

# 1 A-Stable Backwards Difference

In the BDF methods, the time derivative is approximated using one-sided finite differences. The BDF2 method is A-stable, whereas BDF3 is not. Consider a multi-step method in which the time derivative is approximated by the average of the BDF2 and BDF3 time-derivative approximations:

$$\frac{du}{dt} = f \rightarrow \frac{1}{2} \frac{du}{dt}|_{\text{BDF2}} + \frac{1}{2} \frac{du}{dt}|_{\text{BDF3}} = f$$

- a. Determine the coefficients  $\alpha_k$  and  $\beta_k$  that define this method. What is its order of accuracy?

## BDF2:

The expression for the derivative can be expressed as,

$$u_t = \frac{\frac{3}{2}u^{n+1} - 2u^n + \frac{1}{2}u^{n-1}}{\Delta t} = f^{n+1}$$

## BDF3:

The expression for the derivative can be expressed as,

$$u_t = \frac{\frac{11}{6}u^{n+1} - 3u^n + \frac{3}{2}u^{n-1} - \frac{1}{3}u^{n-2}}{\Delta t} = f^{n+1}$$

Averaging BDF2 and BDF3 together will give the time-derivative approximation,

$$\Delta t f^{n+1} = \underbrace{\frac{1}{2} \left( \frac{3}{2}u^{n+1} - 2u^n + \frac{1}{2}u^{n-1} \right)}_{\text{BDF2}} + \underbrace{\frac{1}{2} \left( \frac{11}{6}u^{n+1} - 3u^n + \frac{3}{2}u^{n-1} - \frac{1}{3}u^{n-2} \right)}_{\text{BDF3}}$$

Combining like terms results in,

$$\frac{5}{3}u^{n+1} - \frac{5}{2}u^n + u^{n-1} - \frac{1}{6}u^{n-2} = \Delta t f^{n+1}$$

This results in the coefficients  $\alpha$  and  $\beta$  to be,

$$\alpha_1 = \frac{5}{3}, \quad \alpha_0 = -\frac{5}{2}, \quad \alpha_{-1} = 1, \quad \alpha_{-2} = -\frac{1}{6}, \quad \beta_1 = 1$$

## Order of Accuracy

Firstly, is to start with the Taylor-Series expansion expression,

$$u^{n+k} = u^n + (k\Delta t)u_t^n + \frac{1}{2}(k\Delta t)^2u_{tt}^n + \frac{1}{6}(k\Delta t)^3u_{ttt}^n + \frac{1}{24}(k\Delta t)^4u_{t(4)}^n + \dots \mathcal{O}(\Delta t^5)$$

$$f^{n+k} = u_t^{n+k} = u_t^n + (k\Delta t)u_{tt}^n + \frac{1}{2}(k\Delta t)^2u_{ttt}^n + \frac{1}{6}(k\Delta t)^3u_{t(4)}^n + \frac{1}{24}(k\Delta t)^4u_{t(5)}^n + \dots \mathcal{O}(\Delta t^5)$$

Conducting Taylor-Expansions:

$$u^{n+1} = u^n + \Delta t u_t^n + \frac{1}{2} \Delta t^2 u_{tt}^n + \frac{1}{6} \Delta t^3 u_{ttt}^n + \frac{1}{24} \Delta t^4 u_{t(4)}^n + \dots \mathcal{O}(\Delta t^5)$$

$$u^n = u^n$$

$$u^{n-1} = u^n - \Delta t u_t^n + \frac{1}{2} \Delta t^2 u_{tt}^n - \frac{1}{6} \Delta t^3 u_{ttt}^n + \frac{1}{24} \Delta t^4 u_{t(4)}^n + \dots \mathcal{O}(\Delta t^5)$$

$$u^{n-2} = u^n - 2\Delta t u_t^n + 2\Delta t^2 u_{tt}^n - \frac{4}{3} \Delta t^3 u_{ttt}^n + \frac{2}{3} \Delta t^4 u_{t(4)}^n + \dots \mathcal{O}(\Delta t^5)$$

$$f^{n+1} = u_t^n + \Delta t u_{tt}^n + \frac{1}{2} \Delta t^2 u_{ttt}^n + \frac{1}{6} \Delta t^3 u_{t(4)}^n + \frac{1}{24} \Delta t^4 u_{t(5)}^n + \mathcal{O}(\Delta t^5)$$

Then for the order of accuracy the error gives,

$$\begin{aligned} \epsilon^{n+1} &= \frac{5}{3} u^{n+1} - \frac{5}{2} u^n + u^{n-1} - \frac{1}{6} u^{n-2} - \Delta t f^{n+1} \\ &= \frac{5}{3} \left( u^n + \Delta t u_t^n + \frac{1}{2} \Delta t^2 u_{tt}^n + \frac{1}{6} \Delta t^3 u_{ttt}^n + \frac{1}{24} \Delta t^4 u_{t(4)}^n \right) + \dots \\ &\quad - \frac{5}{2} u^n + \dots \\ &\quad + \left( u^n - \Delta t u_t^n + \frac{1}{2} \Delta t^2 u_{tt}^n - \frac{1}{6} \Delta t^3 u_{ttt}^n + \frac{1}{24} \Delta t^4 u_{t(4)}^n \right) + \dots \\ &\quad - \frac{1}{6} \left( u^n - 2\Delta t u_t^n + 2\Delta t^2 u_{tt}^n - \frac{4}{3} \Delta t^3 u_{ttt}^n + \frac{2}{3} \Delta t^4 u_{t(4)}^n \right) + \dots \\ &\quad - \Delta t \left( u_t^n + \Delta t u_{tt}^n + \frac{1}{2} \Delta t^2 u_{ttt}^n + \frac{1}{6} \Delta t^3 u_{t(4)}^n + \frac{1}{24} \Delta t^4 u_{t(5)}^n \right) \end{aligned}$$

Then using Matlab to simplify gives that the error is,

$$\epsilon^{n+1} = -\frac{1}{6} \Delta t^3 u_{ttt}^n - \frac{1}{6} \Delta t^4 u_{t(4)}^n + \frac{1}{40} \Delta t^5 u_{t(5)}^n$$

Then from the leading term this gives that the convergence is,

$$|\epsilon^{n+1}| = \mathcal{O}(\Delta t^{p+1}) = \mathcal{O}(\Delta t^3)$$

This gives that the order of accuracy is,

$$\boxed{p = 2}$$

Since  $p = 2$ , the order of accuracy for this scheme is second-order accurate.

- b. Perform an eigenvalue-stability analysis and *prove* (analytically) that this method is A-stable. Plot its stability boundary in the  $\lambda\Delta t$  complex number plane, and overlay BDF2 and BDF3.

Starting with the expression for this averaged time-derivative,

$$\Delta t f^{n+1} = \frac{5}{3}u^{n+1} - \frac{5}{2}u^n + u^{n-1} - \frac{1}{6}u^{n-2}$$

Then from here substituting in  $g^{n+k}u_0$  for  $u^{n+k}$  and  $\lambda g^{n+k}u_0$  for  $f^{n+k}$ ,

$$\lambda\Delta t g^{n+1}u_0 = \frac{5}{3}g^{n+1}u_0 - \frac{5}{2}g^nu_0 + g^{n-1}u_0 - \frac{1}{6}g^{n-2}u_0$$

Then taking this expression and dividing by  $g^nu_0$  results in,

$$\lambda\Delta t g = \frac{5}{3}g - \frac{5}{2} + g^{-1} - \frac{1}{6}g^{-2}$$

Isolating the  $\lambda\Delta t$  term then results in,

$$\lambda\Delta t = \frac{5}{3} - \frac{5}{2}g^{-1} + g^{-2} - \frac{1}{6}g^{-3}$$

Further simplifications without loss of generality gives,

$$\lambda\Delta t = \frac{5}{3} + \frac{1}{6g^3}(-15g^2 + 6g - 1)$$

Then by definition, this scheme must be stable if the un-stable region (the regions *inside* the marked plots do not extend into the left-hand plan). In this limiting case, this can be re-written as the limit as  $\lambda\Delta t \rightarrow 0^-$  and solve for the  $\theta$  value at which this occurs,

$$\lim_{\lambda\Delta t \rightarrow 0} = 0 = \frac{5}{3} + \frac{1}{6g^3}(-15g^2 + 6g - 1)$$

Taking this further, isolating and solving for  $g$  gives

$$\lim_{\lambda\Delta t \rightarrow 0} = -10g^3 = -15g^2 + 6g - 1$$

Pulling all  $g$  terms to one side results in,

$$0 = 10g^3 - 15g^2 + 6g - 1$$

Conducting simple factorization gives,

$$0 = (g - 1)(10g^2 - 5g + 1)$$

Using quadratic formula this gives that  $g$  is equivalent to,

$$g = 1, \frac{5 \pm i\sqrt{15}}{20}$$

Solving for the values at which these occurs gives,

$$g = \frac{5 \pm i\sqrt{15}}{20} = \exp[i\theta] = \cos \theta + i \sin \theta$$

$$\theta = 1.1513i \pm 0.6591$$

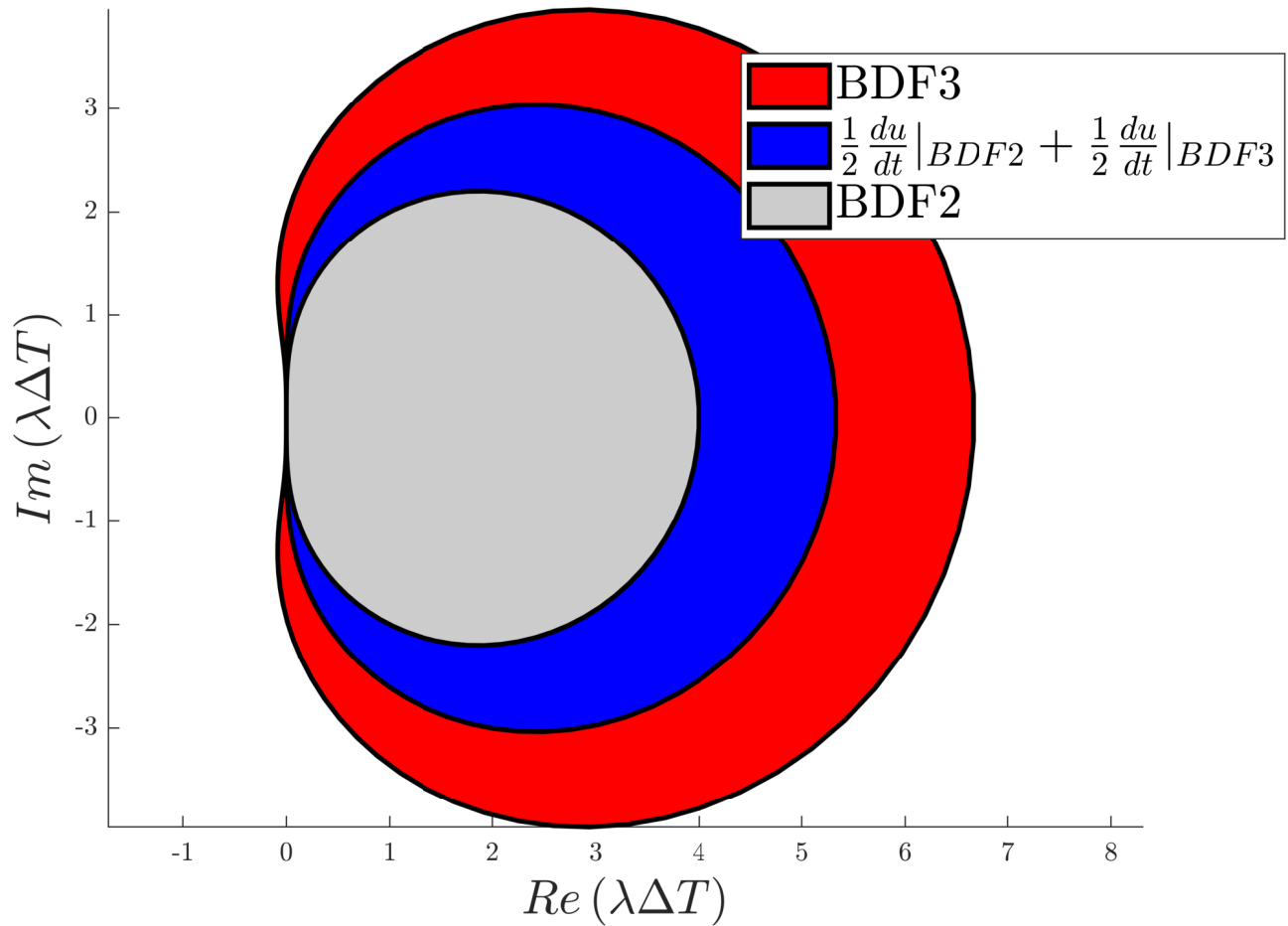
Again for  $g = 1$ ,

$$g = 1 = \exp[i\theta] = \cos \theta + i \sin \theta$$

$$\theta = 0, \quad \text{Physical answer}$$

As shown above, there are three approximated answers in which this averaged time-derivative scheme will cross into the unstable region. However, two of these three are not physical answers as  $\theta \in [0, 2\pi] \mid \theta \in \mathbb{R} \therefore$  the limiting case occurs at  $\theta = 0$  where  $g = 1$  resulting in  $\lambda\Delta t = \frac{5}{3} - \frac{5}{3} = 0$  – resulting in an A-stable scheme.

Plotting these eigenvalue stability regions can show and confirm that  $\theta = 0$  is the limiting case and for the averaged time-derivative that it is indeed A-stable as the unstable region never crosses into the left-hand plane like in BDF3 scheme. Plotting the un-stable regions gives Figure 1 shown below,



**Figure 1:** Eigenvalue stability region for BDF2, BDF3, and the averaged time-derivative approximation.

Shown above in Figure 1 are the un-stable regions for BDF2, BDF3, and the averaged time-derivative of the two. Outside of these regions the schemes remain stable where the left-hand plane where  $\text{Re}(\lambda\Delta T) < 0$  is the A-stable region of the schemes.

- c. Calculate the temporal truncation error of this method,  $\tau = \text{LHS} - \text{RHS}$  of the multistep formula, and show that the leading term is half the magnitude of that of BDF2.

From part a. of this question, I found that the local error was,

$$\epsilon^{n+1} = -\frac{1}{6}\Delta t^3 u_{ttt} - \frac{1}{6}\Delta t^4 u_{t(4)} + \frac{1}{40}\Delta t^5 u_{t(5)}$$

Thus, the truncation error is

$$\epsilon^{n+1} = \underbrace{-\frac{1}{6}\Delta t^3 u_{ttt} - \frac{1}{6}\Delta t^4 u_{t(4)} + \frac{1}{40}\Delta t^5 u_{t(5)} + \dots}_{\text{truncation error: } \mathcal{O}(\Delta t^3)} \mathcal{O}(\Delta t^6)$$

However, proving that the leading term is half the magnitude of that of BDF2, I will use the LHS and RHS definitions from BDF2 and use the Taylor-Series expansions from part a. and simplify as,

$$\begin{aligned} \text{LHS} &= \frac{3}{2}u^{n+1} - 2u^n + \frac{1}{2}u^{n-1} \\ &= \frac{3}{2} \left( u^n + \Delta t u_t^n + \frac{1}{2}\Delta t^2 u_{tt}^n + \frac{1}{6}\Delta t^3 u_{ttt}^n + \frac{1}{24}\Delta t^4 u_{t(4)}^n + \frac{1}{120}\Delta t^5 u_{t(5)}^n \right) + \dots \\ &\quad - 2u^n + \dots \\ &\quad + \frac{1}{2} \left( u^n - \Delta t u_t^n + \frac{1}{2}\Delta t^2 u_{tt}^n - \frac{1}{6}\Delta t^3 u_{ttt}^n + \frac{1}{24}\Delta t^4 u_{t(4)}^n - \frac{1}{120}\Delta t^5 u_{t(5)}^n \right) \\ &= \Delta t u_t^n + \Delta t^2 u_{tt}^n + \frac{1}{6}\Delta t^3 u_{ttt}^n + \frac{1}{12}\Delta t^4 u_{t(4)}^n + \frac{1}{120}\Delta t^5 u_{t(5)}^n \\ \text{RHS} &= \Delta t f^{n+1} \\ &= \Delta t \left( u_t^n + \Delta t u_{tt}^n + \Delta t^2 u_{ttt}^n + \Delta t^3 u_{t(4)}^n + \Delta t^4 u_{t(5)}^n \right) \end{aligned}$$

Taking the difference between the two gives,

$$\begin{aligned} \text{LHS} - \text{RHS} &= \left( \frac{1}{6} - \frac{1}{2} \right) \Delta t^3 u_{ttt}^n + \left( \frac{1}{12} - \frac{1}{6} \right) \Delta t^4 u_{t(4)}^n + \left( \frac{1}{120} - \frac{1}{24} \right) \Delta t^5 u_{t(5)}^n \\ \tau = \text{LHS} - \text{RHS} &= -\frac{1}{3}\Delta t^3 u_{ttt}^n - \frac{1}{12}\Delta t^4 u_{t(4)}^n - \frac{1}{30}\Delta t^5 u_{t(5)}^n \end{aligned}$$

Then re-writing both truncation errors gives,

$$\begin{aligned} \tau_{BDF2} &= -\frac{1}{3}\Delta t^3 u_{ttt}^n - \frac{1}{12}\Delta t^4 u_{t(4)}^n - \frac{1}{30}\Delta t^5 u_{t(5)}^n \\ \tau_{\text{Avg}} &= -\frac{1}{6}\Delta t^3 u_{ttt}^n - \frac{1}{6}\Delta t^4 u_{t(4)}^n + \frac{1}{40}\Delta t^5 u_{t(5)}^n \end{aligned}$$

**By inspection of the truncation errors above, we see that the leading term for the averaged time-derivative is indeed half that of the BDF2 scheme.**

## 2 The Beam-Warming Method

Consider the Beam-Warming (BW) method applied to the one-dimensional advection equation,  $u_t + au_x = 0$ ,  $a > 0$ , with initial condition  $u(x, 0) = u_0(x)$ ,  $x \in [0, L]$  and periodic boundaries.

- a. Derive the modified equation for the BW method and express it in the form

$$u_t + au_x = \alpha u_{xx} - \beta u_{xxx}$$

Use this equation to determine the order of accuracy of the BW method, and discuss the dispersion relation.

Starting with the modified equation for Beam-Warming method,

$$u_j^{n+1} = u_j^n - \frac{\sigma}{2} (3u_j^n - u_{j-1}^n + u_{j-2}^n) + \frac{\sigma^2}{2} (u_{j-2}^n - 2u_{j-1}^n + u_j^n)$$

Conducting the Taylor series expansions for these nodes gives,

$$u_{j-2}^n = u_j^n - 2\Delta x u_x + 2\Delta x^2 u_{xx} - \frac{4}{3}\Delta x^3 u_{xxx} + \frac{2}{3}\Delta x^4 u_{x(4)} + \dots \mathcal{O}(\Delta x^5)$$

$$u_{j-1}^n = u_j^n - \Delta x u_x + \frac{1}{2}\Delta x^2 u_{xx} - \frac{1}{6}\Delta x^3 u_{xxx} + \frac{1}{24}\Delta x^4 u_{x(4)} + \dots \mathcal{O}(\Delta x^5)$$

$$u_j^{n+1} = u_j^n + \Delta t u_t + \frac{1}{2}\Delta t^2 u_{tt} + \frac{1}{6}\Delta t^3 u_{ttt} + \frac{1}{24}\Delta t^4 u_{t(4)} + \dots \mathcal{O}(\Delta t^5)$$

Expanding the right-hand side of the expression I get,

$$\text{RHS} = u_j^n - \frac{\sigma}{2} (3u_j^n - 4u_{j-1}^n + u_{j-2}^n) + \frac{\sigma^2}{2} (u_{j-2}^n - 2u_{j-1}^n + u_j^n)$$

Expressing each quantity I get,

$$\begin{aligned} 3u_j^n - 4u_{j-1}^n + u_{j-2}^n &= 3u_j^n + \dots \\ &\quad - 4 \left( u_j^n - \Delta x u_x + \frac{1}{2}\Delta x^2 u_{xx} - \frac{1}{6}\Delta x^3 u_{xxx} + \frac{1}{24}\Delta x^4 u_{x(4)} + \dots \mathcal{O}(\Delta x^5) \right) + \dots \\ &\quad + u_j^n - 2\Delta x u_x + 2\Delta x^2 u_{xx} - \frac{4}{3}\Delta x^3 u_{xxx} + \frac{2}{3}\Delta x^4 u_{x(4)} + \dots \mathcal{O}(\Delta x^5) \\ &= 2\Delta x u_x - \frac{2}{3}\Delta x^3 u_{xxx} + \frac{1}{2}\Delta x^4 u_{x(4)} + \mathcal{O}(\Delta x^5) \\ u_{j-2}^n - 2u_{j-1}^n + u_j^n &= u_j^n - 2\Delta x u_x + 2\Delta x^2 u_{xx} - \frac{4}{3}\Delta x^3 u_{xxx} + \frac{2}{3}\Delta x^4 u_{x(4)} + \dots \mathcal{O}(\Delta x^5) + \dots \\ &\quad - 2 \left( u_j^n - \Delta x u_x + \frac{1}{2}\Delta x^2 u_{xx} - \frac{1}{6}\Delta x^3 u_{xxx} + \frac{1}{24}\Delta x^4 u_{x(4)} + \dots \mathcal{O}(\Delta x^5) \right) + \dots \\ &\quad + u_j^n \\ &= \Delta x^2 u_{xx} - \Delta x^3 u_{xxx} + \frac{7}{12}\Delta x^4 u_{x(4)} - \frac{1}{4}\Delta x^5 u_{x(5)} + \mathcal{O}(\Delta x^6) \end{aligned}$$

Setting up the relationships I get that,

$$\begin{aligned} u_j^n + \Delta t u_t + \frac{1}{2}\Delta t^2 u_{tt} + \frac{1}{6}\Delta t^3 u_{ttt} + \frac{1}{24}\Delta t^4 u_{t(4)} + \dots \mathcal{O}(\Delta t^5) &= u_j^n - \dots \\ &\quad \frac{\sigma}{2} \left( 2\Delta x u_x - \frac{2}{3}\Delta x^3 u_{xxx} + \frac{1}{2}\Delta x^4 u_{x(4)} + \mathcal{O}(\Delta x^5) \right) + \dots \\ &\quad \frac{\sigma^2}{2} \left( \Delta x^2 u_{xx} - \Delta x^3 u_{xxx} + \frac{7}{12}\Delta x^4 u_{x(4)} - \frac{1}{4}\Delta x^5 u_{x(5)} + \mathcal{O}(\Delta x^6) \right) \\ &= u_j^n - \sigma \Delta x u_x + \frac{\sigma^2}{2} \Delta x^2 u_{xx} - \frac{1}{6}(3\sigma^2 - 2\sigma) \Delta x^3 u_{xxx} + \frac{1}{24}(7\sigma^2 - 6\sigma) \Delta x^4 u_{x(4)} \end{aligned}$$

Starting with subtracting the  $u_j^n$  terms and expanding  $\sigma$  gives,

$$\begin{aligned} \Delta t u_t + \frac{1}{2} \Delta t^2 u_{tt} + \frac{1}{6} \Delta t^3 u_{ttt} + \frac{1}{24} \Delta t^4 u_{t(4)} + \dots \mathcal{O}(\Delta t^5) \\ = \frac{a \Delta t}{\Delta x} \left( -\Delta x u_x + \frac{\sigma}{2} \Delta x^2 u_{xx} - \frac{1}{6} (3\sigma - 2) \Delta x^3 u_{xxx} + \frac{1}{24} (7\sigma - 6) \Delta x^4 u_{x(4)} \right) \end{aligned}$$

From here I will simplify by dividing through by  $\Delta t$  and distributing  $\Delta x$ ,

$$\begin{aligned} u_t + \frac{1}{2} \Delta t u_{tt} + \frac{1}{6} \Delta t^2 u_{ttt} + \frac{1}{24} \Delta t^3 u_{t(4)} + \dots \mathcal{O}(\Delta t^4) \\ = a \left( -u_x + \frac{\sigma}{2} \Delta x u_{xx} + \frac{a}{6} (3\sigma - 2) \Delta x^2 u_{xxx} + \frac{1}{24} (7\sigma - 6) \Delta x^4 u_{x(4)} \right) \end{aligned}$$

Collecting the one-dimensional advection term to the same side,

$$\begin{aligned} u_t + a u_x = -\frac{1}{2} \Delta t u_{tt} - \frac{1}{6} \Delta t^2 u_{ttt} - \frac{1}{24} \Delta t^3 u_{t(4)} + \frac{\sigma a}{2} \Delta x u_{xx} + \dots \\ + \frac{a}{6} (3\sigma - 2) \Delta x^2 u_{xxx} + \frac{1}{24} (7\sigma - 6) \Delta x^4 u_{x(4)} \end{aligned}$$

Now with the expression for the one-dimensional advection solved for, I will relate temporal derivatives to spatial indices by conducting expansions,

$$\begin{aligned} u_{tt} &= -\frac{1}{2} \Delta t u_{ttt} - a u_{xt} + \frac{\sigma a}{2} \Delta x u_{xxt} + \mathcal{O}(\Delta x^2, \Delta t^2) \\ u_{tx} &= -\frac{1}{2} \Delta t u_{ttx} - a u_{xx} + \frac{\sigma a}{2} \Delta x u_{xxx} + \mathcal{O}(\Delta x^2, \Delta t^2) \\ u_{ttt} &= -a u_{xtt} + \mathcal{O}(\Delta x, \Delta t) \\ u_{txx} &= -a u_{xxx} + \mathcal{O}(\Delta x, \Delta t) \\ u_{ttx} &= -a u_{xxt} \end{aligned}$$

With the higher mixed-derivatives solved for, backtracking will find the  $\alpha$  and  $\beta$  coefficients,

$$\begin{aligned} u_{txx} &= -a u_{xxx} \\ u_{ttx} &= a^2 u_{xxx} \\ u_{ttt} &= -a^3 u_{xxx} \\ u_{tx} &= -\frac{1}{2} \Delta t a^2 u_{xxx} - a u_{xx} + \frac{\sigma a}{2} \Delta x u_{xxx} \\ &= -a u_{xx} + \left( \frac{\sigma a}{2} \Delta x - \frac{a^2 \Delta t}{2} \right) u_{xxx} \xrightarrow{0} -a u_{xx} \\ u_{tt} &= \frac{1}{2} \Delta t (a^3 u_{xxx}) + a^2 u_{xx} - \frac{\sigma a^2}{2} \Delta x u_{xxx} = a^2 u_{xx} \\ u_t + a u_x &= -\frac{1}{2} \Delta t a^2 u_{xx} + \frac{1}{6} \Delta t^2 a^3 u_{xxx} + \frac{\sigma a}{2} \Delta x u_{xx} + \frac{a}{6} (3\sigma - 2) \Delta x^2 u_{xxx} + \mathcal{O}(\Delta x^3, \Delta t^3) \\ &= \underbrace{\left( -\frac{1}{2} \Delta t a^2 + \frac{\sigma a}{2} \Delta x \right)}_{\alpha} u_{xx} + \underbrace{\left( \frac{1}{6} \Delta t^2 a^3 + \frac{a}{6} (3\sigma - 2) \Delta x^2 \right)}_{-\beta} u_{xxx} \\ &= 0 \cdot u_{xx} + a \left( \frac{1}{6} \frac{\Delta x^2}{\Delta x^2} \Delta t^2 a^2 + \frac{a}{6} (3\sigma - 2) \Delta x^2 \right) \\ &= 0 \cdot u_{xx} + a \left( \frac{\Delta x^2}{6} \sigma^2 + \frac{a}{6} (3\sigma - 2) \Delta x^2 \right) u_{xxx} \end{aligned}$$

After further simplifications,

$$u_t + au_x = 0 \cdot u_{xx} + \frac{a\Delta x^2}{6} (\sigma^2 - 3\sigma + 2) u_{xxx}$$

This gives that the  $\alpha$  and  $\beta$  expressions are,

$$\alpha = 0, \quad \beta = -\frac{a\Delta x^2}{6} (\sigma^2 - 3\sigma + 2)$$

**Looking above to the dispersion (the coefficient  $\beta$ ) will denote how waves of different frequencies will move at different speeds. This dispersion term will be the cause of oscillations where they were not present before.**

- b. Perform a von-Neumann stability analysis of the Beam-Warming method. What is the stability limit for the CFL number  $\sigma$ ?

In order to complete the von-Neumann stability analysis, first look to the  $\beta$  coefficient to be greater than or equal to zero as the limiting case,

$$\sigma^2 - 3\sigma + 2 \geq 0$$

Performing simple factorization,

$$(\sigma - 2)(\sigma - 1) \geq 0$$

This gives the roots to be,

$$\sigma = 1, 2$$

Since these are the roots, and this is a concave-up parabolic function then from  $\sigma \in [1, 2] < 0$  thus the actual limits for the CFL number  $\sigma$  as,

$$\sigma \in [0, 1] \cup [2, \infty)$$

- c. Implement the BW method in a computer program using  $L = 2$ ,  $a = 0.5$ ,  $u_0(x) = \exp[-100(x/L - 0.5)^2]$  and a final time of  $T = L/a$  (1 period). Perform spatial and temporal convergence studies to demonstrate the order of accuracy in space and time.



### 3 Manufactured Solutions

Discretize the one-dimensional advection-diffusion equation,  $u_t + au_x - \nu u_{xx} = 0$ , using the trapezoidal method in time and second-order central differences in space. Assume a grid of length  $L$ , periodic boundaries, and  $N$  spatial intervals.

- Write a computer program that implements the given method, and run a simulation using the parameters given in problem 2c,  $\nu = 0.1$ ,  $N = 64$ , and a CFL number of  $\sigma = 0.5$ . Plot the state at the final time,  $u(x, T)$ .
- Apply the method of manufactured solutions to your discretization. The PDE will now need a source term:  $u_t + au_x - \nu u_{xx} = s(x, t)$ . Derive the form of  $s(x, t)$  for the manufactured solution  $u^{MS} = \sin(kx - \omega t)$ , with  $k = 4\pi/L$  and  $\omega = 0.5a/L$ .
- Implement the method of manufactured solutions in your discretization, and present the solution at  $t = T = L/a$  for  $N = 64$ ,  $\sigma = 0.5$ .
- Using the manufactured solution, perform spatial and temporal convergence studies of your discretization and verify that the orders of accuracy match your expectations.

# Matlab Code for A-Stable Backwards Difference

**Algorithm 1:** Matlab Code for determining A-Stable backwards differences.

```

1  %-----
2  clear all; clc; close all
3  set(groot,'defaulttextinterpreter','latex');
4  set(groot,'defaultAxesTickLabelInterpreter','latex');
5  set(groot,'defaultLegendInterpreter','latex');
6  %-----
7  % a
8  syms u ut utt uttt utttt uttttt dt
9
10 unpm1 = u + dt*ut + 1/2*dt^2*utt + 1/6*dt^3*uttt + 1/24*dt^4*utttt + 1/120*dt^5*uttttt;
11 un = u;
12 unmm1 = u - dt*ut + 1/2*dt^2*utt - 1/6*dt^3*uttt + 1/24*dt^4*utttt + 1/120*dt^5*uttttt;
13 unmm2 = u - 2*dt*ut + 2*dt^2*utt - 4/3*dt^3*uttt + 2/3*dt^4*utttt + (-2)^5/120*dt^5*uttttt;
14 fnpm1 = ut + dt*utt + 1/2*dt^2*uttt + 1/6*dt^3*utttt + 1/24*dt^4*uttttt;
15
16 epsi = 5/3*unpm1 - 5/2*un + unmm1 - 1/6*unmm2 - dt*fnpm1;
17 pretty(simplify(epsi))
18 %-----
19 % b
20 num = 100;
21 bdf2 = zeros(num, 1);
22 bdf3 = zeros(num, 1);
23 avg_der = zeros(num, 1);
24 thetlin = linspace(0, 2*pi, num);
25 g = exp(thetlin.*1i);
26 syms ldt
27 for i = 1:num
28     eqn = g(i) == 4/3 - 1/(3*g(i)) + 2/3*ldt*g(i);
29     sol = double(solve(eqn, ldt));
30     bdf2(i) = sol;
31
32     eqn = g(i) == 18/11 - 9/(11*g(i)) + 2/(11*g(i)^2) + 6/11*ldt*g(i);
33     sol = double(solve(eqn, ldt));
34     bdf3(i) = sol;
35
36     eqn = 5/3*g(i) - 5/2 + g(i)^(-1) - 1/6*g(i)^(-2) == ldt*g(i);
37     sol = double(solve(eqn, ldt));
38     avg_der(i) = sol;
39 end
40
41 figure()
42 hold on
43 fill(real(bdf3), imag(bdf3), [1,1,1], 'facealpha', 1, 'FaceColor', [1,0,0], 'EdgeColor', 'k', 'linewidth', 1.8)
44 fill(real(avg_der), imag(avg_der), [1,1,1], 'facealpha', 1, 'FaceColor', [0,0,1], 'EdgeColor', 'k', 'linewidth', 1.8)
45 fill(real(bdf2), imag(bdf2), [1,1,1], 'facealpha', 1, 'FaceColor', [0.8,0.8,0.8], 'EdgeColor', 'k', 'linewidth', 1.8)
46 xlim([-2.5, 0])
47 axis equal
48 xlabel('$Re \left( \lambda \Delta T \right)$', 'fontsize', 18)
49 ylabel('$Im \left( \lambda \Delta T \right)$', 'fontsize', 18)
50 legend({'BDF3', '$\frac{1}{2}\frac{du}{dt}|_{BDF2} + \frac{1}{2}\frac{du}{dt}|_{BDF3}$', 'BDF2'}, 'fontsize', 18, 'location', 'best', 'interpreter', 'latex')
51 set(gcf, 'Color', 'w', 'Position', [100 100 1000 500]);
52 export_fig('eigs.eps')
53
54 syms ldt theta
55 g = exp(1i*theta);
56 eqn = 0 == 5/3 - 5/2*g^-1 + g^-2 - 1/6*g^-3;
57 sol = double(solve(eqn, theta));

```