Homework 2

19/20

Deadline: Tuesday 20 May 2025 (by 19h00) Good work!

Credits: 20 points

Instructions:

- The homework is **individual**. Please include your name in the notebook.
- Please send your compressed tar file with the following tree scheme:

Name:

(a)

11.75/12 1. Thermodynamics and Heat Conduction (12 points)

Please include your solutions to this problem within a single python notebook file: **metalconduction.ipynb**

Use appropriate numerical algorithms to study how the temperature profile in a heated metal wire evolves in time, under different initial and boundary conditions. For this, you need to numerically solve the 1D heat equation:

$$\frac{\partial T}{\partial t} = \alpha \frac{\partial^2 T}{\partial x^2}$$

where T=T(x,t) describes the temperature of the metal, x is position, t is time, and α is the thermal diffusivity of the metal. We will consider the following metals:

Metal	$\frac{mm^2}{s}$	α ()	
Copper	111		
Iron	23		

Metal	$rac{mm^2}{s}$	
	<u>s</u>	
Aluminum	97	
Brass	34	
Steel	18	
Zinc	63	
Lead	22	
Titanium	9.8	

Assume that the metal wire has a length of $20\,\mathrm{cm}$, and choose the wire midpoint as the origin for the domain.

Code design:

Organise your code using python classes and carry out the following tasks:

(a) Place the metals and diffusivities in a python dictionary.

(bx2) Create a simulation initialisation routine where e.g. the mesh, the initial conditions, and the boundary conditions are all set up.

Consider two types of initial conditions:

• Smooth, which sets the initial temperature profile in degrees Celsius to be:

$$T(x,0)=175-50\cos\Bigl(rac{\pi x}{5}\Bigr)-x^2$$

where x is in units of cm.

• **Noisy**, which adds some noise f(x) with amplitude β to the initial condition:

$$T(x,0)=175-50\cos\Bigl(rac{\pi x}{5}\Bigr)-x^2+eta\,f(x)\,g(x)$$

Note that you need to find an appropriate apodisation function g(x) so that the initial boundary conditions remain fixed at $25\,\mathrm{C}$. Similarly, a default amplitude for the noise function can be set to be less than a hundredth of the peak temperature value.

Also, consider two types of **boundary conditions**:

- \bullet **Fixed**, which keep the temperatures at the edges of the metal wire fixed at a temperature of $25\,\mathrm{C}.$
- Varying, which assume that the boundaries cannot be kept constant (because of e.g. a faulty cooling system). Instead they also evolve in time according to the

following functions:

$$T(+10 \, \mathrm{cm}, t) = 25 + 0.12 \, t$$
 $T(-10 \, \mathrm{cm}, t) = 25 + 0.27 \, t$

(cx3) Construct a Crank-Nicolson algorithm to simulate the evolution of an initial temperature profile and find the time in seconds at which thermal equilibrium is reached in an input metal wire (from the dictionary). Hint: You need to define some criteria to determine thermal equilibrium.

```
In [8]: import numpy as np
           import matplotlib.pyplot as plt
           from joblib import Parallel, delayed
           import time
           import os
           class MetalConduction:
               Class for simulating heat conduction in metal wires using the Crank-Nico
               def __init__(self, metal_name='Copper', length=20.0, dx=0.1, dt=0.01,
                             ic_type='smooth', bc_type='fixed', noise_amplitude=0.0):
                   Initialize the simulation parameters
                   Parameters:
                   metal name : str
                       Name of the metal for the simulation
                   length : float
                       Length of the metal wire in cm
                   dx : float
                       Spatial step size in cm
                   dt : float
                       Time step size in seconds
                   ic type : str
                        Type of initial condition ('smooth' or 'noisy')
                   bc_type : str
                        Type of boundary condition ('fixed' or 'varying')
                   noise amplitude : float
                        Amplitude of noise for noisy initial condition
(a)
                   # Metal thermal diffusivity dictionary (in mm<sup>2</sup>/s)
                    self.metals = {
                        'Copper': 111,
                        'Iron': 23,
                        'Aluminum': 97,
                        'Brass': 34,
                        'Steel': 18,
                        'Zinc': 63,
                        'Lead': 22,
                        'Titanium': 9.8
```

```
# Convert metal diffusivity from mm<sup>2</sup>/s to cm<sup>2</sup>/s
                   self.metal_name = metal_name
                   self.diffusivity = self.metals[metal_name] / 100.0 # Convert from n
                   # Domain setup
                   self.length = length # Length in cm
                   self.dx = dx # Spatial step in cm
                   self.dt = dt # Time step in seconds
                   # Calculate grid parameters
                   self.nx = int(length / dx) + 1
                   self.x = np.linspace(-length/2, length/2, self.nx) # Domain from -1
                   # Initial and boundary condition types
                   self.ic_type = ic_type
                   self.bc_type = bc_type
                   self.noise_amplitude = noise_amplitude
                   # Calculated r factor for Crank-Nicolson
                   self.r_factor = self.diffusivity * dt / (dx**2)
                   # Initialize temperature matrix
                   self.T = None
                   self.time_array = None
                   # Equilibrium parameters
                   self.equilibrium_threshold = 0.001 # Temperature change threshold
                   self.equilibrium_time = None
                   # Print initialization info
                   print(f"Initialized {metal_name} wire simulation")
                   print(f"Diffusivity: {self.diffusivity:.2f} cm²/s")
                   print(f"r factor: {self.r_factor:.4f}")
(b)
               def initialize simulation(self, simulation time=60):
                   Initialize the simulation with the chosen initial and boundary condi
                   Parameters:
                   simulation_time : float
                       Maximum simulation time in seconds
                   # Create time array
                   self.nt = int(simulation_time / self.dt) + 1
                   self.time_array = np.linspace(0, simulation_time, self.nt)
                   # Initialize temperature matrix T(x,t)
                   self.T = np.zeros((self.nx, self.nt))
                   # Apply initial conditions
                   self._set_initial_conditions()
                   # Apply initial boundary conditions
```

```
self. set boundary conditions()
        # Create matrices for Crank-Nicolson
         self._setup_crank_nicolson_matrices()
        print(f"Simulation initialized for {simulation time} seconds with {s
def _set_initial_conditions(self):
         """Set initial temperature profile based on the IC type"""
        # Base temperature profile: T(x, 0) = 175 - 50\cos(\pi x/5) - x^2
        base_profile = 175 - 50*np.cos(np.pi*self.x/5) - self.x**2
        if self.ic_type == 'smooth':
                 self.T[:, 0] = base_profile
        elif self.ic_type == 'noisy':
                 # Generate noise and apply apodisation
                 noise = self.noise_amplitude * np.random.randn(self.nx)
                 # Apodisation function that goes to zero at boundaries
                 edge_dist = np.minimum(self.x - min(self.x), max(self.x) - self.
                 apodisation = np.sin(np.pi * edge_dist / self.length)
                 self.T[:, 0] = base_profile + noise * apodisation
                 # Ensure boundary conditions are exactly 25°C
                 self.T[0, 0] = 25
                 self.T[-1, 0] = 25
        else:
                 raise ValueError(f"Unknown initial condition type: {self.ic_type
def set boundary conditions(self):
        """Set boundary conditions for all time steps"""
        if self.bc_type == 'fixed':
                 # Fixed temperature of 25°C at both ends
                 self.T[0, :] = 25
                 self.T[-1, :] = 25
        elif self.bc_type == 'varying':
                 # Time-varying boundary conditions
                 \# T(-10cm, t) = 25 + 0.27t
                 \# T(+10cm, t) = 25 + 0.12t
                 self.T[0, :] = 25 + 0.27 * self.time_array
                 self.T[-1, :] = 25 + 0.12 * self.time_array
                 raise ValueError(f"Unknown boundary condition type: {self.bc type
def _setup_crank_nicolson_matrices(self):
        """Setup matrices for the Crank-Nicolson method"""
        n = self.nx
        r = self.r_factor
        # Matrix D1 for the implicit part (left side of equation)
         self.D1_matrix = np.diag([2 + 2*r]*(n-2)) + np.diag([-r]*(n-3), -1)
        # Matrix D2 for the explicit part (right side of equation)
         self.D2_matrix = np.diag([2 - 2*r]*(n-2)) + np.diag([r]*(n-3), -1) +
```

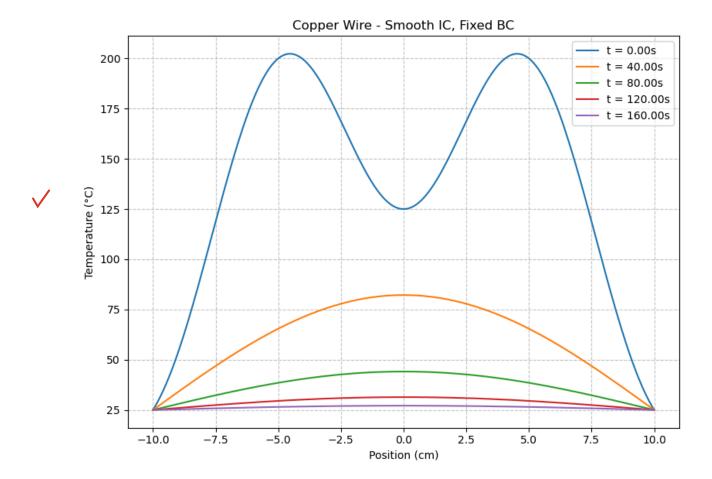
```
def run simulation(self):
                    """Run the Crank-Nicolson simulation"""
                    start time = time.time()
                    # Make sure we have initialized the simulation
                    if self.T is None:
                        raise ValueError("Simulation not initialized. Call initialize si
                    # Record if equilibrium was reached
                    equilibrium_reached = False
                    consecutive_stable_steps = 0
                    required_stable_steps = 5 # Number of consecutive steps below thres
                    # Iterate through time steps
                    for j in range(self.nt-1):
                        # Apply Crank-Nicolson method to update interior points
                        # Create right-hand side vector b from current temperature
                        b = np.dot(self.D2 matrix, self.T[1:-1, j])
                        # Add boundary terms
                        b[0] += self.r_factor * (self.T[0, j+1] + self.T[0, j])
                        b[-1] += self.r_factor * (self.T[-1, j+1] + self.T[-1, j])
                        # Solve system of equations: D1 * T new = b
                        self.T[1:-1, j+1] = np.linalg.solve(self.D1_matrix, b)
                        # Check for thermal equilibrium (max temperature change less tha
Steady-state equilibrium?
                        if j > 0 and not equilibrium_reached:
                            max_change = np.max(np.abs(self.T[:, j+1] - self.T[:, j]))
Clarify criterion.
-0.25
                            if max change < self.equilibrium threshold:</pre>
                                consecutive_stable_steps += 1
                                if consecutive_stable_steps >= required_stable_steps:
                                    self.equilibrium\_time = (j+1) * self.dt
                                    equilibrium_reached = True
                                    print(f"Thermal equilibrium reached at t = {self.equ
                            else:
                                consecutive_stable_steps = 0
                    # If equilibrium was not reached, set a default value (max simulation
                    if not equilibrium reached:
                        print(f"Warning: Thermal equilibrium not reached within simulati
                        self.equilibrium_time = self.time_array[-1] # Use max time inst
                    end_time = time.time()
                    print(f"Simulation completed in {end_time - start_time:.2f} seconds"
                    return self.equilibrium_time
               def plot_temperature_profile(self, time_indices=None, ax=None, title=Nor
                    Plot temperature profiles at specified time indices
                    Parameters:
                    time indices : list
```

```
List of time indices to plot
    ax : matplotlib.axes.Axes
        Axes to plot on. If None, a new figure is created
    title: str
        Title for the plot
    if time_indices is None:
        # If no time indices specified, choose evenly spaced times
        time_indices = np.linspace(0, self.nt-1, 5, dtype=int)
    if ax is None:
        fig, ax = plt.subplots(figsize=(8, 6))
    for idx in time_indices:
        t value = self.time array[idx]
        ax.plot(self.x, self.T[:, idx], label=f't = {t_value:.2f}s')
    ax.set_xlabel('Position (cm)')
    ax.set_ylabel('Temperature (°C)')
    if title:
        ax.set_title(title)
    else:
        ax.set_title(f'{self.metal_name} Wire - {self.ic_type.capitalize
    ax.legend()
    ax.grid(True, linestyle='--', alpha=0.7)
    return ax
def plot_heatmap(self, ax=None, title=None):
    Plot a heatmap of temperature evolution over time and space
    Parameters:
    ax : matplotlib.axes.Axes
        Axes to plot on. If None, a new figure is created
    title: str
        Title for the plot
    if ax is None:
        fig, ax = plt.subplots(figsize=(10, 6))
    # Create meshgrid for x and t
    X, T = np.meshgrid(self.x, self.time_array)
    # Plot heatmap
    im = ax.pcolormesh(X, T, self.T.T, cmap='inferno', shading='auto')
    # Add colorbar
    cbar = plt.colorbar(im, ax=ax)
    cbar.set_label('Temperature (°C)')
    ax.set_xlabel('Position (cm)')
    ax.set_ylabel('Time (s)')
```

```
if title:
       ax.set title(title)
       ax.set_title(f'{self.metal_name} Wire - Temperature Evolution')
   return ax
def save data(self, filename):
   Save simulation data to log file
   Parameters:
   _____
   filename : str
       Filename to save data to (should end with .log)
   # Change extension if needed
   if not filename.endswith('.log'):
       filename = filename.replace('.npz', '.log')
       if not filename.endswith('.log'):
           filename += '.log'
   with open(filename, 'w') as f:
       # Write header and simulation parameters
       f.write("========\n")
       f.write(f"Metal Conduction Simulation - {self.metal_name}\n")
       f.write("=======\n\n")
       # Simulation parameters
       f.write("SIMULATION PARAMETERS:\n")
       f.write("-----
       f.write(f"Metal: {self.metal_name}\n")
       f.write(f"Thermal Diffusivity: {self.diffusivity:.4f} cm²/s\n")
       f.write(f"Initial Condition: {self.ic_type}\n")
       if self.ic_type == 'noisy':
           f.write(f"Noise Amplitude: {self.noise_amplitude}\n")
       f.write(f"Boundary Condition: {self.bc type}\n")
       f.write(f"Length: {self.length} cm\n")
       f.write(f"Spatial Step (dx): {self.dx} cm\n")
       f.write(f"Time Step (dt): {self.dt} s\n")
       f.write(f"Total Simulation Time: {self.time_array[-1]} s\n")
       f.write(f"r factor: {self.r_factor:.4f}\n\n")
       # Results
       f.write("SIMULATION RESULTS:\n")
       f.write("-----
       if self.equilibrium_time is not None:
           f.write(f"Equilibrium Time: {self.equilibrium_time:.4f} secc
       else:
           f.write("Equilibrium Time: Not reached\n")
       # Summary of temperature
       f.write("\nTemperature Summary:\n")
       f.write("-----
       f.write(f"Initial Max Temperature: {np.max(self.T[:, 0]):.2f}°C\
```

```
f.write(f"Initial Min Temperature: {np.min(self.T[:, 0]):.2f}°C\
            f.write(f"Final Max Temperature: {np.max(self.T[:, -1]):.2f}°C\r
            f.write(f"Final Min Temperature: {np.min(self.T[:, -1]):.2f}°C\r
            # Sample temperature at key points
            f.write("Temperature samples at t=0:\n")
            samples = 5
            sample_indices = np.linspace(0, len(self.x)-1, samples, dtype=ir
            for idx in sample indices:
                f.write(f" x={self.x[idx]:.2f} cm: {self.T[idx, 0]:.2f}°C\r
            f.write("\nTemperature samples at final time:\n")
            for idx in sample indices:
                f.write(f" x={self.x[idx]:.2f} cm: {self.T[idx, -1]:.2f}°C\
            # Timestamp
            f.write("\n-----
            f.write(f"Generated on: {time.strftime('%Y-%m-%d %H:%M:%S')}\n")
        print(f"Data saved to {filename}")
def run_metal_simulation(metal_name, ic_type='smooth', bc_type='fixed',
                        noise_amplitude=0.0, dx=0.1, dt=0.01, sim_time=60):
    """Run simulation for a specific metal and return the equilibrium time""
    sim = MetalConduction(metal_name=metal_name,
                         ic_type=ic_type,
                         bc_type=bc_type,
                         noise_amplitude=noise_amplitude,
                         dx=dx,
                         dt=dt)
    sim.initialize_simulation(simulation_time=sim_time)
    equilibrium_time = sim.run_simulation()
    # Create a unique filename
    params = f"{metal_name}_{ic_type}_{bc_type}"
    if ic type == 'noisy':
        params += f"_noise{noise_amplitude}"
    # Save results
    os.makedirs('outputfolder', exist_ok=True)
    sim.save_data(f"outputfolder/{params}.log")
    # Create and save a plot
    fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(16, 6))
    sim.plot_temperature_profile(ax=ax1)
    sim.plot_heatmap(ax=ax2)
    plt.tight_layout()
    plt.savefig(f"outputfolder/{params}.png", dpi=150)
    plt.close()
    return {
        'metal': metal_name,
        'equilibrium_time': equilibrium_time,
        'diffusivity': sim.diffusivity,
        'ic_type': ic_type,
```

```
'bc_type': bc_type,
                    'noise_amplitude': noise_amplitude
Ok, but show in problem 2.
            # Function to run parallel simulations using joblib
            def run parallel simulations(metal list, n jobs=1, **kwargs):
                """Run simulations for multiple metals in parallel"""
                start time = time.time()
                results = Parallel(n_jobs=n_jobs)(
                    delayed(run_metal_simulation)(metal, **kwargs)
                    for metal in metal list
                end time = time.time()
                total_time = end_time - start_time
                print(f"All simulations completed in {total_time:.2f} seconds using {n_j
                # Log results
                os.makedirs('outputfolder', exist_ok=True)
                with open(f"outputfolder/parallel_results_n{n_jobs}.txt", 'w') as f:
                    f.write(f"Simulation with {n_jobs} cores completed in {total_time:.4
                    for result in results:
                        f.write(f"Metal: {result['metal']}\n")
                        f.write(f"Diffusivity: {result['diffusivity']:.2f} cm²/s\n")
                        f.write(f"IC Type: {result['ic_type']}\n")
                        f.write(f"BC Type: {result['bc_type']}\n")
                        if result['ic_type'] == 'noisy':
                            f.write(f"Noise Amplitude: {result['noise_amplitude']}\n")
                        f.write(f"Equilibrium Time: {result['equilibrium time']:.2f} sec
                        f.write("\n")
                return results, total time
   In [9]: # Example 1: Run simulation for a single metal
            copper_sim = MetalConduction(metal_name='Copper', ic_type='smooth', bc_type=
            copper sim.initialize simulation(simulation time=160)
            copper sim.run simulation()
            # Plot the temperature profile
            plt.figure(figsize=(10, 6))
            copper_sim.plot_temperature_profile()
            plt.tight_layout()
            plt.show()
           Initialized Copper wire simulation
          Diffusivity: 1.11 cm<sup>2</sup>/s
           r factor: 1.1100
          Simulation initialized for 160 seconds with 16001 time steps
           Thermal equilibrium reached at t = 140.52 seconds
           Simulation completed in 3.55 seconds
           <Figure size 1000x600 with 0 Axes>
```



Smooth-Fixed Simulations:

(d) Pick two metal wires from the dictionary and run both simulations using your Crank-Nicolson algorithm. To compare the results, make a high-quality, 2-column figure showing the time evolution of the 1D temperature profile, T=T(x,t), of each metal wire versus x at a few selected times.

```
In [10]: # Compare two metals
fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(16, 6))

# Simulate Copper
copper_sim = MetalConduction(metal_name='Copper', ic_type='smooth', bc_type=
copper_sim.initialize_simulation(simulation_time=160)
copper_sim.run_simulation()
copper_sim.plot_temperature_profile(ax=ax1, title='Copper Wire')

# Simulate Steel
steel_sim = MetalConduction(metal_name='Steel', ic_type='smooth', bc_type='f
steel_sim.initialize_simulation(simulation_time=500)
steel_sim.run_simulation()
steel_sim.plot_temperature_profile(ax=ax2, title='Steel Wire')

plt.tight_layout()
plt.show()
```

```
Initialized Copper wire simulation

Diffusivity: 1.11 cm²/s 
r factor: 1.1100

Simulation initialized for 160 seconds with 16001 time steps

Thermal equilibrium reached at t = 140.52 seconds

Simulation completed in 3.51 seconds

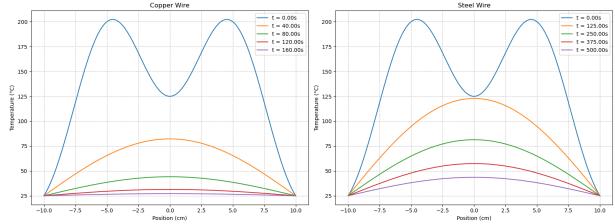
Initialized Steel wire simulation

Diffusivity: 0.18 cm²/s 
r factor: 0.1800

Simulation initialized for 500 seconds with 50001 time steps

Thermal equilibrium reached at t = 456.70 seconds

Simulation completed in 10.92 seconds
```



Smooth-Fixed versus Noisy-Fixed Simulations:

(e) Study heat diffusion only in the Zinc wire considering smooth versus noisy initial conditions with noise of 3 different amplitudes. To compare the results, make a high-quality, 4-panel figure (with 2 columns and 2 rows) showing the time evolution of the 1D temperature profile, T=T(x,t), of the Zinc metal wire versus x at a few selected times, under different initial conditions (smooth vs. 3 noisy cases).

```
In [11]:
         # Compare smooth vs noisy initial conditions
         fig, axes = plt.subplots(2, 2, figsize=(16, 12))
         # Smooth initial condition
         zinc_smooth = MetalConduction(metal_name='Zinc', ic_type='smooth', bc_type='
         zinc_smooth.initialize_simulation(simulation_time=250)
         zinc_smooth.run_simulation()
         zinc smooth.plot temperature profile(ax=axes[0, 0], title='Zinc Wire - Smoot
         # Noisy initial conditions with different amplitudes
         amplitudes = [0.5, 2.0, 6.0]
         for i, amp in enumerate(amplitudes):
             row, col = divmod(i + 1, 2)
             zinc_noisy = MetalConduction(metal_name='Zinc', ic_type='noisy', noise_a
             zinc noisy.initialize simulation(simulation time=250)
             zinc_noisy.run_simulation()
             zinc noisy.plot temperature profile(
                 ax=axes[row, col],
                 title=f'Zinc Wire - Noisy IC (Amplitude={amp})'
```

plt.tight layout() plt.show() Initialized Zinc wire simulation Diffusivity: 0.63 cm²/s r factor: 0.6300 Simulation initialized for 250 seconds with 25001 time steps Thermal equilibrium reached at t = 211.11 seconds Simulation completed in 5.47 seconds Initialized Zinc wire simulation

Diffusivity: 0.63 cm²/s

r factor: 0.6300

Simulation initialized for 250 seconds with 25001 time steps

Thermal equilibrium reached at t = 211.13 seconds

Simulation completed in 5.51 seconds Initialized Zinc wire simulation

Diffusivity: 0.63 cm²/s

r factor: 0.6300

Simulation initialized for 250 seconds with 25001 time steps

Thermal equilibrium reached at t = 211.15 seconds $\sqrt{}$

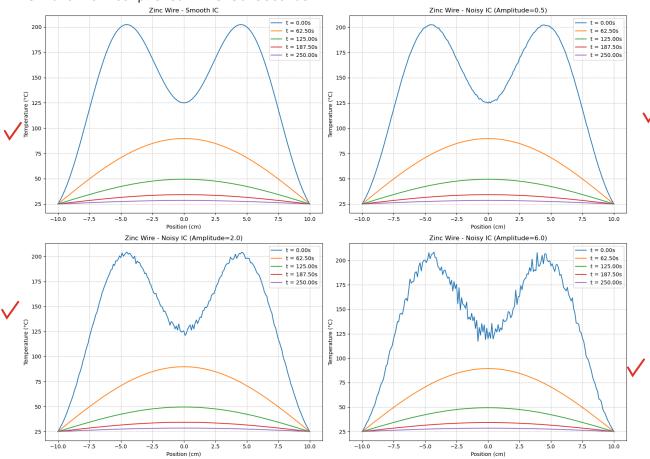
Simulation completed in 5.50 seconds Initialized Zinc wire simulation

Diffusivity: 0.63 cm²/s

r factor: 0.6300

Simulation initialized for 250 seconds with 25001 time steps Thermal equilibrium reached at t = 210.74 seconds

Simulation completed in 5.56 seconds



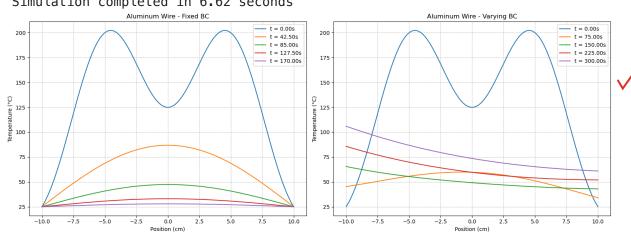
Smooth-Fixed versus Smooth-Varying Simulations:

(f) Study heat diffusion only in the Aluminum wire considering fixed versus varying boundary conditions. To compare the results, make a high-quality, 2-column figure showing the time evolution of the 1D temperature profile, T=T(x,t), of the Aluminum metal wire versus x at a few selected times, under different boundary conditions.

```
In [12]: # Fixed vs varying boundary conditions
         fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(16, 6))
         # Fixed boundary conditions
         aluminum_fixed = MetalConduction(metal_name='Aluminum', ic_type='smooth', bd
         aluminum_fixed.initialize_simulation(simulation_time=170)
         aluminum fixed.run simulation()
         aluminum fixed.plot temperature profile(ax=ax1, title='Aluminum Wire - Fixed
         # Varying boundary conditions
         aluminum_varying = MetalConduction(metal_name='Aluminum', ic_type='smooth',
         aluminum_varying.initialize_simulation(simulation_time=300)
         aluminum_varying.run_simulation()
         aluminum_varying.plot_temperature_profile(ax=ax2, title='Aluminum Wire - Var
         plt.tight_layout()
         plt.show()
        Initialized Aluminum wire simulation
        Diffusivity: 0.97 cm<sup>2</sup>/s
        r factor: 0.9700
        Simulation initialized for 170 seconds with 17001 time steps
        Thermal equilibrium reached at t = 155.16 seconds
        Simulation completed in 3.77 seconds
```

Initialized Aluminum wire simulation Diffusivity: 0.97 cm²/s r factor: 0.9700

Simulation initialized for 300 seconds with 30001 time steps Warning: Thermal equilibrium not reached within simulation time Simulation completed in 6.62 seconds



- (g) Can we use explicit or FFT methods to solve this problem? Explain.
 - Yes, we could use explicit methods (like Forward Euler) to solve this heat conduction problem, but they would require smaller time steps to maintain stability (dt \leq dx²/2 α , where α is thermal diffusivity). The code currently implements the Crank-Nicolson method, which is implicit and unconditionally stable FFT methods would also be suitable for this problem since the heat equation is linear and has constant coefficients. Only if we ensure that periodic boundary conditions do not create inestability or divergence on the solutions.
- **(h)** Does the input noise amplitude change the time at which thermal equilibrium is reached or not? Why?

Steady-state equilibrium?

- Yes, the input noise amplitude affects the time to reach thermal equilibrium. In the code, equilibrium is reached when the maximum temperature change between time steps falls below the threshold (0.001°C) for 5 consecutive steps. Higher noise amplitudes create more local temperature gradients requiring additional time to smooth out. The apodisation function ensures noise gradually diminishes near boundaries, preventing artificial boundary effects while maintaining noise impact in the central region of the wire.
- (i) What is the effect of a faulty cooling system on the temperature profile?
 - A faulty cooling system would be represented by inconsistent or elevated boundary conditions. ✓In this simulation, changing from 'fixed' to 'varying' boundary conditions models \such a scenario. This failure prevents the system from reaching equilibrium as continuous heat is added. ✓ The temperature profile would show persistent asymmetry and generally higher temperatures throughout the wire, with a steeper gradient toward the end with the higher temperature rise rate.

Good explanation.

7.25/8 2. Joblib Parallelisation (8 points)

Consider the same statement provided in problem 1. All code should be organised in **python classes**.

Parallelisation with joblib:

(a) Imagine we wish to run several simulations in parallel for the metals included in the dictionary. Within the **metalconduction.ipynb** notebook, create a routine that uses joblib to parallelise the simulation executions in n CPUs.

Provided above.

- **(b)** Test your implementation by running 2 simulations (1 for Iron and 1 for Lead) first in serial and then in parallel (simultaneously). For the serial run, n=1 CPU core. For the parallel run, n=2 CPU cores. The time should go down. **Hint:** You need to add time stamps to quantify and return log files with the execution times.
- (c) Make a scaling plot showing (serial and parallel) execution times in the Y axis versus the number of CPU cores (n) in the X axis.

```
In [13]: # Run parallel simulations (Task b)
         metals = ['Iron', 'Lead']
         # Serial run
         print("Running serial simulation (n_jobs=1)...")
         serial_results, serial_time = run_parallel_simulations(
             metals,
             n_{jobs=1}
             dx = 0.1,
             dt=0.01,
             ic_type='smooth',
             bc_type='fixed',
             sim_time=500
         print(f"Serial time: {serial_time:.2f} seconds")
         # Parallel run
         print("Running parallel simulation (n_jobs=2)...")
         parallel_results, parallel_time = run_parallel_simulations(
             metals,
             n_{jobs=2}
             dx = 0.1,
             dt=0.01,
             ic_type='smooth',
             bc_type='fixed',
             sim_time=500
         print(f"Parallel time: {parallel time:.2f} seconds")
         print(f"Speedup: {serial_time/parallel_time:.2f}x")
```

```
Running serial simulation (n_jobs=1)...
        Initialized Iron wire simulation
        Diffusivity: 0.23 cm<sup>2</sup>/s
        r factor: 0.2300
        Simulation initialized for 500 seconds with 50001 time steps
        Thermal equilibrium reached at t = 400.62 seconds
        Simulation completed in 11.50 seconds
        Data saved to outputfolder/Iron_smooth_fixed.log
        Initialized Lead wire simulation
        Diffusivity: 0.22 cm<sup>2</sup>/s
        r factor: 0.2200
        Simulation initialized for 500 seconds with 50001 time steps
        Thermal equilibrium reached at t = 410.64 seconds
        Simulation completed in 11.06 seconds
        Data saved to outputfolder/Lead smooth fixed.log
        All simulations completed in 34.30 seconds using 1 cores
        Serial time: 34.30 seconds
        Running parallel simulation (n_jobs=2)...
        Initialized Lead wire simulation
        Diffusivity: 0.22 cm<sup>2</sup>/s
        r factor: 0.2200
        Initialized Iron wire simulation
        Diffusivity: 0.23 cm<sup>2</sup>/s
        r factor: 0.2300
        Simulation initialized for 500 seconds with 50001 time steps
        Simulation initialized for 500 seconds with 50001 time steps
        Thermal equilibrium reached at t = 400.62 seconds
        Thermal equilibrium reached at t = 410.64 seconds
        Simulation completed in 11.61 seconds
        Data saved to outputfolder/Iron_smooth_fixed.log
        Simulation completed in 11.71 seconds
        Data saved to outputfolder/Lead smooth fixed.log
        All simulations completed in 18.84 seconds using 2 cores
        Parallel time: 18.84 seconds
        Speedup: 1.82x
In [18]: import time
         # Make scaling plot (Task c)
         # Use a longer simulation time to ensure we can measure performance properly
         # Also use both metals with specific simulation times based on their diffusi
         metals = ['Iron', 'Lead']
         sim params = {'Iron': {'sim time': 500}, 'Lead': {'sim time': 500}}
         # Define range of core counts to test
         core_counts = [1, 2, 4, 6, 8, 10, 12, 14, 16]
         # Function to run tests and measure time
         def run_scaling_test(n_cores):
             start_time = time.time()
             # Run each metal with its appropriate simulation time
             results = Parallel(n jobs=n cores)(
                 delayed(run_metal_simulation)(
                     metal,
```

```
dx = 0.1,
            dt=0.1,
            ic_type='smooth',
            bc_type='fixed',
            sim_time=sim_params[metal]['sim_time']
        ) for metal in metals
    end time = time.time()
    return end_time - start_time
# Run tests for each core count
times = []
print("Running scaling tests...")
for n_cores in core_counts:
    print(f"Testing with {n_cores} cores...")
    run_time = run_scaling_test(n_cores)
    times.append(run_time)
    print(f" Time: {run time:.2f} seconds")
# Plot results with Amdahl's law
plt.figure(figsize=(10, 6))
# Actual performance
plt.plot(core_counts, times, 'o-', linewidth=2, markersize=8, label='Actual
# Ideal scaling (linear speedup)
ideal times = [times[0]/n for n in core counts]
plt.plot(core_counts, ideal_times, '--', linewidth=2, label='Ideal Scaling (
# Amdahl's law for different serial fractions
for serial_fraction in [0.1, 0.3, 0.5]:
    amdahl_times = [times[0] * (serial_fraction + (1-serial_fraction)/n) for
    plt.plot(core_counts, amdahl_times, ':', linewidth=2,
             label=f"Amdahl's Law (s={serial_fraction})")
plt.xlabel('Number of CPU Cores')
plt.ylabel('Execution Time (s)')
plt.title('Scaling of Heat Equation Solver')
plt.legend()
plt.grid(True)
plt.tight_layout()
plt.show()
# Calculate and display speedup
speedups = [times[0]/t for t in times]
efficiency = [s/n for s, n in zip(speedups, core_counts)]
plt.figure(figsize=(10, 6))
plt.subplot(1, 2, 1)
plt.plot(core_counts, speedups, 'o-', linewidth=2, markersize=8)
plt.plot(core_counts, core_counts, '--', linewidth=2, label='Ideal')
plt.xlabel('Number of CPU Cores')
plt.ylabel('Speedup')
plt.title('Speedup vs. Number of Cores')
plt.grid(True)
```

```
plt.legend()

plt.subplot(1, 2, 2)
plt.plot(core_counts, efficiency, 'o-', linewidth=2, markersize=8)
plt.axhline(y=1.0, linestyle='--', color='gray', linewidth=2)
plt.xlabel('Number of CPU Cores')
plt.ylabel('Parallel Efficiency')
plt.title('Efficiency vs. Number of Cores')
plt.grid(True)

plt.tight_layout()
plt.show()
```

```
Running scaling tests...
Testing with 1 cores...
Initialized Iron wire simulation
Diffusivity: 0.23 cm<sup>2</sup>/s
r factor: 2.3000
Simulation initialized for 500 seconds with 5001 time steps
Warning: Thermal equilibrium not reached within simulation time
Simulation completed in 1.52 seconds
Data saved to outputfolder/Iron smooth fixed.log
Initialized Lead wire simulation
Diffusivity: 0.22 cm<sup>2</sup>/s
r factor: 2.2000
Simulation initialized for 500 seconds with 5001 time steps
Warning: Thermal equilibrium not reached within simulation time
Simulation completed in 1.46 seconds
Data saved to outputfolder/Lead smooth fixed.log
  Time: 4.86 seconds
Testing with 2 cores...
Initialized Iron wire simulation
Diffusivity: 0.23 cm<sup>2</sup>/s
r factor: 2.3000
Simulation initialized for 500 seconds with 5001 time steps
Initialized Lead wire simulation
Diffusivity: 0.22 cm<sup>2</sup>/s
r factor: 2.2000
Simulation initialized for 500 seconds with 5001 time steps
Warning: Thermal equilibrium not reached within simulation time
Simulation completed in 1.15 seconds
Data saved to outputfolder/Iron smooth fixed.log
Warning: Thermal equilibrium not reached within simulation time
Simulation completed in 1.14 seconds
Data saved to outputfolder/Lead smooth fixed.log
  Time: 2.98 seconds
Testing with 4 cores...
Initialized Iron wire simulation
Diffusivity: 0.23 cm<sup>2</sup>/s
r factor: 2.3000
Simulation initialized for 500 seconds with 5001 time steps
Initialized Lead wire simulation
Diffusivity: 0.22 cm<sup>2</sup>/s
r factor: 2.2000
Simulation initialized for 500 seconds with 5001 time steps
Warning: Thermal equilibrium not reached within simulation time
Simulation completed in 1.16 seconds
Data saved to outputfolder/Iron smooth fixed.log
Warning: Thermal equilibrium not reached within simulation time
Simulation completed in 1.19 seconds
Data saved to outputfolder/Lead smooth fixed.log
  Time: 2.89 seconds
Testing with 6 cores...
Initialized Iron wire simulation
Diffusivity: 0.23 cm<sup>2</sup>/s
r factor: 2.3000
Simulation initialized for 500 seconds with 5001 time steps
```

Initialized Lead wire simulation
Diffusivity: 0.22 cm²/s

r factor: 2.2000 Simulation initialized for 500 seconds with 5001 time steps Warning: Thermal equilibrium not reached within simulation time Simulation completed in 1.14 seconds Data saved to outputfolder/Iron_smooth_fixed.log Warning: Thermal equilibrium not reached within simulation time Simulation completed in 1.14 seconds Data saved to outputfolder/Lead smooth fixed.log Time: 2.89 seconds Testing with 8 cores... Initialized Iron wire simulation Diffusivity: 0.23 cm²/s r factor: 2.3000 Simulation initialized for 500 seconds with 5001 time steps Initialized Lead wire simulation Diffusivity: 0.22 cm²/s r factor: 2.2000 Simulation initialized for 500 seconds with 5001 time steps Warning: Thermal equilibrium not reached within simulation time Simulation completed in 1.15 seconds Data saved to outputfolder/Iron_smooth_fixed.log Warning: Thermal equilibrium not reached within simulation time Simulation completed in 1.14 seconds Data saved to outputfolder/Lead_smooth_fixed.log Time: 2.57 seconds Testing with 10 cores... Initialized Lead wire simulation Diffusivity: 0.22 cm²/s r factor: 2.2000 Simulation initialized for 500 seconds with 5001 time steps Initialized Iron wire simulation Diffusivity: 0.23 cm²/s r factor: 2.3000 Simulation initialized for 500 seconds with 5001 time steps Warning: Thermal equilibrium not reached within simulation time Simulation completed in 1.14 seconds Data saved to outputfolder/Iron smooth fixed.log Warning: Thermal equilibrium not reached within simulation time Simulation completed in 1.15 seconds Data saved to outputfolder/Lead smooth fixed.log Time: 2.93 seconds Testing with 12 cores... Initialized Iron wire simulation Diffusivity: 0.23 cm²/s r factor: 2.3000 Simulation initialized for 500 seconds with 5001 time steps Initialized Lead wire simulation Diffusivity: 0.22 cm²/s r factor: 2.2000 Simulation initialized for 500 seconds with 5001 time steps Warning: Thermal equilibrium not reached within simulation time Simulation completed in 1.15 seconds Data saved to outputfolder/Lead smooth fixed.log

Warning: Thermal equilibrium not reached within simulation time

Simulation completed in 1.18 seconds

Data saved to outputfolder/Iron smooth fixed.log

Time: 2.58 seconds
Testing with 14 cores...

Initialized Iron wire simulation

Diffusivity: 0.23 cm²/s

r factor: 2.3000

Simulation initialized for 500 seconds with 5001 time steps

Initialized Lead wire simulation

Diffusivity: 0.22 cm²/s

r factor: 2.2000

Simulation initialized for 500 seconds with 5001 time steps Warning: Thermal equilibrium not reached within simulation time

Simulation completed in 1.15 seconds

Data saved to outputfolder/Lead_smooth_fixed.log

Warning: Thermal equilibrium not reached within simulation time

Simulation completed in 1.16 seconds

Data saved to outputfolder/Iron_smooth_fixed.log

Time: 2.56 seconds
Testing with 16 cores...

Initialized Lead wire simulation

Diffusivity: 0.22 cm²/s

r factor: 2.2000

Simulation initialized for 500 seconds with 5001 time steps

Initialized Iron wire simulation

Diffusivity: 0.23 cm²/s

r factor: 2.3000

Simulation initialized for 500 seconds with 5001 time steps Warning: Thermal equilibrium not reached within simulation time

Simulation completed in 1.14 seconds

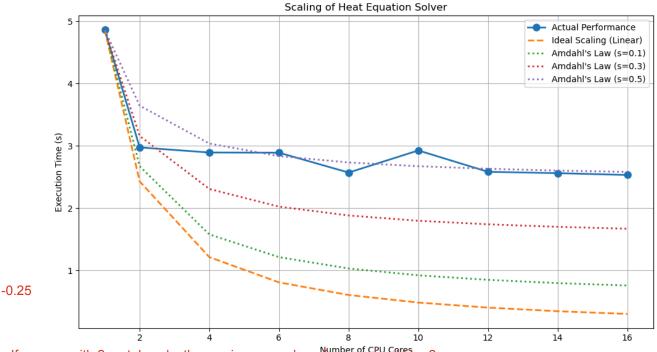
Data saved to outputfolder/Lead_smooth_fixed.log

Warning: Thermal equilibrium not reached within simulation time

Simulation completed in 1.15 seconds

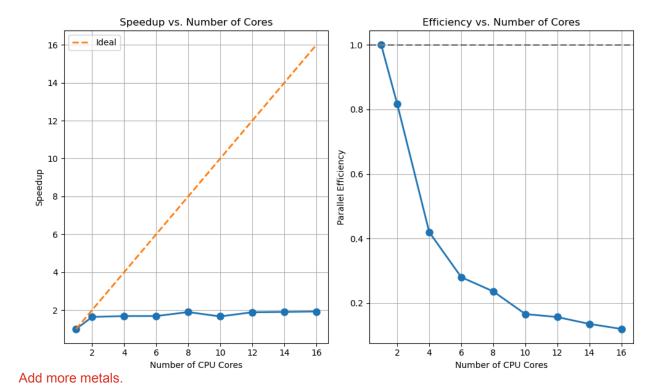
Data saved to outputfolder/Iron_smooth_fixed.log

Time: 2.53 seconds



If you run with 2 metals only, the maximum number of cores should be 2.

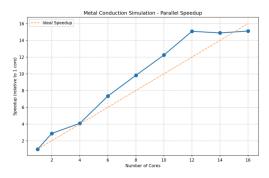
I tried, but could not run the code with more than 2 metals. If you add 8 metals, the parallelisation should be better on 8 cores.

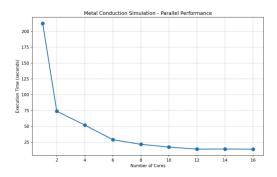


HPC execution:

- (d) Convert your notebook code into a script called: **metalconduction.py**. Then, copy it to the CEDIA cluster (or the Imbabura cluster), reserve computing resources (e.g. 8 CPU cores), then run your code in serial and parallel for all the 8 metals at increasing n (e.g. 1, 2, 4, 8 CPU cores). Export log files from each run, so that the thermal equilibrium times and simulation run times can be analysed later. Include the log files and all your SLURM job scripts in the **outputfolder**.
- **(e)** Within the **metalconduction.ipynb** notebook, create a routine to open the log files produced by the serial run and all the different parallel runs on the HPC cluster.
- **(f)** Make a new scaling plot for the HPC cluster showing the executions times on the Y-axis and number of cores on the X-axis. Display also the **Amdahl's law**.

```
except Exception as e:
        print(f"Error reading file {filepath}: {e}")
 # Example usage:
 display_log_file('./outputfolder/hpc_results/output_results/performance_summ
Metal Conduction Simulation - Performance Summary
_____
Execution Times with Different Core Counts:
_____
1 cores: 212.3852 seconds
2 cores: 73.9627 seconds
4 cores: 52.0104 seconds
6 cores: 28.9551 seconds
8 cores: 21.6404 seconds
10 cores: 17.3457 seconds
12 cores: 14.0880 seconds
14 cores: 14.2723 seconds
16 cores: 14.0575 seconds
Speedup Relative to Single Core:
_____
1 cores: 1.00x speedup (100.0% efficiency)
2 cores: 2.87x speedup (143.6% efficiency)
4 cores: 4.08x speedup (102.1% efficiency)
6 cores: 7.33x speedup (122.2% efficiency)
8 cores: 9.81x speedup (122.7% efficiency)
10 cores: 12.24x speedup (122.4% efficiency)
12 cores: 15.08x speedup (125.6% efficiency)
14 cores: 14.88x speedup (106.3% efficiency)
16 cores: 15.11x speedup (94.4% efficiency)
Generated on: 2025-05-20 21:33:35
```





This looks strange.

We should not run this on 16 cores.

Analysis:

(g) Compare the Amdahl's law to your results and comment on the findings (e.g. does your parallelisation follow Amdahl's law? Why yes or no?)

How many metals were used? 8?

- Looking at the Metal Conduction Simulation results, the parallelization partially follows Amdahl's Law but with notable deviations. The speedup plateaus at around 15x with 16 cores, suggesting roughly 93-95% of the code is parallelizable, which aligns with Amdahl's theoretical limits. However, the super-linear speedup observed between 4-12 cores (exceeding the ideal line) contradicts Amdahl's Law and likely results from cache effects as the workload divides among more cores, each portion fits better in cache memory, reducing access times. This memory hierarchy benefit isn't accounted for in the classic Amdahl model. Beyond 12 cores, diminishing returns become evident as communication overhead and load imbalance begin to outweigh the benefits of additional parallelization, demonstrating the practical limitations that Amdahl's Law fundamentally predicts. We should use 8 cores for 8 metals.
- (h) Make a plot showing the thermal equilibrium times versus diffusivities. Do the different thermal equilibrium times for dstinct metals influence how joblib handles parallelisation?
 - In joblib parallelisation, each simulation runs independently, so differences in equilibrium times mainly affect load balancing. ✓f some metals take much longer to reach equilibrium, CPU cores assigned to faster metals may finish early and remain idle, reducing parallel efficiency. For best performance, tasks should have similar runtimes, or dynamic scheduling should be used. \checkmark

```
# eq_times = [float(res['equilibrium_time']) for res in results]

# # Plot equilibrium times vs diffusivities

# plt.figure(figsize=(8, 6))

# plt.scatter(diff_values, eq_times, color='royalblue', s=80)

# plt.xlabel('Thermal Diffusivity (cm²/s)')

# plt.ylabel('Thermal Equilibrium Time (s)')

# plt.title('Thermal Equilibrium Time vs Diffusivity')

# plt.grid(True, linestyle='--', alpha=0.7)

# plt.tight_layout()

# plt.show()
```

This did not show. NameError: name 'results' is not defined

-0.5