Caleb Logemann AER E 546 Fluid Mechanics and Heat Transfer I Homework 1

#1

(a) How many 'data' points are needed to obtain a third order accurate polynomial approximation? Derive a finite difference formula for $\partial T/\partial x$ that is third order accurate in Δx . Use only the minimum number of points.

Four data points are needed to obtain a third order accurate polynomial approximation, as the taylor series for a function, f with four coefficients is of the following form.

$$f(x_0) + f'(x_0)(x - x_0) + \frac{1}{2}f''(x_0)(x - x_0)^2 + \frac{1}{6}f'''(x_0)(x - x_0)^3 + O((x - x_0)^4)$$

Note that this polynomial has errors of order $(x-x_0)^4$, so the approximation is third order accurate.

In order to derive a finite difference formula for $\frac{\partial T}{\partial x}$ that is third order accurate I will first find a third order accurate polynomial approximation using four points. The four points I will use will be equally spaces with spacing Δx and the will be labeled x_{-2}, x_{-1}, x_0, x_1 with function values f_{-2}, f_{-1}, f_0, f_1 respectively. The polynomial approximation will solve the following equations for a, b, c, and d.

$$f_{-2} = a + b(-2\Delta x) + c(-2\Delta x)^{2} + d(-2\Delta x)^{3}$$

$$f_{-1} = a + b(-\Delta x) + c(-\Delta x)^{2} + d(-\Delta x)^{3}$$

$$f_{0} = a$$

$$f_{1} = a + b\Delta x + c(\Delta x)^{2} + d(\Delta x)^{3}$$

Clearly $a = f_0$. The 2nd and 4th equations can be added to solve for c. Summing these equations gives

$$f_{-1} + f_1 = 2f_0 + 2c(\Delta x)^2$$
$$c = \frac{f_{-1} + f_1 - 2f_0}{2(\Delta x)^2}$$

Summing the first equation and -2 times the second equation gives

$$f_{-2} - 2f_{-1} = -f_0 + 2c(\Delta x)^2 + -6d(\Delta x)^3$$

$$f_{-2} - 2f_{-1} = -f_0 + f_{-1} + f_1 - 2f_0 + -6d(\Delta x)^3$$

$$f_{-2} - 3f_{-1} + 3f_0 - 1f_1 = -6d(\Delta x)^3$$

$$d = \frac{f_{-2} - 3f_{-1} + 3f_0 - 1f_1}{-6(\Delta x)^3}$$

Plugging all these values into the final equation allows for b to be found.

$$\begin{split} f_1 &= f_0 + b\Delta x + \frac{f_{-1} + f_1 - 2f_0}{2} + \frac{-f_{-2} + 3f_{-1} - 3f_0 + 1f_1}{6} \\ \frac{6f_1}{6} &= \frac{6f_0}{6} + b\Delta x + \frac{3f_{-1} + 3f_1 - 6f_0}{6} + \frac{-f_{-2} + 3f_{-1} - 3f_0 + 1f_1}{6} \\ \frac{f_{-2} - 6f_{-1} + 3f_0 + 2f_1}{6} &= b\Delta x \\ b &= \frac{f_{-2} - 6f_{-1} + 3f_0 + 2f_1}{6\Delta x} \end{split}$$

Now that we have a polynomial approximation of

$$p(x) = a + b(x - x_0) + c(x - x_0)^2 + d(x - x_0)^3$$

where the values of a, b, c, and d were computed above. We can now compute the first derivative of this approximation, which is

$$p'(x) = b + 2c(x - x_0) + 3d(x - x_0)^2.$$

The first derivative at x_0 is thus b, or $p'(x_0) = b$. Therefore a third order approximation of the first derivative at the point x_0 is

$$b = \frac{f_{-2} - 6f_{-1} + 3f_0 + 2f_1}{6\Delta x}$$

(b) Derive the second order accurate centered difference formula for $\frac{\partial^2 T}{\partial x^2}$.

First I will derive the second order polynomial approximation for three points centered around x_0 . I will label the points x_{-1} , x_0 , and x_1 with function values f_{-1} , f_0 , and f_1 respectively. The third order accurate polynomial approximation will be of the form

$$a + b(x - x_0) + c(x - x_0)^2$$

Note that the second derivative of this approximation is always c, so the formula for c will also be the centered finite difference for the second derivative of second order. Finding this approximation amounts to solving the following three equations.

$$f_{-1} = a - b\Delta x + c(\Delta x)^{2}$$
$$f_{0} = a$$
$$f_{1} = a + b\Delta x + c(\Delta x)^{2}$$

Clearly $a = f_0$. The first and third equations can be summed to find c.

$$f_{-1} + f_1 = 2f_0 + 2c(\Delta x)^2$$
$$c = \frac{f_{-1} - f_0 + f_1}{2(\Delta x)^2}$$

Thus we don't even need to solve for b because the second order central finite difference for the second derivative is

$$\frac{f_{-1}-f_0+f_1}{2(\Delta x)^2}$$

#2

(a) The equation for a damped oscillator is

$$\ddot{Y} + \sigma \dot{Y} + \omega^2 Y = 0.$$

Let the non-dimensional frequency be $\omega = 1$. Consider the two damping rates $\sigma = 0.0$ and $\sigma = 0.5$. Solve this by RK2, out to t = 32, with the intial conditions Y(0) = 1 and $\dot{Y}(0) = 0$. The time-step can be $\Delta t = 32/N$, where N is the number of integration points. Plot solutions with N = 21, 101, 301. What is the analytical solution? Compare your numerical solutions to the exact result.

First I will compute the analytical solution to this differential equation. This can be done by finding the characteristic polynomial of the equation, which is

$$r^2 + \sigma r + 1 = 0.$$

Using the quadractic formula, we see that the roots of this polynomial are $r = -\frac{\sigma}{2} \pm \frac{\sqrt{\sigma^2 - 4}}{2}$. When $\sigma = 0.0$, the roots are $r = \pm i$. In the case of complex roots the general solution will be

$$Y(t) = c_1 \cos(t) + c_2 \sin(t).$$

Using the intial conditions we see that the exact solution is

$$Y(t) = \cos(t)$$
.

When $\sigma = 0.5$ the roots are $r = -\frac{1}{4} \pm \frac{\sqrt{15}}{4}i$. In this case the general solution is

$$Y(t) = e^{-\frac{1}{4}t} \left(c_1 \cos\left(\frac{\sqrt{15}}{4}t\right) + c_2 \sin\left(\frac{\sqrt{15}}{4}t\right) \right)$$

and the exact solution with boundary conditions is

$$Y(t) = e^{-\frac{1}{4}t}\cos\left(\frac{\sqrt{15}}{4}t\right).$$

Now in order to solve this equation numerically with RK2, we first need to transform this second order differential equation into a system of first order differential equations. To do this let $Z = \dot{Y}$, then the system becomes

$$\dot{Y} = Z$$
$$\dot{Z} = -\sigma Z - \omega^2 Y$$

This is in the form $\dot{x} = RHS(x)$ where

$$x = [Y, Z]^{T}$$

$$RHS(x) = brx_{2}, -\sigma x_{2} - \omega^{2} x_{1}^{T}.$$

The following is a method for running RK2 given a function to evaluate the RHS.

```
function [result] = RK2(RHSFunc, x0, nTimeSteps, tFinal)
    nEquations = length(x0);
    result = zeros(nTimeSteps+1, nEquations);
    result(1,:) = x0;

deltaT = tFinal/nTimeSteps;
    t = (i-1)*deltaT;
    temp = result(i,:) + 0.5*deltaT*RHSFunc(t, result(i,:));
    result(i+1,:) = result(i,:) + deltaT*RHSFunc(t + 1/2*deltaT, temp);
    end
end
```

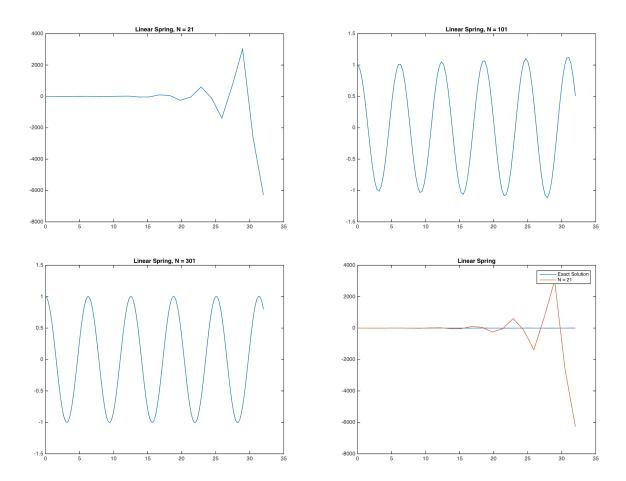
The following script now uses the previous function to run RK2 for the undamped and damped linear spring.

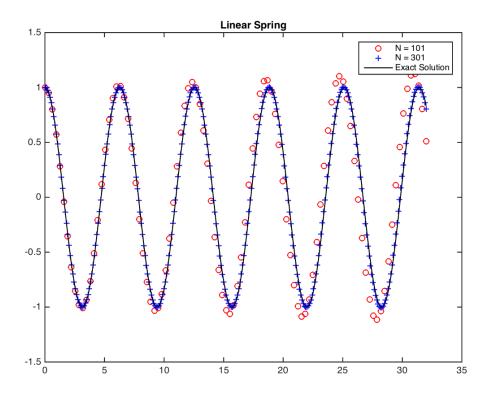
```
% Problem 2a
tFinal = 32:
% initial conditions
x0 = [1, 0];
% damping rate
sigma = 0.0;
RHSFunc = @(t, x) [x(2), -sigma*x(2) - x(1)];
n1 = 21;
sol1 = RK2 (RHSFunc, x0, n1, tFinal);
plot(linspace(0, 32, n1+1), sol1(:,1));
title('Linear Spring, N = 21');
saveas(gcf, 'Figures/01_01.png', 'png');
n2 = 101;
sol2 = RK2 (RHSFunc, x0, n2, tFinal);
plot(linspace(0, 32, n2+1), sol2(:,1));
title('Linear Spring, N = 101');
saveas(gcf,'Figures/01_02.png', 'png');
n3 = 301;
sol3 = RK2 (RHSFunc, x0, n3, tFinal);
plot(linspace(0, 32, n3+1), sol3(:,1));
title('Linear Spring, N = 301');
saveas(gcf, 'Figures/01_03.png', 'png');
exactSolFunc = @(t) cos(t);
exactSol = exactSolFunc(linspace(0, 32, 1000));
plot(linspace(0, 32, 1000), exactSol, linspace(0, 32, n1+1), sol1(:,1));
title('Linear Spring');
legend('Exact Solution', 'N = 21');
saveas(gcf,'Figures/01_04.png', 'png');
plot(linspace(0, 32, n2+1), sol2(:,1), 'ro',...
    linspace(0, 32, n3+1), sol3(:,1), 'b+',...
    linspace(0, 32, 1000), exactSol, 'k');
legend('N = 101', 'N = 301', 'Exact Solution');
title('Linear Spring');
saveas(gcf,'Figures/01_05.png', 'png');
sigma = 0.5;
RHSFunc = @(t, x) [x(2), -sigma*x(2) - x(1)];
n1 = 21;
sol1 = RK2 (RHSFunc, x0, n1, tFinal);
plot(linspace(0, 32, n1+1), sol1(:,1));
title('Damped Linear Spring, N = 21');
saveas(gcf,'Figures/01_06.png', 'png');
n2 = 101;
sol2 = RK2 (RHSFunc, x0, n2, tFinal);
plot(linspace(0, 32, n2+1), sol2(:,1));
title('Damped Linear Spring, N = 101');
saveas(gcf,'Figures/01_07.png', 'png');
n3 = 301;
sol3 = RK2 (RHSFunc, x0, n3, tFinal);
plot(linspace(0, 32, n3+1), sol3(:,1));
title('Damped Linear Spring, N = 301');
saveas(gcf,'Figures/01_08.png', 'png');
exactSolFunc = @(t) exp(-sigma/2*t).*cos(sqrt(15)/4 * t);
```

```
exactSol = exactSolFunc(linspace(0, 32, 1000));
plot(linspace(0, 32, 1000), exactSol, linspace(0, 32, n1+1), sol1(:,1));
title('Damped Linear Spring');
legend('Exact Solution', 'N = 21');
saveas(gcf,'Figures/01_09.png', 'png');

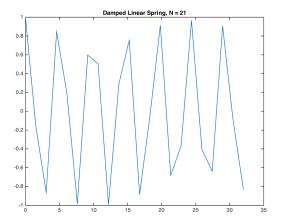
plot(linspace(0, 32, n2+1), sol2(:,1), 'ro',...
    linspace(0, 32, n3+1), sol3(:,1), 'b+',...
    linspace(0, 32, 1000), exactSol, 'k');
legend('N = 101', 'N = 301', 'Exact Solution');
title('Damped Linear Spring');
saveas(gcf,'Figures/01_10.png', 'png');
```

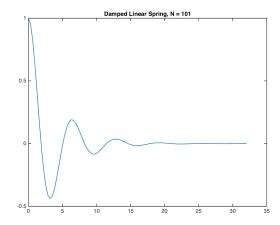
The following images are produced for the undamped linear spring, that is when $\sigma = 0.0$. Note that for N = 21, the numerical solution diverges from the exact solution, but for N = 101 and N = 301, the numerical solution is close to exact solution and gets more accurate as N is increased.

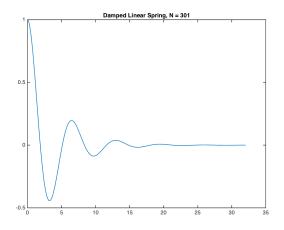


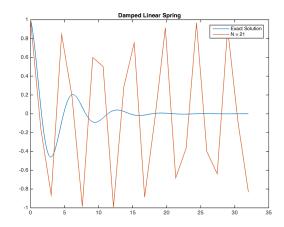


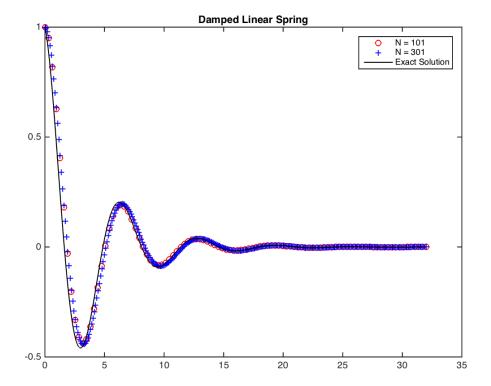
For the damped case, i.e. when $\sigma=0.5$, the following images are produced. Note that the numerical solution for N=21 doesn't grow rapidly and diverge, but is doesn't accurately represent the exact solution. Again as N is increased the accuracy of the numerical solution increases.











(b) The equation for a nonlinear spring (without damping) is

$$\ddot{Y} + Y - BY^3 = 0.$$

Solve by RK2 out to t = 32 with the intial conditions Y(0) = 1 and $\dot{Y}(0) = 0$. Plot Y(t) for B = 0.2, 0.6, 0.9, 0.999. Chose N large enough to get an accurate solution; that will depend on the value of B.

First in order to apply RK2 to this problem we must turn this second order ODE into a system of first order ODEs. In order to accomplish this let $\dot{Y} = Z$, then we have the following system

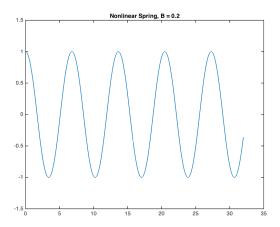
$$\dot{Y} = Z$$

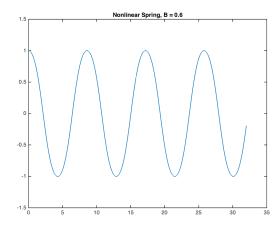
$$\dot{Z} = -Y - BY^3$$

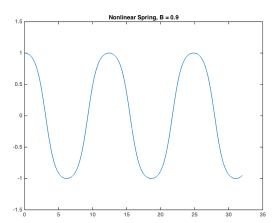
The following script now uses the same RK2 method shown in part (a), but with the RHS given above.

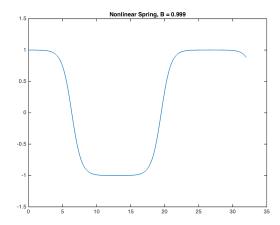
```
% Problem 2b
tFinal = 32;
x0 = [1, 0];
B = 0.2;
RHSFunc = 0(t, x) [x(2), B*x(1)^3 - x(1)];
n = 1000;
sol = RK2(RHSFunc, x0, n, tFinal);
plot(linspace(0,32,n+1), sol(:,1));
title('Nonlinear Spring, B = 0.2')
saveas(gcf, 'Figures/01_11.png', 'png');
B = 0.6;
RHSFunc = @(t, x) [x(2), B*x(1)^3 - x(1)];
n = 1000;
sol = RK2 (RHSFunc, x0, n, tFinal);
plot(linspace(0,32,n+1), sol(:,1));
title('Nonlinear Spring, B = 0.6')
saveas(gcf, 'Figures/01_12.png', 'png');
B = 0.9;
RHSFunc = @(t, x) [x(2), B*x(1)^3 - x(1)];
n = 1000;
sol = RK2(RHSFunc, x0, n, tFinal);
plot(linspace(0,32,n+1), sol(:,1));
title('Nonlinear Spring, B = 0.9')
saveas(gcf, 'Figures/01_13.png', 'png');
B = 0.999;
RHSFunc = @(t, x) [x(2), B*x(1)^3 - x(1)];
n = 3000;
sol = RK2(RHSFunc, x0, n, tFinal);
plot(linspace(0,32,n+1), sol(:,1));
title('Nonlinear Spring, B = 0.999')
saveas(gcf, 'Figures/01_14.png', 'png');
```

In this script I used N = 1000 for B = 0.2, 0.6, 0.9 and N = 3000 for B = .999. B is controlling the nonlinearity of the spring and the larger that contribution is the smaller the timestep needs to be. The following images are produced.









#3 Repeat the linear spring computation (ex. 2.a) with AB2. What does the solution for $\sigma = 0.0$ tell you about the stability of AB2?

I implemented the following function to run AB2 method for any RHS function.

```
function [result] = AB2(RHSFunc, x0, nTimeSteps, tFinal)
    nEquations = length(x0);
    result = zeros(nTimeSteps+1, nEquations);
    result(1,:) = x0;
    deltaT = tFinal/nTimeSteps;

    * take first step with explicit Euler method
    rhsOld = RHSFunc(0, result(1,:));
    result(2,:) = x0 + deltaT*rhsOld;

for i = 2:nTimeSteps
    t = (i-1)*deltaT;
    rhsNew = RHSFunc(t, result(i,:));
    result(i+1, :) = result(i,:) + deltaT*(1.5*rhsNew - 0.5*rhsOld);
    rhsOld = rhsNew;
end
end
```

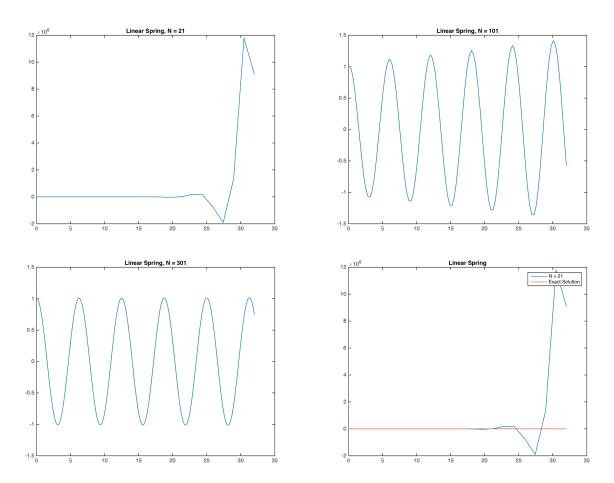
The following script now repeats exercise 2.a with this function instead of RK2.

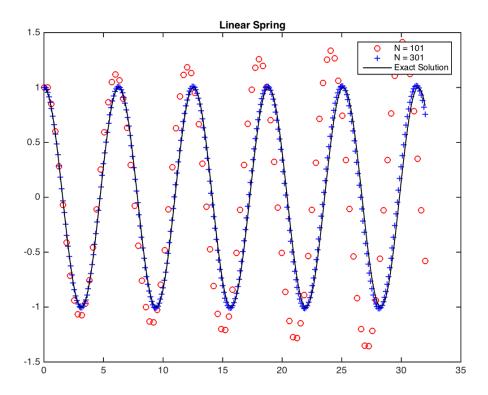
```
% Problem 3
tFinal = 32;
% initial conditions
x0 = [1, 0];
% damping rate
sigma = 0.0;
RHSFunc = @(t, x) [x(2), -sigma*x(2) - x(1)];
n1 = 21;
sol1 = AB2(RHSFunc, x0, n1, tFinal);
plot(linspace(0, 32, n1+1), sol1(:,1));
title('Linear Spring, N = 21');
saveas(gcf, 'Figures/01_15.png', 'png');
n2 = 101;
sol2 = AB2(RHSFunc, x0, n2, tFinal);
```

```
plot(linspace(0, 32, n2+1), sol2(:,1));
title('Linear Spring, N = 101');
saveas(gcf,'Figures/01_16.png', 'png');
n3 = 301;
sol3 = AB2 (RHSFunc, x0, n3, tFinal);
plot(linspace(0, 32, n3+1), sol3(:,1));
title('Linear Spring, N = 301');
saveas(gcf,'Figures/01_17.png', 'png');
exactSolFunc = @(t) cos(t);
exactSol = exactSolFunc(linspace(0, 32, 1000));
plot(linspace(0, 32, n1+1), sol1(:,1),...
linspace(0, 32, 1000), exactSol);
title('Linear Spring');
legend('N = 21', 'Exact Solution');
saveas(gcf,'Figures/01_18.png', 'png');
plot(linspace(0, 32, n2+1), sol2(:,1), 'ro',...
    linspace(0, 32, n3+1), sol3(:,1), 'b+',...
    linspace(0, 32, 1000), exactSol, 'k');
legend('N = 101', 'N = 301', 'Exact Solution');
title('Linear Spring');
saveas(gcf, 'Figures/01_19.png', 'png');
sigma = 0.5;
RHSFunc = @(t, x) [x(2), -sigma*x(2) - x(1)];
n1 = 21;
sol1 = AB2(RHSFunc, x0, n1, tFinal);
plot(linspace(0, 32, n1+1), sol1(:,1));
title('Damped Linear Spring, N = 21');
saveas(gcf,'Figures/01_20.png', 'png');
n2 = 101;
sol2 = AB2 (RHSFunc, x0, n2, tFinal);
plot(linspace(0, 32, n2+1), sol2(:,1));
title('Damped Linear Spring, N = 101');
saveas(gcf,'Figures/01_21.png', 'png');
n3 = 301;
sol3 = AB2 (RHSFunc, x0, n3, tFinal);
plot(linspace(0, 32, n3+1), sol3(:,1));
title('Damped Linear Spring, N = 301');
saveas(gcf,'Figures/01_22.png', 'png');
exactSolFunc = @(t) exp(-sigma/2*t).*cos(sqrt(15)/4 * t);
exactSol = exactSolFunc(linspace(0, 32, 1000));
plot(linspace(0, 32, n1+1), sol1(:,1), 'ro',...
    linspace(0, 32, 1000), exactSol, 'k');
title('Damped Linear Spring');
legend('N = 21', 'Exact Solution');
saveas(gcf,'Figures/01_23.png', 'png');
plot(linspace(0, 32, n2+1), sol2(:,1), 'b+',...
    linspace(0, 32, n3+1), sol3(:,1), 'ro',...
    linspace(0, 32, 1000), exactSol, 'k');
title('Damped Linear Spring');
legend('N = 101', 'N = 301', 'Exact Solution');
saveas(gcf,'Figures/01_24.png', 'png');
```

The following images were produced in the undamped case. Note that for N=21 the solution

diverges rapidly, much more rapidly than for RK2. This shows that adams-bashforth has a smaller area of stability as the time step needs to be much smaller in order to get an accurate solution. However for N = 101 and N = 301 the solutions are relatively accurate if less so than for RK2.





The following images were produced for the damped case Note that the solution diverges for N=21.

