

# Project Report

## 2. Root finding

### 2.1

Given that  $\tan(\theta) = 2 \cot(\beta) \frac{M^2 \sin^2(\beta) - 1}{M^2(\gamma + \cos(2\beta)) + 2}$ , when  $\theta = 0$ ,  $\tan(\theta) = 0$ :

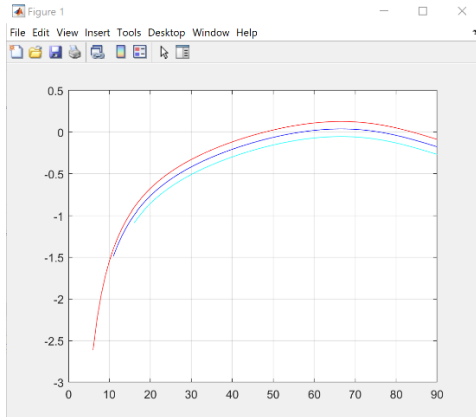
$$0 = 2 \cot(\beta) \frac{M^2 \sin^2(\beta) - 1}{M^2(\gamma + \cos(2\beta)) + 2}$$

Hence, either  $2 \cot(\beta) = 0$  or  $M^2 \sin^2(\beta) - 1 = 0$

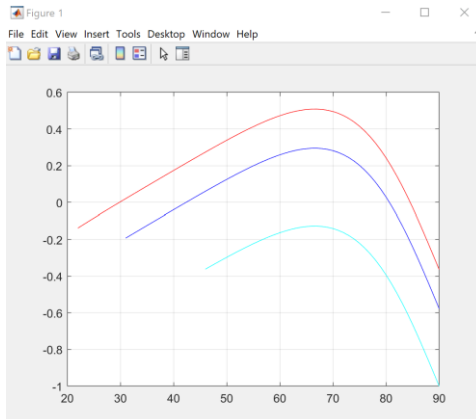
1.  $2 \cot(\beta) = 0$ ,  $\beta = 90^\circ \times Z$ ,  $Z = \text{integer}$ . Since in this question,  $\beta$  stands for the shock wave angle,  $0 < \beta < 180$ , thus  $\beta_1 = 90^\circ$
2.  $M^2 \sin^2(\beta) - 1 = 0$ ,  $M^2 \sin^2(\beta) = 1$ ,  $\sin^2(\beta) = \frac{1}{M^2}$ ,  $\sin(\beta) = \pm \frac{1}{M}$ ,  $\beta_2 = \arcsin\left(\frac{1}{M}\right)$   
hence,  $\beta_U = 90^\circ$ ,  $\beta_L = \arcsin\left(\frac{1}{M}\right)$

### 2.2

- a)  $M = 1.5$ ,  $\theta = 5^\circ$  (red),  $10^\circ$  (blue),  $15^\circ$  (cyan)



- b)  $M = 5$ ,  $\theta = 20^\circ$  (red),  $30^\circ$  (blue),  $45^\circ$  (cyan)



Observed from the graph, when **both  $\beta_U$  and  $\beta_L$  exist**, for the **same  $M$** , when  $\theta$  increases,  $\beta_U$  decreases and  $\beta_L$  increases.

**In terms of  $\theta_{max}$** , for  $M=1.5$ , when  $\theta = 15^\circ$ , there's not a root,  $\theta_{max}$  must be between  $10^\circ$  and  $15^\circ$ .

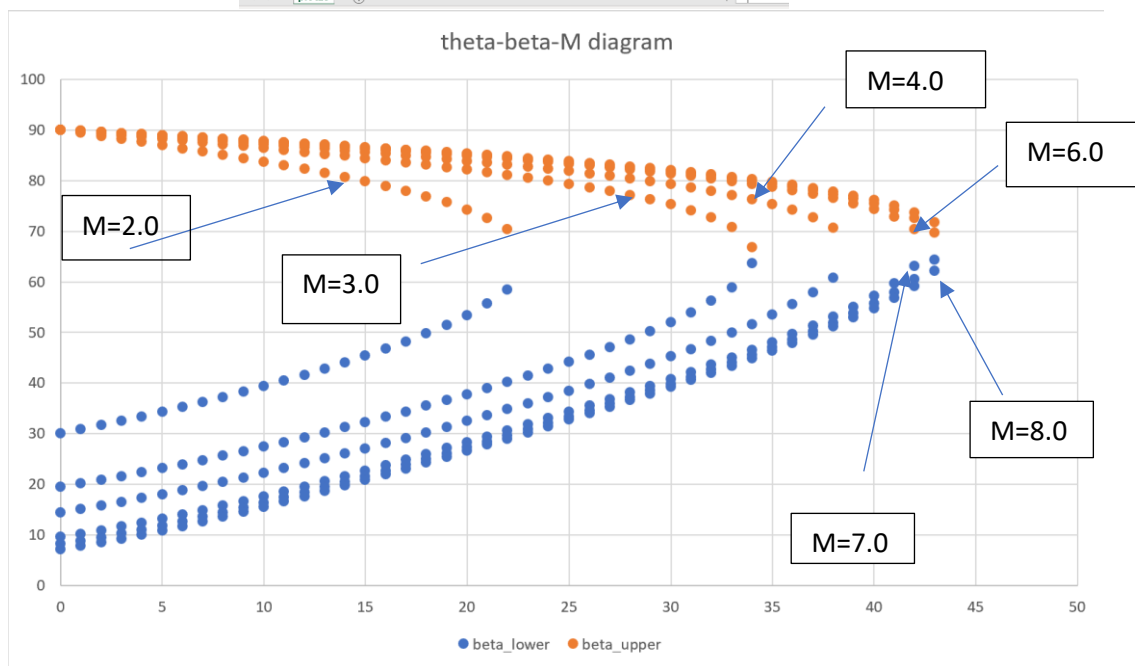
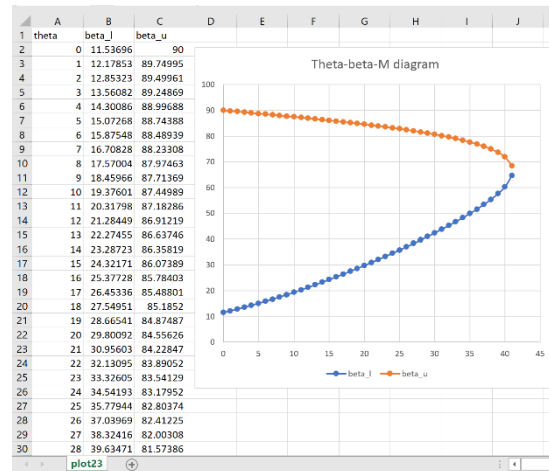
for  $M=5$ , when  $\theta = 45^\circ$ , there's no root,  $\theta_{max}$  must be above  $30^\circ$  below  $45^\circ$ .

### 2.3

- a) For  $M = 5.0, \theta = 20^\circ$ , I chose the two extremes:  $\beta_U = 90^\circ, \beta_L = \arcsin\left(\frac{1}{M}\right) = 11.54$  as the initial guesses for the roots.

```
/* set the initial guess */
ibeta_l = asin(1.0 / m);
ibeta_u = IBETAU * FACTOR;
```

b)



c)

## 3. Regression

First, focus on the second row of matrix in each side, expand the equation:

$$a \sum_{i=1}^N x_i + b \cdot N = \sum_{i=1}^N y_i$$

Divide by N

$$a \sum_{i=1}^N \frac{x_i}{N} + b = \sum_{i=1}^N \frac{y_i}{N}$$

Simplify the equation

$$a\bar{x} + b = \bar{y}$$

$$1. \quad b = \bar{y} - a\bar{x}$$

Recall from lecture that:

$$S = \sum_{i=1}^N (y_i - ax_i - b)^2$$

To minimize S, find the derivative of S:

$$\frac{\partial S}{\partial a} = \sum_{i=1}^N 2(y_i - ax_i - b)(-x_i) = 0$$

Substitute b from equation 1

$$\sum_{i=1}^N 2(y_i - ax_i - \bar{y} + a\bar{x})(-x_i) = 0$$

$$\sum_{i=1}^N -x_i y_i + ax_i^2 + x_i \bar{y} - a\bar{x} x_i = 0$$

Separate into two sums:

$$\sum_{i=1}^N (x_i y_i - x_i \bar{y}) = a \sum_{i=1}^N (x_i^2 - \bar{x} x_i)$$

$$a = \frac{\sum_{i=1}^N (x_i y_i - x_i \bar{y})}{\sum_{i=1}^N (x_i^2 - \bar{x} x_i)}$$

Since

$$\sum_{i=1}^N (x_i / N) = \bar{x}$$

So that we can obtain:

$$\sum_{i=1}^N (\bar{x}^2 - \bar{x} x_i) = 0 \quad \text{and} \quad \sum_{i=1}^N (\bar{x} \bar{y} - \bar{x} y_i) = 0$$

Add these two zero terms to both denominator and numerator:

$$a = \frac{\sum_{i=1}^N (x_i y_i - x_i \bar{y}) + 0}{\sum_{i=1}^N (x_i^2 - \bar{x} x_i) + 0} = \frac{\sum_{i=1}^N (x_i y_i - x_i \bar{y}) + \sum_{i=1}^N (\bar{x} \bar{y} - \bar{x} y_i)}{\sum_{i=1}^N (x_i^2 - \bar{x} x_i) + \sum_{i=1}^N (\bar{x}^2 - \bar{x} x_i)}$$

Refine and combine the terms:

$$a = \frac{\sum_{i=1}^N (x_i y_i - 2x_i \bar{y} + \bar{x} \bar{y})}{\sum_{i=1}^N (x_i^2 - 2\bar{x} x_i + \bar{x}^2)} = \frac{\sum_{i=1}^N (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^N (x_i - \bar{x})^2}$$

When  $\sum_{i=1}^N (x_i - \bar{x})^2 = 0$ , this linear regression will fail.

## 4. Linear Algebraic Systems

**Start from the first row**, Multiply each term by

$$c_2/a_1$$

the new row becomes

$$[c_1, c_2 b_1/a_1 | b_1 Q_1/a_1]$$

then minus the second row by this new row, the second row now becomes

$$[a_2 - c_2 b_1/a_1, b_2 | Q_2 - Q_1 c_2/a_1]$$

from now on each processed row only has **two terms left**.

for  $i_{th}$  ( $i = 2, 3, \dots, N$ ) row  $c_i, a_i, b_i$  continue applying **Gauss elimination**:

1. Multiply the  $(i-1)_{th}$  row by  $c_i/a_{i-1}^*$ , the first term in  $(i-1)_{th}$  row is  $c_i$ , the second term becomes  $c_i b_{i-1}/a_{i-1}^*$ , Q becomes

$$Q_i - Q_{i-1}^* c_i/a_{i-1}.$$

2. Minus the  $i_{th}$  row by the new row, now the  $i_{th}$  row become

$$\left[ a_i - \frac{c_i b_{i-1}}{a_{i-1}^*}, b_i \mid Q_i - Q_{i-1}^* c_i/a_{i-1} \right]$$

3. Now the matrix becomes the form of the second given matrix.

$$\begin{bmatrix} a_1^* & b_1 & 0 & 0 & \dots & 0 \\ 0 & a_2^* & b_2 & 0 & \dots & 0 \\ 0 & 0 & a_3^* & b_3 & \dots & \vdots \\ \vdots & \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & \dots & 0 & 0 & a_{N-1}^* & b_{N-1} \\ 0 & \dots & 0 & 0 & 0 & a_N^* \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_N \end{Bmatrix} = \begin{Bmatrix} Q_1^* \\ Q_2^* \\ Q_3^* \\ \vdots \\ Q_N^* \end{Bmatrix}$$

4. Hence, start from the last row, going from bottom to top, for  $i = N$ , since only one term left,

$$a_i^* x_i = Q_i^*$$

$$x_i = Q_i^*/a_i^*$$

5. For other rows,  $i = 1, 2, \dots, N-1$ ,

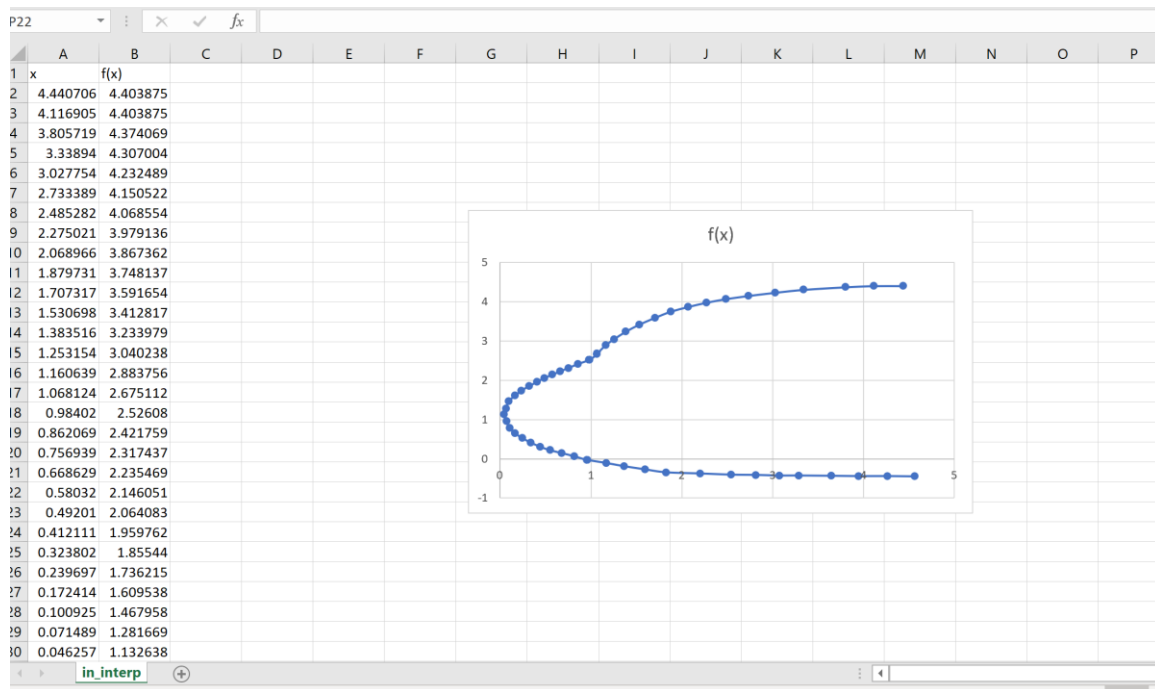
$$a_i^* x_i + b_i x_{i+1} = Q_i^*$$

6. Hence

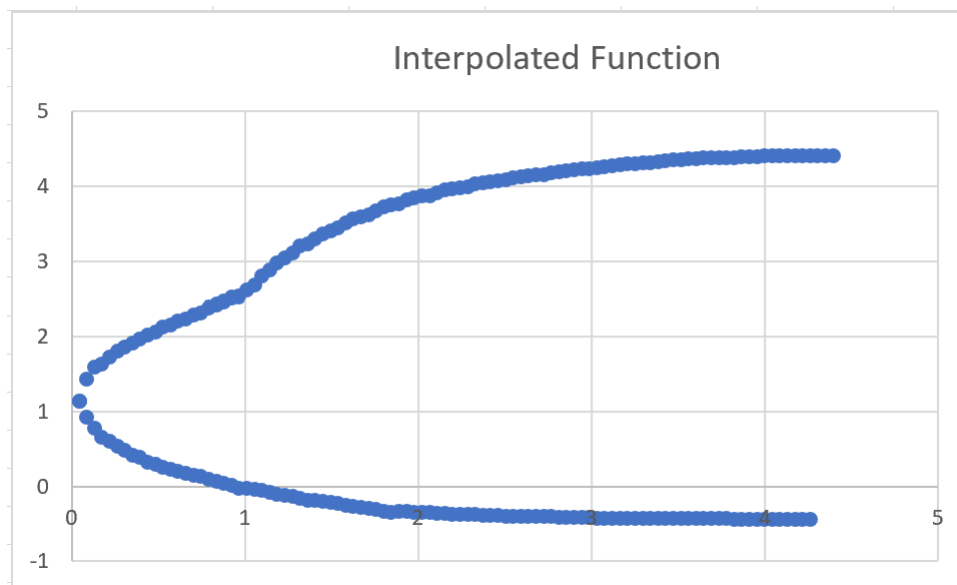
$$x_i = (Q_i^* - b_i x_{i+1})/a_i^*$$

## 5. Interpolation

Plotted Data in Excel



The interpolated functions plotted:

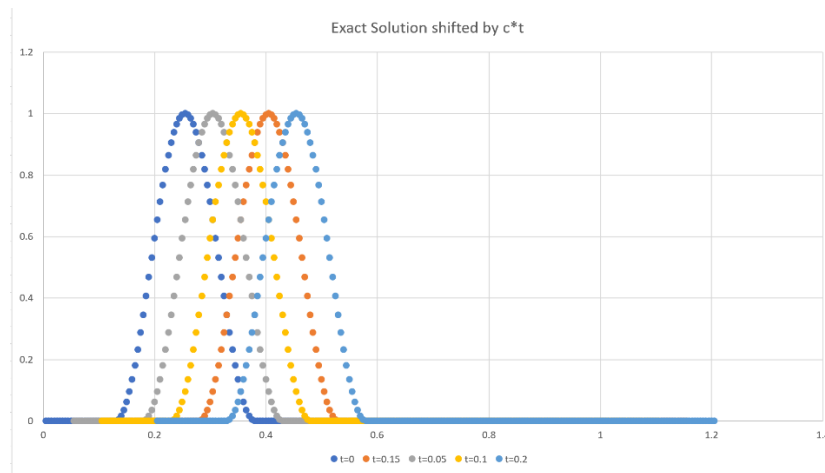


In order to search the corresponded interval by  $x$ , two guarding mechanisms were introduced to check if  $x$  is within the interval that we desired:

- `target >= interp[i].xo) && (target <= (interp[i].xo + interp[i].h))`  
one is applied when the order of  $x$  variables is ascending in interpolated functions
- `target <= interp[i].xo) && (target >= (interp[i].xo + interp[i].h))`  
The other one is applied when the order of  $x$  variable is descending.

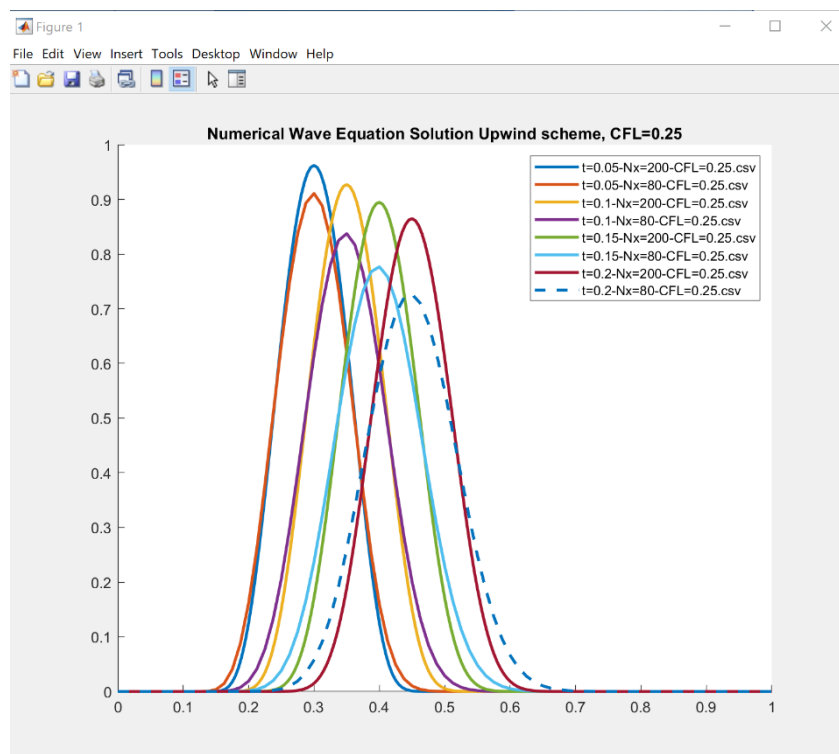
## 6. Differentiation, differential equations

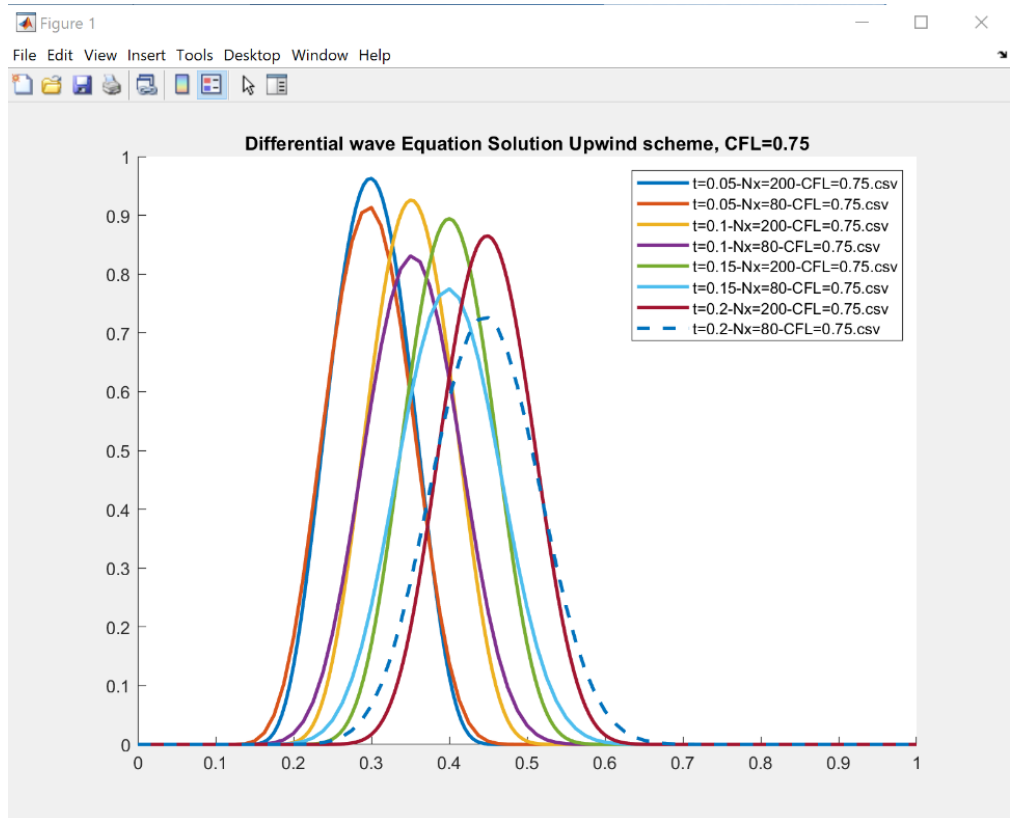
Below is the graph of the **exact solution**:



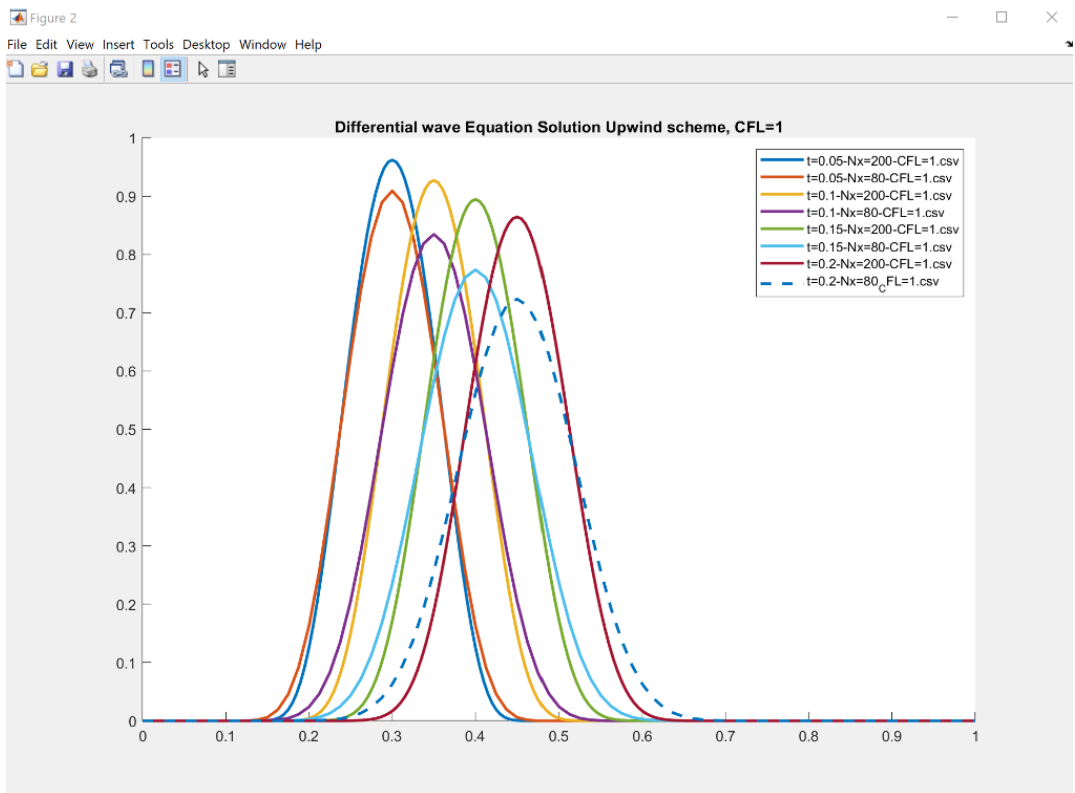
It is easily seen that it shows a wave that shifts from left to right along  $x$  axis during a time. Compared with the exact solution, the **central scheme** gives a **more accurate prediction** than **upwind method**.

- **How does the agreement of the numerical prediction with the exact solution depend on  $\Delta x$ ?**  
Apply variable controlling method, keep CFL value unchanged, plot the curve with  $Nx = 80$  and  $Nx = 200$  in the same grid, here are the results:
  - **The Upwind scheme:**

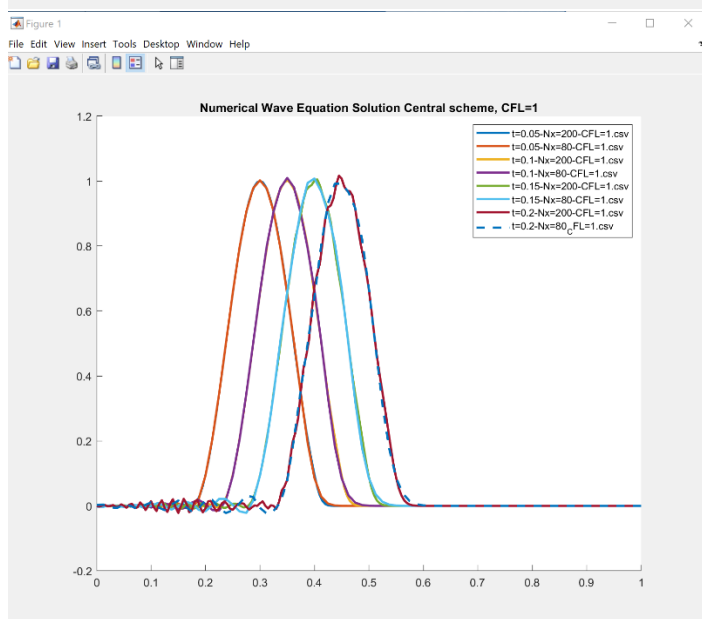
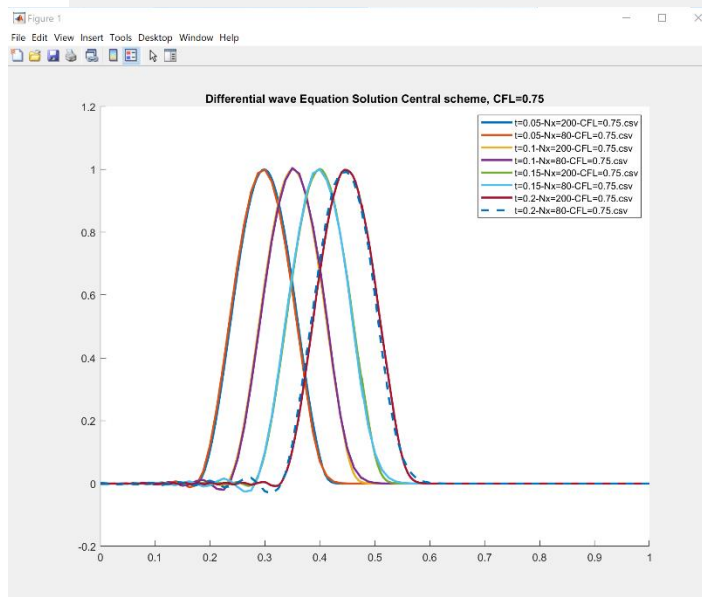
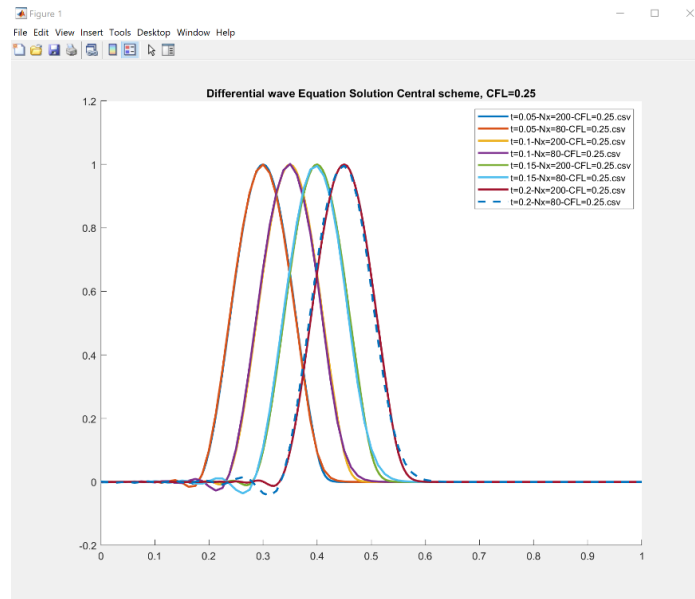




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## The central scheme:

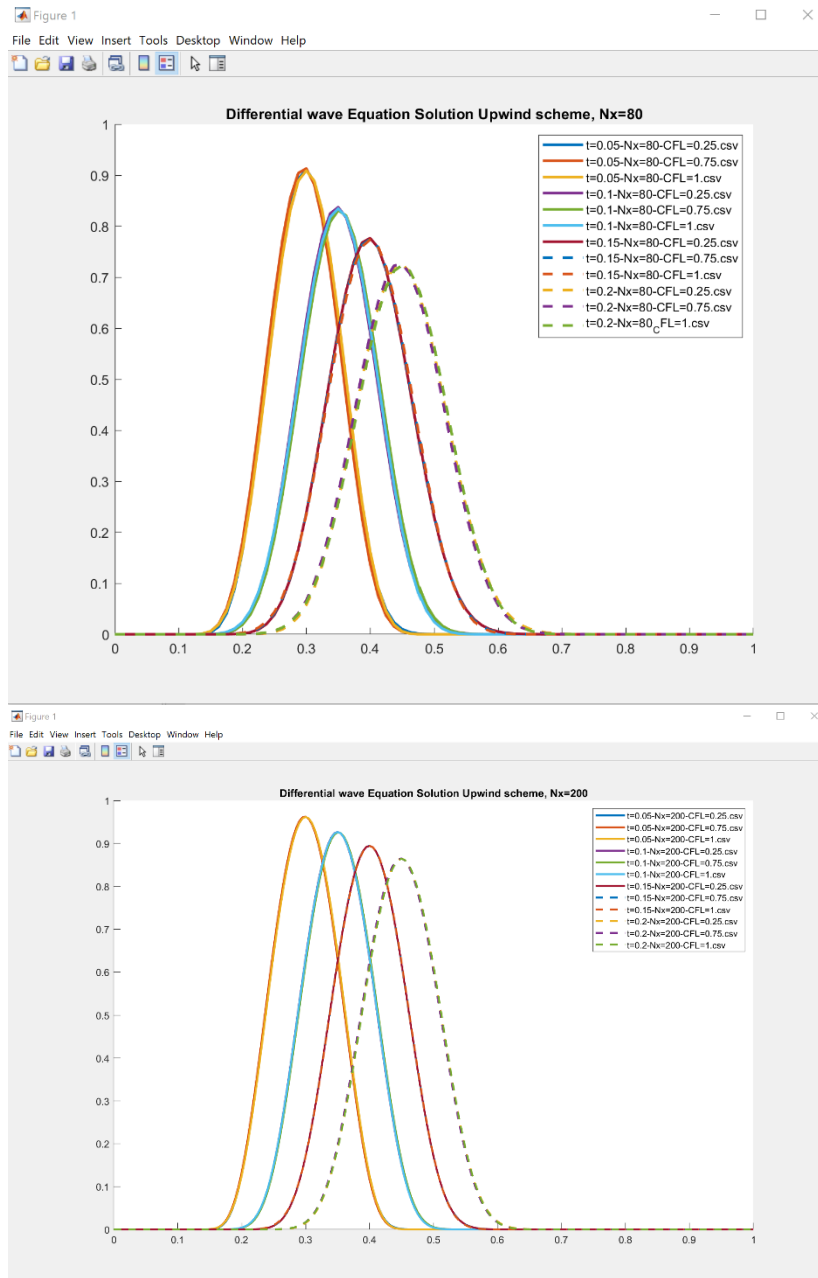




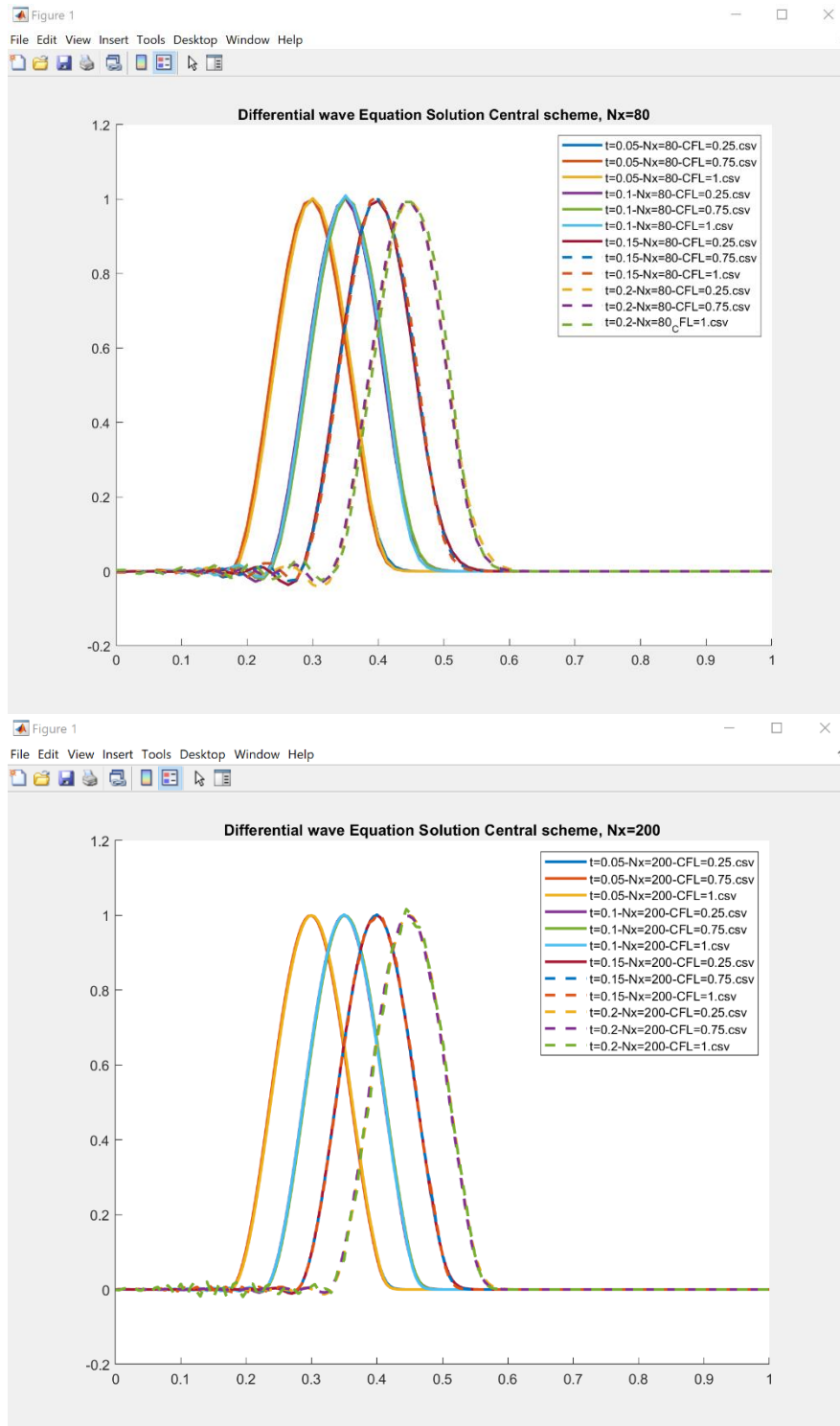
**Discussion:** The grid resolution determines the accuracy of the numerical predicted solution. For all CFL values, the curves with  $Nx = 200$ , have closer peak of wave values with the exact solution than those with  $Nx = 80$  as well as smoother curves. This effect is more obvious in Upwind scheme. In addition, higher resolution also generates more stable curves in horizontal direction, observed from Central scheme graphs. Therefore, higher resolution provides more accurate prediction than lower resolution.

- How does the agreement of the numerical prediction with the exact solution depend on the CFL number?

- Upwind Scheme



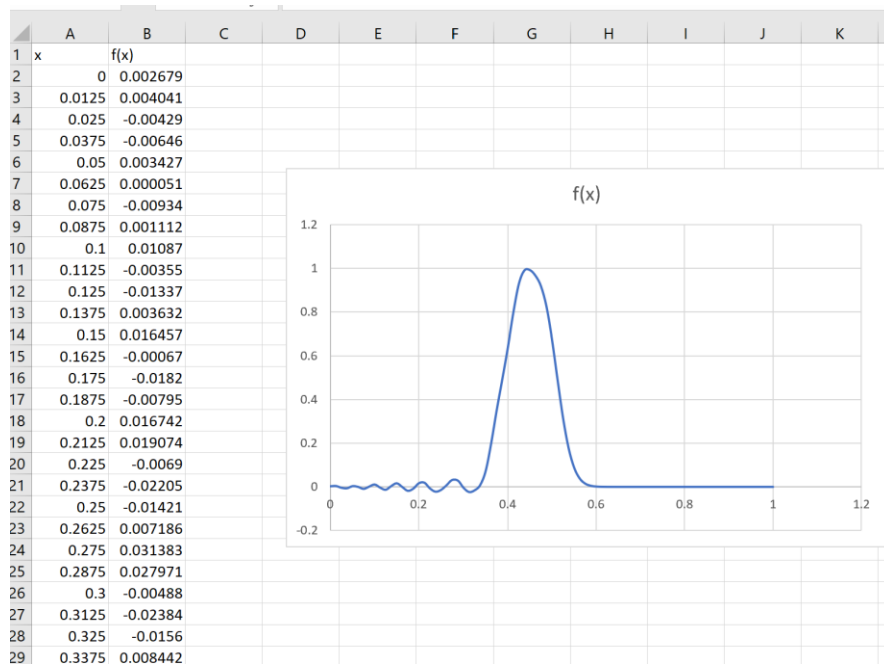
- Central Scheme



**Discussion:** CFL value also determines the accuracy, not for  $\Delta x$  but for  $\Delta t$ , with lower CFL value comes higher time resolution. From the graphs, especially in Central scheme, the effect of different CFL values is significant, the curves with lower CFL value possess **smaller, more stable error area** before entering the wave. Hence, CFL number provides more accurate prediction, more agreement with the exact solution.

- What happens if I chose  $CFL > 1$ ?

Discussion: An appropriate CFL Number is meant to make the code run stably. Once the CFL number exceeds 1, the code can not generate stable curve predictions anymore. When CFL is higher, so as the influence. Below it the case **when CFL=1.002**:



**When CFL = 2:**

