

# Math 317 Assignment 3

Due in class: November 17th, 2016

**Instructions:** Submit a hard copy of your solution with your name and student number. (**No name = zero grade!**) You must include all relevant program code, electronic output and explanations of your results. Write your own codes and comment them. Late assignment will not be graded and will receive a grade of zero.

1. Consider the integral  $I(f) = \int_0^1 f(x)dx$ .

(a) (5 marks) Determine the two point Gauss quadrature for  $I$  on  $[0, 1]$ . What is its degree of accuracy?

**Solution:** By the change of variable,  $x = \frac{1-t}{2} + \frac{1+t}{2} = \frac{1+t}{2}$ , the two point Gauss quadrature on  $[0, 1]$  is

$$I_{Gauss}(f) = \frac{1}{2} \left( f \left( \frac{1}{2} \left( 1 - \frac{1}{\sqrt{3}} \right) \right) + f \left( \frac{1}{2} \left( 1 + \frac{1}{\sqrt{3}} \right) \right) \right),$$

with degree of accuracy of 3.

(b) (5 marks) Determine  $c_0, c_1$  and  $x_1$  so that the quadrature  $I_h(f) = c_0 f(0) + c_1 f(x_1)$  has the highest degree of accuracy possible. State this degree.

**Solution:** Denote the quadrature as  $I_h(f) := c_0 f(0) + c_1 f(x_1)$ . The degree of accuracy conditions imposes

$$\begin{cases} I(1) = I_h(1) \\ I(x) = I_h(x) \\ I(x^2) = I_h(x^2) \end{cases} \iff \begin{cases} 1 = c_0 + c_1 \\ \frac{1}{2} = c_1 x_1 \\ \frac{1}{3} = c_1 x_1^2 \end{cases} \iff \begin{cases} x_1 = \frac{2}{3} \\ c_1 = \frac{3}{4} \\ c_0 = \frac{1}{4} \end{cases}$$

Thus

$$I_h(f) := \frac{1}{4} \left( f(0) + 3f\left(\frac{2}{3}\right) \right)$$

Moreover,  $I_h(f)$  has degree of accuracy of 2, since

$$\frac{1}{4} = I(x^3) \neq I_h(x^3) = \frac{1}{4} \left( 0 + 3 \frac{8}{27} \right) = \frac{2}{9}$$

(c) (5 marks) Recall from probability, the Gaussian distribution with mean  $\mu = 0$  and standard deviation  $\sigma = 1$  is  $f(x) = \frac{e^{-x^2/2}}{\sqrt{2\pi}}$ . In this case, compare the approximate values for  $I$  using the two point Gauss quadrature and your quadrature in part (b). (The exact value of  $I$  is 0.3413447...).

**Solution:** As expected, Gauss quadrature is more accurate than the quadrature from part (b):

$$I_{Gauss}(f) = \frac{1}{2\sqrt{2\pi}} \left( \exp \left( -\frac{1}{2} \left( \frac{1}{2} \left( 1 - \frac{1}{\sqrt{3}} \right)^2 \right) \right) + \exp \left( -\frac{1}{2} \left( \frac{1}{2} \left( 1 + \frac{1}{\sqrt{3}} \right)^2 \right) \right) \right)$$

$$\approx 0.34122114 \dots$$

$$I_h(f) = \frac{1}{4\sqrt{2\pi}} \left( \exp(0) + 3 \exp \left( -\frac{1}{2} \left( \frac{2}{3} \right)^2 \right) \right) \approx 0.33932157 \dots$$

2. Consider the integral  $I(f) = \int_0^1 f(x)dx$ .

(a) (8 marks) Derive the formula for the composite trapezoidal rule and its error.

**Solution:** Let  $x_k = x_0 + kh$  where  $h = \frac{b-a}{n}$ . On each  $[x_k, x_{k+1}]$ , there is some  $\xi_k \in (x_k, x_{k+1})$ ,

$$\int_{x_k}^{x_{k+1}} f(x) dx = h(f(x_k) + f(x_{k+1})) - f^{(2)}(\xi_k) \frac{h^3}{12}$$

Hence

$$\begin{aligned} I(f) &= h \sum_{k=0}^{n-1} (f(x_k) + f(x_{k+1})) + \frac{h^3}{12} \sum_{k=0}^{n-1} f^{(2)}(\xi_k) \\ &= h \sum_{k=0}^{n-1} (f(x_k) + f(x_{k+1})) + \frac{b-a}{12} f^{(2)}(\xi) h^2 \end{aligned}$$

since

$$\frac{1}{n} \sum_{k=0}^{n-1} f^{(2)}(\xi_k) = f^{(2)}(\xi)$$

for some  $\xi \in (a, b)$  by the intermediate value theorem.

(b) (10 marks) Using Richard's extrapolation method, we can use the composite trapezoidal rule to derive a more accurate quadrature. Such quadrature is given by

$$I_h^R(f) = \frac{4I_{\frac{h}{2}}(f) - I_h(f)}{3}$$

where  $I_h$  denotes here the composite trapezoidal rule and has error  $O(h^4)$ . The goal of this question is to perform a convergence analysis of the composite Trapezoidal rule and the improved quadrature for  $I(f) = \int_0^1 e^{-x} dx$ . In order to do so write a program to approximate using both quadratures for  $h = 1, 2^{-1}, \dots, 2^{-8}$ , plot  $\log(\text{error})$  versus  $\log(h)$  confirm their convergence rate by estimating the slopes of the lines in the loglog plot.

**Solution:** For small  $h$ , we expect error  $\approx Ch^p$  where  $p = 2$  for the composite trapezoidal rule and  $p = 4$  for the Richard's quadrature. Then

$$\begin{aligned} \text{error} &\approx Ch^p \\ \log(\text{error}) &\approx p \log(h) + \log(C) \end{aligned}$$

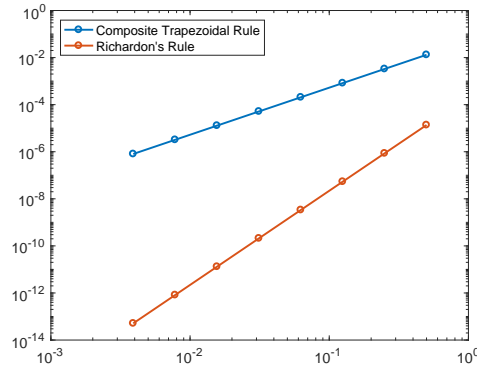
For sufficiently small  $h$ , the plot of  $\log(\text{error})$  versus  $\log(h)$  should be a line with slope  $p$ . Moreover, we can estimate the slope  $p$  by

$$p \approx \frac{\log(\text{error}_{i+1}) - \log(\text{error}_i)}{\log(h_{i+1}) - \log(h_i)}$$

where  $\text{error}_i = I(f) - I_{h_i}(f)$  and  $h_i = 2^{-i}$ . The estimated slopes are display in the next table.

$h$	Composite Trapezoidal Rule		Richardson's quadrature	
$2^{-1}$	$1.311 \times 10^{-2}$	-	$1.362 \times 10^{-5}$	-
$2^{-2}$	$3.289 \times 10^{-3}$	1.996	$8.558 \times 10^{-7}$	3.992
$2^{-3}$	$8.229 \times 10^{-4}$	1.999	$5.356 \times 10^{-8}$	3.998
$2^{-4}$	$2.058 \times 10^{-4}$	2.000	$3.349 \times 10^{-9}$	3.999
$2^{-5}$	$5.144 \times 10^{-5}$	2.000	$2.093 \times 10^{-10}$	4.000
$2^{-6}$	$1.286 \times 10^{-5}$	2.000	$1.308 \times 10^{-11}$	4.000
$2^{-7}$	$3.215 \times 10^{-6}$	2.000	$8.172 \times 10^{-13}$	4.001
$2^{-8}$	$8.038 \times 10^{-7}$	2.000	$5.118 \times 10^{-14}$	3.997

Errors and estimated order of convergence for both quadratures.



Loglog plot of the error versus  $h$  for both quadratures.

3. This exercise is to derive the order conditions for linear multistep method. Recall a  $k$ -step linear multi-step method for first-order initial value problems has the form:

$$\Phi_h := y_{n+1} + \sum_{i=0}^{k-1} a_i y_{n-k+1+i} - h \sum_{i=0}^k b_i f_{n-k+1+i} = 0,$$

where  $a_i, b_i$  are constants and  $f_i := f(t_i, y_i)$ .

- (a) (5 marks) Denoting  $a_k = 1$  and  $y$  as the exact solution to the initial value problem, show that the local truncation error is,

$$\tau_h(t_n) = \sum_{i=0}^k a_i y(t_{n-k+1} + ih) - h \sum_{i=0}^k b_i y'(t_{n-k+1} + ih).$$

**Solution:** By taking  $a_k = 1$ , we have that

$$y_{n+1} = a_k y_{n+1} = a_k y_{n-k+1-k}$$

and so

$$\Phi_h = \sum_{i=0}^k a_i y_{n-k+1+i} - h \sum_{i=0}^k b_i f_{n-k+1+i}.$$

Since  $y$  is the exact solution to the initial value problem, we have that for  $i = 0, \dots, k$ ,

$$f(t_{n-k+1} + ih, y(t_{n-k+1} + ih)) = y'(t_{n-k+1} + ih).$$

Hence

$$\tau_h(t_n) = \sum_{i=0}^k a_i y(t_{n-k+1} + ih) - h \sum_{i=0}^k b_i y'(t_{n-k+1} + ih).$$

(b) (5 marks) By Taylor expanding  $y, y'$  around  $t_{n-k+1}$ , show that for some  $\xi_i, \eta_i \in [t_{n-k+1}, t_{n-k+1} + ih]$

$$\tau_h(t_n) = \sum_{i=0}^k a_i \left( \sum_{q=0}^p \frac{(ih)^q}{q!} y^{(q)}(t_{n-k+1}) + \frac{(ih)^{p+1}}{(p+1)!} y^{(p+1)}(\xi_i) \right) - h \sum_{i=0}^k b_i \left( \sum_{q=1}^p \frac{(ih)^{q-1}}{(q-1)!} y^{(q)}(t_{n-k+1}) + \frac{(ih)^p}{p!} y^{(p+1)}(\eta_i) \right).$$

**Solution:** By Taylor expanding  $y$  around  $t_{n-k+1}$ , we have for  $i = 0, \dots, k$

$$y(t_{n-k+1} + ih) = \sum_{q=0}^p \frac{(ih)^q}{q!} y^{(q)}(t_{n-k+1}) + \frac{(ih)^{p+1}}{(p+1)!} y^{(p+1)}(\xi_i)$$

for some  $\xi_i \in [t_{n-k+1}, t_{n-k+1} + ih]$ .

By Taylor expanding  $y'$  around  $t_{n-k+1}$ , we have for  $i = 0, \dots, k$

$$y'(t_{n-k+1} + ih) = \sum_{q=1}^p \frac{(ih)^{q-1}}{(q-1)!} y^{(q)}(t_{n-k+1}) + \frac{(ih)^p}{p!} y^{(p+1)}(\eta_i)$$

for some  $\eta_i \in [t_{n-k+1}, t_{n-k+1} + ih]$ .

Plugging the above two expressions in the formula derived for  $\tau_h(t_n)$  in a) leads to

$$\begin{aligned} \tau_h(t_n) &= \sum_{i=0}^k a_i \left( \sum_{q=0}^p \frac{(ih)^q}{q!} y^{(q)}(t_{n-k+1}) + \frac{(ih)^{p+1}}{(p+1)!} y^{(p+1)}(\xi_i) \right) \\ &\quad - h \sum_{i=0}^k b_i \left( \sum_{q=1}^p \frac{(ih)^{q-1}}{(q-1)!} y^{(q)}(t_{n-k+1}) + \frac{(ih)^p}{p!} y^{(p+1)}(\eta_i) \right), \end{aligned}$$

as desired.

(c) (5 marks) Show that the local truncation error can be written in the form,

$$\tau_h(t_n) = \sum_{q=0}^p \left( \frac{h^q}{q!} y^{(q)}(t_{n-k+1}) C_q \right) + \frac{h^{p+1}}{(p+1)!} D,$$

where

$$C_q = \sum_{i=0}^k i^q a_i - q \sum_{i=0}^k i^{q-1} b_i, \quad D = \sum_{i=0}^k \left( i^{p+1} a_i y^{(p+1)}(\xi_i) - (p+1) i^p b_i y^{(p+1)}(\eta_i) \right).$$

**Solution:** It's enough to see that

$$\begin{aligned} &\sum_{i=0}^k a_i \left( \sum_{q=0}^p \frac{(ih)^q}{q!} y^{(q)}(t_{n-k+1}) + \frac{(ih)^{p+1}}{(p+1)!} y^{(p+1)}(\xi_i) \right) \\ &= \sum_{q=0}^p \frac{h^q}{q!} y^{(q)}(t_{n-k+1}) \left( \sum_{i=0}^k i^q a_i \right) + \frac{h^{p+1}}{(p+1)!} \sum_{i=0}^k i^{p+1} a_i y^{(p+1)}(\xi_i) \end{aligned}$$

and

$$\begin{aligned}
& h \sum_{i=0}^k b_i \left( \sum_{q=1}^p \frac{(ih)^{q-1}}{(q-1)!} y^{(q)}(t_{n-k+1}) + \frac{(ih)^p}{p!} y^{(p+1)}(\eta_i) \right) \\
&= \sum_{q=1}^p \frac{h^q}{(q-1)!} y^{(q)}(t_{n-k+1}) \left( \sum_{i=0}^k i^{q-1} b_i \right) + \frac{h^{p+1}}{p!} \sum_{i=0}^k i^p b_i y^{(p+1)}(\eta_i) \\
&= \sum_{q=0}^p \frac{h^q}{q!} y^{(q)}(t_{n-k+1}) \left( q \sum_{i=0}^k i^{q-1} b_i \right) + \frac{h^{p+1}}{(p+1)!} \sum_{i=0}^k (p+1) i^p b_i y^{(p+1)}(\eta_i)
\end{aligned}$$

- (d) (3 marks) Conclude that a  $k$ -step linear multi-step method is of order  $p$  if and only if  $a_i, b_i$  satisfies  $C_q = 0$  for all  $q = 0, \dots, p$ . Or equivalently,  $a_i, b_i$  satisfies for all  $q = 0, \dots, p$ ,

$$q \sum_{i=0}^k i^{q-1} b_i = k^q + \sum_{i=0}^{k-1} i^q a_i. \text{ (i.e. order conditions)}$$

**Solution:** A  $k$ -step linear multi-step method is of order  $p$  if and only if the lowest power of  $h$  in the local truncation error is  $p+1$ , i.e., if and only if  $C_q = 0$  for all  $q = 0, \dots, p$ .

4. The implicit 2-step Milne-Simpson method is:

$$y_{n+1} = y_{n-1} + \frac{h}{3}(f_{n+1} + 4f_n + f_{n-1}).$$

- (a) (10 marks) Show that the local truncation error is  $O(h^5)$ .

**Solution:** There's two ways to solve this: use Taylor expansion or the order conditions to show that the method is of order 4 (recall that the order of a method is always the order of the local truncations error minus 1). Here we provide both proofs.

**Taylor expansion approach:** The local truncation error is given by

$$\tau_h(t_n) = y(t_{n+1}) - y(t_{n-1}) - \frac{1}{3}hf(t_{n+1}, y(t_{n+1})) - \frac{4}{3}hf(t_n, y(t_n)) - \frac{1}{3}hf(t_{n-1}, y(t_{n-1}))$$

for  $n = 1, 2, \dots, N-1$ . The idea is to Taylor expand every term in the above expression which is not evaluated at  $t = t_n$ . We

$$y(t_{n\pm 1}) = y(t_n) \pm hy'(t_n) + \frac{h^2}{2}y''(t_n) \pm \frac{h^3}{6}y^{(3)}(t_n) + \frac{h^4}{24}y^{(4)}(t_n) + \mathcal{O}(h^5), \quad f(t_n, y(t_n)) = y'(t_n),$$

and

$$f(t_{n\pm 1}, y(t_{n\pm 1})) = y'(t_{n\pm 1}) = y'(t_n) \pm hy''(t_n) + \frac{h^2}{2}y^{(3)}(t_n) \pm \frac{h^3}{6}y^{(4)}(t_n) + \mathcal{O}(h^4).$$

Plugging in the above formulas into the expression of  $\tau_{n+1}(h)$  and simplifying shows that  $\tau_h(t_n) = \mathcal{O}(h^5)$  as desired.

**Order conditions approach:** We have

$$a_1 = 0, \quad a_0 = -1, \quad b_2 = \frac{1}{3}, \quad b_1 = \frac{4}{3}, \quad b_0 = \frac{1}{3}.$$

We have to check that

$$q \sum_{i=0}^k i^{q-1} = k^q + \sum_{i=0}^{k-1} i^q a_i$$

for  $q = 0, \dots, 4$  and that it fails for  $q = 5$ . Indeed, we have

- $q = 0$

$$q \sum_{i=0}^k i^{q-1} = 0$$

$$k^q + \sum_{i=0}^{k-1} i^q a_i = 2^0 + 0^0 \times (-1) + 1^0 \times 0 = 0$$

- $q = 1$

$$q \sum_{i=0}^k i^{q-1} = 1 \left( 0^0 \times \frac{1}{3} + 1^0 \times \frac{4}{3} + 2^0 \times \frac{1}{3} \right) = 2$$

$$k^q + \sum_{i=0}^{k-1} i^q a_i = 2^1 + 0^1 \times (-1) + 1^1 \times 0 = 2$$

- $q = 2$

$$q \sum_{i=0}^k i^{q-1} = 2 \left( 0^1 \times \frac{1}{3} + 1^1 \times \frac{4}{3} + 2^1 \times \frac{1}{3} \right) = 4$$

$$k^q + \sum_{i=0}^{k-1} i^q a_i = 2^2 + 0^2 \times (-1) + 1^2 \times 0 = 4$$

- $q = 3$

$$q \sum_{i=0}^k i^{q-1} = 3 \left( 0^2 \times \frac{1}{3} + 1^2 \times \frac{4}{3} + 2^2 \times \frac{1}{3} \right) = 8$$

$$k^q + \sum_{i=0}^{k-1} i^q a_i = 2^3 + 0^3 \times (-1) + 1^3 \times 0 = 8$$

- $q = 4$

$$q \sum_{i=0}^k i^{q-1} = 4 \left( 0^3 \times \frac{1}{3} + 1^3 \times \frac{4}{3} + 2^3 \times \frac{1}{3} \right) = 16$$

$$k^q + \sum_{i=0}^{k-1} i^q a_i = 2^4 + 0^4 \times (-1) + 1^4 \times 0 = 16$$

- $q = 5$

$$q \sum_{i=0}^k i^{q-1} = 5 \left( 0^4 \times \frac{1}{3} + 1^4 \times \frac{4}{3} + 2^4 \times \frac{1}{3} \right) = \frac{100}{3}$$

$$k^q + \sum_{i=0}^{k-1} i^q a_i = 2^5 + 0^5 \times (-1) + 1^5 \times 0 = 32$$

- (b) (5 marks) Show that the method is zero-stable and conclude that it is convergent.

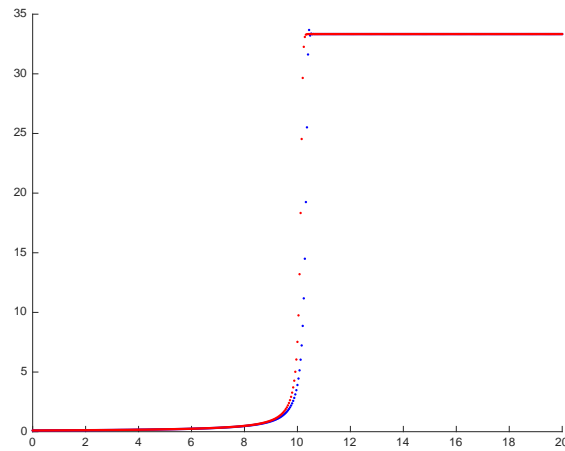
**Solution:** The characteristic polynomial is given by  $p(\lambda) = \lambda^2 - 1$  which has roots  $-1$  and  $1$ . Therefore the method is zero stable. In part *a*), we saw that the method has order 4 and therefore it is consistent. We can then conclude that it is convergent.

5. Consider the I.V.P. on  $t \in [0, T]$ :

$$y' = y^2(1 - \epsilon y).$$

- (a) (10 marks) For  $T = 20, y(0) = 0.1, \epsilon = 0.03$ , use the forward Euler and Trapezoidal method to solve the I.V.P. with  $N = 500$  and plot both solutions versus  $t$ .

**Solution:** See Figure.



Plot of the solution versus  $t$  with  $N = 500$  for the Euler method (blue) and the Trapezoidal method (red).

- (b) (4 marks) For the equilibrium solution  $y^* = 1/\epsilon$ , show that the I.V.P. is approximately,

$$y' \approx -\frac{1}{\epsilon}(y - y^*) \text{ when } |y - y^*| \text{ is small.}$$

*Hint: Taylor expand  $f(y) = y^2(1 - \epsilon y)$  around  $y = y^*$ .*

**Solution:** We have

$$f'(y) = -y^2\epsilon + 2y(1 - y\epsilon)$$

and so

$$f'(y^*) = -\frac{1}{\epsilon}.$$

Hence we have

$$\begin{aligned} f(y) &= f(y^*) + f'(y^*)(y - y^*) + \mathcal{O}(|y - y^*|^2) \\ &\approx -\frac{1}{\epsilon}(y - y^*) \end{aligned}$$

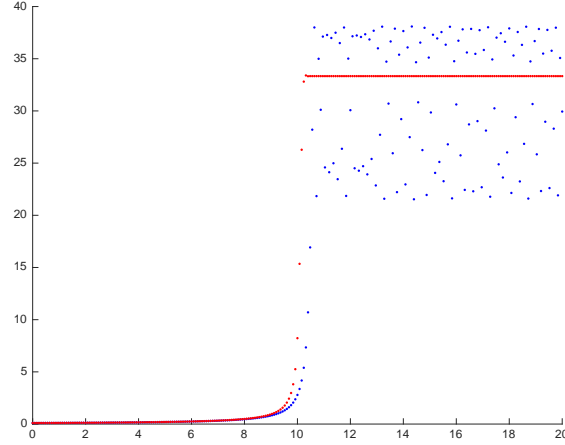
when  $|y - y^*|$  is small.

- (c) (4 marks) Plot both solutions when  $N = 250$  and use part (b) to explain what is happening.

**Solution:** See Figure. From the plot it's clear the Euler method didn't converge. The reason for this is that, as shown in b) the problem behaves like a stiff problem with  $\lambda = -\frac{1}{\epsilon}$  and so for the Euler method to converge we need

$$h < -\frac{2}{\lambda} = 2\epsilon = 0.06.$$

However, with  $N = 250$ ,  $h$  does not satisfy the above condition since  $h = 0.08$ . Note as well, that in part a), since  $N = 500$ ,  $h = 0.04$ , thus explaining why the Euler method behaved better.



Plot of the solution versus  $t$  with  $N = 250$  for the Euler method (blue) and the Trapezoidal method (red).

6. Consider the non-dimensionalized pendulum problem

$$\begin{cases} \theta''(t) + \sin(\theta(t)) = 0, & t \in [0, T], \\ \theta(0) = a, \\ \theta'(0) = b. \end{cases}$$

Let  $\theta(t)$  denote the exact solution.

(a) (2 marks) Write the second order equation as a system of first order equations.

**Solution:** Let  $y_1 = \theta$  and  $y_2 = \theta'$ . Hence the second order equation can be rewritten as

$$\mathbf{y}'(t) = \begin{bmatrix} y_1'(t) \\ y_2'(t) \end{bmatrix} = \begin{bmatrix} \theta'(t) \\ \theta''(t) \end{bmatrix} = \begin{bmatrix} \theta'(t) \\ -\sin(\theta(t)) \end{bmatrix} = \begin{bmatrix} y_2(t) \\ -\sin(y_1(t)) \end{bmatrix} = \mathbf{F}(t, \mathbf{y}(t))$$

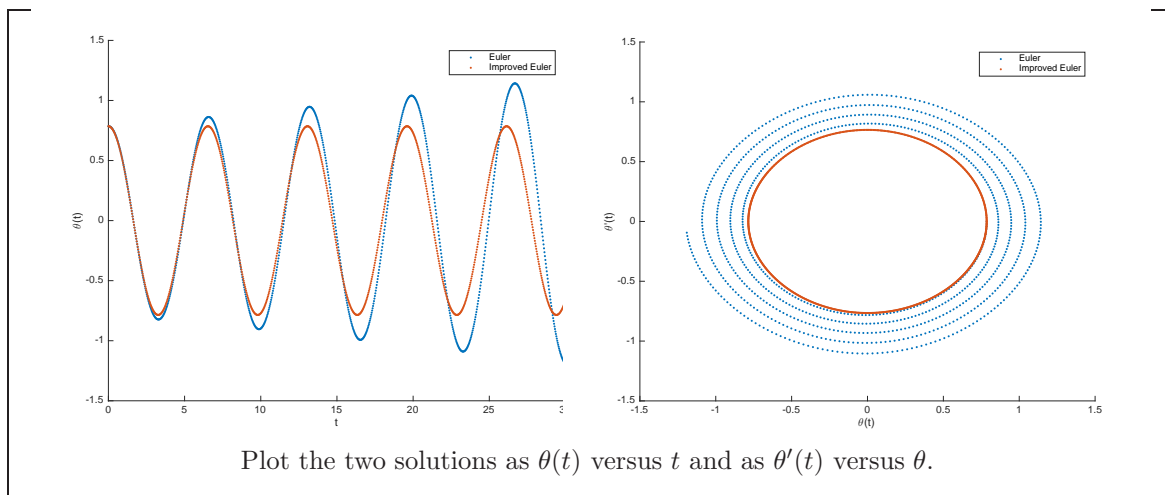
with

$$\mathbf{y}(0) = \begin{bmatrix} y_1(0) \\ y_2(0) \end{bmatrix} = \begin{bmatrix} a \\ b \end{bmatrix} = \mathbf{y}_0$$

(b) (10 marks) For  $N = 1000$ , use the forward Euler and improved Euler's method to solve the first order system for  $a = \pi/4, b = 0$  up to  $T = 30$ . Plot the two solutions as  $\theta(t)$  versus  $t$  and as  $\theta'(t)$  versus  $\theta(t)$ .

**Solution:** See Figure.





- (c) (2 marks) Let  $E(t) = \frac{(\theta'(t))^2}{2} - \cos(\theta(t))$  denote the energy of the pendulum. Show that the energy is conserved, i.e.,

$$\frac{d}{dt}E(t) = 0.$$

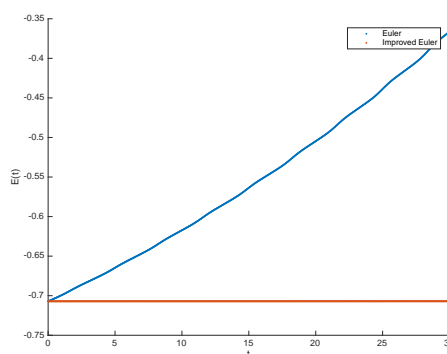
**Solution:** We have

$$\frac{d}{dt} \left( \frac{(\theta')^2}{2} - \cos(\theta) \right) = \theta''\theta' + \theta' \sin(\theta) = \theta'(\theta'' + \sin(\theta)) = 0,$$

where in the last equation we used the fact that  $\theta$  is the exact solution of the pendulum problem.

- (d) (2 marks) Plot the energy computed using the two methods as a function of  $t$ . Which method has the least “energy drift”?

**Solution:** See Figure. The method that has the least “energy drift” is the improved Euler; the energy remains approximately constant.



Plot the two solutions as  $\theta$  versus  $t$  and as  $\theta'$  versus  $\theta$ .