

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, \dots, x_n be $n+1$ distinct points in $[a, b]$. Let $f : [a, b] \rightarrow \mathbb{R}$ 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 $i \in \{0, \dots, n\}$.

Towards this end, we define the polynomials $I_k(x)$ for $k \in \{0, \dots, n\}$ in the following manner:

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

(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, \dots, x_n be $n+1$ distinct points in $[a, b]$. Let $f : [a, b] \rightarrow \mathbb{R}$ 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:

$$\begin{aligned} f[x_0] &:= f(x_0) \\ f[x_0, x_1, \dots, x_k] &:= \frac{f[x_1, \dots, x_k] - f[x_0, \dots, x_{k-1}]}{x_k - x_0} \end{aligned} \quad \text{for all } 1 < k \leq n$$

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

$$\begin{aligned} P_n(x) &:= f[x_0] + f[x_0, x_1](x - x_0) \\ &\quad + f[x_0, x_1, x_2](x - x_0)(x - x_1) \\ &\quad + \dots \\ &\quad \vdots \\ &\quad + f[x_0, x_1, \dots, x_n](x - x_0)(x - x_1) \cdots (x - x_{n-1}) \end{aligned}$$

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, \dots, x_k in $[a, b]$ and a function $f : [a, b] \rightarrow \mathbb{R}$ which is sufficiently differentiable.

Suppose we are given the following values:

$$\begin{array}{c} f^{(0)}(x_0), f^{(1)}(x_0), \dots, f^{(m_1-1)}(x_0) \\ f^{(0)}(x_1), f^{(1)}(x_1), \dots, f^{(m_2-1)}(x_1) \\ \vdots \\ f^{(0)}(x_k), f^{(1)}(x_k), \dots, f^{(m_k-1)}(x_k) \end{array}$$

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

Thus, we are given $m_1 + m_2 + \dots + m_k =: n + 1$ 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:

$$M = N_1(h) + k_1 h + k_2 h^2 + k_3 h^3 + \dots,$$

for some constants k_1, k_2, k_3, \dots .

Define the following:

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

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 $N_j(h)$ as follows:

| h | $N_1(h)$ | $N_2(h)$ | $N_3(h)$ | $N_4(h)$ |
|-------|------------|------------|------------|----------|
| h | $N_1(h)$ | | | |
| $h/2$ | $N_1(h/2)$ | $N_2(h)$ | | |
| $h/4$ | $N_1(h/4)$ | $N_2(h/2)$ | $N_3(h)$ | |
| $h/8$ | $N_1(h/8)$ | $N_2(h/4)$ | $N_3(h/2)$ | $N_4(h)$ |

$N_4(h)$ 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:

$$M = N_1(h) + k_2 h^2 + k_4 h^4 + \dots.$$

In this case, we define:

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

Then, we do the remaining stuff as before.

2 Numerical Integration

$$I = \int_a^b f(x)dx$$

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

$$I \approx \frac{b-a}{6} \left\{ f(a) + 4f\left(\frac{a+b}{2}\right) + f(b) \right\}$$

$$E^S = -\frac{1}{90}f^{(4)}(\eta) \left(\frac{b-a}{2}\right)^5, \text{ for some } \eta \in [a, b]$$

7. Composite Simpson's

$$I \approx \frac{h}{6} \left[f(x_0) + 2 \sum_{i=1}^{N-1} f(x_i) + 4 \sum_{i=1}^N f\left(x_{i-1} + \frac{h}{2}\right) + f(x_N) \right]$$

$$E_C^S = -f^{(4)}(\xi) \frac{(h/2)^4(b-a)}{180}, \text{ for some } \xi \in [a, b]$$

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

8. Gaussian Quadrature

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

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

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

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

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:

$n = 0$: $Q_1(x) = x$ and $x_0 = 0$. $c_0 = 2$.

$n = 1$: $Q_2(x) = x^2 - \frac{1}{3}$ and $x_0 = -\frac{1}{\sqrt{3}}$, $x_1 = \frac{1}{\sqrt{3}}$. $c_0 = c_1 = 1$.

$n = 2$: $Q_3(x) = x^3 - \frac{3}{5}x$ 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 $0 < p < 1$ and are asked to calculate $I = \int_a^b f(x) dx$.

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

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

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

Then, $I = I_1 + I_2$.

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 $I = \int_a^b f(x) dx$ be the integral that we want to approximate.

Suppose that ϵ 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: $[x_0, x_1]$, $[x_1, x_2]$, \dots , $[x_{n-1}, x_n]$. (Naturally, $a = x_0 \leq x_1 \leq \dots \leq x_n = b$.)

Now, for each subinterval, compute the following values:

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

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

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 $\int_{x_{i-1}}^{x_i} f(x)dx$.

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:

| N | T_N | T'_N | T_N^2 | T_N^3 |
|-----|-------|---------|---------|---------|
| 1 | T_1 | | | |
| 2 | T_2 | T_2^1 | | |
| 4 | T_4 | T_4^1 | T_4^2 | |
| 8 | T_8 | T_8^1 | T_8^2 | T_8^3 |

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 $I = T_N + c_2 h^2 + c_4 h^4 + \dots$. This is why we used the special case formula of 1. 4.