

# NUMERICAL METHODS ASSIGNMENT № 5

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## Problem 1

### Part (a)

Formula for absolute error when  $v$  is approximating  $u$  is:

$$|u - v|$$

Formula for relative error when  $v$  is approximating  $u$  is:

$$\frac{|u - v|}{|u|}$$

Relative error is more meaningful especially if  $|u| \gg 1$  as it doesn't change depending on the value of  $u$ .

Example:

u	v	Absolute Error	Relative Error
1	0.99	0.01	0.01
100	99	1	0.01

### Part (b)

$$f(x + h) = f(x) + hf'(x) + \frac{h^2}{2!}f''(x) + \dots + \frac{h^n}{n!}f^n(x)$$

$$f(x + 2h) = f(x) + 2hf'(x) + \frac{(2h)^2}{2!}f''(x) + \dots + \frac{(2h)^n}{n!}f^n(x)$$

$$2f(x + h) = 2f(x) + 2hf'(x) + 2\frac{h^2}{2!}f''(x) + \dots + 2\frac{h^n}{n!}f^n(x)$$

$$2f(x + h) - f(x + 2h) = f(x) + \frac{(2 - 2^2)}{2!}h^2f''(x) + \dots + \frac{(2 - 2^n)}{n!}h^nf^n(x)$$

$$h^2f''(x) = f(x) - 2f(x + h) + f(x + 2h) + \frac{(2 - 2^3)}{3!}h^3f'''(x) + \dots + \frac{(2 - 2^n)}{n!}h^nf^n(x)$$

$$f''(x) = \frac{f(x) - 2f(x + h) + f(x + 2h)}{h^2} + \frac{(2 - 2^3)}{3!}hf'''(x) + \dots + \frac{(2 - 2^n)}{n!}h^{n-2}f^n(x)$$

$$\left| f''(x) - \frac{f(x) - 2f(x + h) + f(x + 2h)}{h^2} \right| \approx -h|f'''(x)|$$

So long as  $f'''(x)$  is bounded, the discretization error is  $\mathcal{O}(h)$ .

### Part (c)

$$(\tan(x))'' = \frac{2\sin x}{\cos^3 x}$$
$$f''(x) \approx \frac{\tan(x) - 2\tan(x+h) + \tan(x+2h)}{h^2}$$

Actual error is:

$$\left| \frac{2\sin x}{\cos^3 x} - \frac{\tan(x) - 2\tan(x+h) + \tan(x+2h)}{h^2} \right|$$

Since expected error is  $O(h)$ , the graphic for difference between expected error and actual error is:

Listing 1: Python code to generate the graph below.

```
1 import matplotlib.pyplot as plt
2 import numpy as np
3
4 x = np.pi / 4
5 h = [1e-1, 1e-2, 1e-3, 1e-4, 1e-5, 1e-6, 1e-7, 1e-8, 1e-9, 1e-10]
6
7 arr = []
8 for i in range(0, 10):
9     derivative = 2 * np.sin(x) / (np.cos(x) * np.cos(x) * np.cos(x))
10    approx = (np.tan(x) - 2 * np.tan(x + h[i]) + np.tan(x + 2 * h[i])) / (↵
        h[i] * h[i])
11    err = abs(derivative - approx)
12    diff = err - h[i]
13    arr.append(diff)
14 plt.plot(h, arr)
15 plt.xscale('log')
16 plt.yscale('log')
```

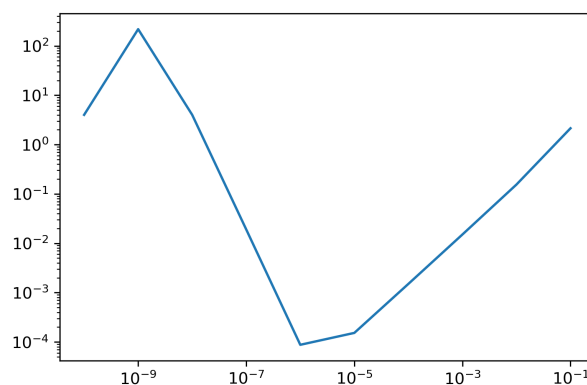


Figure 1: Difference between the expected error ( $O(h)$ ) and the actual error

### Part (d)

The actual error consists of discretization errors and roundoff errors. When  $h$  is relatively large, discretization error dominates roundoff errors but as  $h$  gets smaller roundoff error starts to dominate. For smaller  $h$  values actual error gets very large because of roundoff errors and as a result produces bigger errors than expected.

### Part (e)

The problem is well conditioned because small perturbations in value of the function produces similar results near  $\frac{\pi}{4}$ .

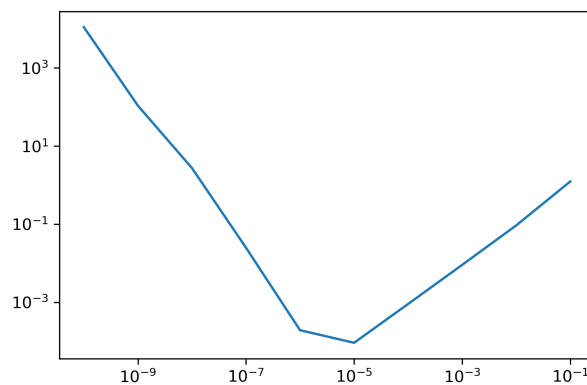


Figure 2: Actual error when  $x = \frac{\pi}{4} - 0.1$

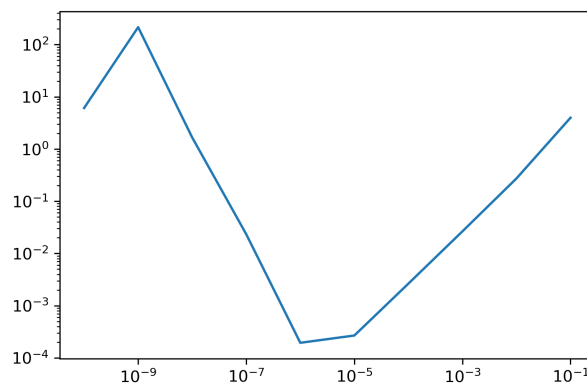


Figure 3: Actual error when  $x = \frac{\pi}{4} + 0.1$

As we can see from the graphs, small perturbations in  $\frac{\pi}{4}$  does not cause big changes in calculation.

## Problem 2

### Part (a)

$$\begin{array}{ll} 0.03754 \times 2 = 0.07508 & 0 \\ 0.07508 \times 2 = 0.15016 & 0 \\ 0.15016 \times 2 = 0.30032 & 0 \\ 0.30032 \times 2 = 0.60064 & 0 \\ 0.60064 \times 2 = 1.20128 & 1 \\ 0.20128 \times 2 = 0.40256 & 0 \\ 0.40256 \times 2 = 0.80512 & 0 \\ 0.80512 \times 2 = 1.61024 & 1 \\ 0.61024 \times 2 = 1.22048 & 1 \end{array}$$

Normalized floating point representation of the number is  $1.0011 \times 2^{-5}$

Ten bit word for the number is 0101010011. First bit is 0 because number is positive, second bit is 1 because exponent is negative, next four bits are 0101 because exponent is 5, final four bits are 0011 because mantissa is 0011.

### Part (b)

Decimal equivalent of the ten bit word is  $(1 + \frac{1}{8} + \frac{1}{16}) \times 2^{-5} = 0.037109375$

### Part (c)

Magnitude of the relative true error is:

$$\begin{aligned} |\epsilon_a| &= \left| \frac{0.03754 - 0.037109375}{0.03754} \right| \\ &= 0.01147097762 \end{aligned}$$

Machine epsilon  $\eta$  is:

$$\begin{aligned} \eta &= \frac{1}{2} \times 2^{-4} \\ &= 0.03125 \end{aligned}$$

$|\epsilon_a| < \eta$  is true.

## Problem 3

### Part (a)

Fixed points of a function are points where  $h(x) = x$ .

$$\begin{aligned}x^2 + \frac{4}{25} &= x \\x^2 - x + \frac{4}{25} &= 0 \\(x - \frac{1}{5})(x - \frac{4}{5}) &= 0\end{aligned}$$

Points  $(0.2, 0.2)$  and  $(0.8, 0.8)$  are the fixed points.

### Part (b)

Since  $|h'(0.2)| = 0.4$  and there exists a constant  $\rho < 1$  where  $|h'(0.2)| < \rho$ , the point  $(0.2, 0.2)$  will converge if  $-\infty < x_0 < 0.8$ .

Since  $|h'(0.8)| = 1.6$  and there is no constant  $\rho < 1$  where  $|h'(0.8)| < \rho$ , the point  $(0.8, 0.8)$  will diverge regardless of  $x_0$ .

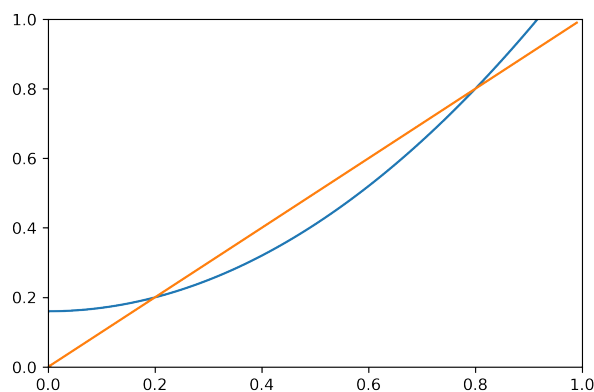


Figure 4: Graph of  $h(x)$

### Part (c)

$$\rho = |h'(0.2)|, \rho = 0.4$$

$$|x_k - x^*| \approx \rho |x_{k-1} - x| \approx \dots \approx \rho^k |x_0 - x|$$

$$|x_k - x^*| \approx 0.1 |x_0 - x|$$

$$\rho^k = 0.1$$

$$k \log_{10} \rho \approx -1$$

$$k = \left\lceil \frac{1}{-\log_{10} \rho} \right\rceil$$

$$k = 3$$

## Part (d)

Newton's method is a special case of fixed point iteration where  $f(x) = x - \frac{g(x)}{g'(x)}$ . This relationship is useful because everything we know about fixed point iteration holds true for Newton's method. Newton's method converges fast but requires  $f$  to be differentiable.

## Problem 4

If we denote first product as  $x_1$ , second product as  $x_2$  and third product as  $x_3$ , the system of equations is as follows:

$$200x_1 + 600x_2 + 250x_3 = 55500$$

$$250x_1 + 500x_2 + 200x_3 = 51000$$

$$300x_1 + 350x_2 + 400x_3 = 51000$$

$$\begin{bmatrix} 200 & 600 & 250 & 55500 \\ 250 & 500 & 200 & 51000 \\ 300 & 350 & 400 & 51000 \end{bmatrix}$$

Multiply  $R_1$  by  $\frac{1}{200}$ .

$$\begin{bmatrix} 1 & 3 & \frac{5}{4} & \frac{555}{2} \\ 250 & 500 & 200 & 51000 \\ 300 & 350 & 400 & 51000 \end{bmatrix}$$

Add  $-250 \times R_1$  to  $R_2$ .

$$\begin{bmatrix} 1 & 3 & \frac{5}{4} & \frac{555}{2} \\ 0 & -250 & -\frac{225}{2} & -18375 \\ 300 & 350 & 400 & 51000 \end{bmatrix}$$

Add  $-300 \times R_1$  to  $R_3$ .

$$\begin{bmatrix} 1 & 3 & \frac{5}{4} & \frac{555}{2} \\ 0 & -250 & -\frac{225}{2} & -18375 \\ 0 & -550 & 25 & -32250 \end{bmatrix}$$

Multiply  $R_2$  by  $-\frac{1}{250}$ .

$$\begin{bmatrix} 1 & 3 & \frac{5}{4} & \frac{555}{2} \\ 0 & 1 & \frac{9}{20} & \frac{147}{2} \\ 0 & -550 & 25 & -32250 \end{bmatrix}$$

Add  $550 \times R_2$  to  $R_3$ .

$$\begin{bmatrix} 1 & 3 & \frac{5}{4} & \frac{555}{2} \\ 0 & 1 & \frac{9}{20} & \frac{147}{2} \\ 0 & 0 & \frac{545}{2} & 8175 \end{bmatrix}$$

Multiply  $R_3$  by  $\frac{2}{545}$ .

$$\begin{bmatrix} 1 & 3 & \frac{5}{4} & \frac{555}{2} \\ 0 & 1 & \frac{9}{20} & \frac{147}{2} \\ 0 & 0 & 1 & 30 \end{bmatrix}$$

Plugging in  $x_3 = 30$  to second row gives us:

$$x_2 + \frac{9}{20} \cdot 30 = \frac{147}{2}$$

$$x_2 = 60$$

Plugging in  $x_3 = 30$  and  $x_2 = 60$  to first row gives us:

$$x_1 + 3 \cdot 60 + \frac{5}{4} \cdot 30 = \frac{555}{2}$$

$$x_1 = 60$$

Unit price of first product is \$ 60, unit price of the second product is \$60, unit price of the third product is \$ 30.

$$200 \cdot 60 + 600 \cdot 60 + 250 \cdot 30 = 55500$$

$$250 \cdot 60 + 500 \cdot 60 + 200 \cdot 30 = 51000$$

$$300 \cdot 60 + 350 \cdot 60 + 400 \cdot 30 = 51000$$

All equations are true for the values we have found.