

## Mathematics 25: Numerical Analysis

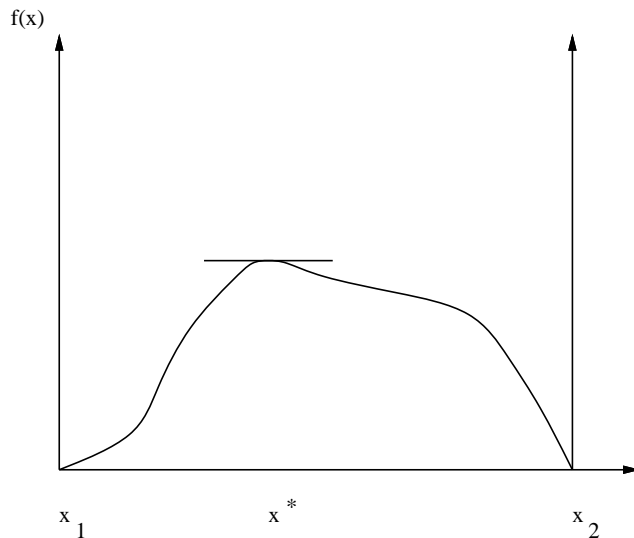
Numerical analysis of most differential algorithms is founded on *three* major mathematical results:

- (1) Rolle's Theorem
- (2) Taylor's Theorem
- (3) Interpolation Formulae

Rolle's Theorem is the fundamental, and the other two follow fairly directly.

### (1) Rolle's Theorem

If a continuously differentiable function,  $f(x)$  say, vanishes at two points,  $x_1$  and  $x_2$ , then  $f^{(1)}(x^*) = 0$ , for some  $x^* \in (x_1, x_2)$ .



$f(x)$  starts out either up or down. Eventually this initial direction must reverse, and this point marks  $f^{(1)}(x^*) = 0$ .

### (2) Taylor's Theorem

$$f(x+h) = \sum_{m=0}^n \frac{h^m}{m!} f^{(m)}(x) + \frac{h^{n+1}}{(n+1)!} f^{(n+1)}(x+\theta h)$$

for some  $\theta \in (0, 1)$ . To prove this result we set:

$$g(y) = f(x+y) - \sum_{m=0}^n \frac{y^m}{m!} f^{(m)}(x) - Ay^{n+1}$$

and choose  $A$  so that  $g(h) = 0$ . Note that:

$$g(0) = g^{(1)}(0) = \dots = g^{(n)}(0) = 0$$

We now employ Rolle's Theorem  $n$  times. Now  $g(0) = 0 = g(h)$  and so there is a point where  $g^{(1)}(\theta_1 h) = 0$ , for some  $\theta_1 \in (0, 1)$ . Now  $g^{(1)}(0) = 0 = g^{(1)}(\theta_1 h)$  and so there is a point where  $g^{(2)}(\theta_2 \theta_1 h) = 0$  for some  $\theta_2 \in (0, 1)$ . This argument proceeds one at a time

until, finally, we find a point where,  $g^{(n+1)}(\theta_n \theta_{n-1} \dots \theta_2 \theta_1 h) = 0$  for some  $\theta_n \in (0, 1)$ . We set  $\theta = \theta_n \theta_{n-1} \dots \theta_2 \theta_1$  and then we immediately see that:

$$g^{(n+1)}(\theta h) = 0 = f^{(n+1)}(x + \theta h) - (n+1)!A$$

and Taylor's Theorem is proven.

### (3) Interpolation Formulae

The basic idea here is to use *difference formulae* to construct the interpolating polynomial. A function,  $f(x)$ , is central and then subsequent differences and their inverses are defined by:

$$\begin{aligned} [x_0 x_1] &= \frac{f(x_0) - f(x_1)}{x_0 - x_1} & f(x_0) &= f(x_1) + (x_0 - x_1)[x_0 x_1] \\ [x_0 x_1 x_2] &= \frac{[x_0 x_1] - [x_1 x_2]}{x_0 - x_2} & [x_0 x_1] &= [x_1 x_2] + (x_0 - x_2)[x_0 x_1 x_2] \\ [x_0 x_1 x_2 x_3] &= \frac{[x_0 x_1 x_2] - [x_1 x_2 x_3]}{x_0 - x_3} & [x_0 x_1 x_2] &= [x_1 x_2 x_3] + (x_0 - x_3)[x_0 x_1 x_2 x_3] \end{aligned}$$

and so on. If we include a special, *variable* point,  $x$  into the description, then:

$$\begin{aligned} f(x) &= f(x_0) + (x - x_0)([x_0 x_1] + (x - x_1)([x_0 x_1 x_2] + (\dots (x - x_{n-1})[x_0 x_1 \dots x_n])) \dots) \\ &\quad + (x - x_0)(x - x_1) \dots (x - x_n)[x_0 x_1 \dots x_n] \equiv P(x) + R(x) \end{aligned}$$

where  $P(x)$  is the initial polynomial, depending only on the values of  $f(x)$  evaluated at the points  $x_m$ , and  $R(x)$  is the final term:

$$R(x) = (x - x_0)(x - x_1) \dots (x - x_n)[x_0 x_1 \dots x_n]$$

Each difference reduces a polynomial by one degree, so if the original function were a polynomial of degree  $n$ , the corresponding  $R(x)$  would vanish. This shows that  $P(x)$  is the *unique* interpolating polynomial which agrees with the function  $f(x)$  at the special points  $x_m$ .

Unless the function is 'pathological', then  $R(x_m) = 0$  for each  $x_m$ . This enables us to apply Rolle's Theorem. There are initially  $n+1$  zeroes, and so between each neighbouring pair there is a point for which  $R^{(1)}(x^*)$  vanishes. There are now  $n$  zeroes in  $R^{(1)}(x)$ , and so between each neighbouring pair there is a point for which  $R^{(2)}(x^*) = 0$ . We can sequentially proceed with this argument, until eventually we find a single point for which,  $R^{(n)}(x^*) = 0$ . A direct differentiation of the explicit representation then provides us with:

$$R^{(n)}(x^*) = 0 = f^{(n)}(x^*) - n![x_0 x_1 x_2 \dots x_n]$$

We finally employ a rather subtle idea, and note that  $R(x)$  itself is composed of an  $n+1$ 'th difference, and so if we include the point  $x$  itself in amongst the  $x_m$  and repeat the argument we find:

$$[x x_0 x_1 x_2 \dots x_n] = \frac{f^{(n+1)}(x^*)}{(n+1)!}$$

for some  $x^*$  contained in the region covered by the  $x_m$  and  $x$ ! The final interpolation formula therefore becomes:

$$f(x) = f(x_0) + (x - x_0)([x_0 x_1] + (x - x_1)([x_0 x_1 x_2] + (\dots (x - x_{n-1})[x_0 x_1 \dots x_n])) \dots) \\ + (x - x_0)(x - x_1) \dots (x - x_n) \frac{f^{(n+1)}(x^*)}{(n+1)!}$$

For equal spaced points at  $x_n = nh$ , we find:

$$[x_0 x_1 \dots x_r] = \frac{[x_1 x_2 \dots x_r] - [x_0 x_1 \dots x_{r-1}]}{rh} \equiv \frac{(\Delta - 1)[x_0 x_1 \dots x_{r-1}]}{rh} = \frac{(\Delta - 1)^r}{r!h^r} f_0$$

in terms of the operator  $\Delta$ , which raises all the indicies by one. In these terms:

$$f(x) = f_0 + h \frac{x}{h} (\Delta - 1) f_0 + \frac{h^2}{2!} \frac{x}{h} \left( \frac{x}{h} - 1 \right) (\Delta - 1)^2 f_0 + \frac{h^3}{3!} \frac{x}{h} \left( \frac{x}{h} - 1 \right) \left( \frac{x}{h} - 2 \right) (\Delta - 1)^3 f_0 + \dots$$

in terms of:

$$(\Delta - 1)^n f_0 = \sum_{m=0}^n \frac{n!}{m!(n-m)!} (-1)^{n-m} f_m$$

Since all errors from a truncation to this sequence at order  $n$ , are order  $h^{n+1}$ , we can form approximations to derivatives of arbitrary accuracy by using as many more terms than derivatives as is required to achieve our desired accuracy.