

1) All we need to do is show that $RHS \leq 1/8$
 $| \frac{df}{dy} |$ is bounded with $f(t, y(t)) = y'(t)$. So
 for $t \geq 1$ we have

$$f(t, y(t)) = \frac{1}{t^4 + 2y^2}$$

So,

note this derivative exists as $t \geq 1$, never $(t=0, y=0)$

$$\left| \frac{\partial f}{\partial y} \right| = \left| \frac{-4y}{(t^4 + 2y^2)^2} \right| = \frac{4|y|}{(t^4 + 2y^2)^2} \leq \frac{4y}{(t^4 + 2y^2)^2}$$

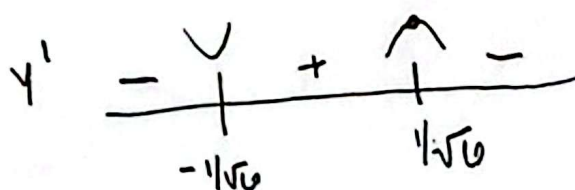
To show this function is bounded, we compute its derivative, set it equal to 0, and find the maximum. As $t \geq 1$, the above is bounded above by $\frac{4y}{(1+2y^2)^2}$, so we can actually find the maximum of this instead.

$$\frac{d}{dy} \rightarrow \frac{-2(4y) \cdot 4y}{(1+2y^2)^3} + \frac{4}{(1+2y^2)^2} = 0$$

$$\rightarrow -32y^2 + 4(1+2y^2) = 0$$

$$\rightarrow -24y^2 + 4 = 0$$

$$y^2 = 1/6 \rightarrow y = \pm \sqrt{1/6}$$



checking to see
which is the
maximum

So $\left| \frac{df}{dy} \right|$ is bounded above by

$$C = \frac{4\sqrt{16}}{(1 + 2\sqrt{\frac{1}{6}})^2}$$

$$= \frac{4}{\sqrt{6}} \frac{1}{(1 + \frac{1}{3})^2} = \frac{4}{\sqrt{6}} \frac{3^2}{4^2}$$
$$= \frac{9}{4\sqrt{6}}$$

Hence f is Lipschitz, hence the system

$$y'(t) = f(t, y(t))$$

has a unique solution.

2) We have

$$\begin{aligned} \bullet \quad u'' &= -K u \\ u(t_0) &= u_0 \\ u'(t_0) &= v_0 \end{aligned}$$

If we put $v = u'$, then the above becomes

$$\begin{aligned} u' &= v \\ v' &= -K u \end{aligned} = \begin{pmatrix} \overbrace{0 \quad 1}^A \\ \underbrace{-K \quad 0}_{\overbrace{A}^{\vec{x}}} \end{pmatrix} \begin{pmatrix} u \\ v \end{pmatrix} = A \vec{x} = f(\vec{x})$$

Now take $\vec{x}, \vec{y} \in \mathbb{R}^2$, then \star

$$\|f(\vec{x}) - f(\vec{y})\| = \|A\vec{x} - A\vec{y}\|$$

$$= \|A\| \|\vec{x} - \vec{y}\|$$

So to satisfy the Lipschitz condition, we need

$$\|A\| \|\vec{x} - \vec{y}\| \leq C \|\vec{x} - \vec{y}\| \rightarrow C \geq \|A\|$$

Hence $C = \|A\|$ is the smallest Lipschitz constant of f , for a given norm.

The smallest constants for various norms are computed next. we have

(next page)

$$C_1 = \|A\|_1 = \text{"max absolute column sum"}$$

$$= \max(1, K)$$

$$C_2 = \|A\|_2 = \left\{ \begin{array}{l} \text{By noticing that } A \text{ is Hermitian} \\ (A = A^T), \text{ the singular values} \\ \text{are the absolute values of the} \\ \text{eigenvalues of } A. \text{ So} \end{array} \right.$$

$$\det(A - \lambda I) = \det \begin{bmatrix} -\lambda & 1 \\ -K & -\lambda \end{bmatrix}$$

$$= \lambda^2 + K = 0$$

$$\rightarrow \lambda_{\pm} = \pm i\sqrt{K}, \text{ so that } |\lambda_{\pm}| = \sqrt{K}$$

So as $\|A\|_2$ is the largest singular value of A ,

$$C_2 = \|A\|_2 = \sqrt{K}$$

$$C_{\infty} = \|A\|_{\infty} = \text{"max absolute row sum"}$$

$$= \max(1, K)$$

3) Firstly, by Taylor's theorem

$$f_{n+1} = f(t_{n+1}, y_{n+1}) = f(t_n + h, y_{n+1}) = f_n + h \overbrace{\frac{df}{dt} \bigg|_{(t_n, y_{n+1})}}^{g_n} + O(h^2)$$

So our step is (numerical solution)

$$\begin{aligned} y_{n+1} &= y_n + h [\theta f_n + (1-\theta) f_{n+1}] \\ &= y_n + h [\theta f_n + (1-\theta) [f_n + h g_n + O(h^2)]] \\ &= y_n + h [f_n + (1-\theta) h g_n + O(h^2)] \end{aligned}$$

The actual solution is

$$\begin{aligned} y(t_{n+1}) &= y(t_n + h) = y(t_n) + h y'(t_n) + \frac{h^2}{2} y''(t_n) + O(h^3) \\ &= y(t_n) + h f_n + \frac{h^2}{2} g_n + O(h^3) \end{aligned}$$

So the local truncation error is (replacing $y(t_j)$ w/ y_j)

$$\begin{aligned} \frac{y_{n+1} - y(t_{n+1})}{h} &= \frac{1}{h} \left(y_n + h f_n + h^2 (1-\theta) g_n - y_n - h f_n - \frac{h^2}{2} g_n + O(h^3) \right) \\ &= \frac{1}{h} \left[\frac{(1/2 - \theta)}{2} h^2 g_n + O(h^3) \right] \\ &= \underbrace{\frac{1}{2} h g_n (1/2 - \theta)}_{(*)} + O(h^2) \end{aligned}$$

So when $\theta = 1/2$ it is $O(h^2)$ as $(*) = 0$, and
when $\theta \neq 1/2$ it is $O(h)$ as $(*) \neq 0$.

4) The numerical solution is

$$\begin{aligned} y_{n+1} &= y_n + \frac{h}{2} [f(t_{n+1}, \tilde{y}_{n+1}) + f_n] \\ &= y_n + \frac{h}{2} \underbrace{[f(t_n+h, y_n+hf_n)]}_{A} + f_n \quad (*) \end{aligned}$$

Using Taylor's thm, centered at (t_n, y_n) ,
A becomes

$$= f_n + h \underbrace{[f_t + f f_y]}_{\substack{a_n \\ y'(t_n, y_n)}} + \frac{h^2}{2} \underbrace{[f_{tt} + 2f f_{ty} + f_y^2 f_{yy}]}_{\substack{b_n \\ y''(t_n, y_n)}} + O(h^3)$$

So

$$(*) = y_n + hf_n + \frac{h^2}{2} a_n + \frac{h^3}{4} b_n + O(h^4)$$

The exact solution is

$$\begin{aligned} y(t_{n+1}) &= y(t_n+h) \\ &= y(t_n) + hf_n + \frac{h^2}{2} a_n + \frac{h^3}{6} b_n + O(h^4) \end{aligned}$$

So the local truncation error is (putting $y_n = y(t_n)$)

$$\begin{aligned} \frac{y_{n+1} - y(t_{n+1})}{h} &= \frac{[y_n + hf_n + \frac{h^2}{2} a_n + \frac{h^3}{4} b_n - y_n - hf_n - \frac{h^2}{2} a_n - \frac{h^3}{6} b_n + O(h^4)]}{h} \\ &= \left(\frac{1}{4} - \frac{1}{6}\right) h^2 b_n + O(h^3) \end{aligned}$$

So the error is $O(h^2)$ with leading term
 $\frac{1}{12} b_n(t_n, y_n) = \frac{1}{12} y'''(t_n)$