

16811 A1

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1.

- (a) The implementation is in the `q1.py`. Run `python q1.py` to get the interpolation result.
- (b) By running `python q1.py`, I got the interpolate result for $\sin(2\pi x)$ at $x = \frac{1}{10}$ as 0.5877882222892551.
- (c) By running `python q1.py`, I got the estimate results:
 at $n = 2$, 1.0882;
 at $n = 4$, 1.048946673207547;
 at $n = 40$, 0.7157134629956571.
 The actual value of $f(0.09)$ is 0.7158708568974157.
- (d) the interpolation error for $n = 2$ is 1.9589807426941546
 the interpolation error for $n = 4$ is 5.418917597832672
 the interpolation error for $n = 6$ is 15.731031225174867
 the interpolation error for $n = 8$ is 45.68235961893594
 the interpolation error for $n = 10$ is 131.19950253756573
 the interpolation error for $n = 12$ is 372.06087588229434
 the interpolation error for $n = 14$ is 1043.1298787688495
 the interpolation error for $n = 16$ is 2896.750208459296
 the interpolation error for $n = 18$ is 7981.511500692966
 the interpolation error for $n = 20$ is 21851.558191480788
 the interpolation error for $n = 40$ is 427327258.30265874

The error estimate results show that the error increases as n increases. This demonstrates that larger n does not necessary leads to a better result. This make sense, because the accuracy of polynomial interpolation depends on how close the interpolated point is to the middle of the x values of the set of points used. In the case for example we are interpolating $f(x)$ where $x \in [-1, 1]$, as n increases, the middle becomes farther from the interpolated points, and thus the error gets larger.

2.

$$f'(x) = 2\pi \cos(2\pi x); f''(x) = -(2\pi)^2 \sin(2\pi x) f'''(x) = -(2\pi)^3 \cos(2\pi x)$$

1. linear interpolation

$$\begin{aligned} \max_{0 \leq x \leq 1} |f''(x)| &\leq (2\pi)^2 \\ \max_{0 \leq x \leq 1} (x - x_i)(x - x_{i+1}) &= \max_{0 \leq y \leq h} |y(y - h)| = \frac{h^2}{4} \\ \max_{0 \leq x \leq 1} |e_1(x)| &\leq \frac{1}{4} \cdot (2\pi)^2 \cdot \frac{h^2}{4} = \frac{(\pi)^2 h^2}{4} \\ \text{want } \frac{(\pi)^2 h^2}{4} &\leq 5 \cdot 10^{-7} \Rightarrow h \leq 0.000112539 \\ \Rightarrow N = \frac{1}{h} &= 8885.8 \end{aligned}$$

Therefore, need 8886 data points (or 8885 intervals).

2. quadratic interpolation

$$\begin{aligned} \max_{0 \leq x \leq 1} |f'''(x)| &\leq (2\pi)^3 \\ \max_{0 \leq x \leq 1} (x - x_{i-1})((x - x_i)(x - x_{i+1})) &= \max_{-h \leq y \leq h} |(y + h)y(y - h)| = \frac{2h^3}{3\sqrt{3}} \\ \max_{0 \leq x \leq 1} |e_1(x)| &\leq \frac{1}{6} \cdot (2\pi)^3 \cdot \frac{2h^3}{3\sqrt{3}} = \frac{8\pi^3 h^3}{9\sqrt{3}} \\ \text{want } \frac{8\pi^3 h^3}{9\sqrt{3}} &\leq 5 \cdot 10^{-7} \Rightarrow h \leq 0.0031555702 \\ \Rightarrow N = \frac{1}{h} &= 316.9 \end{aligned}$$

Therefore, need 317 data points (or 316 intervals).

3.

The implementation is in the q3.py. Run *python q3.py* to get the result.
The solutions are: $x_{low} = 14.0661939, x_{high} = 17.2208$.

4.

(a) Define $h(x) = \frac{f(x)}{f'(x)}$.

By Taylor's series, $h(\xi + \varepsilon) = h(\xi) + h'(\xi)\varepsilon + \frac{1}{2}h''(\xi)\varepsilon^2 + \dots$

$$\begin{aligned} h(\xi) &= \frac{f(\xi)}{f'(\xi)} = 0 \\ h'(\xi) &= 1 - f(\xi) \frac{f''(\xi)}{(f'(\xi))^2} = 1 - \frac{f'(\xi)f''(\xi) + f(\xi)f'''(\xi)}{2f'(\xi)f''(\xi)} \\ &= 1 - \frac{f''(\xi)f''(\xi) + f'(\xi)f'''(\xi) + f(\xi)f''''(\xi) + f'(\xi)f'''(\xi)}{2f''(\xi)f''(\xi) + 2f'(\xi)f'''(\xi)} = \frac{f''(\xi)f''(\xi)}{2f''(\xi)f''(\xi)} = \frac{1}{2} \end{aligned}$$

Rewrite $x_{n+1} = x_n - h(x_n)$ as $\xi + \varepsilon_{n+1} = \xi + \varepsilon_n - h(x_n)$.

$\Rightarrow \varepsilon_{n+1} = \varepsilon_n - h(x_n)$.

Then when $n \rightarrow \infty, x \rightarrow \xi$, we can get

$$\begin{aligned} \varepsilon_{n+1} &= \varepsilon_n - h(\xi) \\ &= \varepsilon_n - (h(\xi) + h'(\xi)\varepsilon_n) \\ &= \varepsilon_n - (0 + \frac{1}{2}\varepsilon_n) \\ &= \varepsilon_n - (0 + \frac{1}{2}\varepsilon_n) \\ &= \frac{1}{2}\varepsilon_n \end{aligned}$$

Then we can get $\lim_{n \rightarrow \infty} \frac{\varepsilon_{n+1}}{\varepsilon_n} = \frac{1}{2}$

Thus, in this case ξ is a root of order 2 of $f(x)$, the Newton's method converges linearly, and no longer converges quadratically.

(b) From (a), we get $h(\xi) = 0, h'(\xi) = \frac{1}{2}$.

$$\begin{aligned} h''(\xi) &= -\frac{(f'(\xi))^2 f''(\xi) + f(\xi) f'(\xi) f'''(\xi) - 2f(\xi) (f''(\xi))^2}{(f'(\xi))^3} \\ &= -\frac{(2f'(\xi))^2 f'''(\xi) + f(\xi) f'(\xi) f'''(\xi) - 3f(\xi) f''(\xi) f'''(\xi)}{3(f'(\xi))^2 f''(\xi)} \\ &= -\frac{(3f'(\xi))^2 f''''(\xi) + f'(\xi) f''(\xi) f'''(\xi) - 3f(\xi) (f'''(\xi))^2 - 2f'(\xi) f''(\xi) f''''(\xi) + f(\xi) f'(\xi) f''''(\xi)}{3(f'(x))^2 f'''(\xi) + f'(\xi) (f''(\xi))^2} \\ &= -\frac{1}{6} \frac{f'''(\xi)}{f''(\xi)} \end{aligned}$$

Similarly, if modify Netwon as $x_{n+1} = x_n - 2h(x_n)$, then rewrite as $\varepsilon_{n+1} = \varepsilon_n - 2h(x_n)$.

Then when $n \rightarrow \infty, x \rightarrow \xi, \varepsilon_n \rightarrow 0$, we can get

$$\begin{aligned}
 \varepsilon_{n+1} &= \varepsilon_n - 2h(\xi) \\
 &= \varepsilon_n - 2(h(\xi) + h'(\xi)\varepsilon_n + \frac{1}{2}h''(\xi)(\varepsilon_n)^2) \\
 &= \varepsilon_n - (0 + \frac{1}{2}\varepsilon_n + \frac{1}{2}(-\frac{1}{6}\frac{f'''(\xi)}{f''(\xi)})(\varepsilon_n)^2) \\
 &= -\frac{1}{6}\frac{f'''(\xi)}{f''(\xi)}(\varepsilon_n)^2
 \end{aligned}$$

Then we can get $\lim_{n \rightarrow \infty} \frac{\varepsilon_{n+1}}{(\varepsilon_n)^2} = -\frac{1}{6}\frac{f'''(\xi)}{f''(\xi)}$
 Since $f''(\xi) \neq 0$, in this case ξ is a root of order 2 of $f(x)$, the Newton's method does converge quadratically.

5.

The implementation is in the q5.py. Run *python q5.py* to get the approximation result.

The approximation result for $x^5+x^4+x^3+x^2+x+1$ is $-1.0, -0.5+0.866i, -0.5-0.866i, 0.5+0.866i, 0.5-0.866i$.

6.

(a) Define $Q = \begin{pmatrix} 1 & -3 & 1 & -3 & 0 \\ 0 & 1 & -3 & 1 & -3 \\ 1 & -1 & -6 & 0 & 0 \\ 0 & 1 & -1 & -6 & 0 \\ 0 & 0 & 1 & -1 & -6 \end{pmatrix}$

$$\mathbf{x} = \begin{pmatrix} x^4 \\ x^3 \\ x^2 \\ x \\ 1 \end{pmatrix}$$

Using the method of resultants, and $Q\mathbf{x} = 0$ iff $\det(Q) = 0$. Since $\det(Q) = 0$, then $p(x)$ and $q(x)$ share common root.

(b) Using the ratio method, let Q be the an $(n-1)*n$ matrix of rank $n-1$, then a solution to $Q\mathbf{x} = 0$ satisfies:

$$x = \frac{x^4}{x^3} = (-1)^{1+2} \frac{\det(Q_i)}{\det(Q_j)} = (-1) \frac{405}{-135} = 3$$

where Q_i is the $(n-1)*(n-1)$ matrix obtained by deleting the i -th column.

7.

Treat x as constant, and rewrite:

$$p(y) = 1 \cdot y^2 + 0 \cdot y + (x^2 - 1)$$

$$q(y) = 0 \cdot y^2 + 1 \cdot y + x^2$$

Then using the method of resultants, define $Q =$

$$\begin{pmatrix} 1 & 0 & x^2 - 1 & 0 \\ 0 & 1 & 0 & x^2 - 1 \\ 1 & x^2 & 0 & 0 \\ 0 & 1 & x^2 & 0 \end{pmatrix}$$

let $\det(Q) = (x^4 + x^2 - 1)(x^2 - 1) = 0$, we find the projection points $x = -1, 1, \pm\sqrt{\frac{-1+\sqrt{5}}{2}}, \pm\sqrt{\frac{-1-\sqrt{5}}{2}}$.

By plugging into $p(y) = q(y)$, we see roots $y = \frac{1}{2}(1 + \sqrt{5}), \frac{1}{2}(1 - \sqrt{5})$.

8.

- (a) The implementation is in the *get_w* function at q8.py.

Explanation:

To determine whether a 2D point (x, y) falls within the triangle formed by three 2D points:

$$\text{Define } A = \begin{pmatrix} x^{(i)} & x^{(j)} & x^{(k)} \\ y^{(i)} & y^{(j)} & y^{(k)} \\ 1 & 1 & 1 \end{pmatrix}$$

$$b = \begin{pmatrix} x \\ y \\ 1 \end{pmatrix}$$

Then calculate v such that $Av = b$. If the components of vector v are all positive, then point (x, y) falls within the triangle formed by $(x^{(i)}, y^{(i)})$, $(x^{(j)}, y^{(j)})$, $(x^{(k)}, y^{(k)})$.

- (b) The implementation is in q8.py. Run *python q8.py* to get the result and plots.
- (c) I first load the data from paths.txt. Then picked the 3 optimal paths: 1. pick 3 paths from left or right side, and the picking criteria is that the starting points of the 3 paths should be closet to the starting point; the starting point falls within the triangle formed by the three starting points of the path selected.
2. Decide left or right 3 paths, which are the better choice. Calculate the weights using the method in (a): $Av = b$, and v contains the three weights we want to solve.
3. The time scale for t is $[0, 1, \dots, 49]$, the same as the interpolated paths.
4. I used linear interpolation and found it already gave a good result, therefore, there's no need to use a more complex method for simplicity
- (d) the green lines are the selected 3 paths, and back line the interpolated result. Figures are in the last pages of this pdf.
- (e) If there are more obstacles or different shapes of obstacles, then I need to modify my code that I can no longer just devide to left or right cases. There would be more complicated cases to consider, e.g. the paths to be selected will have more complex shapes.





