-1}} = p_1(t_i) \end{aligned}$$

Hence  $\frac{d}{dx} B_i^2(x)$  is continuous at  $t_i$ . Similarly,

$$\begin{aligned} p_3(t_{i+1}) &= \frac{2(t_{i+1}-t_{i+2})}{(t_{i+2}-t_i)(t_{i+2}-t_{i+1})} = -\frac{2}{t_{i+2}-t_i} \\ p_2(t_{i+1}) &= \frac{t_{i+1}+t_{i-1}-2t_{i+1}}{(t_{i+1}-t_{i-1})(t_{i+1}-t_i)} + \frac{t_{i+2}+t_i-2t_{i+1}}{(t_{i+2}-t_i)(t_{i+1}-t_i)} = -\frac{2}{t_{i+2}-t_i} = p_3(t_{i+1}) \end{aligned}$$

Hence  $\frac{d}{dx} B_i^2(x)$  is continuous at  $t_{i+1}$ .

(c) We know  $\frac{d}{dx}B_i^2(x)$  is continuous, and is a linear function at each interval  $(t_{i-1}, t_i]$ ,  $(t_i, t_{i+1}]$  and  $(t_{i+1}, t_{i+2}]$ . And we have that

$$\frac{d}{dx}B_i^2(t_{i-1}) = 0, \quad \frac{d}{dx}B_i^2(t_i) = \frac{2}{t_{i+1} - t_{i-1}} > 0.$$

So by the property of linear function,

$$\frac{d}{dx}B_i^2(x) > 0, \quad x \in (t_{i-1}, t_i]$$

Moreover,

$$\frac{d}{dx}B_i^2(t_{i+1}) = -\frac{2}{t_{i+2} - t_i} < 0$$

Hence by the property of linear function, there is unique  $x^* \in (t_i, t_{i+1})$  such that  $\frac{d}{dx}B_i^2(x^*) = 0$ . Follows from (3.2) we have the following equation.

$$\frac{t_{i+1} + t_{i-1} - 2x^*}{t_{i+1} - t_{i-1}} + \frac{t_{i+2} + t_i - 2x^*}{t_{i+2} - t_i} = 0$$

Solve it and we got

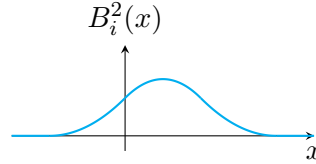
$$x^* = \frac{t_{i+2}t_{i+1} - t_it_{i-1}}{(t_{i+2} + t_{i+1}) - (t_i + t_{i-1})}.$$

(d) By (c) we know that:

$$\begin{aligned} \frac{d}{dx}B_i^2(x) &> 0, \quad x \in (t_{i-1}, x^*) \\ \frac{d}{dx}B_i^2(x) &< 0, \quad x \in (x^*, t_{i+2}) \end{aligned}$$

Also  $B_i^2(t_{i-1}) = B_i^2(t_{i+2}) = 0$ . And  $B(x^*) < 1$  could be verified by a trivial computation. Hence  $B_i^2(x) \in [0, 1]$ .

(e) Clearly the image of  $B_i^2(x)$  with different  $i$  could be obtained by translation. So we just draw with  $i = 0$ .



**Problem 3.6** Verify Theorem 3.32 algebraically for the case of  $n = 2$ , i.e.

$$(t_{i+2} - t_{i-1})[t_{i-1}, t_i, t_{i+1}, t_{i+2}](t - x)_+^2 = B_i^2(x).$$

**Solution** For  $x \in (t_{i-1}, t_i]$ , by Lagrange's formula we have:

$$\begin{aligned} [t_{i-1}, t_i, t_{i+1}, t_{i+2}](t - x)_+^2 &= \frac{(t_i - x)^2}{(t_i - t_{i-1})(t_i - t_{i+1})(t_i - t_{i+2})} + \frac{(t_{i+1} - x)^2}{(t_{i+1} - t_{i-1})(t_{i+1} - t_i)(t_{i+1} - t_{i+2})} \\ &\quad + \frac{(t_{i+2} - x)^2}{(t_{i+2} - t_{i-1})(t_{i+2} - t_i)(t_{i+2} - t_{i+1})} \\ &= \frac{(x - t_{i-1})^2}{(t_{i+2} - t_{i-1})(t_{i+1} - t_{i-1})(t_i - t_{i-1})} = \frac{B_i^2(x)}{t_{i+2} - t_{i-1}} \end{aligned}$$

For  $x \in (t_i, t_{i+1}]$ , by Lagrange's formula we have:

$$\begin{aligned} [t_{i-1}, t_i, t_{i+1}, t_{i+2}](t - x)_+^2 &= \frac{(t_{i+1} - x)^2}{(t_{i+1} - t_{i-1})(t_{i+1} - t_i)(t_{i+1} - t_{i+2})} + \frac{(t_{i+2} - x)^2}{(t_{i+2} - t_{i-1})(t_{i+2} - t_i)(t_{i+2} - t_{i+1})} \\ &= \frac{B_i^2(x)}{t_{i+2} - t_{i-1}} \end{aligned}$$



For  $x \in (t_{i+1}, t_{i+2}]$ , by Lagrange's formula we have:

$$\begin{aligned} [t_{i-1}, t_i, t_{i+1}, t_{i+2}](t-x)_+^2 &= \frac{(t_{i+2}-x)^2}{(t_{i+2}-t_{i-1})(t_{i+2}-t_i)(t_{i+2}-t_{i+1})} \\ &= \frac{B_i^2(x)}{t_{i+2}-t_{i-1}} \end{aligned}$$

Hence we verified

$$(t_{i+2}-t_{i-1})[t_{i-1}, t_i, t_{i+1}, t_{i+2}](t-x)_+^2 = B_i^2(x)$$

in the support of  $B_i^2(x)$ . And clearly, the equation is also right when  $B_i^2(x)$  vanishes.

### Problem 3.7 Scaled integral of B-splines.

Deduce from the Theorem on derivatives of B-splines that the scaled integral of a B-spline  $B_i^n(x)$  over its support is independent of its index  $i$  even if the spacing of the knots is not uniform.

**Solution** By the Theorem on derivatives of B-splines, we have

$$\frac{d}{dx} B_i^{n+1}(x) = \frac{(n+1)B_i^n(x)}{t_{i+n}-t_{i-1}} - \frac{(n+1)B_{i+1}^n(x)}{t_{i+n+1}-t_i}, \quad n = 1, 2, \dots$$

Integral to both side, we have:

$$\int_{t_{i-1}}^{t_{i+n+1}} \frac{d}{dx} B_i^{n+1}(x) dx = \int_{t_{i-1}}^{t_{i+n+1}} \left( \frac{(n+1)B_i^n(x)}{t_{i+n}-t_{i-1}} - \frac{(n+1)B_{i+1}^n(x)}{t_{i+n+1}-t_i} \right) dx, \quad n = 1, 2, \dots$$

For the left side, we have:

$$LHS = B_i^{n+1}(t_{i+n+1}) - B_i^{n+1}(t_{i-1}) = 0 - 0 = 0.$$

For the right side, we have:

$$RHS = (n+1) \left( \int_{t_{i-1}}^{t_{i+n}} \frac{B_i^n(x)}{t_{i+n}-t_{i-1}} dx - \int_{t_i}^{t_{i+n+1}} \frac{B_{i+1}^n(x)}{t_{i+n+1}-t_i} dx \right)$$

Then we got

$$\int_{t_{i-1}}^{t_{i+n}} \frac{B_i^n(x)}{t_{i+n}-t_{i-1}} dx = \int_{t_i}^{t_{i+n+1}} \frac{B_{i+1}^n(x)}{t_{i+n+1}-t_i} dx$$

Hence the scaled integral of  $B_i^n(x)$  over its support is independent of  $i$ .

### Problem 3.8 Symmetric Polynomials.

We have a theorem on expressing complete symmetric polynomials as divided difference of monomials.

- Verify this theorem for  $m = 4$  and  $n = 2$  by working out the table of divided difference and comparing the result to the definition of complete symmetric polynomials.
- Prove this theorem by the lemma on the recursive relation on complete symmetric polynomials.

**Solution**

(a) By the definition,

$$\tau_2(x_i, x_{i+1}, x_{i+2}) = x_i^2 + x_{i+1}^2 + x_{i+2}^2 + x_i x_{i+1} + x_i x_{i+2} + x_{i+1} x_{i+2}.$$

Make a table of divided difference as following.

$x_i$	$x_i^4$		
$x_{i+1}$	$x_{i+1}^4$	$(x_{i+1}^2 + x_i^2)(x_{i+1} + x_i)$	
$x_{i+2}$	$x_{i+2}^4$	$(x_{i+2}^2 + x_{i+1}^2)(x_{i+2} + x_{i+1})$	$\frac{(x_{i+2}^2 + x_{i+1}^2)(x_{i+2} + x_{i+1}) - (x_{i+1}^2 + x_i^2)(x_{i+1} + x_i)}{x_{i+2} - x_i}$

Then the result follows from

$$\begin{aligned}
& \frac{(x_{i+2}^2 + x_{i+1}^2)(x_{i+2} + x_{i+1}) - (x_{i+1}^2 + x_i^2)(x_{i+1} + x_i)}{x_{i+2} - x_i} \\
&= \frac{(x_{i+2}^3 - x_i^3) + x_{i+1}(x_{i+2}^2 - x_i^2) + x_{i+1}^2(x_{i+2} - x_i)}{x_{i+2} - x_i} \\
&= (x_{i+2}^2 + x_{i+2}x_i + x_i^2) + x_{i+1}(x_{i+2} + x_i) + x_{i+1}^2 \\
&= \tau_2(x_i, x_{i+1} + x_{i+2}).
\end{aligned}$$

(b) By the lemma on recursive relations of complete symmetric polynomials, we have

$$\begin{aligned}
& (x_{i+n+1} - x_i)\tau_{m-n-1}(x_i, \dots, x_{i+n+1}) \\
&= \tau_{m-n}(x_i, \dots, x_{i+n+1}) - \tau_{m-n}(x_i, \dots, x_{i+n}) - x_i\tau_{m-n-1}(x_i, \dots, x_{i+n+1}) \\
&= \tau_{m-n}(x_{i+1}, \dots, x_{i+n+1}) + x_i\tau_{m-n-1}(x_i, \dots, x_{i+n+1}) - \tau_{m-n}(x_i, \dots, x_{i+n}) - x_i\tau_{m-n-1}(x_i, \dots, x_{i+n+1}) \\
&= \tau_{m-n}(x_{i+1}, \dots, x_{i+n+1}) - \tau_{m-n}(x_i, \dots, x_{i+n}).
\end{aligned}$$

Now we prove the theorem by induction. For  $n = 0$ , clearly

$$\tau_m(x_i) = [x_i]x^m = x_i^m.$$

Now we suppose the theorem is true for some  $0 \leq n < m$ . Then for  $n + 1$ , we have

$$\begin{aligned}
\tau_{m-n-1}(x_i, \dots, x_{i+n+1}) &= \frac{\tau_{m-n}(x_{i+1}, \dots, x_{i+n+1}) - \tau_{m-n}(x_i, \dots, x_{i+n})}{x_{i+n+1} - x_i} \\
&= \frac{[x_{i+1}, \dots, x_{i+n+1}]x^m - [x_i, \dots, x_{i+n}]}{x_{i+n+1} - x_i} \\
&= [x_i, \dots, x_{i+n+1}]x^m
\end{aligned}$$

Then the theorem is proved by induction.

## Chapter 4 Computer Arithmetic

**Problem 4.1** Convert the decimal integer 477 to a normalized FPN with  $\beta = 2$ .

**Solution**  $477 = (111011101)_2 = (1.11011101)_2 \times 2^8$ .

**Problem 4.2** Convert the decimal fraction  $\frac{3}{5}$  to a normalized FPN with  $\beta = 2$ .

**Solution** Calculate by the following table.

Arithmetic	Decimal Part	Integer Part
$\frac{3}{5} \times 2 = \frac{6}{5}$	$\frac{1}{5}$	1
$\frac{1}{5} \times 2 = \frac{2}{5}$	$\frac{2}{5}$	0
$\frac{2}{5} \times 2 = \frac{4}{5}$	$\frac{4}{5}$	0
$\frac{4}{5} \times 2 = \frac{8}{5}$	$\frac{3}{5}$	1
$\vdots$	$\vdots$	$\vdots$

Hence we got that

$$\frac{3}{5} = (1.0011001 \dots)_2 \times 2^{-1}$$

**Problem 4.3** Let  $x = \beta^e$ ,  $e \in \mathbb{Z}$ ,  $L < e < U$  be a normalized FPN in  $\mathbb{F}$  and  $x_L, x_R \in \mathbb{F}$  the two normalized FPNs adjacent to  $x$  such that  $x_L < x < x_R$ . Prove  $x_R - x = \beta(x - x_L)$ .

**Solution** We represent  $x, x_L, x_R$  in the form of normalized FPN as following.

$$\begin{aligned} x &= (1.00 \dots 0)_\beta \times \beta^e \\ x_L &= ([\beta - 1].[\beta - 1] \dots [\beta - 1])_\beta \times \beta^{e-1} \\ x_R &= (1.00 \dots 01)_\beta \times \beta^e \end{aligned}$$

And hence we have:

$$\begin{aligned} x_R - x &= (0.00 \dots 01)_\beta \times \beta^e = \beta^{e-p+1} \\ x - x_L &= (0.00 \dots 01)_\beta \times \beta^{e-1} = \beta^{e-p} \end{aligned}$$

That is  $x_R - x = \beta(x - x_L)$ .

**Problem 4.4** By reusing your result of II, find out the two normalized FPNs adjacent to  $x = \frac{3}{5}$  under the IEEE 754 single-precision protocol. What is  $\text{fl}(x)$  and the relative roundoff error?

**Solution** Recall that  $x = (1.0011001 \dots)_2 \times 2^{-1}$ , find  $x_L$  and  $x_R$  under IEEE 754 single-precision protocol:

$$\begin{aligned} x_L &= (1.0011001 \ 10011001 \ 1001100)_2 \times 2^{-1}, \\ x_R &= (1.0011001 \ 10011001 \ 1001101)_2 \times 2^{-1}. \end{aligned}$$

We calculate that:

$$\begin{aligned} x - x_L &= (1.10011001 \dots)_2 \times 2^{-23} = \frac{8}{5} \times 2^{-23}, \\ x_R - x_L &= 2^{-22}, \\ x_R - x &= (x_R - x_L) - (x - x_L) = 2^{-22} - \frac{8}{5} \times 2^{-23} = \frac{2}{5} \times 2^{-23}. \end{aligned}$$

Clearly that  $x_R - x < x - x_L$ , hence  $\text{fl}(x) = x_R$  and the relative roundoff error is  $\frac{|x_R - x|}{|x|} = \frac{2}{3} \times 2^{-23}$ .

**Problem 4.5** If the IEEE 754 single-precision protocol did not round off numbers to the nearest, but simply dropped excess bits, what would the unit roundoff be?

**Solution** It would be  $\epsilon_u^* = \epsilon_M = \beta^{1-p} = 2^{-23}$ .

To prove it, we should prove that for  $x \in \mathcal{R}(\mathcal{F})$ , we have

$$fl^*(x) = x(1 + \delta), \quad |\delta| < \epsilon_u^* \quad (4.1)$$

where  $fl^*(x)$  is the approximate of  $x$  got by the discription of the problem.

we could find  $x_L, x_R \in \mathcal{F}$  s.t.

- $x_L$  and  $x_R$  are adjacent.
- $x_L \leq x \leq x_R$ .

If  $x = x_L$  or  $x = x_R$ , then  $fl^*(x) - x = 0$  and (4.1) clearly holds. Otherwise  $x_L < x < x_R$ . Then Lemma 4.23, Definition 4.22 yield

$$|fl^*(x) - x| \leq |x_R - x_L| \leq \epsilon_u^* \min(|x_L|, |x_R|) < \epsilon_u^* |x|.$$

which yields (4.1). And the upper bound of the error can be reached as  $x \rightarrow x_{R-}$ . Hence  $\epsilon_u^*$  is the unit roundoff.

**Problem 4.6** How many bits of precision are lost in the subtraction  $1 - \cos x$  when  $x = \frac{1}{4}$ ?

**Solution** For  $x = \frac{1}{4}$ , we know that  $1 > \cos x$ , hence by the theorem on the loss of most significant digits, we should calculate:

$$1 - \frac{\cos x}{1} = 0.0310875783 \dots \in [2^{-6}, 2^{-5}].$$

(The result above is calculated with long double, which is accurate enough.)

Hence we lost at most 6 and at least 5 significant bits.

**Problem 4.7** Suggest at least two ways to compute  $1 - \cos x$  to avoid catastrophic cancellation caused by subtraction.

**Solution**

(1) We can use Taylor series:

$$\begin{aligned} 1 - \cos x &= 1 - \left( 1 - \frac{x^2}{2!} + \frac{x^4}{4!} - \frac{x^6}{6!} + \dots \right) \\ &= \frac{x^2}{2!} - \frac{x^4}{4!} + \frac{x^6}{6!} - \dots \end{aligned}$$

(2) We can use sum-to-product identities:

$$1 - \cos x = \cos 0 - \cos x = 2 \sin^2 \left( \frac{x}{2} \right).$$

**Problem 4.8** What are the condition numbers of the following functions? Where are they large?

- $f_1(x) = (x - 1)^\alpha$ ,
- $f_2(x) = \ln x$ ,
- $f_3(x) = e^x$ ,
- $f_4(x) = \arccos x$ .

**Solution**

- We should discuss the value of  $\alpha$ .

(i)  $\alpha \neq 0$ . The condition number of  $f_1$  is  $C_{f_1}(x) = \left| \frac{x f_1'(x)}{f_1(x)} \right| = \left| \frac{\alpha x (x-1)^{\alpha-1}}{(x-1)^\alpha} \right| = \left| \frac{\alpha x}{x-1} \right|$ .

Hence  $C_{f_1}(x) \rightarrow +\infty$  as  $x \rightarrow 1$ .

(ii)  $\alpha = 0$ . The condition number of  $f_1$  is  $C_{f_1}(x) = \left| \frac{xf'_1(x)}{f_1(x)} \right| = \left| \frac{0}{1} \right| = 0$ .

Hence  $C_{f_1}(x)$  will never be large.

- The condition number of  $f_2$  is  $C_{f_2}(x) = \left| \frac{xf'_2(x)}{f_2(x)} \right| = \left| \frac{x \cdot \frac{1}{x}}{\ln x} \right| = \left| \frac{1}{\ln x} \right|$ .

Hence  $C_{f_2}(x) \rightarrow +\infty$  as  $x \rightarrow 0_+$ .

- The condition number of  $f_3$  is  $C_{f_3}(x) = \left| \frac{xf'_3(x)}{f_3(x)} \right| = \left| \frac{xe^x}{e^x} \right| = |x|$ .

Hence  $C_{f_3}(x) \rightarrow +\infty$  as  $x \rightarrow \pm\infty$ .

- The condition number of  $f_4$  is  $C_{f_4}(x) = \left| \frac{xf'_4(x)}{f_4(x)} \right| = \left| \frac{x}{\sqrt{1-x^2} \arccos x} \right|$

Hence  $C_{f_4}(x) \rightarrow +\infty$  as  $x \rightarrow \pm 1$ .

**Problem 4.9** Consider the function  $f(x) = 1 - e^{-x}$  for  $x \in [0, 1]$ .

- Show that  $C_f(x) \leq 1$  for  $x \in [0, 1]$ .
- Let  $A$  be the algorithm that evaluates  $f(x)$  for the machine number  $x \in \mathbb{F}$ . Assume that the exponential function is computed with relative error within machine roundoff. Estimate  $\text{cond}_A(x)$  for  $x \in [0, 1]$ .
- Plot  $\text{cond}_f(x)$  and the estimated upper bound of  $\text{cond}_A(x)$  as a function of  $x$  on  $[0, 1]$ . Discuss your results.

**Solution**

- (1) The condition number of  $f$  is  $C_f(x) = \left| \frac{xe^{-x}}{1-e^{-x}} \right| = \left| \frac{x}{e^x-1} \right|$ .

Notice that  $C_f(x)$  decreases in  $x \in [0, 1]$ , and  $\lim_{x \rightarrow 0} C_f(x) = 1$ . Hence  $C_f(x) \leq 1$  for  $x \in [0, 1]$ .

- (2) See that

$$\epsilon_u > |f_A(x) - f(x)| = |f(x_A) - f(x)| = |f'(\xi)| \cdot |x - x_A|, \quad \text{for } \xi \text{ between } x \text{ and } x_A.$$

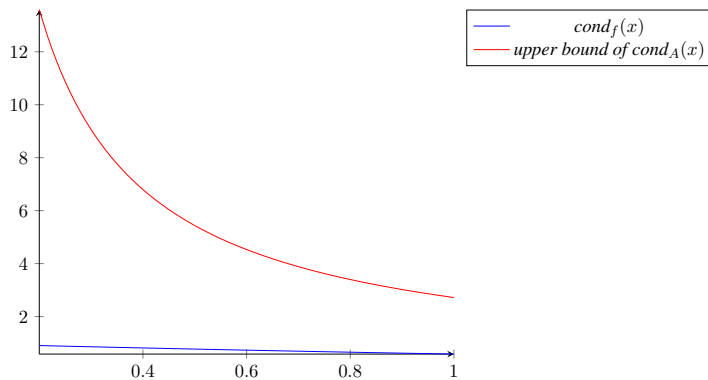
So we have

$$|x - x_A| < \frac{\epsilon_u}{|f'(\xi)|} = \frac{\epsilon_u}{e^{-\xi}} \leq e\epsilon_u.$$

Hence by the definition,

$$\text{cond}_A(x) = \frac{1}{\epsilon_u} \min_{\{x_A\}} \frac{|x - x_A|}{x} < \frac{e}{x}.$$

- (3) See the figure of  $C_f(x)$  and the upper bound of  $\text{cond}_A(x)$  here.



The upper bound of  $\text{cond}_A(x)$  goes large as  $x \rightarrow 0_+$ , which means calculating  $f$  with algorithm  $A$  as  $x$  is small will cause catastrophic cancellation.

**Problem 4.10** The math problem of root finding for a polynomial

$$q(x) = \sum_{i=0}^n a_i x^i, \quad a_n = 1, a_0 \neq 0, a_i \in \mathbb{R} \quad (4.2)$$

can be considered as a vector function  $f : \mathbb{R}^n \rightarrow \mathbb{C}$ :

$$r = f(a_0, a_1, \dots, a_{n-1}).$$

Derive the componentwise condition number of  $f$  base on the 1-norm. For the Wilkinson example, compute your condition number, and compare your result with that in the Wilkinson Example. What does the comparison tell you?

**Solution** Notice that  $r$  satisfies equation

$$\sum_{i=0}^n a_i r^i = 0.$$

Compute partial differential to each side, we have

$$r^j + \sum_{i=0}^n i a_i r^{i-1} \frac{\partial r}{\partial a_j} = 0, \quad j = 1, 2, \dots, n-1.$$

That implies

$$\nabla r = -\frac{(1, r, \dots, r^{n-1})}{q'(r)}.$$

Hence we have

$$\text{cond}_f(\mathbf{a}) = \frac{\|\mathbf{a}\|_1 \|\nabla r\|_1}{|r|} = \frac{\sum_{i=1}^{n-1} |a_i| \cdot \sum_{i=0}^{n-1} |r|^i}{|\sum_{i=0}^n i a_i r^i|} \left( \geq \frac{\sum_{i=1}^{n-1} |a_i r^i|}{|\sum_{i=0}^n i a_i r^i|} \right)$$

In Wilkinson example,  $q(x) = \prod_{i=1}^n (x - i)$ ,  $r = n$  is a root. Then we have

$$\sum_{i=0}^{n-1} |a_n r^i| \geq -\sum_{i=0}^{n-1} a_i r^i = r^n = n^n,$$

and

$$\left| \sum_{i=0}^n i a_i r^i \right| = |r| \cdot |q'(r)| = n |q'(n)| < n^2 n!.$$

Hence we have  $\text{cond}_f(\mathbf{a}) \geq \frac{n^{n-2}}{n!}$ , which goes  $+\infty$  as  $n \rightarrow \infty$ . It supports the Wilkinson Example. And it tells us that finding the root of high-order polynomial equation is very hard.

**Problem 4.11** Suppose the division of two FPNs is calculated in a register of precision  $2p$ . Give an example that contradicts the conclusion of the model of machine arithmetic.

**Solution** Consider the FPN system:  $\beta = 2, p = 2, L = -1, U = 1$ . The number:

$$a = (1.0)_2 \times 2^0, \quad b = (1.1)_2 \times 2^0.$$

We have the theoretical result:

$$\frac{a}{b} = \frac{2}{3} = (0.101010 \dots)_2.$$

If calculating in a register with precision  $2p = 4$ , then:

$$\text{fl}\left(\frac{a}{b}\right) = \text{fl}((0.101)_2) = (0.10)_2 \times 2^0 = 0.5$$

The relative error is

$$\left| \frac{0.5}{\frac{2}{3}} - 1 \right| = 0.25$$

And the machine precision is

$$\epsilon_u = 2^{-2} = 0.25$$

So the result contradicts to the model of machine arithmetic.



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**Problem 4.12** If the bisection method is used in single precision FPNs of IEEE 754 starting with the interval  $[128, 129]$ , can we compute the root with absolute accuracy  $< 10^{-6}$ ? Why?

**Solution** In this system,  $\epsilon_M = 2^{-23}$ . Hence the distance of two adjacent floating numbers in  $[128, 129]$  is

$$2^7 \epsilon_M = 2^{-16} \approx 1.5259 \times 10^{-5} > 2 \times 10^{-6}.$$

Hence we can't compute the root with absolute accuracy less than  $10^{-6}$ .

**Problem 4.13** In fitting a curve by cubic splines, one gets inaccurate results when the distance between two adjacent points is much smaller than those of other adjacent pairs. Use the condition number of a matrix to explain this phenomenon.

**Solution** Consider calculate  $s(x) = ax^3 + bx^2 + cx + d$  on  $[x_i, x_{i+1}]$  by  $s(x_i), s(x_{i+1}), s'(x_i), s'(x_{i+1})$ . We should solve a linear equation where the coefficient matrix is

$$\begin{pmatrix} x_i^3 & x_i^2 & x_i & 1 \\ x_{i+1}^3 & x_{i+1}^2 & x_{i+1} & 1 \\ 3x_i^2 & 2x_i & 1 & 0 \\ 3x_{i+1}^2 & 2x_{i+1} & 1 & 0 \end{pmatrix}.$$

It has large condition number. So the result will be very inaccurate.