

# Numerical Solution to the 2D Wave Equation on a Complex Domain

MACM 416 - Final Project Presentation

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# The Wave Equation

## Problem Statement

The Wave Equation:

$$u_{tt} = c^2 \Delta u, \quad \vec{x} \in \Omega$$

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The Wave Equation:

$$u_{tt} = c^2 \Delta u, \vec{x} \in \Omega$$

In 2D:

$$u_{tt} = c^2 \left( \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} \right), (x, y) \in \Omega$$

# The Wave Equation

## Properties

The constant  $c$  represents the “wave speed”.



**Figure:** Water droplet and propagating waves

# The Wave Equation

## Properties

Boundary conditions determine wave interactions at the boundary  $\partial\Omega$

$$u \mid_{\partial\Omega} = 0$$

Homogeneous Dirichlet conditions, wave inverts at the boundary.

$$\nabla u \cdot \hat{n} \mid_{\partial\Omega} = 0$$

Homogeneous Neumann conditions, wave reflects without inverting.

# The Wave Equation

## The Hard Problem

$$u_{tt} = \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2}, \quad (x, y) \in \Omega$$

Boundary conditions

$$\nabla u \cdot \hat{n} \big|_{\partial\Omega} = 0$$

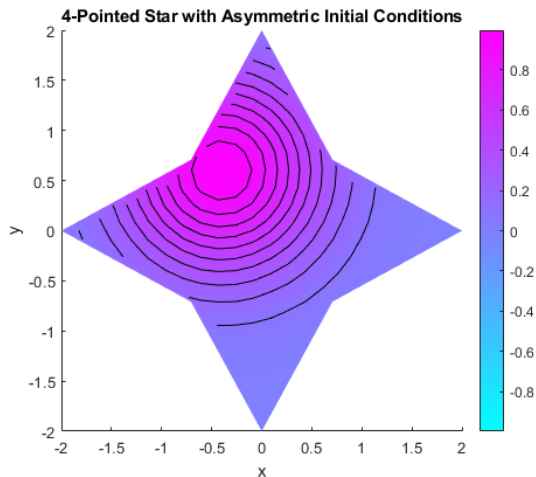
Initial conditions

$$u(x, y, 0) = \exp \left( -(x - a)^2 - (y - b)^2 \right)$$

$$u_t(x, y, 0) = 0$$

# The Wave Equation

Domain  $\Omega$



**Figure:** Asymmetric initial conditions for the wave equation on a 4-point star

# The Test Problem

## Forced Wave in 1-D

$$u_{tt} = u_{xx} + \sin(t)$$

Boundary conditions

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

Initial conditions

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

Exact solution using Fourier series

$$u(x, t) = \sum_{n=1}^{\infty} 2 \left( \frac{1 - (-1)^n}{\pi n (\pi^2 n^2 - 1)} \right) \left( \sin(t) - \frac{1}{\pi n} \sin(n\pi t) \right) \sin(n\pi x)$$



# Numerical Approximation Methods

## Time Discretization

$$\frac{\partial^2 u}{\partial t^2} \Big|_{t=t_n} = \frac{u(t_{n+1}) - 2u(t_n) + u(t_{n-1}))}{\Delta t^2} + O(\Delta t^2).$$

- ▶ Explicit time stepping
- ▶ Low computational cost allowing for larger time ranges
- ▶ Order  $\Delta t^2$  accuracy
- ▶ We take  $\Delta t = 10^{-3}$  for optimal range of stable resolutions

# Numerical Approximation Methods

## Spatial Discretization

### Linear finite elements

Find  $U^{n+1}$  such that for all  $v \in \mathcal{V}$

$$\int_{\Omega} \frac{1}{\Delta t^2} v U^{n+1} dx + \int_{\Omega} -\frac{2}{\Delta t^2} v U^n dx + \frac{1}{\Delta t^2} v U^{n-1} dx - \nabla v \cdot \nabla U^n dx - v f dx = 0.$$

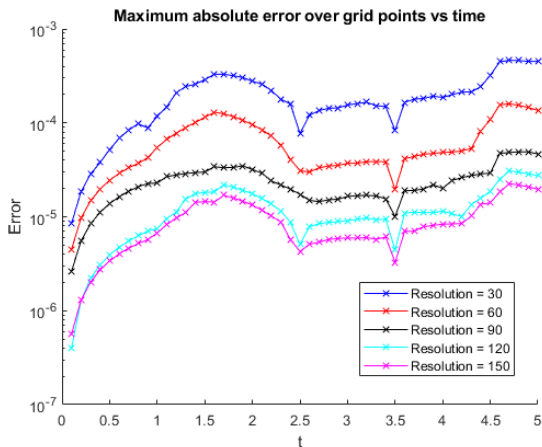
Where  $U^{n+1} \approx u(x, y, t_{n+1})$

- ▶ Adaptable to arbitrary domains
- ▶ Approximately  $O(h^2)$  where  $h$  is the maximum diameter of the triangular mesh
- ▶ Total consistency  $O(\Delta t^2 + h^2)$

# Test Problem Analysis

Approximations compared to truncated Fourier series

$$u(x, t) = \sum_{n=1}^{10^6} 2 \left( \frac{1 - (-1)^n}{\pi n (\pi^2 n^2 - 1)} \right) \left( \sin(t) - \frac{1}{\pi n} \sin(n\pi t) \right) \sin(n\pi x)$$



# Hard Problem Analysis

How do we test convergence without an exact solution?

- ▶ Approximations observe an “energy” principle
- ▶ Change in approximations between resolution values decreases as resolution increases

# Hard Problem Analysis

## Energy Principle

The unforced wave equation

$$u_{tt} = c^2 \Delta u$$

Energy is constant

$$E = \int_{\Omega} u_{tt}^2 + c^2 |\nabla u|^2 dx$$

# Hard Problem Analysis

## Energy Principle

We measure the “total energy variation” as our benchmark.

**Table:** Difference of minimum and maximum energy for the approximation with symmetric initial condition

	Total Energy Variation (Symmetric Initial Conditions)
Resolution 30	1.0000e-05
Resolution 60	3.0000e-05
Resolution 90	4.0000e-05

# Hard Problem Analysis

## Convergence

We measure “Maximum Difference” and “Average Difference” away from the boundary

**Table:** Maximum difference and average difference between approximations with asymmetric initial conditions at varying resolutions

Resolution	Maximum Difference	Average Difference
$R_1 = 15, R_2 = 30$	0.0306	0.0025
$R_1 = 30, R_2 = 60$	0.0197	0.0012
$R_1 = 60, R_2 = 90$	0.0088	4.6632e-04
$R_1 = 90, R_2 = 120$	0.0051	2.6388e-04

# Hard Problem Analysis

## Efficiency

These approximations can take significant computation times

**Table:** Computation time taken for asymmetric initial conditions for  $t = [0, 1]$

Resolution	Compute Time (seconds)
15	61.84
30	56.94
60	156.22
90	393.22
120	702.47
150	1148.51



# Neural Net

## An Overview of the Problem

Initial condition:

$$u(x, y, 0) = \phi(x, y) = e^{-x^2 - y^2}$$

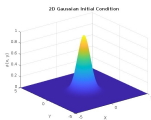


Figure: Gauss Initial condition

Boundary condition:

$$\frac{\partial u}{\partial n} = 0$$

Analytical solution

$$u(x, y, t) = \exp[-(\sqrt{x^2 + y^2} - t)^2]$$

# Neural Net

First attempt: writing my own Loss function

- ▶ Have to use `dlArray` (deep learning array) class
- ▶ Must define its own Loss function.
- ▶ Must define its own Network Architecture.
- ▶ Must manage Numerical Error (rounding) carefully

# Neural Net

Second Attempt: using Levenberg-Marquardt algorithm

Produces somewhat decent answer but the training time is painfully long with just 10 nodes and 3 hidden layers, and 50 training iteration.

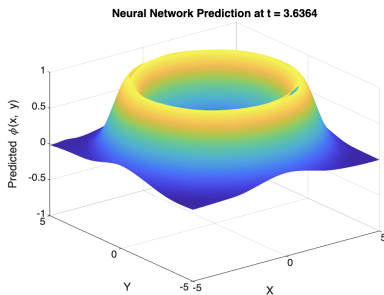


Figure: Prediction

# Frame Title

## Error

The error is super high to be reliable, it is just there to give intuition how the solution looks like.

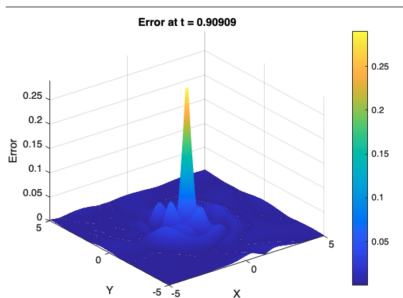


Figure: Error