

High-Performance **C**omputing 2025

PDEs (Heat, Poisson & Laplace Equations) and Mathematical Software

Motivation

PDEs – Heat, Poisson, and Laplace Equation

Poisson, Laplace, and Heat equations are fundamental for modeling essential physical phenomena.

Their relationship can be enlightening ...

Heat, Poisson, and Laplace Equations

Definitions

Heat Equation

Describes the diffusion of a quantity (e.g. heat) in a region over time.

$$\frac{\partial u}{\partial t} = \Delta u$$

Written out as ...

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

A positive coefficient called "thermal diffusivity".

Poisson Equation

Describes potential fields influenced by a given source (**source term**).

$$\Delta u = f$$

$$\left(\frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} + \frac{\partial^2 u}{\partial z^2} \right) = f(x, y, z)$$

If **source term** is equal to zero, the equation reduces to the "**Laplace's Equation**".

Heat, Poisson, and Laplace Equations

We have seen this before ...

Heat Equation: Can be viewed as a special case of *Fisher's Equation*.

Fisher's Equation (reminder):

Used to describe biological populations:
spatial diffusion with **reaction/growth**.

$$\frac{\partial s}{\partial t} = \delta \Delta s + \rho s(1 - s)$$

Heat Equation

$$\frac{\partial s}{\partial t} = \delta \Delta s + \rho s(1 - s)$$

* Notice that "S" is function of space, and time!

Ok but what about *Poisson*, and *Laplace Equations* ...
how are they related?

Heat, Poisson, and Laplace Equations

Steady-State of Heat Equation

Consider the Heat Equation with a **source term**

$$\frac{\partial u(x, y, t)}{\partial t} = \alpha \Delta u(x, y, t) + f(x, y)$$

What will the “*steady-state*” solution (i.e. $t \rightarrow \infty$) look like ?

Heat, Poisson, and Laplace Equations

Steady-State of Heat Equation

Poisson Equation: Steady-state form of “Heat Equation with a source”.

The **steady-state** solution will follow:

$$\boxed{\frac{\partial u(x, y, t)}{\partial t}} = \alpha \Delta u(x, y, t) + f(x, y) = \boxed{0} \Rightarrow \Delta u(x, y, t) = -\frac{1}{\alpha} f(x, y)$$

In steady-state there is **no change induced by time**:

$$\boxed{\Delta u(x, y, \cancel{t}) = -\frac{1}{\alpha} f(x, y)}$$

Heat, Poisson, and Laplace Equations

Steady-State of Heat Equation w. Zero Source Term

Laplace Equation: Steady-state form of “Heat Equation without a source”.

Consider the Heat Equation with **no heat source**:

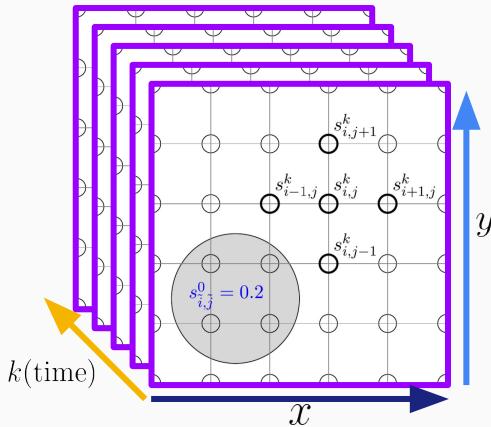
$$\frac{\partial u(x, y, t)}{\partial t} = \alpha \Delta u(x, y, t) + \cancel{f(x, y)}$$

In steady-state there is **no change induced by time**:

$$\boxed{\Delta u(x, y, \cancel{t}) = 0}$$

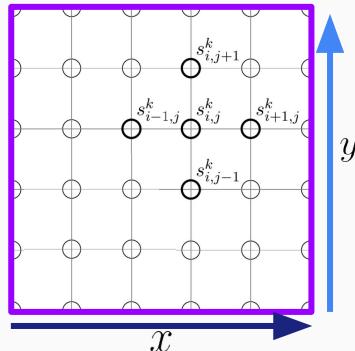
Solution Method

Basic Steps



Time-Dependent (e.g., Heat, Fisher's Eq.)

- Formulation & Representation:** Model the domain, **initial condition**, and **boundaries conditions**.
- Discretization Space & Time:** Convert the continuous problem into a set of discrete equations.
- Solve Space & Time:** Solution over domain per time-step.



Time-Independent (e.g., Poisson's and Laplace's Eq.)

- Formulation & Representation:** Model the domain and **boundaries conditions**.
- Discretization Space:** Convert the continuous problem into a set of discrete equations.
- Solve Space:** Solution over domain.

Solution Method

Linear or Nonlinear?

Fisher's Equation (Nonlinear)

** Note we use \mathbf{s} in place \mathbf{u}

$$\frac{1}{\tau} (s_{i,j}^k - s_{i,j}^{k-1}) = \frac{\delta}{h^2} (-4s_{i,j}^k + s_{i+1,j}^k + s_{i-1,j}^k + s_{i,j+1}^k + s_{i,j-1}^k) + \rho s_{i,j}^k (1 - s_{i,j}^k)$$



The solution is than the root of:

$$f(\mathbf{s}^k | \mathbf{s}^{k-1}, \mathbf{A}, c_1, c_2) := \mathbf{s}^k - \mathbf{s}^{k-1} - c_1 \mathbf{A} \mathbf{s}^k - c_2 \mathbf{s}^k \cdot (1 - \mathbf{s}^k)$$

Use Newton's Method → multiple linear solves

How would this change for the Poisson and Laplace equation?

Solution Method

Linear or Nonlinear?

Poisson's Equation (Linear):

For example consider a forcing function $\sin(x, y)$ which is discretized to
 $\mathbf{b} = [\sin(x_0, y_0), \sin(x_1, y_0), \dots, \sin(x_{N-1}, y_{N-1})]$.

The solution is than the root of:

$$\Delta u(x, y) = \sin(x, y) \Rightarrow f(\mathbf{u} | \mathbf{A}, c) = \boxed{\mathbf{A}\mathbf{u} - \mathbf{cb}}$$

Or simply single linear solve

... and Laplace's Equation?

Solution Method

Pseudo Algorithm

Nonlinear w. time stepping

```
Input u_initail_value, K, iter_max, eps

// Initial Conditions
u_last ← u_initail_value
u      ← u_last

// Time loop
For k = 1 to K
    // Newton loop
    For iter=1 to iter_max
        // Linear Solve
        update ← lin_solve(J(u|u_last),f(u|u_last))
        u ← u - update
        // Convergence Check
        If norm(update)<eps
            break
        Endif
    Endfor
    // Swap Solution
    u_last ← u
Endfor

Return u
```

Linear wo. time stepping

```
Input u_initail_value, K, iter_max, eps

// Initial Conditions
u_last ← u_initail_value
u      ← u_last

// Time loop
For k = 1 to K
    // Newton loop
    For iter=1 to iter_max
        // Linear Solve
        u ← lin_solve(A,b)
        u ← u - update
        // convergence Check
        If norm(update)<eps
            break
        Endif
    Endfor
    // Swap Solution
    u_last ← u
Endfor

Return u
```

Solution Method

Linear Solve

Linear solve is central to the solution of both **linear** and **nonlinear** PDEs.

Direct Methods: Solve matrices in fixed steps with notable stability, especially for well-conditioned systems.

Utilize matrix factorizations (e.g., LU, Cholesky, QR) to reduce the problem to simpler triangular systems, which can then be solved through back-substitution.

Iterative Methods: Memory-efficient with *adjustable accuracy*, though they demand careful considerations for stability. For example:

1. **Conjugate Gradient (CG):** An iterative method for symmetric positive-definite systems of linear equations.
2. **Generalized Minimal Residual (GMRES):** An iterative method for nonsymmetric (possibly indefinite) system of linear equations.

In the examples we have seen, **CG** is very much applicable.

For more on CG, see "[An Introduction to the Conjugate Gradient Method Without the Agonizing Pain](#)"

Steps: (i) assemble matrices/vectors , (ii) solve and (iii) maybe repeat

... a toolkit would really help!



Portable, Extensible Toolkit for Scientific Computation: Scalable (parallel) solution of scientific applications modeled by partial differential equations (PDEs).

Supports **MPI** and GPU acceleration through CUDA, HIP, Kokkos, or OpenCL, as well as hybrid MPI-GPU parallelism

Other frameworks exist such as Trilinos

Toolkits

PETSc Code Structure

```
int main(int argc, char **argv) {
...
// Boilerplate code for initial setup
PetscCall(KSPCreate(PETSC_COMM_WORLD, &ksp));
PetscCall(KSPSetComputeRHS(ksp, ComputeRHS, &user));
PetscCall(KSPSetComputeOperators(ksp, ComputeMatrix, NULL));
PetscCall(KSPSetDM(ksp, da));
PetscCall(KSPSetFromOptions(ksp));
PetscCall(KSPSetUp(ksp));
...
}

PetscErrorCode ComputeMatrix(KSP ksp, Mat A, Mat P, void *ctx) {
    DM da;
    DMDALocalInfo info;

    // Retrieve the distributed array, grid information, and global grid dimensions
    PetscCall(KSPGetDM(ksp, &da));
    PetscCall(DMDAGetLocalInfo(da, &info));
    ...

}

PetscErrorCode ComputeRHS(KSP ksp, Vec b, void *ctx) {
    DM da;
    DMDALocalInfo info;

    // Retrieve the distributed array, grid information, and global grid dimensions
    PetscCall(KSPGetDM(ksp, &da));
    PetscCall(DMDAGetLocalInfo(da, &info));
    ...
}
```

Main Idea:

- “KSP” linear solvers in PETSc.
- **ComputeMatrix** and **ComputeRHS** tells how to assemble the global matrix and RHS vector from the local data owned by each MPI rank.
- You can swap solvers and configurations very easily! *without changing your PDE code.*

Many tutorials available ...
Use them!