

# HUST

**ĐẠI HỌC BÁCH KHOA HÀ NỘI**  
HANOI UNIVERSITY OF SCIENCE AND TECHNOLOGY

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# 2D Wave Equations with CUDA Parallel Model

Course: Parallel & Distributed Programming

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# Problem Specification

# 1. Introduction - Problem

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- **Problem:** Simulate real-time propagation of waves in a 2D environment by updating wave displacement values at each grid point over successive time steps. The output should support smooth visualization suitable for real-time animation.
- **Challenges:** High computational cost for large grids requires efficient parallelization strategies to achieve the necessary runtime performance.



# Mathematical Formula

## 2. Mathematical Formula

### 1 - Dimension Equation

Let  $u(x, t)$  be the displacement at time  $t$  of a point initially at  $x$ .

The displacement function  $u$  is governed by the mathematical model

$$\frac{\partial^2 u}{\partial t^2} = c^2 \frac{\partial^2 u}{\partial x^2}, \quad x \in (0, L), \quad t \in (0, T] \quad (1)$$

$$u(x, 0) = I(x), \quad x \in [0, L] \quad (2)$$

$$\frac{\partial}{\partial t} u(x, 0) = 0, \quad x \in [0, L] \quad (3)$$

$$u(0, t) = 0, \quad u(L, t) = 0, \quad t \in (0, T] \quad (4)$$

$I(x)$  is the initial shape of the string.

The constant  $c$  and the function  $I(x)$  must be prescribed.

- This partial differential equation (PDE) contains a second-order derivative in time, two **initial conditions** (2), (3) are required to determine a **unique** solution.
- PDE need **boundary conditions** (4)

The solution  $u(x, t)$  varies in space and time and describes waves that move with velocity  $c$ .



## 2. Mathematical Formula

### Discretization

$$\frac{\partial^2 u}{\partial t^2} = c^2 \frac{\partial^2 u}{\partial x^2}, \quad x \in (0, L), \quad t \in (0, T]$$

We introduce the **uniformly distributed constant mesh points**, with spacings  $\Delta t$  and  $\Delta x$ . We have that:

$$x_i = i\Delta x, \quad i = 0, \dots, N_x, \quad t_n = n\Delta t, \quad n = 0, \dots, N_t.$$

The temporal domain  $[0, T]$  and spatial domain  $[0, L]$  is represented by a finite number of mesh points:

$$0 = t_0 < t_1 < t_2 < \dots < t_{N_t-1} < t_{N_t} = T.$$

$$0 = x_0 < x_1 < x_2 < \dots < x_{N_x-1} < x_{N_x} = L.$$



## 2. Mathematical Formula

### Discretization

$$\frac{\partial^2 u}{\partial t^2} = c^2 \frac{\partial^2 u}{\partial x^2}, \quad x \in (0, L), \quad t \in (0, T]$$

We relax the strict condition that equation holds everywhere in the space - time domain  $(0, L) \times (0, T]$  to the requirement that the PDE is fulfilled at the interior mesh points only:

$$\frac{\partial^2}{\partial t^2} u(x_i, t_n) = c^2 \frac{\partial^2}{\partial x^2} u(x_i, t_n),$$

For  $n = 0$  we have the **initial conditions**  $u = I(x)$ ,  $u_t = 0$

For  $i = 0$  and  $i = N_x$  we have the **boundary conditions**  $u = 0$

## 2. Mathematical Formula

### Replacing derivatives by finite differences

$$\frac{\partial^2}{\partial t^2} u(x_i, t_n) = c^2 \frac{\partial^2}{\partial x^2} u(x_i, t_n),$$

We approximate the derivatives using finite differences.

Central differences formula:  $f''(x) = \frac{f(x+h) - 2f(x) + f(x-h)}{h^2}$

Applying this approximation to both the time and space derivatives:

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

## 2. Mathematical Formula

### Formulating a recursive algorithm

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

- We introduced parameter:  $C = c \frac{\Delta t}{\Delta x}$   
which also known as the *Courant number*.
- By rearranging the terms, we obtain the recursive update formula:

$$u_i^{n+1} = -u_i^{n-1} + 2u_i^n + C^2 (u_{i+1}^n - 2u_i^n + u_{i-1}^n)$$

## 2. Mathematical Formula

### Formulating a recursive algorithm

$$u_i^{n+1} = -u_i^{n-1} + 2u_i^n + C^2 (u_{i+1}^n - 2u_i^n + u_{i-1}^n)$$

A problem with arises when  $n = 0$  since the formula for  $u_i^1$  involves  $u_i^{-1}$

But we have  $0 = \frac{\partial}{\partial t} u(x_i, t_n) \approx \frac{u_i^1 - u_i^{-1}}{2\Delta t}$  *(Initial condition)*

$$\Rightarrow u_i^{n-1} = u_i^{n+1}, \quad i = 0, \dots, N_x, \quad n = 0$$

$$\Rightarrow u_i^1 = u_i^0 - \frac{1}{2}C^2 (u_{i+1}^0 - 2u_i^0 + u_{i-1}^0)$$



## 2. Mathematical Formula

### Applying in 2-Dimension Equation

$$\frac{\partial^2 u}{\partial t^2} = c^2 \left( \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} \right)$$

Turn into this recursive function for implementation:

$$u^{n+1} = -u_{i,j}^{n-1} + 2u_{i,j}^n + C_x^2(u_{i+1,j}^n - 2u_{i,j}^n + u_{i-1,j}^n) + C_y^2(u_{i,j+1}^n - 2u_{i,j}^n + u_{i,j-1}^n)$$



# Numeric Algorithms

# 3. Numeric Algorithms

## Initialize Variables:

- Initialize the grid for the wave displacement and velocity:  $u(x, y, t)$  and  $v(x, y, t)$  where  $u$  is the displacement and  $v$  is the velocity

## Initial Conditions:

- Set the initial wave field  $u(x, y, t=0)$  (often zero or a specific initial condition).
- Set the initial velocity field  $v(x, y, t=0)$  (often zero).

## Update Rule (Finite Difference):

- Use the central difference scheme for the spatial and time derivatives to update the wave field:

$$u^{n+1} = -u_{i,j}^{n-1} + 2u_{i,j}^n + C_x^2(u_{i+1,j}^n - 2u_{i,j}^n + u_{i-1,j}^n) + C_y^2(u_{i,j+1}^n - 2u_{i,j}^n + u_{i,j-1}^n)$$



# 3. Numeric Algorithms

## Boundary Conditions:

- Apply fixed or periodic boundary conditions depending on the problem setup. Typically, at the edges of the grid, the displacement  $u$  is set to 0 or some predefined value.

## Parallelization (CUDA):

- Use CUDA kernels to parallelize the spatial updates, ensuring the 2D grid is processed efficiently on the GPU.



# Parallel Algorithms

# 4. Parallel Algorithms

We solve the 2D wave equation using the **Finite Difference Method (FDM)**

We employ a parallel algorithm to accelerate the computation using GPU-based CUDA programming



# 4. Parallel Algorithms

## Finite Difference Method (FDM)

- Discretize space and time
- Computes the wave function  $u(x,y,t)$  at each time step

### Formula

$$u_{i,j}^{n+1} = 2u_{i,j}^n - u_{i,j}^{n-1} + r^2(u_{i+1,j}^n + u_{i-1,j}^n + u_{i,j+1}^n + u_{i,j-1}^n - 4u_{i,j}^n)$$

where  $r = \frac{c\Delta t}{\Delta x}$  is Courant number

Each grid point can be updated independently → Ideal for parallelization

# 4. Parallel Algorithms

## Parallelization Strategy

**1. Discretization:** Both the space and time domains are discretized, and each grid point is mapped to a thread.

**2. Initialization:** The wave field is initialized using the given initial conditions

**3. Wave Update:** The wave field is updated for each time step using FDM, depending on the method selected

**4. Boundary Conditions:** Boundary conditions are applied at the edges of the grid.

**5. Parallel Execution:** The wave updates at each grid point are computed in parallel using CUDA kernels, ensuring efficient use of GPU resources

# Experiment & Result

# 4. Experiment

## Device

- **CPU:** Intel(R) Xeon(R) CPU @ 2.20GHz.
- **GPU:** NVIDIA Tesla P100 GPUs.
- **RAM:** 29GB.
- **HDD:** 57.6GB

## CUDA Settings:

```
dim3 blockSize(32, 32);
```

```
dim3 gridSize((NX + blockSize.x - 1) / blockSize.x, (NY + blockSize.y - 1) / blockSize.y);
```



# 4. Experiment

## Evaluation Method

Two methods were compared:

- **CPU-based:** Serial C implementation
- **GPU-based:** CUDA implementation

We evaluate the stimulation under different grid sizes and varying numbers of time steps.

When compared, **faster** simulations mean **better** performance

# 4. Experiment

## Execution Time Comparison Result in seconds

NX	NY	Total step	CPU	CUDA-FDM
1000	1000	1000	11.2886	1.3706
1000	1000	5000	51.0509	1.6098
1000	1000	10000	101.3409	2.8262
3000	3000	10000	907.2015	<b>21.4762</b>



# CONCLUSION

# 5. Conclusion

## Advantage

- Achieved  $\sim 10 - 30x$  speedup over the CPU (serial C) version.
- Demonstrated strong performance gains via parallelism.
- Suitable for large-scale simulations.

## Limitation

- Initialization bug in  $u_{\text{prev}}$  can affect early simulation steps.
- Courant condition imposes strict stability limits (smaller time steps).
- Simple boundary conditions reduce flexibility.
- I/O bottlenecks due to frequent large array transfers between GPU and CPU



# 5. Conclusion

## Future Work

- Fix initialization of  $u_{\text{prev}}$ .
- Use shared memory for better performance.
- Extend to variable wave speed, damping, and complex boundaries.
- Optimize data output and memory transfer strategies.





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