5.3 High-Order Taylor Methods

Consider the IVP

$$y' = f(t, y), \quad a \le t \le b, \ y(a) = \beta.$$

Definition: The difference method

$$w_0 = \beta$$

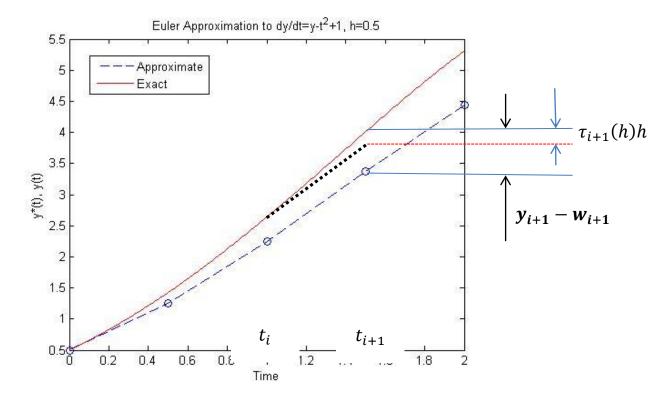
$$w_{i+1} = w_i + h\phi(t_i, w_i), \qquad \text{for each } i = 0, 1, 2, \dots, N-1, \quad \text{with step size } h = \frac{b-a}{N}$$

has Local Truncation Error

$$\tau_{i+1}(h) = \frac{y_{i+1} - (y_i + h\phi(t_i, y_i))}{h} = \frac{y_{i+1} - y_i}{h} - \phi(t_i, y_i) \quad \text{for each } i = 0, 1, 2, \dots, N-1.$$

Note: $y_i := y(t_i)$ and $y_{i+1} := y(t_{i+1})$.

Geometric view of local truncation error



Example. Analyze the local truncation error of Euler's method for solving y' = f(t, y), $a \le t \le b$, $y(a) = \beta$. Assume |y''(t)| < M with M > 0 constant. **Solution**:

$$\tau_{i+1}(h) = \frac{y_{i+1} - (y_i + hf(t_i, y_i))}{h} = \frac{y_{i+1} - y_i}{h} - f(t_i, y_i) = \frac{y(t_i) + hf(t_i, y_i) + \frac{h^2}{2}y''(\xi_i) - y_i}{h} - f(t_i, y_i)$$
with $\xi_i \in (t_i, t_{i+1})$.
$$\tau_{i+1}(h) = \frac{h}{2}y''(\xi_i).$$

Thus $|\tau_{i+1}(h)| \leq \frac{h}{2}M$.

So the local truncation error in Euler's method is O(h).

Consider the IVP

$$y' = f(t, y), \quad a \le t \le b, \ y(a) = \beta.$$

Compute
$$y'', y^{(3)} \cdots$$

First, by IVP:
$$y'' = f'(t, y(t))$$

 $y^{(3)}(t) = f''(t, y(t))$
 \vdots
 $y^{(k)}(t) = f^{(k-1)}(t, y(t))$

Second, by chain rule:

$$y''(=\frac{dy'(t)}{dt}) = \frac{df(t,y(t))}{dt} = \frac{\partial f}{\partial t}(t,y(t)) + \frac{\partial f}{\partial y}(t,y(t)) \cdot y'(t) = \frac{\partial f}{\partial t}(t,y(t)) + \frac{\partial f}{\partial y}(t,y(t)) \cdot f(t,y(t))$$

Derivation of higher-order Taylor methods

Consider the IVP

$$y' = f(t, y),$$
 $a \le t \le b,$ $y(a) = \beta,$ with step size $h = \frac{b-a}{N},$ $t_{i+1} = a + ih.$

Expand y(t) in the nth Taylor polynomial about t_i , evaluated at t_{i+1}

$$y(t_{i+1}) = y(t_i) + hy'(t_i) + \frac{h^2}{2}y''(t_i) + \dots + \frac{h^n}{n!}y^{(n)}(t_i) + \frac{h^{n+1}}{(n+1)!}y^{(n+1)}(\xi_i)$$

$$= y(t_i) + hf(t_i, y(t_i)) + \frac{h^2}{2}f'(t_i, y(t_i)) + \dots + \frac{h^n}{n!}f^{(n-1)}(t_i, y(t_i)) + \frac{h^{n+1}}{(n+1)!}f^{(n)}(\xi_i, y(\xi_i))$$

for some $\xi_i \in (t_i, t_{i+1})$. Delete remainder term to obtain the *n*th Taylor method of order *n*.

Denote
$$T^{(n)}(t_i, w_i) = f(t_i, w_i) + \frac{h}{2}f'(t_i, w_i) + \dots + \frac{h^{n-1}}{n!}f^{(n-1)}(t_i, w_i)$$

Taylor method of order n

$$w_0 = \beta$$

 $w_{i+1} = w_i + hT^{(n)}(t_i, w_i)$ for each $i = 0, 1, 2, \dots, N-1$.

Remark: Euler's method is Taylor method of order one.

Example 1. Use Taylor method of orders (a) two and (b) four with N = 10 to the IVP

$$y' = y - t^2 + 1$$
, $0 \le t \le 2$, $y(0) = 0.5$.

Solution:

$$h = \frac{2-0}{N} = \frac{2-0}{10} = 0.2. \text{ So } t_i = 0 + 0.2i = 0.2i \qquad \text{for each } i = 0, 1, 2, \cdots, 10.$$
(a) $f'(t, y(t)) = \frac{d}{dt}(y - t^2 + 1) = y' - 2t = y - t^2 + 1 - 2t$

$$\text{So } T^{(2)}(t_i, w_i) = f(t_i, w_i) + \frac{h}{2}f'(t_i, w_i) = (w_i - t_i^2 + 1) + 0.1(w_i - t_i^2 + 1 - 2t_i) = 1.1(w_i - t_i^2 + 1) - 0.2t_i$$
The 2^{nd} order Taylor method is

$$w_0 = 0.5$$

 $w_{i+1} = w_i + 0.2(1.1(w_i - t_i^2 + 1) - 0.2t_i)$ for each $i = 0, 1, 2, \dots, 9$

Now compute approximations at each time step:

$$w_0 = 0.5$$

 $w_1 = w_0 + 0.2(1.1(w_0 - (0)^2 + 1) - 0.2(0)) = 0.83;$ abs. eror: $|w_1 - y_1| = 0.000701$
 $w_2 = w_2 + 0.2(1.1(w_2 - (0.2)^2 + 1) - 0.2(0.2)) = 1.2158;$ abs. eror: $|w_2 - y_2| = 0.001712$

(b)
$$f''(t,y(t)) = \frac{d}{dt}(f') = (y-t^2+1-2t)' = y'-2t-2 = y-t^2+1-2t-2 = y-t^2-2t-1$$

 $f^{(3)}(t,y(t)) = \frac{d}{dt}(f'') = (y-t^2-2t-1)' = y'-2t-2 = y-t^2+1-2t-2 = y-t^2-2t-1$
So $T^{(4)}(t_i,w_i) = f(t_i,w_i) + \frac{h}{2}f'(t_i,w_i) + \frac{h^2}{3!}f''(t_i,w_i) + \frac{h^3}{4!}f^{(3)}(t_i,w_i)$
 $= (w_i-t_i^2+1) + \frac{h}{2}(w_i-t_i^2+1-2t_i) + \frac{h^2}{6}(w_i-t_i^2-2t_i-1) + \frac{h^3}{24}(w_i-t_i^2-2t_i-1)$
 $= \left(1 + \frac{h}{2} + \frac{h^2}{6} + \frac{h^3}{24}\right)(w_i-t_i^2) - \left(1 + \frac{h}{3} + \frac{h^2}{12}\right)(ht_i) + 1 + \frac{h}{2} - \frac{h^2}{6} - \frac{h^3}{24}$

The 4th order Taylor method is

$$w_0 = 0.5$$

$$w_{i+1} = w_i + h \left(\left(1 + \frac{h}{2} + \frac{h^2}{6} + \frac{h^3}{24} \right) \left(w_i - t_i^2 \right) - \left(1 + \frac{h}{3} + \frac{h^2}{12} \right) (ht_i) + 1 + \frac{h}{2} - \frac{h^2}{6} - \frac{h^3}{24} \right)$$
for each $i = 0, 1, 2, \dots, 9$.

Now compute approximate solutions at each time step:

$$w_1 = 0.5 + 0.2 \left(\left(1 + \frac{0.2}{2} + \frac{0.2^2}{6} + \frac{0.2^3}{24} \right) (0.5 - 0) - \left(1 + \frac{0.2}{3} + \frac{0.2^2}{12} \right) (0) + 1 + \frac{0.2}{2} - \frac{0.2^2}{6} - \frac{0.2^3}{24} \right) = 0.8293$$
abs. eror of 4th order Taylor at t_1 : $|w_1 - y_1| = 0.000001$

$$w_2 = 0.8293 + 0.2 \left(\left(1 + \frac{0.2}{2} + \frac{0.2^2}{6} + \frac{0.2^3}{24} \right) (0.8293 - 0.2^2) - \left(1 + \frac{0.2}{3} + \frac{0.2^2}{12} \right) (0.2(0.2)) + 1 + \frac{0.2}{2} - \frac{0.2^2}{6} - \frac{0.2^3}{24} \right)$$

$$= 1.214091$$

abs. eror 4th order Taylor at t_2 : $|w_2 - y_2| = 0.000003$

Finding approximations at time other than t_i

Example. (Table 5.4 on Page 259). Assume the IVP $y' = y - t^2 + 1$, $0 \le t \le 2$, y(0) = 0.5 is solved by the 4th order Taylors method with time step size h = 0.2. $w_6 = 3.1799640$ ($t_6 = 1.2$), $w_7 = 3.7324321$ ($t_7 = 1.4$). Find y(1.25). **Solution**:

Method 1: use linear Lagrange interpolation.

$$y(1.25) \approx \frac{1.25 - 1.4}{1.2 - 1.4} w_6 + \frac{1.25 - 1.2}{1.4 - 1.2} w_7 = 3.3180810$$

Method 2: use Hermite polynomial interpolation (more accurate than the result by linear Lagrange interpolation).

First use $y' = y - t^2 + 1$ to approximate y'(1.2) and y'(1.4).

$$y'(1.2) = y(1.2) - (1.2)^2 + 1 \approx 3.1799640 - (1.2)^2 + 1 = 2.7399640$$

 $y'(1.4) = y(1.4) - (1.4)^2 + 1 \approx 3.7324321 - (1.4)^2 + 1 = 2.7724321$

Then use **Theorem 3.9** to construct Hermite polynomial $H_3(x)$.

$$y(1.25) \approx H_3(1.25)$$
.

Error analysis

Theorem 5.12 If Taylor method of order *n* is used to approximate the solution to the IVP

$$y' = f(t, y), \quad a \le t \le b, \quad y(a) = \beta$$

with step size h and if $y \in C^{n+1}[a, b]$, then the **local truncation error** is $O(h^n)$.

Remark:
$$y(t_{i+1}) = y(t_i) + hf(t_i, y(t_i)) + \frac{h^2}{2}f'(t_i, y(t_i)) + \dots + \frac{h^n}{n!}f^{(n-1)}(t_i, y(t_i)) + \frac{h^{n+1}}{(n+1)!}f^{(n)}(\xi_i, y(\xi_i))$$

$$\tau_{i+1}(h) = \frac{y_{i+1} - y_i}{h} - T^{(n)}(t_i, y_i) = \frac{h^n}{(n+1)!}f^{(n)}(\xi_i, y(\xi_i)).$$

 $y^{(n+1)}(t) = f^{(n)}(t, y(t))$ is bounded by $|y^{(n+1)}(t)| \le M$.

Thus
$$|\tau_{i+1}(h)| \le \frac{h^n}{(n+1)!} M$$
.

So the local truncation error in Euler's method is $O(h^n)$.

5.4 Runge-Kutta Methods

Motivation: Obtain high-order accuracy of Taylor's method without knowledge of derivatives of f(t, y).

Theorem 5.13(Taylor's Theorem in Two Variables) Suppose f(t,y) and partial derivative up to order n+1 continuous on $D = \{(t,y) | a \le t \le b, c \le y \le d\}$, let $(t_0,y_0) \in D$. For $(t,y) \in D$, there is $\xi \in [t,t_0]$ and $\mu \in [y,y_0]$ with $f(t,y) = P_n(t,y) + R_n(t,y)$.

Here

$$\begin{split} P_{n}(t,y) &= f(t_{0},y_{0}) + \left[(t-t_{0}) \frac{\partial f}{\partial t}(t_{0},y_{0}) + (y-y_{0}) \frac{\partial f}{\partial y}(t_{0},y_{0}) \right] \\ &+ \left[\frac{(t-t_{0})^{2}}{2} \frac{\partial^{2} f}{\partial t^{2}}(t_{0},y_{0}) + (t-t_{0})(y-y_{0}) \frac{\partial^{2} f}{\partial t \partial y}(t_{0},y_{0}) + \frac{(y-y_{0})^{2}}{2} \frac{\partial^{2} f}{\partial y^{2}}(t_{0},y_{0}) \right] \\ &+ \left[\frac{1}{n!} \sum_{j=0}^{n} \binom{n}{j} (t-t_{0})^{n-j} (y-y_{0})^{j} \frac{\partial^{n} f}{\partial t^{n-j} \partial y^{j}}(t_{0},y_{0}) \right] \end{split}$$

$$R_n(t,y) = \frac{1}{(n+1)!} \sum_{j=0}^{n+1} {n+1 \choose j} (t-t_0)^{n+1-j} (y-y_0)^j \frac{\partial^{n+1} f}{\partial t^{n+1-j} \partial y^j} (t_0, y_0)$$

 $P_n(t, y)$ is the *n*th Taylor polynomial in two variables.

Derivation of Runge-Kutta method of order two

1. Determine a_1 , α_1 , β_1 such that

$$a_1 f(t + \alpha_1, y + \beta_1) \approx f(t, y) + \frac{h}{2} f'(t, y) = T^{(2)}(t, y) \text{ with } O(h^2) \text{ error.}$$
Notice $f'(t, y) = \frac{df(t, y(t))}{dt} = \frac{\partial f}{\partial t} (t, y(t)) + \frac{\partial f}{\partial y} (t, y(t)) \cdot y'(t) = \frac{\partial f}{\partial t} (t, y(t)) + \frac{\partial f}{\partial y} (t, y(t)) \cdot f(t, y(t))$
We have $T^{(2)}(t, y) = f(t, y) + \frac{h}{2} \frac{\partial f}{\partial t} (t, y(t)) + \frac{h}{2} \frac{\partial f}{\partial y} (t, y(t)) \cdot f(t, y(t))$ (1)

2. Expand $a_1 f(t + \alpha_1, y + \beta_1)$ in 1st degree Taylor polynomial:

$$a_1 f(t + \alpha_1, y + \beta_1) = a_1 f(t, y) + a_1 \alpha_1 \frac{\partial f}{\partial t}(t, y) + a_1 \beta_1 \frac{\partial f}{\partial v}(t, y) + a_1 R_1 (t + \alpha_1, y + \beta_1)$$

$$(2)$$

3. Match coefficients of equation (1) and (2) gives

$$a_1 = 1$$
, $a_1 \alpha_1 = \frac{h}{2}$, $a_1 \beta_1 = \frac{h}{2} f(t, y(t))$

with unique solution

$$a_1 = 1, \qquad \alpha_1 = \frac{h}{2}, \qquad \beta_1 = \frac{h}{2} f(t, y(t))$$

4. This gives

$$T^{(2)}(t,y) = f\left(t + \frac{h}{2}, y + \frac{h}{2}f(t,y(t))\right) - R_1\left(t + \frac{h}{2}, y + \frac{h}{2}f(t,y(t))\right)$$
 with $R_1\left(t + \frac{h}{2}, y + \frac{h}{2}f(t,y(t))\right) = O(h^2)$

Midpoint Method (one of Runge-Kutta methods of order two)

Consider the IVP

$$y' = f(t, y), \quad a \le t \le b, \ y(a) = \beta.$$

with step size $h = \frac{b-a}{N}$.

$$w_0 = \beta$$

$$w_{i+1} = w_i + hf\left(t_i + \frac{h}{2}, w_i + \frac{h}{2}f(t_i, w_i)\right), \quad \text{for each } i = 0, 1, 2, \dots, N-1.$$

Local truncation error is $O(h^2)$.

Two stage formula:

$$w_0 = \beta$$

$$k_1 = f(t_i, w_i)$$

$$k_2 = f\left(t_i + \frac{h}{2}, w_i + \frac{h}{2}k_1\right)$$

$$w_{i+1} = w_i + hk_2$$

Example 2. Use the Midpoint method with N = 10, h = 0.2, $t_i = 0.2i$ and $w_0 = 0.5$ to solve the IVP $y' = y - t^2 + 1$, $0 \le t \le 2$, y(0) = 0.5.