

# Finite difference methods for the wave equation

MATMEK-4270

Prof. Mikael Mortensen, University of Oslo

# The wave equation is a partial differential equation (PDE)

$$\frac{\partial^2 u}{\partial t^2} = c^2 \frac{\partial^2 u}{\partial x^2}.$$

where  $u(x, t)$  is the solution and  $c$  is the constant wavespeed. We will consider the time and space domains:  $t \in [0, T]$ ,  $x \in [0, L]$ .

- The wave equation is an initial-boundary value problem!
- Two initial conditions required since two derivatives in time
- Two boundary conditions required since two derivatives in space
- The solutions are waves that can be written as  $u(x + ct)$  and  $u(x - ct)$

# Wave solution with different boundary conditions

# Boundary conditions

## Dirichlet (Fixed end)

$$u(0, t) = u(L, t) = 0$$

The wave will be reflected, but  $u$  will change sign. A nonzero Dirichlet condition is also possible, but will not be considered here.

## Neumann (Loose end)

$$\frac{\partial u}{\partial x}(0, t) = \frac{\partial u}{\partial x}(L, t) = 0$$

The wave will be reflected without change in sign. A nonzero Neumann condition is also possible, but will not be considered here.

# Boundary conditions continued

Open boundary (No end)

$$\frac{\partial u(0, t)}{\partial t} - c \frac{\partial u(0, t)}{\partial x} = 0$$

$$\frac{\partial u(L, t)}{\partial t} + c \frac{\partial u(L, t)}{\partial x} = 0$$

The wave will simply pass undisturbed and unreflected through an open boundary.

Periodic boundary (No end)

$$u(0, t) = u(L, t)$$

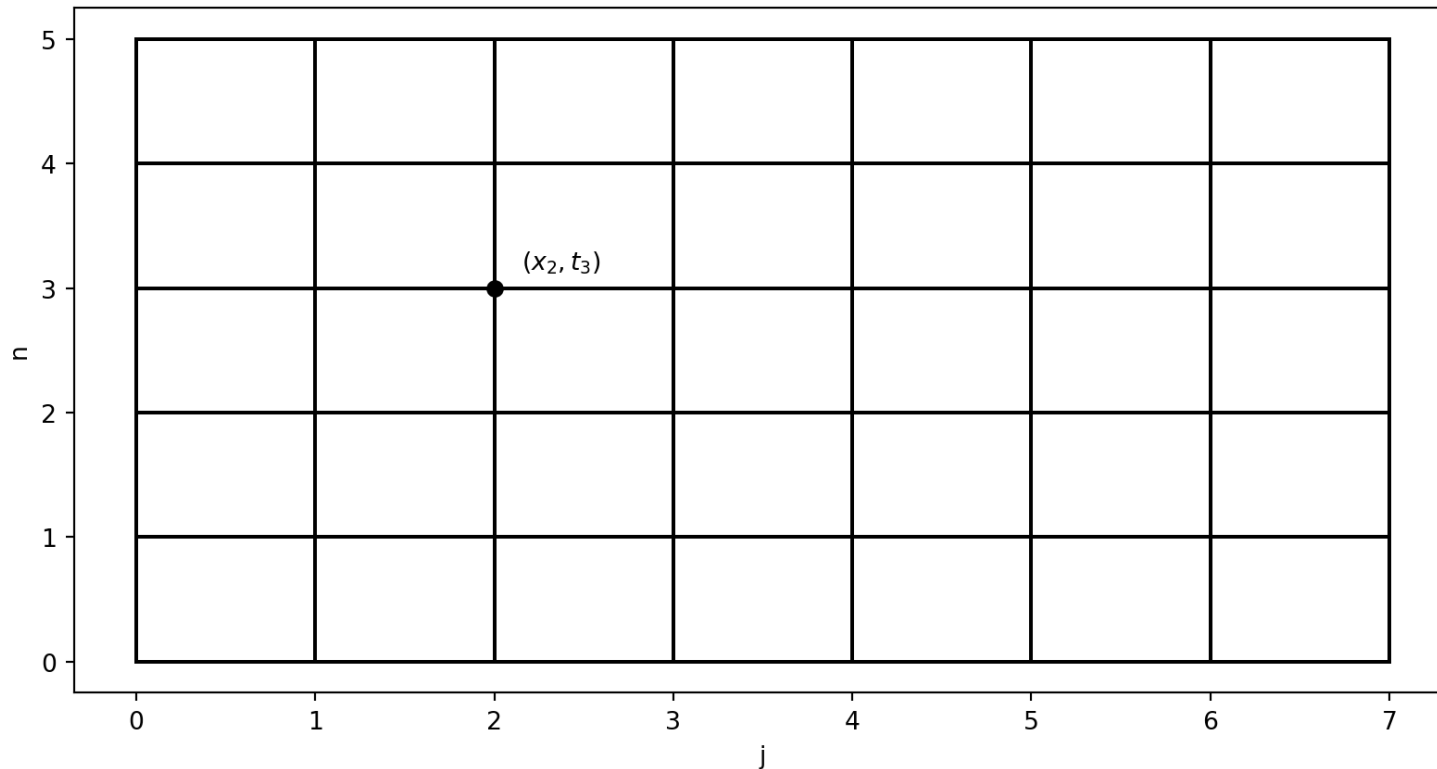
The solution repeats itself indefinitely.

# Discretization

The simplest possible discretization is uniform in time and space

$$t_n = n\Delta t, \quad n = 0, 1, \dots, N_t$$

$$x_j = j\Delta x, \quad j = 0, 1, \dots, N$$



# A mesh function in space and time is defined as

$$u_j^n = u(x_j, t_n)$$

The mesh function has one value at each node in the mesh. For simplicity in later algorithms we will use the vectors

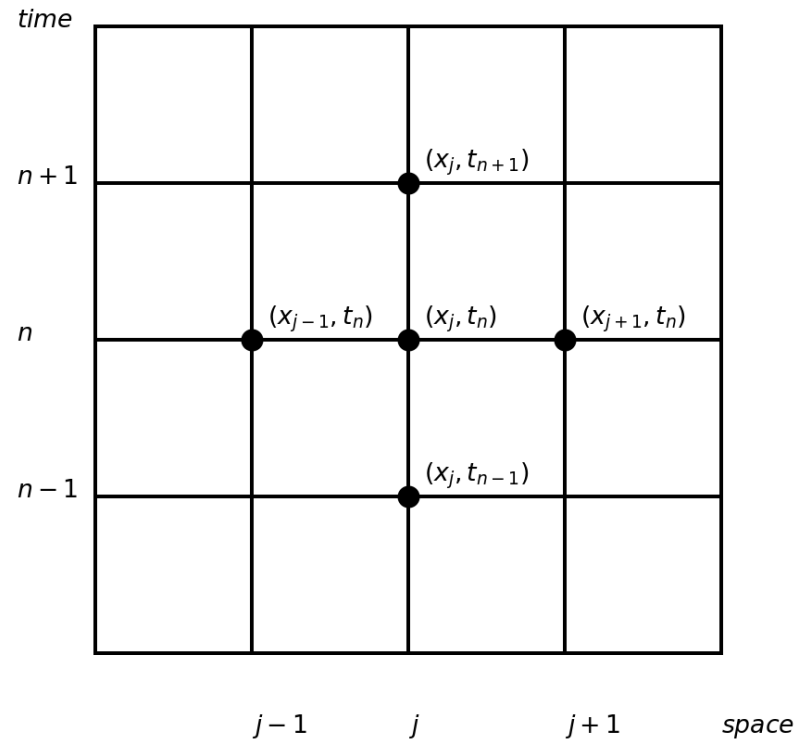
$$u^n = (u_0^n, u_1^n, \dots, u_N^n)^T,$$

which is the solution vector at time  $t_n$ .

A second order accurate discretization of the wave equation is

$$\frac{u_j^{n+1} - 2u_j^n + u_j^{n-1}}{\Delta t^2} = c^2 \frac{u_{j+1}^n - 2u_j^n + u_{j-1}^n}{\Delta x^2}$$

# The finite difference stencil makes use of 5 neighboring points

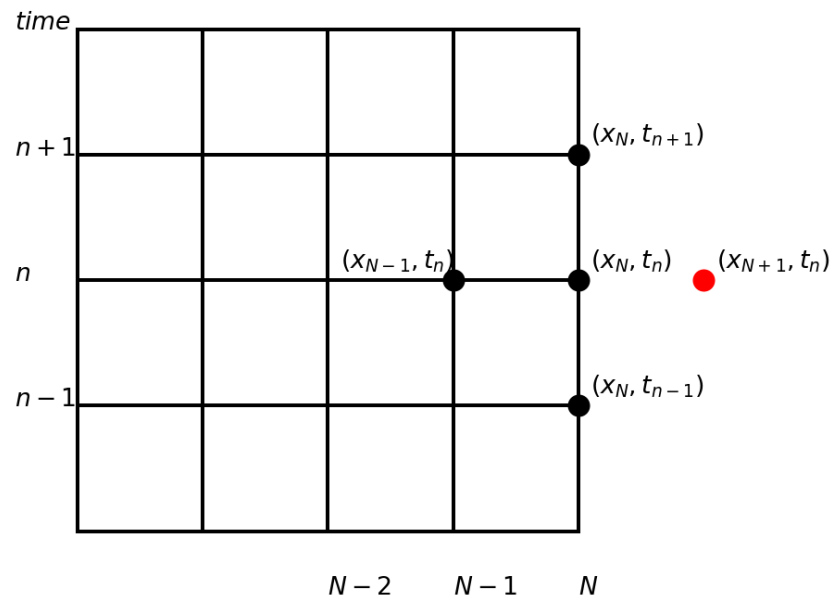


$$\frac{u_j^{n+1} - 2u_j^n + u_j^{n-1}}{\Delta t^2} = c^2 \frac{u_{j+1}^n - 2u_j^n + u_{j-1}^n}{\Delta x^2}$$

Can only be used for **internal points**



# The finite difference stencil is not used at the spatial boundary



$$\frac{u_N^{n+1} - 2u_N^n + u_N^{n-1}}{\Delta t^2} = c^2 \frac{u_{N+1}^n - 2u_N^n + u_{N-1}^n}{\Delta x^2}$$

- Used at the boundary the regular stencil will contain a **ghost node**
- But at the boundary we use boundary conditions and do not solve the PDE!

# We use a marching method in time

1. Initialize  $u^0$  and  $u^1$
2. for  $n$  in range( $1, N_t - 1$ ):
  - for  $j$  in range( $1, N - 1$ ):
    - $u_j^{n+1} = 2u_j^n - u_j^{n-1} + \left(\frac{c\Delta t}{\Delta x}\right)^2 (u_{j+1}^n - 2u_j^n + u_{j-1}^n)$

and apply the chosen boundary conditions.

All the indices makes it a bit messy. Lets make use of a differentiation matrix for the spatial dimension! And the Courant (or CFL) number

$$\bar{c} = \frac{c\Delta t}{\Delta x}$$

# Use differentiation matrix to simplify the notation

We define the second differentiation matrix without the scaling  $1/(\Delta x)^2$  such that

$$D^{(2)} = \begin{bmatrix} -2 & 1 & 0 & 0 & 0 & \cdots & 0 \\ 1 & -2 & 1 & 0 & 0 & \cdots & 0 \\ 0 & 1 & -2 & 1 & 0 & \cdots & 0 \\ \vdots & & \ddots & \ddots & \ddots & & \vdots \\ 0 & \cdots & 0 & 1 & -2 & 1 & 0 \\ 0 & \cdots & 0 & 0 & 1 & -2 & 1 \\ 0 & \cdots & 0 & 0 & 0 & 1 & -2 \end{bmatrix}$$

and thus row  $0 < j < N$  of  $D^{(2)}u^n$  becomes

$$(D^{(2)}u^n)_j = u_{j+1}^n - 2u_j^n + u_{j-1}^n$$

# The vectorized marching method becomes

1. Initialize  $u^0$  and  $u^1$
2. for  $n$  in range( $1, N_t - 1$ ):
  - $u^{n+1} = 2u^n - u^{n-1} + \underline{c}^2 D^{(2)} u^n$
  - Apply boundary conditions to  $u_0^{n+1}$  and  $u_N^{n+1}$
- The boundary step can **often**, but not always, be incorporated into the matrix  $D^{(2)}$
- Very easy to vectorize using the matrix vector product!

# PDE solvers (of time-dependent problems) should use memory carefully

## Note

- At any time we only need to store three vectors:  $u^{n+1}$ ,  $u^n$  and  $u^{n-1}$ .
  - Memory requirement =  $3(N + 1)$  floating point numbers
- Storing all time steps requires  $(N_t + 1) \times (N + 1)$  floating point numbers
- Not a huge problem for our case, but for 2 or 3 spatial dimensions it is very important!

# Implementation - A low-memory marching method needs to update solution vectors

1. Allocate three vectors  $u^{nm1}, u^n, u^{np1}$ , representing  $u^{n-1}, u^n, u^{n+1}$ .
2. Initialize  $u^0$  and  $u^1$  by setting  $u^{nm1} = u^0, u^n = u^1$
3. for  $n$  in range( $1, N_t - 1$ ):
  - $u^{np1} = 2u^n - u^{nm1} + \underline{c}^2 D^{(2)} u^n$
  - Apply boundary conditions to  $u_0^{np1}$  and  $u_N^{np1}$
  - Update to next iteration:
    - $u^{nm1} \leftarrow u^n$
    - $u^n \leftarrow u^{np1}$

# In Python

## Set up solver

```
1 import numpy as np
2 from scipy import sparse
3 import sympy as sp
4 x, t = sp.symbols('x,t')
5 N = 100
6 Nt = 500
7 L = 2
8 c = 1 # wavespeed
9 dx = L / N
10 CFL = 1.0
11 dt = CFL*dx/c
12 xj = np.linspace(0, L, N+1)
13 unml, un, unpr = np.zeros((3, N+1))
14 D2 = sparse.diags([1, -2, 1], [-1, 0, 1], (N+1, N+1))
15 u0 = sp.exp(-200*(x-L/2+t)**2)
```

## Solve by marching method

```
1 unml[:] = sp.lambdify(x, u0.subs(t, 0))(xj)
2 un[:] = sp.lambdify(x, u0.subs(t, dt))(xj)
3 for n in range(Nt):
4     unpr[:] = 2*un - unml + CFL**2 * D2 @ un
5     unpr[0] = 0
6     unpr[-1] = 0
7     unml[:] = un
8     un[:] = unpr
```

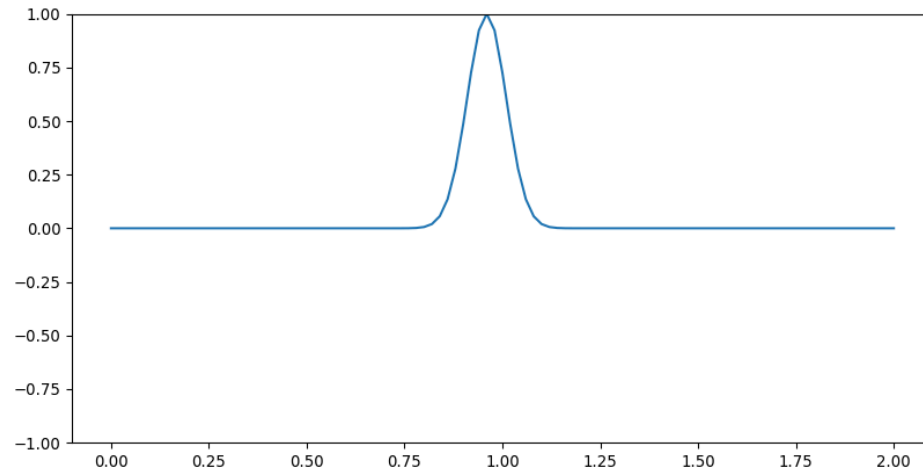
# Store results at intermediate intervals for plotting

```
1 unm1[:] = sp.lambdify(x, u0.subs(t, 0))(xj)
2 un[:] = sp.lambdify(x, u0.subs(t, dt))(xj)
3 plotdata = {0: unm1.copy()}
4 for n in range(Nt):
5     unp1[:] = 2*un - unm1 + CFL**2 * D2 @ un
6     unp1[0] = 0
7     unp1[-1] = 0
8     unm1[:] = un
9     un[:] = unp1
10    if n % 10 == 0:
11        plotdata[n] = unp1.copy()
```

For example every tenth time step. Normally you do not need every time step to get a good animation.



# Create animation after the simulation is finished



```
1 def animation(data):
2     from matplotlib import animation
3     fig, ax = plt.subplots()
4     v = np.array(list(data.values()))
5     t = np.array(list(data.keys()))
6     save_step = t[1]-t[0]
7     line, = ax.plot(xj, data[0])
8     ax.set_ylim(v.min(), v.max())
9     def update(frame):
10         line.set_ydata(data[frame*save_step])
11         return (line,)
12     ani = animation.FuncAnimation(fig=fig, func=update, frames=len(data), blit=True)
13     ani.save('wavemovie.apng', writer='pillow', fps=5) # This animated png opens in a browser
```

# How to implement the initial conditions?

To initialize a mesh function  $u^0$ , we write

$$u^0 = I(x)$$

which represents

$$u_j^0 = I(x_j), \quad \forall j = 0, 1, \dots, N$$

```
1 u0 = sp.exp(-200*(x-L/2+t)**2)
2 unml[:] = sp.lambdify(x, u0.subs(t, 0))(xj)
```

How about the second condition  $\frac{\partial u}{\partial t}(x, 0) = 0$ ?

Just like for the vibration equation there are several options. If you have an analytical solution  $I(x)$  you can specify:

**one wave**

$$u^1 = I(x + c\Delta t)$$

**Two waves**

$$u^1 = 0.5(I(x + c\Delta t) + I(x - c\Delta t))$$

# If you do not have analytic $I(x)$ , then what?

## How to fix $\frac{\partial u}{\partial t}(x, 0) = 0$ , option 1

Use a forward difference

$$\frac{\partial u}{\partial t}(x, 0) \approx \frac{u^1 - u^0}{\Delta t} = 0, \quad \text{such that} \quad u^1 = u^0$$

Only first order accurate, but still a possibility.

Use a second order forward difference

$$\frac{\partial u}{\partial t}(x, 0) \approx \frac{-u^2 + 4u^1 - 3u^0}{2\Delta t} = 0, \quad \text{such that} \quad u^1 = \frac{3u^0 + u^2}{4}$$

Second order accurate, but **implicit**.

# How to implement $\frac{\partial u}{\partial t}(x, 0) = 0$ , option 2

Use a second order central difference

$$\frac{\partial u}{\partial t}(x, 0) = \frac{u^1 - u^{-1}}{2\Delta t} = 0, \quad \text{such that} \quad u^1 = u^{-1}$$

and the PDE at  $n = 0$

$$u^1 = 2u^0 - \textcolor{red}{u}^{-1} + \underline{c}^2 D^{(2)} u^0$$

Insert for  $u^{-1} = u^1$  to obtain

$$u^1 = u^0 + \frac{\underline{c}^2}{2} D^{(2)} u^0$$

Second order accurate and **explicit**

# How to fix boundary conditions?

We will consider 4 different types of boundary conditions

Dirichlet	$u(0, t)$ and $u(L, t)$
Neumann	$\frac{\partial u}{\partial x}(0, t)$ and $\frac{\partial u}{\partial x}(L, t)$
Open	$\frac{\partial u}{\partial t}(0, t) - c \frac{\partial u}{\partial x}(0, t) = 0$ and $\frac{\partial u}{\partial t}(L, t) + c \frac{\partial u}{\partial x}(L, t) = 0$
Periodic	$u(L, t) = u(0, t)$

## Note

Accounting for boundary conditions very often takes more than 50 % of the lines of code in a PDE solver!

# Dirichlet boundary conditions

We need to fix  $u(0, t) = I(0)$  and  $u(L, t) = I(L)$  and start by fixing this at  $t = 0$

$$u_0^0 = I(0) \quad \text{and} \quad u_N^0 = I(L)$$

Next, we compute

$$u^1 = u^0 + \frac{\underline{c}^2}{2} D^{(2)} u^0$$

Here, if the first and last rows of  $D^{(2)}$  are set to zero, then  $u_0^1 = u_0^0$  and  $u_N^1 = u_N^0$ .

Next, for  $n = 1, 2, \dots, N_t - 1$

$$u^{n+1} = 2u^n - u^{n-1} + \underline{c}^2 D^{(2)} u^n$$

Again, if the first and last rows of  $D^{(2)}$  are zero, then  $u_0^{n+1} = u_0^0$  and  $u_N^{n+1} = u_N^0$  for all  $n$ . The boundary values remain as initially set at  $t = 0$ .

# Dirichlet boundary conditions summary

Set  $u^0 = I(x)$  and define a modified differentiation matrix

$$\tilde{D}^{(2)} = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & -2 & 1 & 0 & 0 & 0 & 0 & \dots \\ 0 & 1 & -2 & 1 & 0 & 0 & 0 & \dots \\ \vdots & & & \ddots & & & & \dots \\ \vdots & 0 & 0 & 0 & 1 & -2 & 1 & 0 \\ \vdots & 0 & 0 & 0 & 0 & 1 & -2 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

Now boundary conditions will be ok at all time steps simply by:

1. Initialize  $u^0$  and compute  $u^1 = u^0 + \frac{c^2}{2} \tilde{D}^{(2)} u^0$ . Set  $u^{nm1} = u^0, u^n = u^1$
2. for  $n$  in range( $1, N_t - 1$ ):
  - $u^{np1} = 2u^n - u^{nm1} + \underline{c}^2 \tilde{D}^{(2)} u^n$
  - Update to next iteration:  $u^{nm1} = u^n; u^n = u^{np1}$

# Dirichlet boundary conditions summary

## Note

It is also possible to do nothing with  $D^{(2)}$  and simply fix the boundary conditions after updating all the internal points

1. Initialize  $u^{nm1} = u^0$  and compute  $u^n = u^1 = u^0 + \frac{c^2}{2} D^{(2)} u^0$ .
2. Set  $u_0^{nm1} = u_0^n = 0$  and  $u_N^{nm1} = u_N^n = 0$ .
3. for  $n$  in range(1,  $N_t - 1$ ):
  - $u^{np1} = 2u^n - u^{nm1} + \underline{c}^2 D^{(2)} u^n$
  - Set  $u_0^{np1} = 0$  and  $u_N^{np1} = 0$
  - Update to next iteration:  $u^{nm1} \leftarrow u^n; u^n \leftarrow u^{np1}$

## Note

Regular, unmodified  $D^{(2)}$ , where the first and last rows are completely irrelevant.

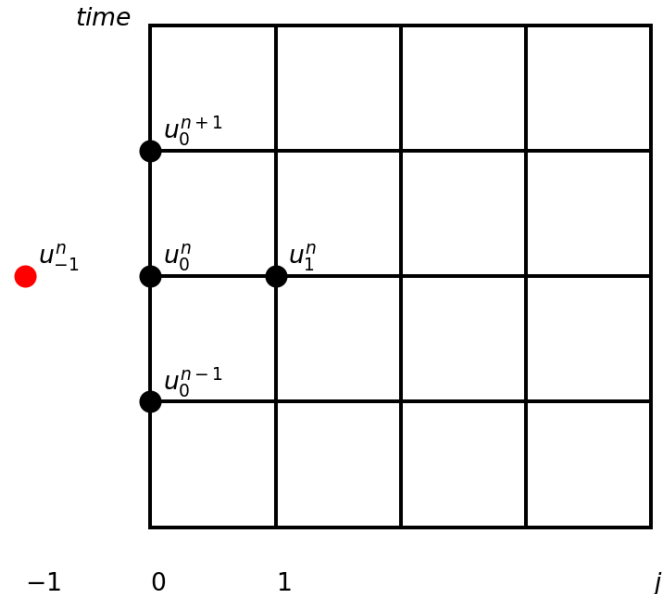


# Neumann boundary conditions

We need to fix  $\frac{\partial u}{\partial x}(0, t) = 0$  and  $\frac{\partial u}{\partial x}(L, t) = 0$ . We already have  $u^0 = I(x)$ .

A second order central scheme at  $x = 0$  is using ghost cell at  $j = -1$

$$\frac{\partial u}{\partial x}(0, t_n) = \frac{u_1^n - u_{-1}^n}{2\Delta x} = 0 \rightarrow u_{-1}^n = u_1^n$$



# Use the ghost cell and the PDE to fix the Neumann condition

The PDE at the left hand side  $j = 0$  using ghost cell:

$$u_0^{n+1} = 2u_0^n - u_0^{n-1} + \underline{c}^2(u_1^n - 2u_0^n + u_{-1}^n)$$

Insert for  $u_{-1}^n = u_1^n$  and obtain

$$u_0^{n+1} = 2u_0^n - u_0^{n-1} + \underline{c}^2(2u_1^n - 2u_0^n)$$

## Note

Second order accurate and **explicit**. Can be implemented by modifying  $D^{(2)}$ !

## Neumann at $x = L$ is the same

$$\frac{\partial u}{\partial x}(L, t_n) = \frac{u_{N+1}^n - u_{N-1}^n}{2\Delta x} = 0 \rightarrow u_{N+1}^n = u_{N-1}^n$$

The PDE at the right hand side  $j = N$  using ghost cell:

$$u_N^{n+1} = 2u_N^n - u_N^{n-1} + \underline{c}^2(u_{N+1}^n - 2u_N^n + u_{N-1}^n)$$

Insert for  $u_{N+1}^n = u_{N-1}^n$  and obtain

$$u_N^{n+1} = 2u_N^n - u_N^{n-1} + \underline{c}^2(2u_{N-1}^n - 2u_N^n)$$

And for  $n = 1$  we similarly get

$$u_0^1 = u_0^0 + \frac{\underline{c}^2}{2}(2u_1^n - 2u_0^n) \quad \text{and} \quad u_N^1 = u_N^0 + \frac{\underline{c}^2}{2}(2u_{N-1}^n - 2u_N^n)$$

# Neumann summary

Set  $u^0 = I(x)$  and define a modified differentiation matrix

$$\tilde{D}^{(2)} = \begin{bmatrix} -2 & 2 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & -2 & 1 & 0 & 0 & 0 & 0 & \dots \\ 0 & 1 & -2 & 1 & 0 & 0 & 0 & \dots \\ \vdots & & & \ddots & & & & \dots \\ \vdots & 0 & 0 & 0 & 1 & -2 & 1 & 0 \\ \vdots & 0 & 0 & 0 & 0 & 1 & -2 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 2 & -2 \end{bmatrix}$$

Now boundary conditions will be ok at all time steps simply by:

1. Initialize  $u^0$  and compute  $u^1 = u^0 + \frac{c^2}{2} \tilde{D}^{(2)} u^0$ . Set  $u^{nm1} = u^0, u^n = u^1$
2. for  $n$  in range( $1, N_t - 1$ ):
  - $u^{np1} = 2u^n - u^{nm1} + \underline{c}^2 \tilde{D}^{(2)} u^n$
  - Update to next iteration:  $u^{nm1} \leftarrow u^n; u^n \leftarrow u^{np1}$

# Open boundary

The wave simply disappears through the boundary

$$\frac{\partial u}{\partial t}(0, t) - c \frac{\partial u}{\partial x}(0, t) = 0 \quad \text{and} \quad \frac{\partial u}{\partial t}(L, t) + c \frac{\partial u}{\partial x}(L, t) = 0$$

As for Neumann there are several ways to implement these boundary conditions. The simplest option is to solve the first order accurate

$$\frac{u_0^{n+1} - u_0^n}{\Delta t} - c \frac{u_1^n - u_0^n}{\Delta x} = 0$$

such that

$$u_0^{n+1} = u_0^n + \frac{c\Delta t}{\Delta x} (u_1^n - u_0^n)$$

## Second order option

$$\frac{u_0^{n+1} - u_0^{n-1}}{2\Delta t} - c \frac{-u_2^n + 4u_1^n - 3u_0^n}{2\Delta x} = 0$$

Solve for the boundary node  $u_0^{n+1}$

$$u_0^{n+1} = u_0^{n-1} + \frac{c\Delta t}{\Delta x} (-u_2^n + 4u_1^n - 3u_0^n)$$

Nice option, but difficult to incorporate in the  $D^{(2)}$  matrix, since there is no way to modify the first and last rows of  $D^{(2)}$  such that

$$u_0^{n+1} = 2u_0^n - u_0^{n-1} + \underline{c}^2 (D^{(2)} u^n)_0$$

# Second second order option

Use central, second order scheme

$$\frac{u_0^{n+1} - u_0^{n-1}}{2\Delta t} - c \frac{u_1^n - u_{-1}^n}{2\Delta x} = 0$$

and isolate the ghost node  $u_{-1}^n$ :

$$\textcolor{red}{u}_{-1}^n = u_1^n - \frac{1}{\underline{c}}(u_0^{n+1} - u_0^{n-1})$$

Use regular PDE at the boundary that includes the ghost node:

$$u_0^{n+1} = 2u_0^n - u_0^{n-1} + \underline{c}^2(u_1^n - 2u_0^n + \textcolor{red}{u}_{-1}^n)$$

This gives an equation for  $u_0^{n+1}$  that fixes the open boundary condition:

$$u_0^{n+1} = 2(1 - \underline{c})u_0^n - \frac{1 - \underline{c}}{1 + \underline{c}}u_0^{n-1} + \frac{2\underline{c}^2}{1 + \underline{c}}u_1^n$$

# Open boundary conditions

Left boundary:

$$u_0^{n+1} = 2(1 - \underline{c})u_0^n - \frac{1 - \underline{c}}{1 + \underline{c}}u_0^{n-1} + \frac{2\underline{c}^2}{1 + \underline{c}}u_1^{n_1}$$

Right boundary:

$$u_N^{n+1} = 2(1 - \underline{c})u_N^n - \frac{1 - \underline{c}}{1 + \underline{c}}u_N^{n-1} + \frac{2\underline{c}^2}{1 + \underline{c}}u_{N-1}^{n_1}$$

Both **explicit** and second order. But not possible to implement into the matrix such that

$$u^{n+1} = 2u^n - u^{n-1} + \underline{c}^2 D^{(2)} u^n$$



# Implementation open boundaries

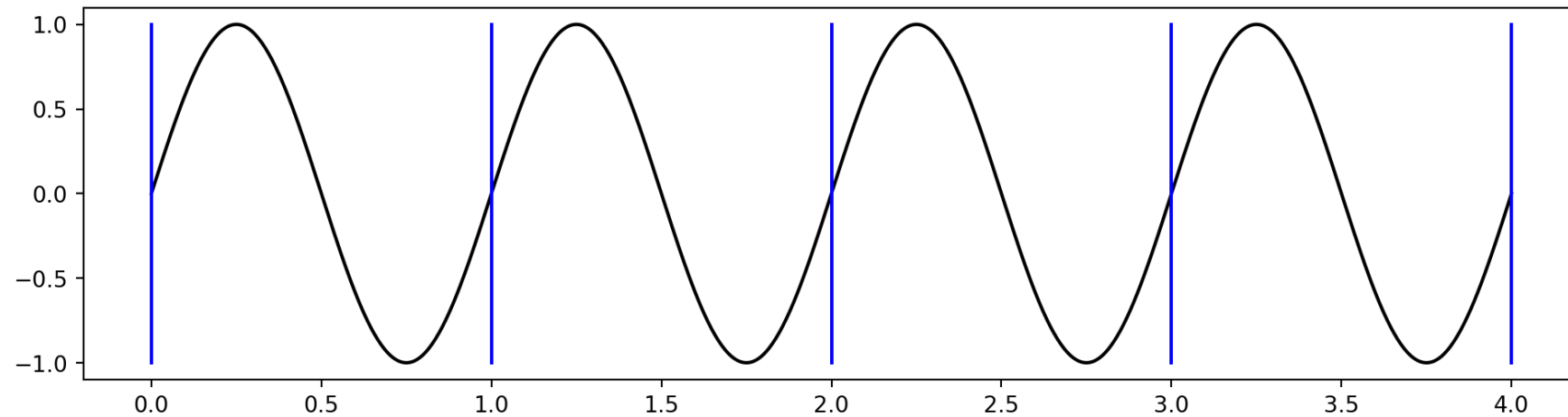
1. Initialize  $u^0$  and compute  $u^1 = u^0 + \frac{\underline{c}^2}{2} D^{(2)} u^0$ . Set  $u^{nm1} = u^0, u^n = u^1$
2. for  $n$  in range(1,  $N_t - 1$ ):
  - $u^{np1} = 2u^n - u^{nm1} + \underline{c}^2 D^{(2)} u^n$
  - $u_0^{np1} = 2(1 - \underline{c})u_0^n - \frac{1-\underline{c}}{1+\underline{c}}u_0^{nm1} + \frac{2\underline{c}^2}{1+\underline{c}}u_1^{nm1}$
  - $u_N^{np1} = 2(1 - \underline{c})u_N^n - \frac{1-\underline{c}}{1+\underline{c}}u_N^{nm1} + \frac{2\underline{c}^2}{1+\underline{c}}u_{N-1}^{nm1}$
  - Update to next iteration:  $u^{nm1} \leftarrow u^n; u^n \leftarrow u^{np1}$

## Note

There is no need to use a modified  $D^{(2)}$ . The two updates of  $u_0^{np1}$  and  $u_N^{np1}$  will overwrite anything computed in the first step.

# Periodic boundary conditions

A periodic solution is a solution that is repeating itself indefinitely. For example  $u(x) = \sin(2\pi x)$ :

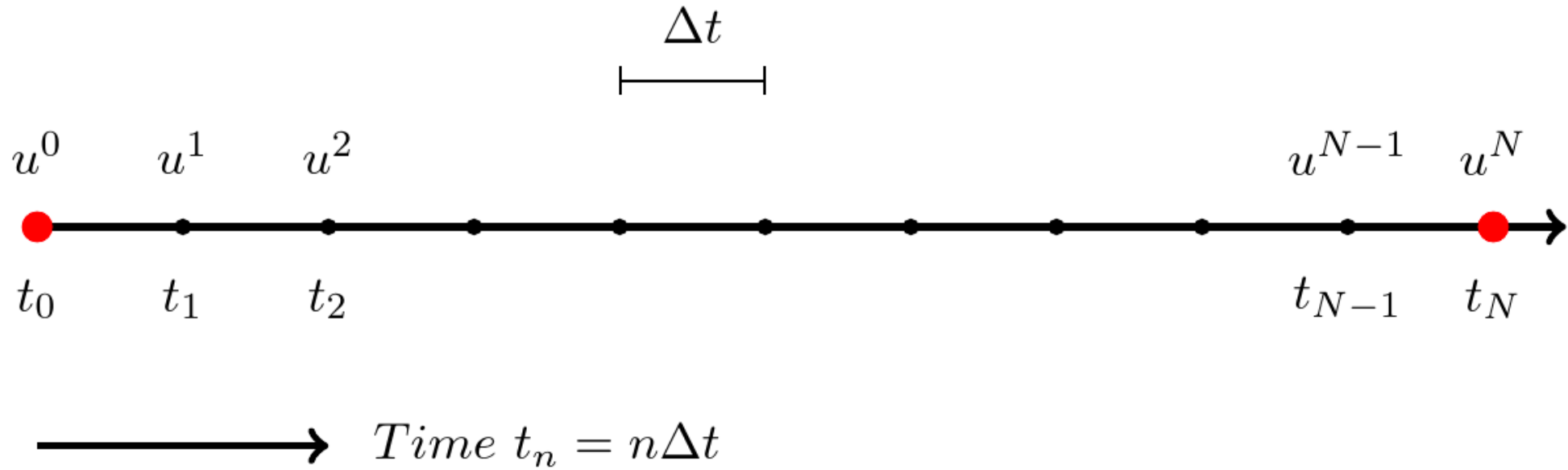


We solve the problem for example for  $x \in [0, 1]$ , but the actual solution will be like above, with no boundaries.

## *i* Note

A periodic domain is also referred to as a domain with no boundaries.

# A periodic mesh in time

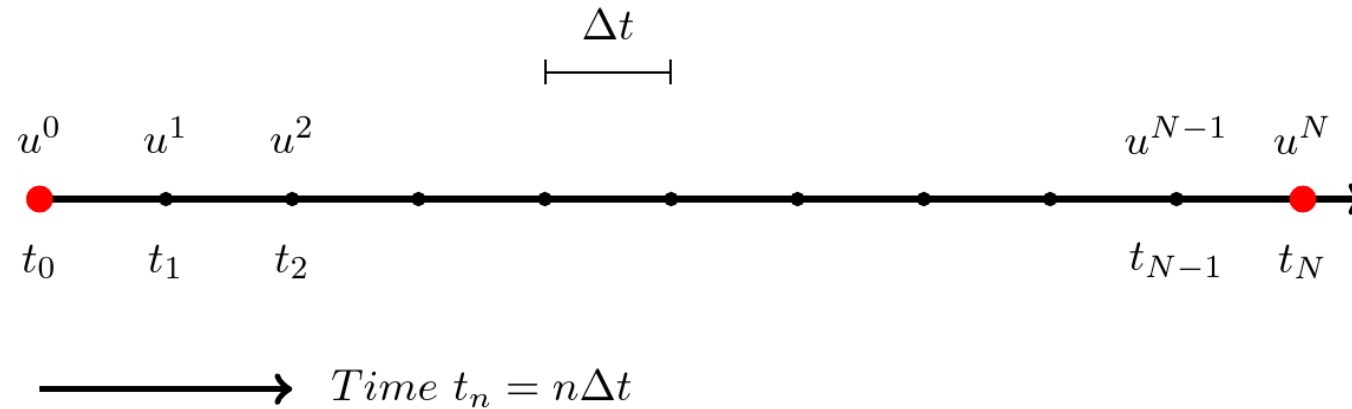


$$u(t_N) = u(0) \quad \text{or} \quad u^N = u^0$$

## Note

There are only  $N$  unknowns  $u^0, u^1, \dots, u^{N-1}$  for a mesh with  $N + 1$  nodes.

# Consider the discretization of $u''$



At the left hand side of the domain, the point to the left of  $u^0$  is  $u^{N-1}$

$$u''(0) \approx \frac{u^1 - 2u^0 + \textcolor{red}{u}^{-1}}{h^2} = \frac{u^1 - 2u^0 + \textcolor{red}{u}^{N-1}}{h^2}$$

At the right hand side of the domain the point to the right of  $u^{N-1}$  is  $u^N = u^0$

$$u''(t_{N-1}) \approx \frac{\textcolor{red}{u}^N - 2u^{N-1} + u^{N-2}}{h^2} = \frac{\textcolor{red}{u}^0 - 2u^{N-1} + u^{N-2}}{h^2}$$

Periodic boundary conditions can be implemented in the matrix  $D^{(2)} \in \mathbb{R}^{N+1 \times N+1}$

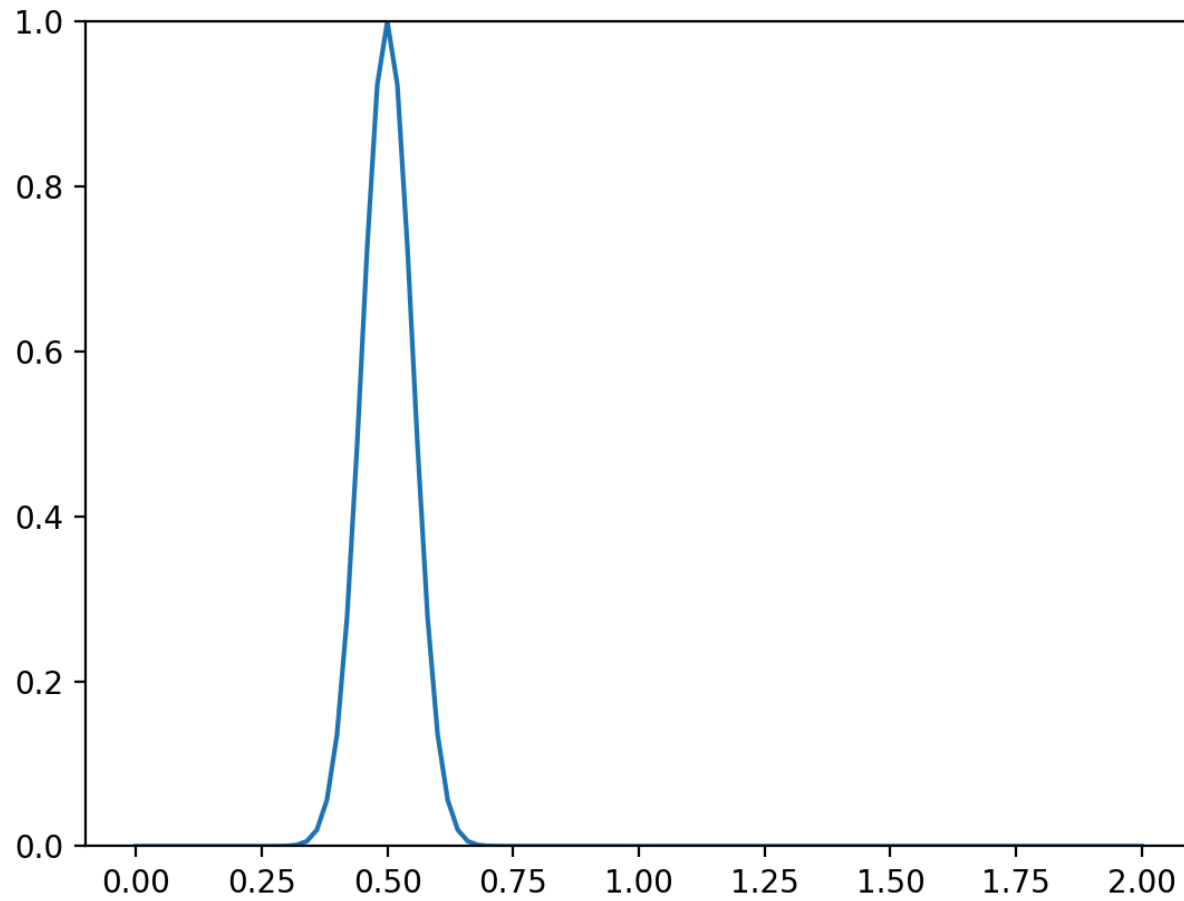
$$\tilde{D}^{(2)} = \begin{bmatrix} -2 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\ 1 & -2 & 1 & 0 & 0 & 0 & 0 & \dots \\ 0 & 1 & -2 & 1 & 0 & 0 & 0 & \dots \\ \vdots & & & \ddots & & & & \dots \\ \vdots & 0 & 0 & 0 & 1 & -2 & 1 & 0 \\ \vdots & 0 & 0 & 0 & 0 & 1 & -2 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & -2 \end{bmatrix}$$

Note that the matrix expects  $\mathbf{u} = (u^0, u^1, \dots, u^{N-1}, u^N)$ , even though  $u^0 = u^N$ . The last row in  $\tilde{D}^{(2)}$  is thus irrelevant, because we will set  $u^0 = u^N$  manually.

# Implementation periodic boundaries

1. Initialize  $u^0$  and compute  $u^1 = u^0 + \frac{\underline{c}^2}{2} \tilde{D}^{(2)} u^0$ . Set  $u^{nm1} = u^0, u^n = u^1$
2. for  $n$  in range( $1, N_t - 1$ ):
  - $u^{np1} = 2u^n - u^{nm1} + \underline{c}^2 \tilde{D}^{(2)} u^n$
  - $u_N^{np1} = u_0^{np1}$
  - Update to next iteration:  $u^{nm1} = u^n; u^n = u^{np1}$

# Periodic wave



# Properties of the wave equation

$$\frac{\partial^2 u}{\partial t^2} = c^2 \frac{\partial^2 u}{\partial x^2}$$

If the initial condition is  $u(x, 0) = I(x)$  and  $\frac{\partial u}{\partial t}(x, 0) = 0$ , then the solution at  $t > 0$  is

$$u(x, t) = \frac{1}{2} (I(x - ct) + I(x + ct))$$

These are two waves - one traveling to the left and the other traveling to the right

If the initial condition  $I(x) = e^{ikx}$ , then

$$u(x, t) = \frac{1}{2} \left( e^{ik(x-ct)} + e^{ik(x+ct)} \right)$$

is a solution



# Representation of waves as complex exponentials

If the initial condition is a sum of waves (superposition, each wave is a solution of the wave equation)

$$I(x) = \sum_{k=0}^K a_k e^{ikx} = \sum_{k=0}^K a_k (\cos kx + i \sin kx)$$

for some  $K$ , then the solution is

$$u(x, t) = \frac{1}{2} \sum_{k=0}^K a_k \left( e^{ik(x-ct)} + e^{ik(x+ct)} \right)$$

We will analyze one component  $e^{ik(x+ct)} = e^{ikx + \omega t}$ , where  $\omega = kc$  is the frequency in time. This is very similar to the investigation we did for the numerical frequency for the vibration equation.

# Assume that the numerical solution is a complex wave

$$u(x_j, t_n) = u_j^n = e^{ik(x_j + \tilde{\omega}t_n)}$$

- How accurate is  $\tilde{\omega}$  compared to the exact  $\omega = kc$ ?
- What can be concluded about stability?

Note that the solution is a recurrence relation

$$u_j^n = e^{ikx_j} e^{i\tilde{\omega}n\Delta t} = (e^{i\tilde{\omega}\Delta t})^n e^{ikx_j}$$

with an amplification factor  $A = e^{i\tilde{\omega}\Delta t}$  such that

$$u_j^n = A^n e^{ikx_j}$$

# Numerical dispersion relation

We can find  $\tilde{\omega}$  by inserting for  $u_j^n = e^{ik(x_j + \tilde{\omega}t_n)}$  in the discretized wave equation

$$\frac{u_j^{n+1} - 2u_j^n + u_j^{n-1}}{\Delta t^2} = c^2 \frac{u_{j+1}^n - 2u_j^n + u_{j-1}^n}{\Delta x^2}$$

This is a lot of work, just like it was for the vibration equation. In the end we should get

$$\tilde{\omega} = \frac{2}{\Delta t} \sin^{-1} \left( C \sin \left( \frac{k\Delta x}{2} \right) \right)$$

where the CFL number is  $C = \frac{c\Delta t}{\Delta x}$

- $\tilde{\omega}(k, c, \Delta x, \Delta t)$  is the numerical dispersion relation
- $\omega = kc$  is the exact dispersion relation
- We can compare the two to investigate numerical accuracy and stability

# Stability

A simpler approach is to insert for  $u_j^n = A^n e^{ikx_j}$  directly in

$$\frac{u_j^{n+1} - 2u_j^n + u_j^{n-1}}{\Delta t^2} = c^2 \frac{u_{j+1}^n - 2u_j^n + u_{j-1}^n}{\Delta x^2}$$

and solve for  $A$ . We get

$$\frac{(A^{n+1} - 2A^n + A^{n-1})e^{ikx_j}}{\Delta t^2} = c^2 A^n \frac{e^{ik(x_j+\Delta x)} - 2e^{ikx_j} + e^{ik(x_j-\Delta x)}}{\Delta x^2}$$

Divide by  $A^n e^{ikx_j}$ , multiply by  $\Delta t^2$  and use  $C = c\Delta t/\Delta x$  to get

$$A - 2 + A^{-1} = C^2(e^{ik\Delta x} - 2 + e^{-ik\Delta x})$$

continue on next slide

# Stability

$$A + A^{-1} = 2 + C^2(e^{ik\Delta x} - 2 + e^{-ik\Delta x})$$

Use  $e^{ix} + e^{-ix} = 2 \cos x$  to obtain

$$A + A^{-1} = 2 + 2C^2(\cos k\Delta x - 1)$$

This is a quadratic equation to solve for A. Using  $\beta = 2(1 + C^2(\cos(k\Delta x) - 1))$  we get that

$$A = \frac{\beta \pm \sqrt{\beta^2 - 4}}{2}$$

We see that  $|A| = 1$  for any real numbers  $-2 \leq \beta \leq 2$ .

**i** For all real numbers  $-2 \leq \beta \leq 2$

$$|\beta \pm \sqrt{\beta^2 - 4}| = 2$$

$$\text{since } |\beta \pm \sqrt{\beta^2 - 4}| = |\beta + i\sqrt{4 - \beta^2}| = \sqrt{\beta^2 + 4 - \beta^2} = 2$$

**For  $|A| \leq 1$  and stability we need  $-2 \leq \beta \leq 2$  and thus**

$$-2 \leq 2(1 + C^2(\cos(k\Delta x) - 1)) \leq 2$$

Rearrange to get that

$$-2 \leq C^2(\cos(k\Delta x) - 1) \leq 0$$

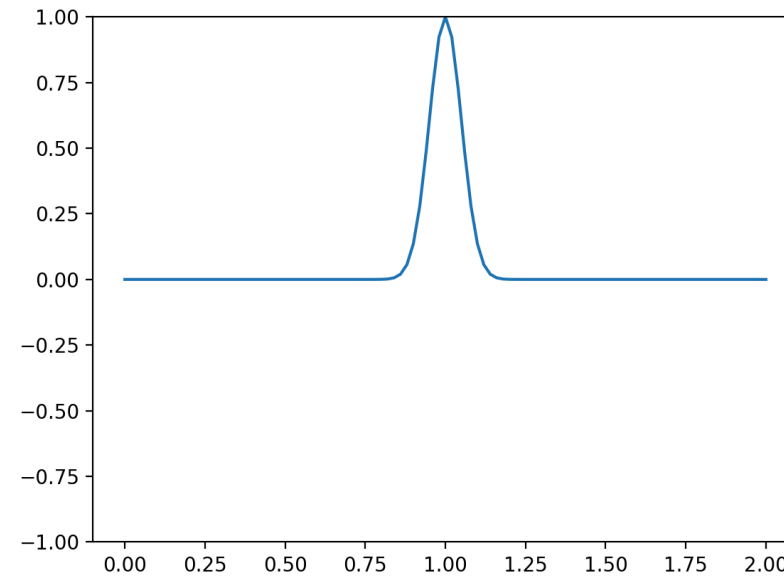
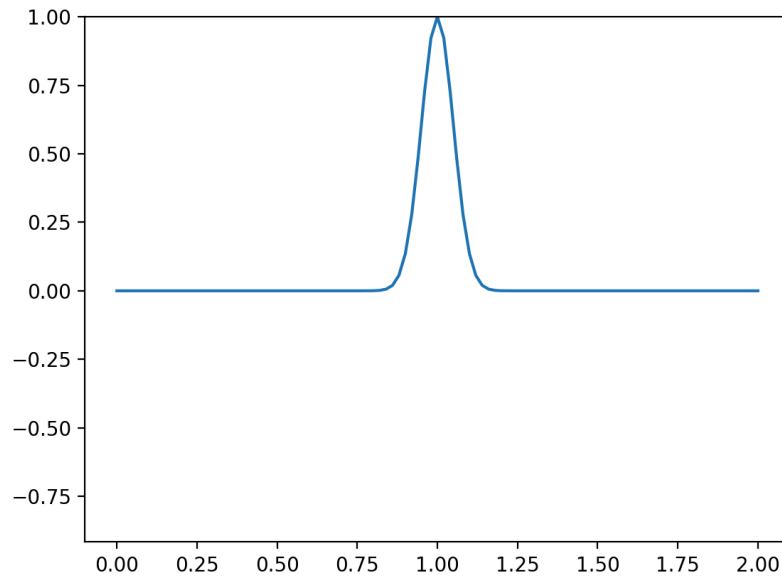
Since  $\cos(k\Delta x)$  can at worst be  $-1$  we get that the positive real CFL number must be smaller than 1

$$C \leq 1$$

Hence (since  $C = c\Delta t/\Delta x$ ) for stability we require that

$$\Delta t \leq \frac{\Delta x}{c}$$

# Test Dirichlet solver using CFL=1.01 vs CFL=1.0



```
1 unm1[:] = sp.lambdify(x, u0.subs(t, 0))(xj)
2 un[:] = sp.lambdify(x, u0.subs(t, dt))(xj)
3 plotdata = {0: unm1.copy()}
4 CFL = 1.01
5 for n in range(Nt):
6     unp1[:] = 2*un - unm1 + CFL**2 * D2 @ un
7     unp1[0] = 0
8     unp1[-1] = 0
9     unm1[:], un[:] = un, unp1
10    if n % 10 == 0:
11        plotdata[n] = unp1.copy()
```