NODEA Note

1 Basic Knowledge

Def of ODE & ODEs: (1st order) ODE: $\frac{dy}{dt} = f(t,y)$ & ODEs: $\frac{d\mathbf{y}}{dt} = \mathbf{f}(t,\mathbf{y})$, $\mathbf{y} = (y_1, \ldots, y_d)^T, \mathbf{f}(t,\mathbf{y}) = (f_1(t,\mathbf{y}), \ldots, f_d(t,\mathbf{y}))^T$

Autonomous: $\frac{d\mathbf{y}}{dt} = \mathbf{f}(\mathbf{y}) \Rightarrow$ autonomous ODE(s). || ⇓ New Autonomous ODEs: $\frac{d\mathbf{y}}{ds} = \mathbf{f}(y_{d+1}, \mathbf{y})$ and $\frac{dy_{d+1}}{ds} = 1$

· **Change to Autonomous**: For $\frac{d\mathbf{y}}{dt} = \mathbf{f}(t,\mathbf{y})$. 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.

$\forall (t,u), (t,v) \in D \quad |f(t,u) - f(t,v)| \le L|u - v|$ (ps:Can use MVT). And Assume that $M_f(T - t_0) \le K, M_f := \max\{|f(t,u)| : (t,u) \in D\}$

⇒ **Then**, ∃ 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 $\frac{dy}{dt} = \mathbf{f}(t,\mathbf{y}), \mathbf{y}(t_0) = \mathbf{y}_0$. 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. ∃ a unique continuously differentiable solution $\mathbf{y}(t)$ to the IVP on $t \in I$.

2 Acknowledge

Notation	Meaning	Notation	Meaning
$[a,b]$	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	**stepsize** ($h = \frac{b-a}{N}$)
t_i	Define $N+1$ points: $t_0, t_1, \ldots, t_N$	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 $t = t_i$

3 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)

⇒ **for** $n = 0, 1, 2, \ldots, N-1$: $y_{n+1} = y_n + hf(t_0 + nh, y_n) = y_n + hf(t_n, y_n)$ **end** (ps: ⇓ Can get $|y''| < M$)

Boundedness Theorem: For $\frac{dy}{dt} = f(t,y), y(a) = y_0$ and suppose there exists a unique, twice differentiable, solution $y(t)$ on $[a,b]$.

Suppose: y is continuous and $|\frac{\partial f}{\partial y}| \le L$. ⇒ the solution y_n 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 $A = 1 + hL, B = h^2M/2$ (suppose $|y''| < M$)

Order Notation ($\mathcal{O}$): we write $z(h) = \mathcal{O}(h^p)$ if $\exists C, h_0 > 0$ s.t. $|z| \le Ch^p, 0 < h < h_0$

Flow Map (Φ): $\Phi_h(y)$ is a flow function if: $\Phi_{t_0,h}(y) = y(t_0 + h; t_0, y_0)$ Approx: $\Psi_h(y) := \hat{\Phi}_h(y)$ 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

· ps: Taylor Series: $y(t_0 + h) = y(t_0) + hy'(t_0) + \frac{h^2}{2}y''(t_0) + \cdots + \frac{h^{n-1}}{(n-1)!}y^{(n-1)}(t_0) + \frac{h^n}{n!}y^{(n)}(t^*), t^* \in [t, t+h]$ $y' = f, y'' = f_t + f_y f$

4 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 $||le(y,h)|| < Ch^{p+1} (< \mathcal{O}(h^{p+1}))$, $C > 0$. ⇒ 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:$\forall T$, $\lim\limits_{h \to 0,\, h = T/N} \max\limits_{n = 0,1,\ldots,N} ||e_n|| = 0$ ⇓ Then the global error satisfies: $\max\limits_{n=0,1,\ldots,N} ||e_n|| = \mathcal{O}(h^p)$ **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 $\hat{L}$ ⇑. (ps:$C = \frac{C}{\hat{L}}(e^{T\hat{L}} - 1)$)

Construction of More General one-step Method: For $y' = f(y), y(t_0) = y_0 \Rightarrow y(t+h) - y(t) = \int_t^{t+h} f(y(\tau))d\tau$

Polynomial Interpolation: For $P(x) \in \mathbb{P}_{s-1}$, if $P(c_i) = g_i, \forall i \in \{1, \ldots, \mathbf{s}\}$. Then $P(x)$ is the *unique* interpolating polynomial.

· **Lagrange interpolating Polynomials**: $\ell_i(x) = \prod\limits_{j=1, j\ne i}^{s} \frac{x - c_j}{c_i - c_j}$ ⇒ $P(x)$ can be written as: $P(x) = \sum\limits_{i=1}^{s} g_i \ell_i(x)$

· **Quadrature Rule**: $\int_{t_0}^{t_0+h} g(t)dt = \int_0^1 g(t_0 + hx)dx \approx h\sum_{i=1}^{s} b_i g(t_0 + hc_i), b_i := \int_0^1 \ell_i(x)dx$ ps: $c_i \to g_i$ 从 $[0,1]$ 中取 If $g(t) \in \mathbb{P}_{p-1}$ ⇒Quadrature Rule has order p

One-Step Collocation Methods: For: $y(t_0) = y_0$, $y'(t_0 + c_ih) = f(y(t_0 + c_ih))$, c_i are chosen nodes in $[0,1]$ $i \in \{1, \ldots, s\}$

⇒ Def $\ell_i(x)$, $a_{ij} := \int_0^{c_i} \ell_j(x)dx$, $b_i := \int_0^1 \ell_i(x)dx$, $F_i := y'(t_0 + c_ih)$ Then: $F_i = f(y_n + h\sum_{j=1}^{s} a_{ij}F_j)$ and $y_{n+1} = y_n + h\sum_{i=1}^{s} b_iF_i$

· **Remark**: For choice of c_i: The optimal choice is attained by *Gauss-Legendre collocation methods*.