#### NODEA Note

### **Basic Knowledge**

**Def of ODE & ODEs**:  $\frac{dy}{dt} = f(t,y)$  & ODEs:  $\frac{dy}{dt} = \mathbf{f}(t,y)$ ,  $\mathbf{y} = (y_1,...,y_d)^T$ ,  $\mathbf{f}(t,y) = (f_1(t,y),...,f_d(t,y))^T$  **Autonomous**:  $\frac{dy}{dt} = \mathbf{f}(\mathbf{y}) \Rightarrow \text{ autonomous ODE}(\mathbf{s})$ .  $|| \Downarrow \text{ New Autonomous ODEs}: \frac{dy}{ds} = \mathbf{f}(y_{d+1},y) \text{ and } \frac{dy_{d+1}}{ds} = 1$ • **Change to Autonomous**: For  $\frac{dy}{dt} = \mathbf{f}(t,y)$ . Let  $y_{d+1} = t$  and *new* independent variable s s.t.  $\frac{dt}{ds} = 1$ 

**Linearity**: ODE:  $\frac{dy}{dt} = f(t, y)$  is linearity if f(t, y) = a(t)y + b(t) || ODEs: If each ODE is linear, then the ODEs are linear.

**Picard's Theorem**: If f(t, y) is continuous in  $D := \{(t, y) : t_0 \le t \le T, |y - y_0| < K\}$  and  $\exists L > 0$  (Lipschitz constant) s.t.

 $\forall (t,u), (t,v) \in D \quad |f(t,u)-f(t,v)| \leq L|u-v| \text{ (ps:Can use MVT)}. \text{ And Assume that } M_f(T-t_0) \leq K, M_f := \max\{|f(t,u)|: (t,u) \in D\}$  $\Rightarrow$  **Then**,  $\exists$  a unique continuously differentiable solution y(t) to the IVP  $\frac{dy}{dt} = f(t,y), y(t_0) = y_0$  on  $t \in [t_0,T]$ .

**Existence & Uniqueness Theorem**: IVP  $\frac{d\mathbf{y}}{dt} = \mathbf{f}(t, \mathbf{y}), \mathbf{y}(t_0) = \mathbf{y}_0$ . If f(t, y) and  $\frac{\partial f}{\partial y_i}$  are continuous in a neighborhood of  $(t_0, \mathbf{y}_0)$ . ⇒ **Then**,  $\exists I := (t_0 - \delta, t_0 + \delta)$  s.t.  $\exists$  a unique continuously differentiable solution  $\mathbf{y}(t)$  to the IVP on  $t \in I$ .

# Acknowledge

Notation	Meaning	Notation	Meaning
[ <i>a</i> , <i>b</i> ]	Approximate function for $t \in [a, b]$	$t_0 = a \mid t_N = b$	Assume that $t_0 = a$ , $t_N = b$
N	number of timesteps (i.e. Break up interval [a, b] into N equal-length sub-intervals)	h	<b>stepsize</b> $(h = \frac{b-a}{N})$
$t_i$	Define $N + 1$ points: $t_0, t_1,, t_N$	$t_m$	$t_m = a + h \cdot m = t_0 + h \cdot m$
$y_i$	Approximation of $y$ at point $t = t_i$ (Except $y_0$ )	$y(t_i)$	Exact value of $y$ at point $t = t_i$

# **Euler's Method and Taylor Series Method**

**Euler's Method Algorithm**: Approximate ODE  $\frac{dy}{dt} = f(t, y), y(t_0) = y_0$  with number of steps N. (Similarly for ODEs)

 $\Rightarrow \textbf{for } n = 0, 1, 2, ..., N-1: \quad y_{n+1} = y_n + hf(t_0 + nh, y_n) = y_n + hf(t_n, y_n) \quad \textbf{end}$   $\text{(ps: $\psi$ Can get } | \text{Boundedness Theorem: For } \frac{dy}{dt} = f(t, y), y(a) = y_0 \text{ and suppose there exists a unique, twice differentiable, solution } y(t) \text{ on } [a, b].$   $\text{Suppose: } y \text{ is continuous and } |\frac{\partial f}{\partial y}| \le L. \Rightarrow \text{ the solution } y_n \text{ given Euler's method satisfies: } e_n = |y_n - y(t_n)| \le Dh, D = e^{(b-a)L} \frac{M}{2L}$ 

• **Lemma**: If  $v_{n+1} \le Av_n + B$ , then  $v_n \le A^n v_0 + \frac{A^n - 1}{A - 1}B$  If  $v_n = e_n := y_n - y(t_n)$ , then A = 1 + hL,  $B = h^2 M/2$  (suppose |y''| < M) **Order Notation (** $\mathcal{O}$ **)**: we write  $z = \mathcal{O}(h^p)$  if  $\exists \mathcal{C}, h_0 > 0$  s.t.  $|z| \le Ch^p$ ,  $0 < h < h_0$ 

**Flow Map (** $\Phi$ **)**:  $\Phi_h(y)$  is a flow function if:  $\Phi_{t_0,h}(y) = y(t_0 + h; t_0, y_0)$  Approx:  $\Psi_h(y) := \widehat{\Phi}_h(y)$  where  $\Psi(y_n) = y_{n+1}$ 

**Taylor Series Method**: Approximate ODE  $\frac{dy}{dt} = f(t,y), y(t_0) = y_0$  with *n-order Methods*: 用 Taylor Series 在  $t_0 + h$  处展开保留到 n 阶  $\cdot$  ps: Taylor Series:  $y(t_0 + h) = y(t_0) + hy'(t_0) + \frac{h^2}{2}y''(t_0) + \dots + \frac{h^{n-1}}{(n-1)!}y^{(n-1)}(t_0) + \frac{h^n}{n!}y^{(n)}(t^*), t^* \in [t, t+h]$   $y' = f, y'' = f_t + f_y f$ 

# **Convergence of One-Step Methods** consider for autonomous y' = f(y)

**Global Error**: global error after n steps:  $e_n := y_n - y(t_n)$  **Local Error**: For *one-step* method is:  $le(y,h) = \Psi_h(y) - \Phi_h(y)$ 

**Consistent**: If  $||le(y,h)|| < Ch^{p+1}(< \mathcal{O}(h^{p+1}))$ , C > 0.  $\Rightarrow$  Consistent at order p. **Stable**: If  $||\Psi_h(u) - \Psi_h(v)|| \le (1 + h\hat{L})||u - v||$ 

**Convergent**: A method is convergent if:  $\forall T$ ,  $\lim_{h \to 0, \ h = T/N} \max_{n = 0, 1, \dots, N} ||e_n|| = 0$   $\downarrow$  Then the global error satisfies:  $\max_{n = 0, 1, \dots, N} ||e_n|| = \mathcal{O}(h^p)$  p-th order **Convergence of One-Step Method**: For y' = f(y), and a one-step method  $\Psi_h(y)$  is  $^1$  consistent at order p and  $^2$  stable with  $\hat{L}$   $\uparrow$ . (ps: $C = \frac{C}{\hat{L}}(e^{T\hat{L}} - 1)$ )