# 1D Heat Equation

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ction Assumptions Numerical Scheme Algorithm

#### Overview

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### Section 1

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# Objective

• Simulate the evolution of the Temperature profile of the 1-D Metal rod shown below, to find the temperature u(x,t) as a function of location and time.



- The rod is said to be insulated at both ends and no internal heat sources or sinks are present.
- **3** An initial temperature profile of f(x) is assumed.

# 1-D Heat Equation

• Consider an arbitrary segment of the rod, whose width is  $\Delta x$  and which is located at x.

$$\dot{e} = \dot{q}_{left} - \dot{q}_{right} \tag{1}$$

$$\dot{q} = -K_0 \frac{\partial u}{\partial x} \tag{2}$$

Fourier's law described in Equation 2 can be applied to find the fluxes on the left and right sides of the segment.

$$c\rho A\Delta x \left( u(x + \Delta x, t) - u(x, t) \right) = -K_0 \Delta t A \left( \left( \frac{\partial u}{\partial x} \right)_x - \left( \frac{\partial u}{\partial x} \right)_{x + \Delta x} \right)$$
 (3)

Simplifying Equation 3 leads to the 1D Heat equation which is a parabolic PDE.

$$\frac{\partial u}{\partial t} = \alpha \frac{\partial^2 u}{\partial x^2} \tag{4}$$

- f 2 This equation is also known as the second order diffusion equation, the co-efficient lpha is known as the Thermal diffusivity and is a material property.
- To make use of the Heat equation, Boundary conditions and Initial condition are required.
- **1** The initial condition in this case is the initial temperature profile u(x,0).
- The solution is effected by the boundary conditions which can be determined from the condition that the rod is insulated at both ends.

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# **Initial and Boundary conditions**

At the boundaries, the rod is insulated at both ends i.e the heat flux at both ends is zero.

$$\left(\frac{\partial u}{\partial x}\right)_{x=0} = \left(\frac{\partial u}{\partial x}\right)_{x=L} = 0 \tag{5}$$

This kind of boundary condition is known as the Neumann Boundary condition as opposed to holding the ends at a certain constant temperature which is known as the Dirichlet boundary condition.

$$u(x,0) = 0.5(\sin(x) + \cos(x)) \tag{6}$$

The Initial condition represented in Equation 6 is a sinosuidal condition.

#### Material

**①** The thermal diffusivity  $\alpha$  is a function of material properties.

$$\alpha = \frac{K_0}{\rho c} \tag{7}$$

- Aluminum was chosen as the material of the rod, arbitrarily.
- For this solution, iti is assumed that all the material properties are constant and do not vary with temperature.
- **1** The specific heat(c) of aluminum at STP is 0.91  $\frac{kJ}{kg-K}$ .
- **1** The density( $\rho$ ) of aluminum at STP is 2710  $\frac{kg}{m^3}$
- **1** The Thermal conductivity( $K_0$ ) of Aluminum at STP is 205  $\frac{W}{m-K}$

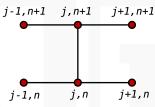
#### Model

- The Finite Difference Method has been chosen to solve the problem as it is computationally least expensive and is suitable to solve parabolic PDEs with relatively simple boundary conditions.
- The rod is divided into a grid of points at which function evaluations would be made.
- ullet The governing equation has been non-dimensionalized and the characteristic length L is assumed to be of unit value.

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This scheme is semi-implicit in nature, as it involves matrix inversion while the previous timestep information is also considered as part of the solution.



② The scheme is second order accurate in space and time i.e  $\mathcal{O}((\Delta x)^2,(\Delta t)^2)$ .

$$\lambda = \frac{\alpha \Delta t}{(\Delta x)^2} \tag{8}$$

3 It is unconditionally stable i.e for any value of the CFL number( $\lambda$ ).

The CN scheme is the most prominent of the Weighted Average schemes.

$$\Theta = 0.5$$

$$u_i^{n+1} = u_i^n + \lambda \left[ (1 - \Theta) \left( u_{i+1}^{n+1} - 2u_i^{n+1} + u_{i-1}^{n+1} \right) + (\Theta) \left( u_{i+1}^n - 2u_i^n + u_{i-1}^n \right) \right]$$

$$u_i^{n+1} = u_i^n + \frac{\lambda}{2} \left[ \left( u_{i+1}^{n+1} - 2u_i^{n+1} + u_{i-1}^{n+1} \right) + \left( u_{i+1}^n - 2u_i^n + u_{i-1}^n \right) \right]$$
(9)

Equation 9 can be represented in a matrix form as follows.

$$\mathbf{A}u^{n+1} = \mathbf{B}u^n + \mathbf{b} \tag{10}$$

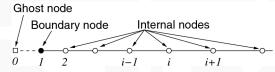
$$\boldsymbol{b} = \begin{bmatrix} 2\lambda \Delta x \dot{q}_0 & \dots & \dots & \dots & 2\lambda \Delta x \dot{q}_L \end{bmatrix}$$

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- The scheme takes information from the present time step as well as the previous time step, making use of the maximum amount of information available.
- However, the scheme does produce oscillations when the gradients are very high in the initial condition or during the first few time steps when this happens.
- **3** This is numerical dispersion is usually observed when the CFL Number  $(\lambda)$  is significantly higher than 0.5 and is caused by the presence of odd order terms in the modified PDE.

# **Boundary Condition Implementation**

To be able to apply the Neumann Boundary conditions described by Equation 5, ghost points are assumed to be present at either ends of the stencil (grid).



② Using the ghost points, the spatial gradients at the boundaries are computed using a first order central difference approach.

$$\frac{\partial u_0}{\partial x} = \frac{u_1 - u_{-1}}{\Delta x} = 0$$

$$u_{-1} = u_1$$

$$\frac{\partial u_N}{\partial x} = \frac{u_{N+1} - u_N - 1}{\Delta x} = 0$$

$$u_{N+1} = u_{N-1}$$

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# **Boundary Condition Implementation**

**①** The relations between  $u_{-1}$ .  $u_1$ ,  $u_{N+1}$  and  $u_{N-1}$  are then used to homogenously implement the CN scheme at boundaries.

# Rannacher Smoothing

- To avoid the dispersion caused by crank nicholson scheme, One strategy is to start the first two solutions with an implicit euler formulation and a timestep that is half the original value.
- Implicit Euler scheme is also unconditionally stable and performs well even when gradients are high.
- **3** Implicit Euler scheme is first order in space and time i.e  $\mathcal{O}((\Delta x), (\Delta t))$ .
- This implementation ensures that the high frequencies at the beginning of the solutio are damped out and then crank nicholson method which is much more accurate can take over.

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### Performance

- The entire project has been written in Python, with the numpy LinAlg solver to invert matrices.
- ② The **np.linalg.solve()** function is of  $\mathcal{O}(n^3)$  which is the highest time complexity you will find in the algorithm.
- On average, a solution with 2000 grid points over 500 time steps takes 28 seconds.
- This implies that each matrix inversion is taking about 0.06 seconds.

#### Structure

There are 6 python files that make up this project.

```
doc plots refs results src temp videos
case.py control.py home.py post.py __pycache__ scheme.py solver.py
```

Figure: The structure of the project and the source directory

- 2 The solutions are solved in the **results** folder
- The figures produced to create the video are saved in temp
- The plots are docs are stored in their respective namesake folders.
- The project is hosted on github and can be found here.

# Setup

- control.py contains all the control parameters such as the grid size, number of time steps and all the variables can be directly altered from here.
- case.py contains material property information as well as the initial condition assumed in the problem, by editing this file, the entire solution can be changed.
- scheme.py contains the pre calculated co-efficient matrices A,B and the vector b for both the Crank Nicholson scheme and the Implicit Euler scheme as both are used in the solution.

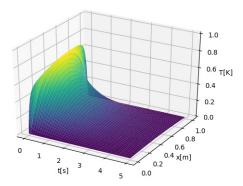
#### function

- The user executes **home.py** with arguments "-s" and "-u0", these confirm whether a new solution must be generated and also which initial solution to apply.
- Here, based on the argument, if a new solution is to be generated, then the relevant parameters are pulled from control.py, case.py and scheme.py
- The parameters are then input into a solve() function written in solver.py that saves csv results in results folder and returns a 2D array.
- The solve() function implements Rannacher time stepping with co-efficients of CN scheme and implicit euler scheme that are pulled from scheme.py.
- In home, this 2D array is passed on to the post processing functions present in post.py.

### Section 5

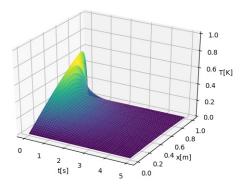
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### Time Evolution of Solutions



**Figure:** Time evolution for  $u(x,0)=0.5(\sin(x)+\cos(x))$ 

#### Time Evolution of Solutions



**Figure:** Time evolution for  $u(x,0)=0.8(\sin(x))$ 

#### Time Evolution of Solutions

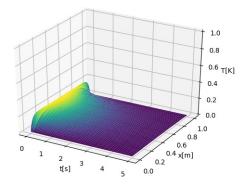
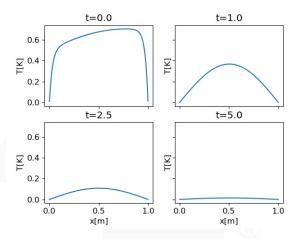


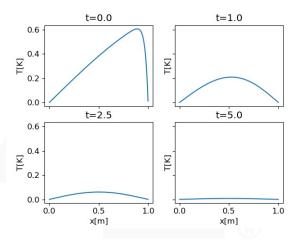
Figure: Time evolution for u(x,0)=0.2 i.e constant

# **Snapshots of Solutions**



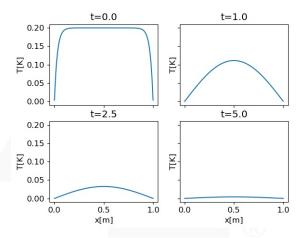
**Figure:** Snapshots of u(x,t) for  $u(x,0)=0.5(\sin(x)+\cos(x))$ 

# **Snapshots of Solutions**



**Figure:** Snapshots of u(x,t) for  $u(x,0)=0.8(\sin(x))$ 

# **Snapshots of Solutions**



**Figure:** Snapshots of u(x,t) for u(x,0)=0.2 i.e constant

# Thank You!