

The Inverse Problem of Thermometry

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Consider a square object of unknown material. Implement and test a program that would estimate thermal conductivity of the object from measurements on the boundary.

Assume the object occupies the domain $\Omega = (0; 1)^2$ and Γ is its boundary. Heat transfer is governed by the equation,

$$\begin{aligned} \frac{\partial u}{\partial t} - \frac{\partial}{\partial x} \left(\sigma \frac{\partial u}{\partial x} \right) - \frac{\partial}{\partial y} \left(\sigma \frac{\partial u}{\partial y} \right) &= f(x, y, t) \quad (x, y) \in \Omega, \quad 0 < t < T \\ \frac{\partial u}{\partial n} &= 0 \quad \text{on } \Gamma \quad 0 < t < T \\ u(x, y) &= 0 \quad \text{at } t = 0 \end{aligned} \quad (1)$$

where $u(x, y, t)$ is the temperature, $\sigma(x, y)$ is the thermal conductivity, $f(x, y, t)$ is the external heating source.

Measurement data $d = d(x, y, t)$ is formed of temperature values at points (x, y) along the domain boundary, Γ , and within the modelling time, $0 < t < T$, Fig. 1, possibly with an additive noise.

Conductivity is estimated by minimizing the functional,

$$L(\sigma) = \int_0^T \int_{\Gamma} (u(x, y, t) - d(x, y, t))^2 dldt + \alpha \int_{\Omega} (\sigma(x, y) - \sigma_0)^2 dxdy \quad (2)$$

with $\sigma(x, y) > 0$ and $u(x, y, t)$ and $\sigma(x, y)$ related via the heat equation (1), σ_0 is the initial guess for $\sigma(x, y)$, and α is a weight.

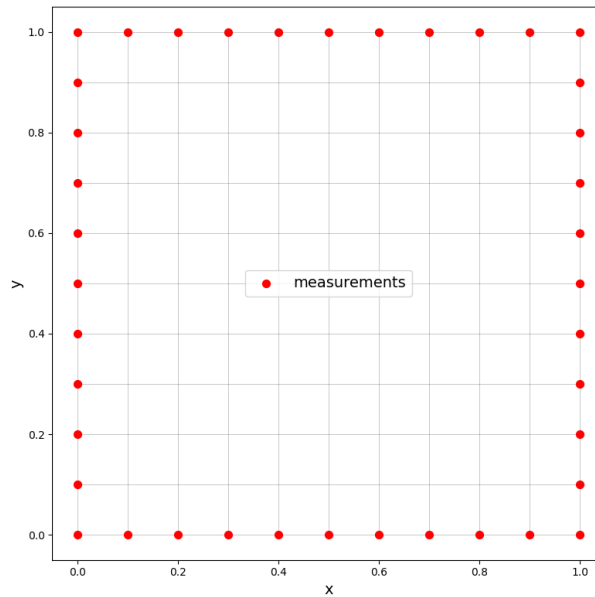


Fig 1. Modelling domain Ω and measurement points.

Part 1 – Differentiable forward solver

Let M be the number of grid cells one direction, then $h = 1/M$ is the grid resolution. Put $x_i = ih$, $y_j = jh$. Similarly, $\tau = T/n$ and $t_k = k\tau$.

Denote as u_{ij}^k approximate value of $u(x_i, y_j, t_k)$ and put $\sigma_{ij} = \sigma(x_i, y_j)$, $f_{ij}^k = f(x_i, y_j, t_k)$.

Discretize (1) with explicit the Euler scheme time-stepping scheme combined with the finite-volume method,

$$\begin{aligned} \frac{u_{ij}^{k+1} - u_{ij}^k}{\tau} + \frac{\sigma_{i+\frac{1}{2}j}(u_{ij}^k - u_{i+1j}^k) + \sigma_{i-\frac{1}{2}j}(u_{ij}^k - u_{i-1j}^k)}{h^2} \\ + \frac{\sigma_{ij+\frac{1}{2}}(u_{ij}^k - u_{ij+1}^k) + \sigma_{ij-\frac{1}{2}}(u_{ij}^k - u_{ij-1}^k)}{h^2} = f_{ij}^k \end{aligned} \quad (3)$$

with $\sigma_{i+\frac{1}{2}j}$ denoting the harmonic average of σ_{ij} and σ_{i+1j} and so on.

Implement (3) in PyTorch or another differentiable framework. **Demonstrate** accuracy of your solver on a test case with known analytical solution for (1), e.g. $u(x, y, t) = (1 - \exp(-t)) \cos(\pi x) \cos(\pi y)$ and $\sigma(x, y) = 1$, $h = 0.1$, $T = 1$.

Hint: With this time-stepping, the time step should satisfy for stability limit,

$$\tau \leq \frac{h^2}{4 \sigma_{\max}}.$$

Part 2 – Optimization

A discrete dataset $d = (d_{sk})$ should be formed of m values of temperature at several points (x_s, y_s) along the domain boundary and at a sequence of n time points, t_k .

Simulate data following (3) with e.g. $\sigma(x, y) = 1 + x + y$, $T = 1$, $h = 0.1$, and for example $f(x, y, t) = \sin(\omega t)$ or other and include 5% noise,

$$d_{sk} = (1 + 0.05\rho)u(x_s, y_s, t_k),$$

where ρ is of uniform random distribution ranging from -1 to 1.

The discrete form of (2) is the following,

$$L(\sigma) = h\tau \sum_{k=1}^n \sum_{s=1}^m (u(x_s, y_s, t_k) - d_{sk})^2 + \alpha h^2 \sum_{j=1}^M \sum_{i=1}^M (\sigma_{ij} - \sigma_0)^2. \quad (4)$$

Starting with $\sigma(x, y) = \sigma_0$ minimize (4) using Adam or another optimizer.

Assuming $\sigma_0 = 2$, **assess** estimated conductivity versus the true one, fitted data versus simulated, functional minimization history.

Consider the impact of the following factors on the quality of the results,

- Source waveform, ω and other kinds of $f(x, y, t)$.
- Weight α .
- transformations of σ , e.g., $\sigma = a + b \tanh(\xi)$.
- Noise level, e.g. 1%, 5%, 10%.