Lecture 2

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1 Lagrange Interpolation

Motivation: for the problem

$$y(t_{k+1}) = y(t_k) + \int_{t_k}^{t_{k+1}} f(t, y(t)) dt$$

we want to construct a method of any order p. When using the theta method with whatever linear combination of only two points (one-step method), we can only achieve order 2. Therefore it is necessary to use more than two points.

Def. Lagrange Interpolation: We want to appoximate g(t) which we know n+1 values at $t_0 < t_1 < t_2 < ... < t_n$. Define

$$p(t) := \sum_{k=0}^{n} g(t_k) L_k(t), \text{ where } L_k(t) := \prod_{j=0, j \neq k}^{n} \frac{t_j - t}{t_j - t_k}$$

 L_k is called the Lagrange interp polynomial. p(t) is called the Lagrange interp of order n, using $\{L_k\}_{0}^{n}$ and the n+1 values of g.

Thm. If $g \in C^{\infty}$, p(t) is Lagrange interp of order n then

$$g(t) - p(t) = \frac{1}{(n+1)!} g^{(n+1)}(\xi) \prod_{k=0}^{n} (t - t_k), \quad t_0 \le \xi \le t_n$$

Further, if assuming $t_i - t_{i-1} = h$, then $\prod_{k=0}^n \le (n+1)!h^{n+1}$, hence

$$g(t) - p(t) \le \max_{t_0 \le \xi \le t_n} |g^{(n+1)}(\xi)| \cdot h^{n+1} \sim O(h^{n+1})$$

2 Multistep Methods

Consider the same IVP, y(t) is its solution. Suppose we know the value of y(t) at s points: $t_n, t_{n+1}, ..., t_{n+s-1}$ (so we also know the value of y'(t) at these points), and we want to approximate $y(t_{n+s})$. We begin with the integral formula from t_{n+s-1} to t_{n+s} :

$$y(t_{n+s}) = y(t_{n+s-1}) + \int_{t_{n+s-1}}^{t_{n+s}} y'(t)dt$$

$$= y(t_{n+s-1}) + \int_{t_{n+s-1}}^{t_{n+s}} p(t)dt + O(h^{s+1})$$

$$= y(t_{n+s-1}) + \int_{t_{n+s-1}}^{t_{n+s}} \sum_{k=n}^{n+s-1} y'(t_k)L_k(t)dt + O(h^{s+1})$$

$$= y(t_{n+s-1}) + \sum_{k=n}^{n+s-1} y'(t_k) \int_{t_{n+s-1}}^{t_{n+s}} L_k(t)dt + O(h^{s+1}) \quad (\dagger)$$
(1)

We hope that $\int_{t_{n+s-1}}^{t_{n+s}} L_k(t)dt$ is a constant times h. Actually we have

$$\int_{t_{n+s-1}}^{t_{n+s}} L_k(t)dt = \int_{t_{s-1}}^{t_s} L_{k'}(t)dt$$

where k = n, n + 1, ..., n + s - 1; k' = 0, 1, ..., s - 1. We find this by translating the lattice $\{t_k\}$ to the left by nh. So we just use k in the following text with k = 0, 1, ..., s - 1. And we denote

$$h\left(\frac{1}{h}\int_{t_{s-1}}^{t_s} L_k(t)dt\right) = hc_k$$

where c_k is indeed a constant. Therefore

$$(\dagger): y(t_{n+s}) = y(t_{n+s-1}) + h \sum_{k=0}^{s-1} c_k f(t_{k+n}, y(t_{k+n})) + O(h^{s+1})$$

Def. Multistep Method: we define the numerical method using (\dagger) , by substituting $y(t_k)$ with discrete approx y_k , i.e.

$$y_{n+s} = y_{n+s-1} + h \sum_{k=0}^{s-1} c_k \cdot f(t_{k+n}, y_{k+n})$$
 (2)

$$c_k(=c_k^{[s]}) = \frac{1}{h} \int_{t_{s-1}}^{t_s} L_k^{[s]}(t)dt = \frac{1}{h} \int_{t_{s-1}}^{t_s} \prod_{j=0, i \neq k}^{s-1} \frac{t_j - t}{t_j - t_k} dt$$
(3)

The iterative formula (2) is called the multistep method with order s. By definition, (†) gives the truncation error of this method, which is $R_k \sim O(h^s)$ when there are s known values of y(t), and the second term is an $(s-1)^{th}$ ordered Lagrange interpolation.

Note that parameter $\{c_k\}$ is a function of k and s (the order of the method). For the methods of different order, c_k 's have different values. We say $\{c_k^{[s]}\}_{k=0}^{s-1}$ in the following text to avoid ambiguity.

 $Ex. 1^{st}$ order multistep:

$$c_0^{[1]} = \int_0^1 L_0^{[1]}(t)dt = \int_0^1 1dt = 1$$
$$y_{n+1} = y_n + hf(t_n, y_n)$$

Degenerates to the Euler method.

 $Ex. 2^{nd}$ order multistep:

$$c_0^{[2]} = \int_1^2 L_0^{[2]}(t)dt = \int_1^2 \frac{1-t}{1-0}dt = -\frac{1}{2}$$

$$c_1^{[2]} = \int_1^2 L_1^{[2]}(t)dt = \int_1^2 \frac{0-t}{0-1}dt = \frac{3}{2}$$

$$y_{n+2} = y_{n+1} + h\left[-\frac{1}{2}f(t_n, y_n) + \frac{3}{2}f(t_{n+1}, y_{n+1})\right]$$

 $Ex. 3^{rd}$ order multistep:

$$c_0^{[3]} = \int_2^3 L_0^{[3]}(t)dt = \int_2^3 \frac{1-t}{1-0} \cdot \frac{2-t}{2-0}dt = \frac{5}{12}$$

$$c_1^{[3]} = \int_2^3 L_1^{[3]}(t)dt = \int_2^3 \frac{0-t}{0-1} \cdot \frac{2-t}{2-1}dt = -\frac{3}{4}$$

$$c_2^{[3]} = \int_2^3 L_2^{[3]}(t)dt = \int_2^3 \frac{0-t}{0-2} \cdot \frac{1-t}{1-2}dt = \frac{23}{12}$$
$$y_{n+3} = y_{n+2} + h \left[\frac{5}{12} f(t_n, y_n) - \frac{4}{3} f(t_{n+1}, y_{n+1}) + \frac{23}{12} f(t_{n+2}, y_{n+2}) \right]$$

The complexity of computing parameters $\{c_k^{[s]}\}$ increases with the order, but we can hardcode the values in the program anyway. The running time of the methods will not involve the computation of $\{c_k^{[s]}\}$.

3 General Formulation of Multistep Methods

We propose a general formulation instead of turning to Lagrange interpolation:

$$\sum_{k=0}^{s} a_k y_{n+k} = h \sum_{k=0}^{s} b_k f(t_{n+k}, y_{n+k})$$
(4)

where $\{a_k\}_{k=0}^s$, $\{b_k\}_{k=0}^s$ are unsolved constants (parameters), independent wrt h, n or the ODE. Let $a_s = 1$, we can obtain an explicit method iff $b_s = 0$:

$$y_{n+s} = -\sum_{k=0}^{s-1} a_k y_{n+k} + h \sum_{k=0}^{s-1} b_k f(t_{n+k}, y_{n+k})$$

Define

$$\psi(n,y) := \sum_{k=0}^{s} a_k y(t_{n+k}) - h \sum_{k=0}^{s} b_k f(t_{n+k}, y(t_{n+k}))$$
(5)

By defintion we want $\psi(n,y) \sim O(h^{p+1})$, then the method is of order $p, 1 \leq p \leq s$. With Taylor expansion of $y(t_{n+k})$ and $y'(t_{n+k})$ at t_n , we have

$$\psi(n,y) = \sum_{k=0}^{s} a_k \sum_{m=0}^{\infty} \left(y^{(m)}(t_n) \cdot \frac{(kh)^m}{m!} \right) - h \sum_{k=0}^{s} b_k \sum_{m=0}^{\infty} \left(y^{(m+1)}(t_n) \cdot \frac{(kh)^m}{m!} \right)$$

$$= \sum_{m=0}^{\infty} \frac{h^m y^{(m)}(t_n)}{m!} \sum_{k=0}^{s} a_k k^m - \sum_{m=1}^{\infty} \frac{h^m y^{(m)}(t_n)}{(m-1)!} \sum_{k=0}^{s} b_k k^{m-1}$$

$$= y(t_n) \sum_{k=0}^{s} a_k + \sum_{m=1}^{\infty} \frac{h^m y^{(m)}(t_n)}{(m-1)!} \sum_{k=0}^{s} (a_k k^m - mb_k k^{m-1})$$
(6)

It is clear that the method is of order p if the m = 0, 1, ..., p order of h^m shrink, i.e. The method (4) is of order $p \iff$

(a)
$$\sum_{k=0}^{s} a_k = 0$$
(b)
$$\sum_{k=0}^{s} (a_k k^m - mb_k k^{m-1}) = 0 \text{ for } m = 1, 2, ..., p$$
(c)
$$\sum_{k=0}^{s} (a_k k^m - mb_k k^{m-1}) \neq 0 \text{ for } m = p + 1 \text{ (or higher)}$$

Now we consider these parameters, define

$$c_0 := \sum_{k=0}^{s} a_k, \quad c_m := \frac{1}{m!} \sum_{k=0}^{s} (a_k k^m - m b_k k^{m-1})$$

The generating polynomial of $\{c_m\}$ is:

$$P(z) := \sum_{m=0}^{\infty} c_m z^m$$

$$\sum_{m=0}^{\infty} c_m z^m = \sum_{k=0}^{s} a_k z^0 + \sum_{m=1}^{\infty} \frac{z^m}{m!} \sum_{k=0}^{s} (a_k k^m - m b_k k^{m-1})$$

$$= \sum_{k=0}^{s} a_k \left(1 + \sum_{m=1}^{\infty} \frac{(kz)^m}{m!} \right) + \sum_{k=0}^{s} b_k z \sum_{m=1}^{\infty} \frac{(kz)^{m-1}}{(m-1)!}$$

$$= \sum_{k=0}^{s} a_k \sum_{m=0}^{\infty} \frac{(kz)^m}{m!} + \sum_{k=0}^{s} b_k z \sum_{m=0}^{\infty} \frac{(kz)^m}{m!}$$

$$= \sum_{k=0}^{s} a_k e^{kz} + \sum_{k=0}^{s} b_k z e^{kz}$$

$$(8)$$

Which gives us

$$P(z) = \sum_{m=0}^{\infty} c_m z^m = \sum_{k=0}^{s} a_k e^{kz} + \sum_{k=0}^{s} b_k z e^{kz}$$

Since z in P(z) has exactly the role of h in $\psi(n,y)$. So the method is of order $p \iff \psi(n,y) = O(h^{p+1}) \iff P(z) = cz^{p+1} + h.o.t$. (higher order terms) (*).

Now let $\omega = e^z$. We have $\log \omega = \log(\omega - 1 + 1) \sim \omega - 1$ as $\omega \to 1$ as $z \to 0$. So $(*) \iff$:

$$P(z) = \sum_{k=0}^{s} a_k \omega^k - \log \omega \sum_{k=0}^{s} b_k \omega^k = c(\log \omega)^{p+1} + h.o.t. = c(\omega - 1)^{p+1} + h.o.t.$$

By using the generating polynomial, we obtain a shortcut to see the order of the (truncation error) of the multistep methods.

Def. Characteristic Polynomial: Define

$$P(\omega) := \sum_{k=0}^{s} a_k \omega^k - \log \omega \sum_{k=0}^{s} b_k \omega^k$$

And further define $\rho(\omega) := \sum_{k=0}^{s} a_k \omega^k$ as the first characteristic polynomial, $\sigma(\omega) := \sum_{k=0}^{s} b_k \omega^k$ as the second characteristic polynomial. Then

$$P(\omega) = \rho(\omega) - \sigma(\omega) \log \omega$$

Thm. The multistep method is of order $p \ge 1 \iff \exists c \ne 0$, such that as $\omega \to 1$,

$$\rho(\omega) - \sigma(\omega) \log \omega = O\left((\omega - 1)^{p-1}\right)$$

Ex. Check the 2^{nd} order multistep method:

$$y_{n+2} = y_{n+1} + h \left[-\frac{1}{2} f(t_n, y_n) + \frac{3}{2} f(t_{n+1}, y_{n+1}) \right]$$

We have

$$\rho(\omega) = \omega^2 - \omega; \quad \sigma(\omega) = \frac{3}{2}\omega - \frac{1}{2}$$

So let $v := \omega - 1$

$$\rho(\omega) - \sigma(\omega) \log \omega = \omega^2 - \omega - \log \omega \left(\frac{3}{2}\omega - \frac{1}{2}\right)$$

$$= v(v+1) - \left(v - \frac{v^2}{2} + \frac{v^3}{3} + O(v^4)\right) \left(\frac{3}{2}v + \frac{1}{2}\right)$$

$$= v^2 + v - \frac{3}{2}v^2 - v + \frac{3}{4}v^3 + \frac{1}{2}v^2 - \frac{1}{3}v^3 + O(v^4)$$

$$= O(v^3)$$
(9)

So by theorem, it is of order 2 indeed.

Ex. Check the multistep method:

$$y_{n+2} - 3y_{n+1} + 2y_n = h \left[\frac{13}{12} f(t_{n+2}, y_{n+2}) - \frac{5}{3} f(t_{n+1}, y_{n+1}) - \frac{5}{12} f(t_n, y_n) \right]$$

We have

$$\rho(\omega) = \omega^2 - 3\omega + 2; \quad \sigma(\omega) = \frac{13}{12}\omega^2 - \frac{5}{3}\omega - \frac{5}{12}\omega^2$$

By letting $v := \omega - 1$,

$$\rho(\omega) - \sigma(\omega) \log \omega = O(v^3)$$

So this method is also of order 2.

4 Convergence of Multistep Methods

Def. Consistency: If a multistep method is at least of order 1, we call it consistent. I.e. $\rho(\omega) - \sigma(\omega) \log \omega = O(v^p), p \ge 2.$

Cor. A method is consistant \iff $c_0 = 0, c_1 = 0$, (recall that $c_0 := \sum_{k=0}^s a_k, c_m := \frac{1}{m!} \sum_{k=0}^s (a_k k^m - m b_k k^{m-1})) \iff$

(a)
$$\sum_{k=0}^{s} a_k = 0$$
(b)
$$\sum_{k=0}^{s} (ka_k - b_k) = 0$$

 $\iff \rho(1) = 0 \text{ and } \rho'(1) = \sigma(1).$

Ex. Consistency, by itself, does not guarantee convergence (a global characteristic). Consider using method

$$y_{n+2} - 3y_{n+1} + 2y_n = h \left[\frac{13}{12} f(t_{n+2}, y_{n+2}) - \frac{5}{3} f(t_{n+1}, y_{n+1}) - \frac{5}{12} f(t_n, y_n) \right]$$

to solve a naive IV problem: y' = 0, y(0) = 1. From our discussion above we know this method is of order 2, hence consistent. And the true solution of this ODE is just $y \equiv 1$.

The method degenerates to recursion $y_{n+2} - 3y_{n+1} + 2y_n = 0$. If we start with $y_0 = 1$, then we get the correct solution. However, if we add a small disturbance to the initial data, i.e. $y_0 = 1 + \epsilon$, ϵ is small, and may be caused by round-off error in computing (which is ubiquitous). We will have, by solving the linear recursion:

$$y_n = C_1 + C_2 \cdot 2^n = 1 - \epsilon + 2^n \epsilon$$

which eventually blows up. And clearly this method does not converge for all ODEs.

Def. Stability: A linear multistep (s-steps) method is said to be zero-stable, if there exists constant K, s.t. for any two sequences $\{y_n\}$ and $\{\hat{y}_n\}$, generated with same method, but with different initial data: $\{y_0, y_1, ..., y_{s-1}\}$ and $\{\hat{y}_0, \hat{y}_1, ..., \hat{y}_{s-1}\}$, we have

$$|y_n - \hat{y}_n| \le K \cdot \max_{0 \le j \le s-1} (|y_j - \hat{y}_j|)$$

for $t_n \leq T$ and $h \searrow 0$. Ususally K depends on T and not depends on h.

Thm. A linear multistep method is zero-stable for any ODE y' = f(y,t) where f satisfies Lipschitz condition \iff The first characteristic polynomial of this method $\rho(z) = \sum_{k=0}^{s} a_k z^k$ has all its zeros (roots) inside the closed unit disc (including the boundary). Moreover, if the zero lie on the unit circle, it must be a simple root (has multiplicity 1).

Proof. (\Rightarrow) We consider the naive IVP y'=0. The method degenerates to linear recursion

$$\sum_{k=0}^{s} a_k y_{n+k} = 0$$

which has general solution $y_n = \sum_r P_r(n) z_r^n$, where z_r is root of $\rho(z) = 0$, $P_r(n)$ is a polynomial whose degree equials $m(z_r) + 1$, $m(z_r)$ is the multiplicity of z_r . It is clear that this problem has solution $y \equiv const$. Hence y_n cannot be unbounded.

Under the other 2 scenarios (1) $|z_r| > 1$ or (2) $|z_r| = 1$, $P_r(n)$ is not constant; clearly y_n unbounded. So the only possibilty is otherwise: $|z_r| < 1$ or $|z_r| = 1$, $P_r(n) \equiv c$ (has degree 0).

 (\Leftarrow) is long, not so relevant to the numerical computing aspect. We just skip that.

Thm. (Convergence) A multistep method is convergent \iff It is consistent and zero-stable. Moreover, if the method is of order p, the global error is also of order p.