

§ 11.2

One - Step Methods.

$$(*) \quad \text{IVP} \quad y'(t) = f(t, y(t)) \quad , \quad \underbrace{t \geq t_0}_{\text{time}} \quad \left. \begin{array}{l} \text{assumption:} \\ \text{well-posed} \end{array} \right\} \quad y(t_0) = y_0$$

Idea: to approximate the solution at time T
 we start from dividing $[t_0, T]$ into small subintervals with $t_0 < t_1 < \dots < t_{N-1} < t_N = T$.

Let y_i = approximation of y at t_i , i.e., $y(t_i)$.
 We will replace $\frac{dy}{dt}$ w/ methods from Ch. 9.

A one-step method is one in which y_{k+1} is determined from y_k . (There are also multistep methods, in which we use approximations from time steps t_{k-1}, t_{k-2}, \dots , to compute y_{k+1} → not studied here, see Section 11.3.)

One-step methods:

We assume

- 1) $h = t_{k+1} - t_k$ is uniform $\forall k$
 - 2) $y(t)$ has as many cont. derivatives as needed; this is for error analysis using Taylor's thm.

§ 11.2.1 Euler's Method.

(Leonhard Euler)

Start w/ $y_0 = y(t_0)$,

for $k = 0, 1, \dots$, set

$$y_{k+1} = y_k + h f(t_k, y_k)$$

(That is,
 $y(t_{k+1}) \approx y_{k+1}$)

From Taylor's thm:

Also:

$$\frac{y_{k+1} - y_k}{t_{k+1} - t_k} \approx \underbrace{y'(t_k)}_{\substack{\text{slope of } y \\ \text{at } t_k}} \stackrel{\text{By ODE}}{=} f(t_k, y(t_k))$$

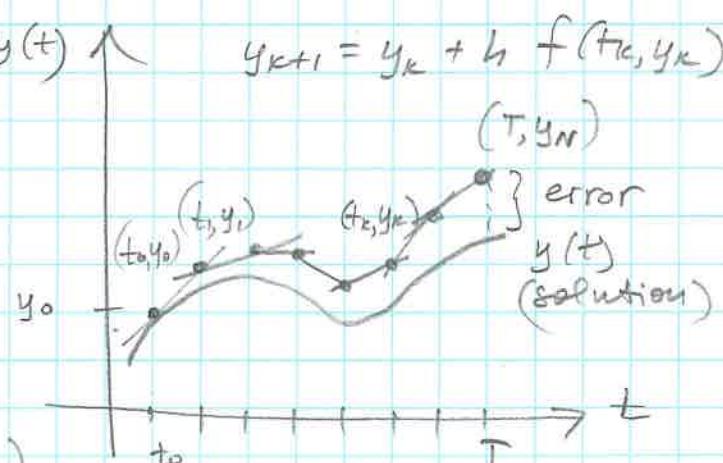
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Geometric Interpretation:

Tang. lines w/ slopes

$f(x_k, y_k)$ are used to approximate sol's.

(moving from one sol. curve to another, corresponding to different initial conditions!)



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Example :

$$\begin{array}{l} \text{IVP} \\ y' = y \\ y(0) = 1 \end{array} \quad \left(t_0 = 0, y_0 = 1 \right)$$

(solution is $y(t) = e^t$)

The IVP is well-posed : $f(t, y) = y$, $\underbrace{\frac{\partial f}{\partial y}}_{\text{continuous}} = 1$

$\Rightarrow f$ is uniformly Lipschitz continuous since

$$|f(t, y_1) - f(t, y_2)| = |y_1 - y_2| \quad \text{with } L = 1$$

$$\Rightarrow |z(t) - y(t)| \leq e^{|t|} |\delta|$$

(Thm 11.1.4) for any $z(t)$ s.t. $z(0) = 1 + \delta$

Euler's method for this IVP:

$$\left\{ \begin{array}{l} y_0 = 1 \end{array} \right.$$

$$\left\{ \begin{array}{l} y_{k+1} = y_k + h \underbrace{f(t_k, y_k)}_{y_k} = y_k + h y_k = (1+h) y_k \end{array} \right.$$

$k = 0, 1, \dots, N$

$$\text{If } h = \frac{T-0}{N} = \frac{T}{N} \quad \text{for some } T$$

$$\Rightarrow y_N = \left(1 + \frac{T}{N}\right)^N \xrightarrow[N \rightarrow \infty]{} e^T$$

Let us estimate $y(4) = e^4$ using Euler's method.

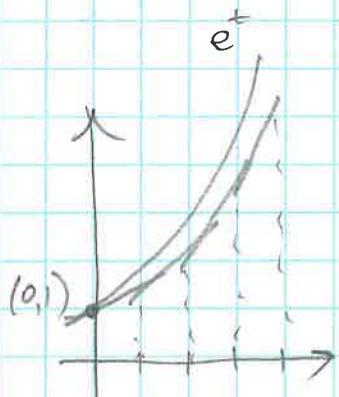
$$\text{If } h = 1 \Rightarrow t_0 = 0, y_0 = 1$$

$$t_1 = 1, \quad y_1 = (1+1)y_0 = 2$$

$$t_2 = 2, \quad y_2 = (1+1)y_1 = 4$$

$$t_3 = 3, \quad y_3 = (1+1)y_2 = 8$$

$$t_4 = 4, \quad y_4 = (1+1)y_3 = 16 \approx y(4) = e^4$$



$$\text{abs. error} = |e^4 - 16| \approx 38.6$$

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Example w/ system of ODE's:

Lotka-Volterra predator-prey system

Given $R(t_0) = R_0$ (rabbits) & $F(t_0) = F_0$ (foxes)

and ODE's $\begin{cases} R' = (\alpha - \beta F)R \\ F' = (\gamma R - \delta)F \end{cases}$

$$\text{w/ } \vec{y} = \begin{pmatrix} R(t) \\ F(t) \end{pmatrix}, \quad \vec{f}(t, \vec{y}) = \begin{pmatrix} (\alpha - \beta F)R \\ (\gamma R - \delta)F \end{pmatrix},$$

we apply Euler's method:

$$\vec{y}_{k+1} = \vec{y}_k + h \vec{f}(t_k, \vec{y}_k), \quad k=0, 1, \dots$$

Error Discussion:

IVP $\begin{cases} y' = f(t, y) \\ y(t_0) = y_0, t \geq t_0 \end{cases}$

① Local Truncation Error (LTE) is the amount by which the true sol. $y(t)$ fails to satisfy the difference equations $\frac{y_{k+1} - y_k}{h} =$

What does it mean? If we use $f(t_k, y_k)$,

$$\frac{y(t_{k+1}) - y(t_k)}{h} = f(t_k, y(t_k)) + \left(\frac{h}{2} y''(\xi_k) \right) = O(h)$$

$(t_k \leq \xi_k \leq t_{k+1})$

$$(y(t_{k+1}) = y(t_k) + h y'(t_k) + \frac{h^2}{2} y''(\xi_k))$$

So, LTE is $O(h) \rightarrow 0$ as $h \rightarrow 0$

(some sources: LTE is $O(h^2)$ as it is the first neglected term in Taylor's series)

If $LTE \rightarrow 0$ as $h \rightarrow 0 \Rightarrow$ the method is consistent.

② Global Error (absolute error):

$|y(t_k) - y_k|$ at mesh nodes t_k .

Consider Thm 11.2.1:

Let $y(t)$ be sol. to the IVP ① and $f(t, y)$ be continuous in the strip $a \leq t \leq b$, $\|f\|_\infty < \infty$, and uniformly Lipschitz continuous in y .

Let $T \in [a, b]$, $t > t_0$, and $h = \frac{T-t_0}{N}$ and (fixed)

$y_{k+1} = y_k + h f(t_k, y_k)$, $k=0, 1, \dots, N-1$. Then, for any k , s.t. $t_k \in [t_0, T]$, $y_k \rightarrow y(t_k)$ as $h \rightarrow 0$ and this convergence is uniform:

$$\max_k |y(t_k) - y_k| \rightarrow 0 \quad [\forall \varepsilon > 0 \exists k_\varepsilon \text{ s.t. } \forall k > k_\varepsilon]$$

and corresp. $t_k \Rightarrow |y(t_k) - y_k| < \varepsilon$

(k_ε depends on ε only)

(Proof: pp. 259-260)

③ Effect of Rounding:

Recall: if $\underbrace{h}_{\text{too small}} \cdot \text{small} \Rightarrow f'(x) \approx \frac{f(\tilde{x}+h) - f(\tilde{x})}{h} = 0$

Consider $\tilde{y}_{k+1} = \tilde{y}_k + h f(t_k, \tilde{y}_k) + \underline{\delta_k}$

computed sol.

roundoff due to
computing
 $f(t_k, \tilde{y}_k)$,
multiplying it
by h , and
adding \tilde{y}_k

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Let $|\delta_k| \leq \delta$

If $\tilde{d}_j = \tilde{y}_j - y_j$ ←
 computed num. sol. using
 exact arithmetic
 num. sol.

then $\underbrace{\tilde{y}_{k+1} - y_{k+1}}_{\tilde{d}_{k+1}} = \underbrace{\tilde{y}_k - y_k}_{\tilde{d}_k} + h (f(t_k, \tilde{y}_k) - f(t_k, y_k))$

$$- f(t_k, y_k)) + \delta_k \Rightarrow$$

$$|\tilde{d}_{k+1}| \leq |\tilde{d}_k| + h |f(t_k, \tilde{y}_k) - f(t_k, y_k)| + \delta$$

if f is Lipschitz
 continuous $\Rightarrow \leq L |\tilde{y}_k - y_k|$
 $\leq |\tilde{d}_k| + Lh |\tilde{d}_k| + \delta = |\tilde{d}_k| (1 + Lh) + \delta.$ $\frac{|\tilde{d}_k|}{|\tilde{d}_k|}$

Using proof of Thm 11.2.1, one can show
 that $|\tilde{d}_{k+1}| \leq e^{(k+1)Lh} |\tilde{d}_0| + \frac{e^{(k+1)Lh} - 1}{Lh} \delta \leq$
 $\leq e^{L(T-t_0)} |\tilde{d}_0| + \frac{e^{L(T-t_0)} - 1}{Lh} \delta$

$$(\tilde{d}_0 = \tilde{y}_0 - y_0)$$

Thus, if h is too small: $\frac{1}{h}$ will dominate!

We have to:
 - choose h small enough to
 reduce the global error, but

- not too small to avoid
 big errors due to roundoff!