

## Problem Set 1

### 1 Problem 1

This problem asked us to calculate the numerical derivative of a function using four points. First, we would Taylor expand all of the terms:

$$f(x - \delta) = f(x) - \delta f'(x) + \frac{\delta^2}{2} f''(x) - \frac{\delta^3}{6} f'''(x) + \frac{\delta^4}{24} f^{(4)}(x) - \frac{\delta^5}{120} f^{(5)}(\xi_1) \quad (1)$$

$$f(x + \delta) = f(x) + \delta f'(x) + \frac{\delta^2}{2} f''(x) + \frac{\delta^3}{6} f'''(x) + \frac{\delta^4}{24} f^{(4)}(x) + \frac{\delta^5}{120} f^{(5)}(\xi_2) \quad (2)$$

$$f(x - 2\delta) = f(x) - 2\delta f'(x) + 2\delta^2 f''(x) - \frac{4\delta^3}{3} f'''(x) + \frac{2\delta^4}{3} f^{(4)}(x) - \frac{4\delta^5}{15} f^{(5)}(\xi_3) \quad (3)$$

$$f(x + 2\delta) = f(x) + 2\delta f'(x) + 2\delta^2 f''(x) + \frac{4\delta^3}{3} f'''(x) + \frac{2\delta^4}{3} f^{(4)}(x) + \frac{4\delta^5}{15} f^{(5)}(\xi_4) \quad (4)$$

Here, in order for us to achieve an estimation with the smallest error term:  $f^{(5)}$ , we would need to factor Equation (1) and (2) by 8, which cancels out the fourth derivative of the terms.

$$\begin{aligned} f(x + 2\delta) - 8f(x - \delta) + 8f(x + \delta) - f(x - 2\delta) = \\ 12\delta f'(x) + \frac{\delta^5}{15} f^{(5)}(\xi_1) + \frac{\delta^5}{120} f^{(5)}(\xi_2) - \frac{4\delta^5}{15} f^{(5)}(\xi_3) - \frac{4\delta^5}{15} f^{(5)}(\xi_4) \end{aligned} \quad (5)$$

The terms on involving  $\xi_n$  on the right side of the equation are the truncation errors and could be estimated by taking the greatest error of all terms of summing up the coefficients, which gives:

$$\left( \frac{\delta^5}{15} + \frac{\delta^5}{15} + \frac{\delta^5}{120} + \frac{\delta^5}{120} \right) \times \max |f^{(5)}(x)| = \frac{2\delta^5}{3} \max |f^{(5)}(x)| \quad (6)$$

Plugging Equation (6) back into Equation (5), and rearranging, we have:

$$f'(x) \leq \frac{f(x + 2\delta) - 8f(x - \delta) + 8f(x + \delta) - f(x - 2\delta)}{12\delta} + \frac{\delta^4}{18} \max |f^{(5)}(x)| \quad (7)$$

Hence, our estimate of the first derivative should be:

$$f'(x) = \frac{f(x+2\delta) - 8f(x-\delta) + 8f(x+\delta) - f(x-2\delta)}{12\delta} \quad (8)$$

Now, besides the truncation error, we also have a round-off error,  $\epsilon$ , which in this case could be written as:

$$E = \frac{f(x+2\delta)\epsilon_1 - 8f(x-\delta)\epsilon_2 + 8f(x+\delta)\epsilon_3 - f(x-2\delta)\epsilon_4}{12\delta} \quad (9)$$

Taking  $\epsilon^*$  as the round off error of  $f(x+\delta)$ , we can rewrite equation (9) as:

$$E \leq \frac{3\epsilon^*}{\delta} |f(x)| \quad (10)$$

Hence the error term including truncation error becomes

$$E(\delta) \leq \frac{\delta^4}{18} \max |f^{(5)}(x)| + \frac{3\epsilon^*}{\delta} |f(x)| \quad (11)$$

Notice that  $\delta$  is not a linear function, but with maxima. Hence, by finding the solution of the first derivative, we could acquire  $\delta$  that minimize  $E$ .

Hence, we have:

$$E'(\delta) = \frac{2\delta^3}{9} |f^{(5)}(x)| - \frac{3\epsilon^*}{\delta^2} |f(x)| = 0 \quad (12)$$

which gives:

$$\delta_{\min} = \sqrt[5]{\frac{27\epsilon^* |f(x)|}{2|f^{(5)}(x)|}} \quad (13)$$

Here, we apply our results to the code at  $x = 42$ , with function  $f(x) = e^x$  and  $f(x) = 0.01e^x$ . And the results from python shows:

```
The derivative of exp(x) at 42 is 1.739274941516718e+18
the fractional error is 2.175037927543144e-12
The derivative of exp(0.01x) at 42 is 0.015219615556183353
the error is 1.9606538614880265e-13
```

In which the fractional error is satisfactory.

## 2 Problem 2

Here, we could like to evaluate the error using of a two point numerical differentiator. Using the same method as Problem one, we have:

$$E(\delta) = \frac{\delta^2}{6} |f'''(x)| - \frac{\epsilon^*}{\delta} |f(x)| = 0 \quad (14)$$

which the optimal value of  $\delta$  to achieve the smallest error would be the zero solution of the first derivative.

$$E'(\delta) = \frac{\delta}{3}|f'''(x)| - \frac{\epsilon^*}{\delta^2}|f(x)| = 0 \quad (15)$$

which gives:

$$\delta_{\min} = \sqrt[3]{\frac{3\epsilon^*|f(x)|}{|f'''(x)|}} \quad (16)$$

### 3 Problem 3

Straightforward, see comment in code

### 4 Problem 4

When using the cosine function, the polynomial and cubic spline fit generates an error on the order of  $10^{-11}$  and  $10^{-12}$  respectively. However when using the rational function fit, the error term is on the order of  $10^{-6}$ . This discrepancy in error should be caused by the fact that the cosine function could not be approximated well by the divisions of two functions.

When evaluating the Lorentz function, I generated an error on the order of  $10^{-9}$  and  $10^{-10}$  respectively using the poly and cubic spline fit. However, when choosing the rational function fit with  $n = 2$  and  $m = 2$ , we have the error term down to  $10^{-16}$ . If we choose  $n = 4$  and  $m = 5$ , we error increase to the order of  $10^{-3}$ . However, if we use `np.linalg.pinv` instead, the error term for  $n = 4$  and  $m = 5$  decreases down to  $10^{-16}$ . This allows python to evaluate a pseudo inverse of the matrices that are singular and non-square. Previously, `np.linalg.inv` is only used to evaluate non-singular squared matrix and failed for other conditions. The values of  $p$  and  $q$  becomes extremely close to 0 with the `np.linalg.pinv` evaluation. And the constant term in the denominator is to ensure we are not dividing by 0.