# MA 214: Numerical Analysis Notes

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#### DISCLAIMER

This is just a collection of formulae/algorithms compiled together.

In the case of algorithms, I explain the procedure concisely. However, do not take this as a substitute for lecture slides as I don't go into the theory at all.

Also, I've modified some things compared to the lecture slides wherever I felt it was an error. So, be warned.

# 1 Interpolation

#### 1. Lagrange Polynomials

Let  $x_0, x_1, \ldots, x_n$  be n+1 distinct points in [a, b]. Let  $f : [a, b] \to \mathbb{R}$  be a function whose value is known at those aforementioned points.

We want to construct a polynomial p(x) of degree  $\leq n$  such that  $p(x_i) = f(x_i)$  for all  $i \in \{0, ..., n\}$ . Towards this end, we define the polynomials  $I_k(x)$  for  $k \in \{0, ..., n\}$  in the following manner:

$$I_k(x) := \prod_{i=0}^n \frac{x - x_i}{x_k - x_i}.$$

(Intuitive understanding:  $I_k$  is a degree n polynomial such that  $I_k(x_j) = 0$  if  $k \neq j$  and  $I_k(x_k) = 1$ .) Now, define p(x) as follows:

$$p(x) := \sum_{i=0}^{n} f(x_i) I_i(x)$$

#### 2. Newton's form

Let  $x_0, x_1, \ldots, x_n$  be n+1 distinct points in [a, b]. Let  $f : [a, b] \to \mathbb{R}$  be a function whose value is known at those aforementioned points.

We want to construct a polynomial  $P_n(x)$  of degree  $\leq n$  such that  $p(x_i) = f(x_i)$  for all  $i \in \{0, \dots, n\}$ .

We define the divided differences (recursively) as follows:

$$f[x_0] := f(x_0)$$
 
$$f[x_0, x_1, \dots, x_k] := \frac{f[x_1, \dots, x_k] - f[x_0, \dots, x_{k-1}]}{x_k - x_0}$$
 for all  $1 < k \le n$ 

With this in place, the desired polynomial  $P_n(x)$  is (not so) simply:

$$P_n(x) := f[x_0] + f[x_0, x_1](x - x_0)$$

$$+ f[x_0, x_1, x_2](x - x_0)(x - x_1)$$

$$+ \dots$$

$$\vdots$$

$$+ f[x_0, x_1, \dots, x_n](x - x_0)(x - x_1) \cdots (x - x_{n-1})$$

Remarks. Note that  $x - x_n$  does not appear in the last term.

Note that given  $P_n(x)$ , it is simple to construct  $P_{n+1}(x)$ .

### 3. Osculatory Interpolation

This is essentially the same as the previous case.

I'll state the problem in the form I think is the simplest. (Any other form can be reduced to this.) Suppose we are given k+1 distinct points  $x_0, \ldots, x_k$  in [a,b] and a function  $f:[a,b] \to \mathbb{R}$  which is sufficiently differentiable.

Suppose we are given the following values:

$$f^{(0)}(x_0), f^{(1)}(x_0), \dots, f^{(m_1-1)}(x_0)$$

$$f^{(0)}(x_1), f^{(1)}(x_1), \dots, f^{(m_2-1)}(x_1)$$

$$\vdots$$

$$f^{(0)}(x_k), f^{(1)}(x_k), \dots, f^{(m_k-1)}(x_k)$$

(Notation:  $f^{(0)}(x) = f(x)$  and  $f^{(n)}(x)$  is the  $n^{\text{th}}$  derivative.)

Thus, we are given  $m_1 + m_2 + \cdots + m_k =: n + 1$  data. As usual, we now want to compute a polynomial  $P_n(x)$  that agrees with f at all the data. (That is, all the given derivatives must also be same.) As it goes without saying,  $P_n(x)$  must have degree  $\leq n$ .

To do this, we list the above points as follows:

$$\underbrace{x_0, x_0, \dots x_0}_{m_1}, \underbrace{x_1, x_1, \dots, x_1}_{m_2}, \dots, \underbrace{x_k, x_k, \dots, x_k}_{m_k}.$$

Now, we just apply the above (Newton's) formula with the following modification in the definition of the divided difference:

$$f[\underbrace{x_i, x_i, \dots, x_i}_{p+1 \text{ times}}] := \frac{f^{(p)}(x_i)}{p!}.$$

#### 4. Richardson Extrapolation

Suppose that for sufficiently small  $h \neq 0$ , we have the formula:

$$M = N_1(h) + k_1h + k_2h^2 + k_3h^3 + \cdots$$

for some constants  $k_1, k_2, k_3, \ldots$ 

Define the following:

$$N_j(h) := N_{j-1}(h/2) + \frac{N_{j-1}(h/2) - N_{j-1}(h)}{2^{j-1} - 1}$$
 for  $j \ge 2$ .

Choose some h sufficiently small (whatever that means). Then,  $N_j(h)$  keeps becoming a better approximation of M as j increases.

We create a table of h and  $N_i(h)$  as follows:

 $N_4(h)$  will be a good approximation, then.

(Look at slide 15 of Lecture 7 for an example.)

### Special case

Sometimes, we may have the following scenario:

$$M = N_1(h) + k_2h^2 + k_4h^4 + \cdots$$

In this case, we define:

$$N_j(h) := N_{j-1}(h/2) + \frac{N_{j-1}(h/2) - N_{j-1}(h)}{4^{j-1} - 1} \quad \text{ for } j \ge 2.$$

Then, we do the remaining stuff as before.

# 2 Numerical Integration

$$I = \int_{a}^{b} f(x) \mathrm{d}x$$

1. Rectangle Rule

$$I\approx (b-a)f(a)$$
 
$$E^R=f'(\eta)\frac{(b-a)^2}{2}, \text{ for some } \eta\in [a,b]$$

2. Midpoint Rule

$$I\approx (b-a)f\left(\frac{a+b}{2}\right)$$
 
$$E^M=\frac{f''(\eta)}{24}(b-a)^3, \text{ for some } \eta\in[a,b]$$

3. Trapezoidal Rule

$$I \approx \frac{1}{2}(b-a)[f(a)+f(b)]$$
 
$$E^T = -f''(\eta)\frac{(b-a)^3}{12}, \text{ for some } \eta \in [a,b]$$

4. Corrected Trapezoidal

$$I \approx \frac{1}{2}(b-a)[f(a)+f(b)] + \frac{(b-a)^2}{12}(f'(a)-f'(b))$$
 
$$E^{CT} = f^{(4)}(\eta)\frac{(b-a)^5}{720}, \text{ for some } \eta \in [a,b]$$

5. Composite Trapezoidal

$$I \approx \frac{h}{2} \left[ f(x_0) + 2 \sum_{i=1}^{N-1} f(x_i) + f(x_N) \right]$$
 
$$E_C^T = -f''(\xi) \frac{h^2(b-a)}{12}, \text{ for some } \xi \in [a,b]$$

Here, Nh = b - a and  $x_i = a + ih$ .

6. Simpson's Rule

$$\begin{split} I &\approx \frac{b-a}{6} \left\{ f(a) + 4f\left(\frac{a+b}{2}\right) + f(b) \right\} \\ E^S &= -\frac{1}{90} f^{(4)}(\eta) \left(\frac{b-a}{2}\right)^5, \text{ for some } \eta \in [a,b] \end{split}$$

7. Composite Simpson's

$$I \approx \frac{h}{6} [f(x_0) + 2 \sum_{i=1}^{N-1} f(x_i) + 4 \sum_{i=1}^{N} f\left(x_{i-1} + \frac{h}{2}\right) + f(x_N)]$$
$$E_C^S = -f^{(4)}(\xi) \frac{(h/2)^4 (b-a)}{180}, \text{ for some } \xi \in [a, b]$$

Here,  $\overline{Nh = b - a}$  and  $x_i = a + ih$ .

#### 8. Gaussian Quadrature

Let  $Q_{n+1}(x)$  denote the  $(n+1)^{\text{th}}$  Legendre polynomial.

Let  $r_0, \ldots, r_{n+1}$  be its roots. (These will be distinct, symmetric about the origin and will lie in the interval [-1,1].

For each  $i \in \{0, ..., n\}$ , we define  $c_i$  as follows:

$$c_i := \int_{-1}^1 \left( \prod_{k=0, k \neq i}^n \frac{x - x_k}{x_i - x_k} \right) \mathrm{d}x.$$

Then, we have

$$\int_{-1}^{1} f(x) dx \approx \sum_{i=0}^{n} f(r_i) c_i.$$

Moreover, if f is a polynomial of degree  $\leq 2n+1$ , then the above is "approximation" is exact.

Standard values:

$$n=0: Q_1(x)=x \text{ and } x_0=0.$$
  $c_0=2.$   $n=1: Q_2(x)=x^2-\frac{1}{3} \text{ and } x_0=-\frac{1}{\sqrt{3}}, \ x_1=\frac{1}{\sqrt{3}}.$   $c_0=c_1=1.$   $n=2: Q_3(x)=x^3-\frac{3}{5}x \text{ and } x_0=-\sqrt{\frac{3}{5}}, \ x_1=0, \ x_2=\sqrt{\frac{3}{5}}.$   $c_0=c_2=5/9, \ c_1=8/9.$ 

### 9. Improper integrals using Taylor series method

Suppose we have  $f(x) = \frac{g(x)}{(x-a)^p}$  for some  $0 and are asked to calculate <math>I = \int_0^b f(x) dx$ .

For the sake of simplicity, I assume a = 0 and b = 1.

Let  $P_4(x)$  denote the fourth Taylor polynomial of g around a. (In this case 0.)

Now, compute 
$$I_1 = \int_0^1 \frac{P_4(x)}{x^p} dx$$
. This can be integrated exactly. (Why?)  
Now, we approximate  $I - I_1$ .

Define

$$G(x) := \begin{cases} \frac{g(x) - P_4(x)}{x^p} & \text{if } 0 < x \le 1\\ 0 & \text{if } x = 0 \end{cases}$$

Then, approximate  $I_2 = \int_0^1 G(x) dx$  using the composite Simpson's rule.

For the case of a = 0, b = 1 and N = 2 for the composite Simpson's part, we get that  $I_2 \approx \frac{1}{12} [2G(0.5) + 4G(0.25) + 4G(0.75) + G(1)].$ 

That is, finally:

$$I \approx \int_0^1 \frac{P_4(x)}{x^p} dx + \frac{1}{12} [2G(0.5) + 4G(0.25) + 4G(0.75) + G(1)].$$

## 10. Adaptive Quadrature

Let  $I = \int_{-\infty}^{0} f(x) dx$  be the integral that we want to approximate.

Suppose that  $\epsilon$  is the accuracy to which we want I. That is, we want a number P such that  $|P-I| < \epsilon$ . Here is what you do:

Subdivide [a, b] into N intervals:  $[x_0, x_1], [x_1, x_2], \ldots, [x_{n-1}, x_n]$ . (Naturally,  $a = x_0 \le x_1 \le \ldots \le x_n = x_n \le x_$ 

Now, for each subinterval, compute the following values:

$$S_{i} = \frac{h}{6} \left( f(x_{i}) + 4f\left(x_{i} + \frac{h}{2}\right) + f\left(x_{i+1}\right) \right), \text{ and}$$

$$\overline{S_{i}} = \frac{h}{12} \left( f(x_{i}) + 4f\left(x_{i} + \frac{h}{2}\right) + 2f\left(x_{i} + \frac{h}{2}\right) + 4f\left(x_{i} + \frac{3h}{4}\right) + f(x_{i+1}) \right).$$

Now, calculate  $E_i = \frac{1}{15} |\overline{S_i} - S_i|$ .

Now, if  $E_i \leq \frac{x_i - x_{i-1}}{b-a} \epsilon$ , then move on to the next interval.

Otherwise, subdivide again to better approximate  $\int_{-\infty}^{x_i} f(x) dx$ .

Finally, sum up all the  $\overline{S_i}$ s and that's the answer. That is,

$$I \approx P = \sum_{i=1}^{n} \overline{S_i}.$$

#### 11. Romberg Integration

Essentially the baby of composite Trapezoidal rule and Romberg integration.

Suppose we want to calculate  $\int_a^b f(x) dx$ . Let N be a power of 2.

$$T_N := \frac{h}{2} \left[ f(x_0) + 2 \sum_{i=1}^{N-1} f(a+ih) + f(x_N) \right], \text{ where } Nh = b - a.$$

Note that  $T_N$  can be computed using  $T_{N/2}$  (assuming  $N \neq 2^0$ ) as:

$$T_N = \frac{T_{N/2}}{2} + h \sum_{i=1}^{N/2} f(a + (2i - 1)h).$$

Now for  $m \geq 1$ , we define:

$$T_N^m = T_N^{m-1} + \frac{T_N^{m-1} - T_{N/2}^{m-1}}{4^m - 1}.$$

(Where  $T_N^0$  is just  $T_N$ .) (Also, for some reason,  $T_N'$  has been used instead of  $T_N^1$ .) Note that  $\frac{N}{2^m}$  must be an integer for  $T_N^m$  to be defined. We create a table as follows:

 $T_8^3$  will be a good approximation, then.

(Look at slide 25 of Lecture 7 for an example.)

Remark. It can be shown that  $I=T_N+c_2h^2+c_4h^4+\cdots$ . This is why we used the special case formula of 1. 4.

#### Numerical Differentiation 3

1.

$$f'(a) \approx \frac{f(a+h) - f(a)}{h}$$
 
$$E(f) = -\frac{1}{2} h f''(\eta) \quad \text{ for some } \eta \in [a,a+h].$$

#### 2. Central Difference Formula

$$f'(a)\approx \frac{f(a+h)-f(a-h)}{2h}$$
 
$$E(f)=-\frac{1}{6}h^3f^{(3)}(\eta)\quad \text{ for some }\eta\in[a-h,a+h].$$

Note that this is an  $O(h^2)$  approximation. Thus, we can use the special case of §1. 4. for better accuracy.

3.  $f'(a) \approx \frac{-3f(a) + 4f(a+h) - f(a+2h)}{2h}$  $E(f) = \frac{1}{2}h^3f^{(3)}(\eta)$  for some  $\eta \in [a, a + 2h]$ . Formula 2 is always the better one whenever applicable. At end points, formula 3 is better than formula 1.

### 4. Central difference for second derivative

$$f''(x_0) = \frac{f(x_0 + h) - 2f(x_0) + f(x_0 - h)}{h^2} - \frac{h^2}{12}f^{(4)}(\xi),$$

for some  $\xi \in (x_0 - h, x_0 + h)$ .

#### 5. Solving boundary-value problems in ODE

Suppose that we want to solve the following (linear) ODE:

$$y''(x) + f(x)y'(x) + g(x)y = q(x)$$

in the interval [a, b] such that we know  $y(a) = \alpha$ , and  $y(b) = \beta$ .

Set  $h := \frac{b-a}{N}$  for some  $N \in \mathbb{N}$  and  $x_0 = a + ih$  for  $h \in \{0, 1, \dots, N\}$ . Using central difference approximation, we set up N-1 linear equations as follows:

$$\frac{y_{i-1} - 2y_0 + y_i}{h^2} + f(x_i) \frac{y_{i+1} - y_{i-1}}{2h} + g(x_i)(y_i) = q(x_i)$$
$$i = 1, 2, \dots, N - 1$$

The above equations can be rearranged as:

$$\left(1 - \frac{hf_i}{2}\right)y_{i-1} + (-2 + h^2g_i)y_i + \left(1 + \frac{hf_i}{2}\right)y_{i+1} = h^2q_i,$$

for i = 1, ..., N - 1; where  $f_i = f(x_i)$  and so on.

#### Solution of non-linear equations 4

Let f be a continuous function on  $[a_0, b_0]$  such that  $f(a_0)f(b_0) < 0$  in all these cases. We want to find a root of f in  $[a_0, b_0]$ . (Existence in implied.)

### 1. Bisection Method

Set n = 0 to start with.

Loop over the following:

Set  $m = \frac{a_n + b_n}{2}$ .

If  $f(a_n)f(m) < 0$ , then set  $a_{n+1} = a_n$  and  $b_{n+1} = m$ .

Else, set  $a_{n+1} = m$  and  $b_{n+1} = b_n$ .

Increase n by one.

We still have a root in  $[a_n, b_n]$ .

#### 2. Regula-falsi or false-position method

Set n = 0 to start with.

Loop over the following:

Set  $w = \frac{f(b_n)a_n - f(a_n)b_n}{f(b_n) - f(a_n)}$ . If  $f(a_n)f(w) < 0$ , then set  $a_{n+1} = a_n$  and  $b_{n+1} = w$ .

Else, set  $a_{n+1} = w$  and  $b_{n+1} = b_n$ .

Increase n by one.

We still have a root in  $[a_n, b_n]$ .

### 3. Modified regula-falsi

Set n = 0 and  $w_0 = a_0$  to start with.

Loop over the following:

Set 
$$F = f(a_n)$$
 and  $G = f(b_n)$ 

Set 
$$F = f(a_n)$$
 and  $G = f(b_n)$ .  
Set  $w_{n+1} = \frac{Ga_n - Fb_n}{G - F}$ .

If  $f(a_n)f(w_{n+1}) \leq 0$ , then set  $a_{n+1} = a_n$  and  $b_{n+1} = w_{n+1}$  and  $G = f(w_{n+1})$ . Furthermore, if we also have  $f(w_n)f(w_{n+1}) > 0$ , set  $F = \frac{F}{2}$ . Else, set  $a_{n+1} = w_{n+1}$  and  $b_{n+1} = b_n$  and  $F = f(w_{n+1})$ . Furthermore, if we also have  $f(w_n)f(w_{n+1}) > 0$ , set  $G = \frac{G}{2}$ .

Increase n by one.

We still have a root in  $[a_n, b_n]$ .

### 4. Secant method

Set  $x_0 = a$ ,  $x_1 = b$  and until satisfied, keep computing  $x_n$  given by

$$x_{n+1} = \frac{f(x_n)x_{n-1} - f(x_{n-1})x_n}{f(x_n) - f(x_{n-1})}$$
 for  $n \ge 1$ .

Remark. This process will be forced to stop if we arrive at  $f(x_n) = f(x_{n-1})$  at some point.

## 5 Iterative methods

1. You are given a function f which is continuously differentiable and you want to find its root. You are also given some  $x_0$ .

Compute the following sequence recursively until satisfied:

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)} \quad \text{for } n \ge 0.$$