

High-Performance Computing 2025

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

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.

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

Poisson Equation

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

$$\Delta u = f$$

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”.

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

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

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

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

Ok but what 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**

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

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**:

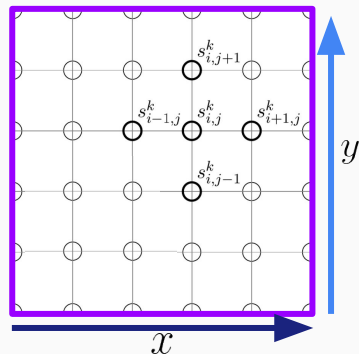
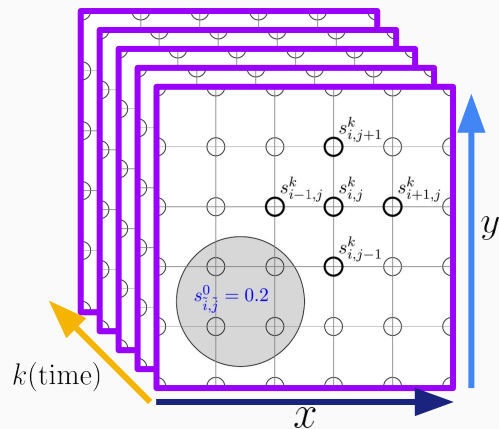
$$\frac{\partial u(x, y, t)}{\partial t} = \alpha \Delta u(x, y, t) + \cancel{f(x, y)}$$

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

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

Solution Method

Basic Steps



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

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

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

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

Solution Method

Linear or Nonlinear?

Fisher's Equation (**Nonlinear**)

** Note we use **s** in place **u**

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

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 then the root of:

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

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
```

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!