

# Numerical Differentiation

Hung Doan

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## Abstract

When calculating a derivative directly is not possible, or numerically cumbersome, it may be better to numerically approximate the derivative. We can do this by using a Taylor series to approximate the function. By rearranging the terms in the Taylor series, we are able to obtain a formula that approximates the error in the derivative of a function. In this case, we will be approximating  $f(x) = \sin(x)$ . After obtaining the error function, we can take the derivative of the function to find optimal solution of  $h$ ,  $h_{opt}$  which minimizes the error in our approximation of the derivative. In this project, I will be using  $h$  values  $h = 10^{-i}$  for  $i = 1, 2, \dots, 16$  to calculate the error in our approximation of the derivative. I will then compare the results to the optimal solution of  $h$ .

## 1 Introduction

In this project, we will be numerically approximating the derivative of the function  $f(x) = \sin(x)$  at  $x = a$  for  $a = 0, 1$ . The Taylor expansion is given by

$$f(a+h) = f(a) + f'(a)h + \frac{f''(c)}{2}h^2, \quad c \in [a, a+h].$$

Using the definition of derivative, we can define  $D_h$ , to be

$$D_h = \frac{f(a+h) - f(a)}{h}. \quad (*)$$

As  $h \rightarrow 0$ ,  $D_h \rightarrow f'(a)$ . From the Taylor series, we can define  $D_h$  to be

$$D_h = f'(a) + \frac{f''(c)}{2}h. \quad (**)$$

## 2 Error in the Approximation of the Derivative

By subtracting  $f'(a)$  on both sides of (\*\*), and accounting for the machine error along with the magnification of the error created by the  $\frac{1}{h}$  term in the  $D_h$  equation, we have our error equation given by

$$|D_h - f'(a)| \approx \frac{|f''(c)|}{2}h + \frac{2eps}{h}.$$

Next, we would like to find a bound  $M_2$  such that  $|f''(x)| \leq M_2$  for all  $x \in [a, a + h]$ . Since  $|f''(x)| = |-\sin(x)|$  and  $0 \leq |-\sin(x)| \leq 1$  for all  $x$ , we can choose  $M_2 = 1$ . Now, define the error,  $errD(h)$  as

$$errD(h) = M_2 \frac{h}{2} + \frac{2\epsilon ps}{h}. \quad (***)$$

Using  $(***)$  with  $h = 10^i$  for  $i = 1, 2, \dots, 16$ , and  $M_2 = 1$ , we obtain Figure 1 and Table 1.

$h$	error (a=0)	error (a=1)
1.0000e-01	5.0000e-02	5.0000e-02
1.0000e-02	5.0000e-03	5.0000e-03
1.0000e-03	5.0000e-04	5.0000e-04
1.0000e-04	5.0000e-05	5.0000e-05
1.0000e-05	5.0000e-06	5.0000e-06
1.0000e-06	5.0044e-07	5.0044e-07
1.0000e-07	5.4441e-08	5.4441e-08
1.0000e-08	4.9409e-08	4.9409e-08
1.0000e-09	4.4459e-07	4.4459e-07
1.0000e-10	4.4409e-06	4.4409e-06
1.0000e-11	4.4409e-05	4.4409e-05
1.0000e-12	4.4409e-04	4.4409e-04
1.0000e-13	4.4409e-03	4.4409e-03
1.0000e-14	4.4409e-02	4.4409e-02
1.0000e-15	4.4409e-01	4.4409e-01
1.0000e-16	4.4409e+00	4.4409e+00

Table 1: Error in Approximation of the derivative when  $a = 0$  and  $a = 1$ .

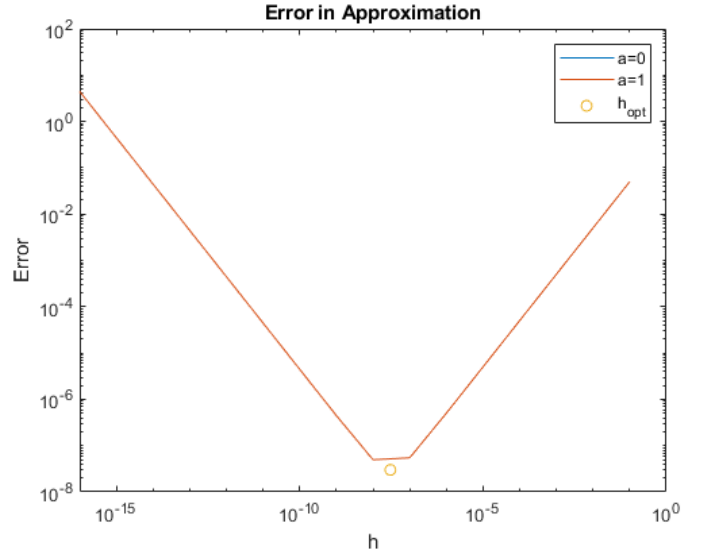


Figure 1: Error in Approximation of the derivative when  $a = 0$  and  $a = 1$ .

We can see that error is as low as  $4.9409 \cdot 10^{-8}$  at  $h = 10^{-8}$  for both  $a = 0$  and  $a = 1$ .

### 3 Finding the Optimal $h$ Value

Equation  $(***)$  is given by

$$errD(h) = M_2 \frac{h}{2} + \frac{2\epsilon ps}{h}.$$

The optimal  $h$  value is given when  $errD(h)$  is 0. Then

$$\begin{aligned}
0 &= \frac{d}{dh} \left( M_2 \frac{h}{2} + \frac{2eps}{h} \right) \\
0 &= \frac{M_2}{2} - \frac{2eps}{h^2} \\
\frac{M_2}{2} &= \frac{2eps}{h^2} \\
h^2 \cdot M_2 &= 4eps \\
h^2 &= \frac{4eps}{M_2} \\
h_{opt} &= 2\sqrt{\frac{eps}{M_2}}
\end{aligned}$$

We can now use this to find the optimal  $h$  value that minimizes the error in the derivative. Using our derived formula for  $h_{opt}$ , we have

$$h_{opt} = 2\sqrt{\frac{eps}{M_2}} = 2\sqrt{\frac{eps}{1}} = 2\sqrt{eps} = 2.9802 \cdot 10^{-8}.$$

This is in agreement to our observations in Figure 1.

## 4 Complex Step Differentiation (CSD)

Complex step differentiation is a method of numerically providing accurate approximations for the derivatives of functions. By slightly manipulating the definition of derivative, we obtain the scheme

$$\frac{f(x+h) - f(x-h)}{2h} = f'(x) + \frac{h^2}{6} f^{(3)}(x) + \dots$$

We can subtract  $f'(x)$  on both sides to obtain the error. Using the same methods as from above, we arrive at an error equation of

$$errD(h) = \frac{h^2}{6} M_2 + \frac{2eps}{h}$$

Setting the error to 0, we can obtain our  $h_{opt}$ .

$$\begin{aligned}
0 &= \frac{d}{dh} \left( M_2 \frac{h^2}{6} + \frac{2eps}{h} \right) \\
0 &= \frac{2M_2h}{6} - \frac{2eps}{h^2} \\
\frac{M_2h}{3} &= \frac{2eps}{h^2} \\
h^3 \cdot M_2 &= 6eps \\
h^3 &= \frac{6eps}{M_2} \\
h_{opt} &= \sqrt[3]{\frac{6eps}{M_2}}
\end{aligned}$$

Again, I will be using  $h$  values  $h = 10^{-i}$  for  $i = 1, 2, \dots, 16$  to calculate the error in our approximation of the derivative. I will then compare the results to the optimal solution of  $h$ .

$h$	error (a=0)	error (a=1)
1.0000e-01	1.6667e-03	1.6667e-03
1.0000e-02	1.6667e-05	1.6667e-05
1.0000e-03	1.6667e-07	1.6667e-07
1.0000e-04	1.6711e-09	1.6711e-09
1.0000e-05	6.1076e-11	6.1076e-11
1.0000e-06	4.4426e-10	4.4426e-10
1.0000e-07	4.4409e-09	4.4409e-09
1.0000e-08	4.4409e-08	4.4409e-08
1.0000e-09	4.4409e-07	4.4409e-07
1.0000e-10	4.4409e-06	4.4409e-06
1.0000e-11	4.4409e-05	4.4409e-05
1.0000e-12	4.4409e-04	4.4409e-04
1.0000e-13	4.4409e-03	4.4409e-03
1.0000e-14	4.4409e-02	4.4409e-02
1.0000e-15	4.4409e-01	4.4409e-01
1.0000e-16	4.4409e+00	4.4409e+00

Table 2: Error in Approximation of the derivative when  $a = 0$  and  $a = 1$ .

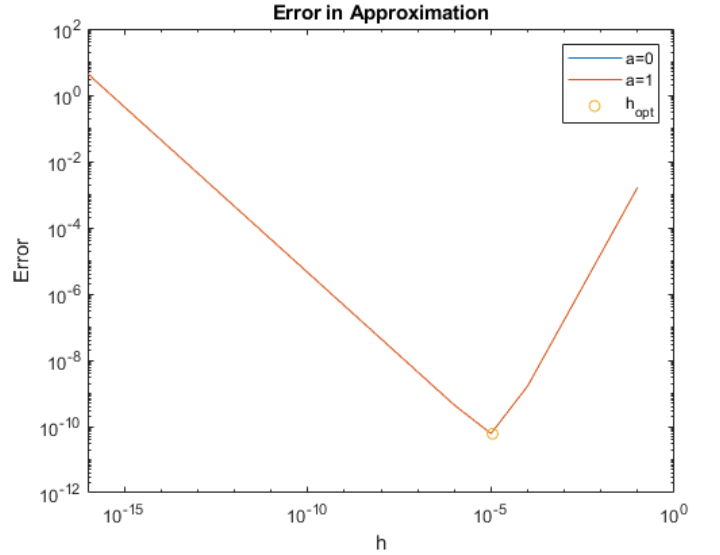


Figure 2: Error in Approximation of the derivative when  $a = 0$  and  $a = 1$ .

This time, we can see that the error is a lot lower than with our previous methods. Here we have error as low as  $6.1076 \cdot 10^{-11}$  and  $h_{opt} = 1.1003 \cdot 10^{-5}$ .