



High-Performance
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Solving the 2d Poisson Equation using the FEM and a CG Method

Finite Element Method

Strong Formulation of the Poisson Equation

Goal: Find the unknown function u (e.g., temperature, potential) on a domain Ω , given a source f (e.g., heat source).

This is described by the **strong form**. Find $u : \Omega \rightarrow \mathbb{R}$ such that

$$-\Delta u = f \quad \in \Omega \quad (\text{Poisson's Equation}) \quad (1)$$

$$u = 0 \quad \in \partial\Omega \quad (\text{Boundary Condition}) \quad (2)$$

Laplace operator Δ :

$$\Delta u = \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} \quad (3)$$

Problem: Requires u to be twice-differentiable. This is too restrictive and difficult to work with computationally.

Notation for the L^2 -inner product:

$$(f, g) := \int_{\Omega} f \cdot g \, dx \quad (4)$$

Weak Formulation of the Poisson Equation I

Idea: We „weaken“ the requirement from C^2 to C^1 by testing the equation in an integral sense.

Define function space V for both the solution u and „test functions“ v :

$$V := H_0^1(\Omega), \quad (5)$$

This space V contains all functions that are one time (weakly) differentiable and are zero on the boundary $\partial\Omega$.

Multiply the PDE by a test function $v \in V$ and integrate over Ω :

$$(-\Delta u, v) = (f, v) \quad \forall v \in V \quad (6)$$

Apply integration by parts (Green's theorem) to the left-hand side to move a derivative from u onto v :

$$(\nabla u, \nabla v) - \int_{\partial\Omega} \partial_n u \phi \, ds = (f, \phi) \quad \forall \phi \in V \quad (7)$$

Since our test functions $v \in V$ are 0 on the boundary ($v = 0$ on $\partial\Omega$), the boundary integral vanishes.

This gives the **weak formulation**: Find $u \in V$ such that

$$(\nabla u, \nabla v) = (f, v) \quad \forall v \in V \quad (8)$$

This is often written abstractly as $a(u, v) = L(v)$, where:

- $a(u, v) := (\nabla u, \nabla v)$ is the **bilinear form** (handles stiffness).
- $L(v) := (f, v)$ is the **linear form** (handles the source/load).

Problem: The space $V = H_0^1$ is infinite-dimensional.

Solution: We approximate V with a finite-dimensional subspace $V_h \subset V$.

Step 1: Meshing

We discretize the domain Ω into a finite number of simple shapes (elements), like triangles. The collection of these elements is the **mesh**, \mathcal{T}_h . The parameter h represents the mesh size (e.g., max triangle width).

Step 2: Define V_h

We define V_h as the space of functions that are:

- Continuous across the whole domain Ω .
- A simple polynomial (e.g., linear) on each triangle $K \in \mathcal{T}_h$.
- Still zero on the boundary $\partial\Omega$.

This is the space of continuous, piecewise-linear functions.

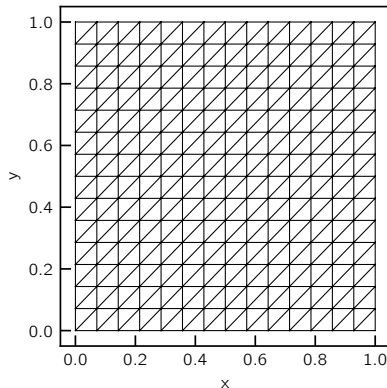


Figure: A mesh \mathcal{T}_h of the unit square.

How do we build a basis for V_h ? We use **local basis functions** (or **shape functions**) ϕ_i associated with each node N_i of the mesh.

For P1 (piecewise-linear) elements, ϕ_i has a key property:

- ϕ_i is 1 at its „own“ node N_i .
- ϕ_i is 0 at all other nodes N_j ($j \neq i$).
- ϕ_i is linear on each triangle.

This is the „hat function“.

Area Coordinates

On a single triangle with nodes (x_1, y_1) , (x_2, y_2) , (x_3, y_3) , these shape functions are identical to the area coordinates L_i :

$$x = L_1 x_1 + L_2 x_2 + L_3 x_3 \quad (9)$$

$$y = L_1 y_1 + L_2 y_2 + L_3 y_3 \quad (10)$$

$$1 = L_1 + L_2 + L_3 \quad (11)$$

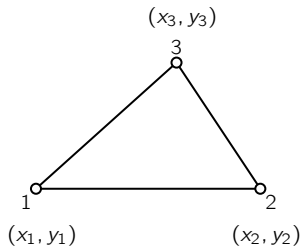


Figure: A single linear triangular element.

Explicit Formulas

The area coordinates L_i are linear functions of (x, y) :

$$L_1 = \frac{a_1 + b_1x + c_1y}{2A}, \quad a_1 = x_2y_3 - x_3y_2, \quad b_1 = y_2 - y_3, \quad c_1 = x_3 - x_2 \quad (12)$$

$$L_2 = \frac{a_2 + b_2x + c_2y}{2A}, \quad a_2 = x_3y_1 - x_1y_3, \quad b_2 = y_3 - y_1, \quad c_2 = x_1 - x_3 \quad (13)$$

$$L_3 = \frac{a_3 + b_3x + c_3y}{2A}, \quad a_3 = x_1y_2 - x_2y_1, \quad b_3 = y_1 - y_2, \quad c_3 = x_2 - x_1 \quad (14)$$

Local Shape Functions

On this linear triangle element, the three local shape functions are just:

$$\phi_1(x, y) = L_1 \quad \phi_2(x, y) = L_2 \quad \phi_3(x, y) = L_3 \quad (15)$$

Gradients of Shape Functions

Since the shape functions $\phi_i = L_i$ are linear, their gradients are constant on each element:

$$\nabla \phi_i = \begin{bmatrix} \frac{\partial \phi_i}{\partial x} \\ \frac{\partial \phi_i}{\partial y} \end{bmatrix} \quad (16)$$

$$\frac{\partial \phi_1}{\partial x} = \frac{y_2 - y_3}{2A} \quad (17)$$

$$\frac{\partial \phi_2}{\partial x} = \frac{y_3 - y_1}{2A} \quad (18)$$

$$\frac{\partial \phi_3}{\partial x} = \frac{y_1 - y_2}{2A} \quad (19)$$

$$\frac{\partial \phi_1}{\partial y} = \frac{x_3 - x_2}{2A} \quad (20)$$

$$\frac{\partial \phi_2}{\partial y} = \frac{x_1 - x_3}{2A} \quad (21)$$

$$\frac{\partial \phi_3}{\partial y} = \frac{x_2 - x_1}{2A} \quad (22)$$

The Galerkin Problem (Discrete Form)

We now solve the weak form on the finite-dimensional space V_h :

Find $u_h \in V_h$ such that

$$(\nabla u_h, \nabla v_h) = (f, v_h) \quad \forall v_h \in V_h \quad (23)$$

As the mesh size $h \rightarrow 0$, the discrete solution u_h converges to the true solution u .

Any function u_h in our space V_h can be written as a linear combination of its basis functions ϕ_j :

$$u_h(x, y) = \sum_{j=1}^n u_j \phi_j(x, y) \quad (24)$$

Here, n is the total number of nodes in the mesh. Because $\phi_j(N_i) = \delta_{ij}$, the unknown coefficients u_j are simply the **unknown values of the solution at the nodes**: $u_j = u_h(N_j)$.

Converting into a Linear Equation System II

Substitute this „ansatz“ into the Galerkin problem:

$$(\nabla u_h, \nabla \phi_h) = (f, \phi_h) \quad \forall \phi_h \in V_h \quad (25)$$

$$\left(\nabla \left(\sum_{j=1}^n u_j \phi_j \right), \nabla \phi_h \right) = (f, \phi_h) \quad \forall \phi_h \in V_h \quad (26)$$

$$\sum_{j=1}^n u_j (\nabla \phi_j, \nabla \phi_h) = (f, \phi_h) \quad \forall \phi_h \in V_h \quad (27)$$

This must hold for *all* $\phi_h \in V_h$. It is sufficient to test it against *each basis function* ϕ_i (for $i = 1, \dots, n$):

$$\sum_{j=1}^n u_j (\nabla \phi_j, \nabla \phi_i) = (f, \phi_i) \quad \forall i \in \{1, \dots, n\} \quad (28)$$

Converting into a Linear Equation System III

This is a linear system of equations $\mathbf{A}\mathbf{u} = \mathbf{F}$!

$$\underbrace{\begin{bmatrix} (\nabla\phi_1, \nabla\phi_1) & \dots & (\nabla\phi_n, \nabla\phi_1) \\ \vdots & \ddots & \vdots \\ (\nabla\phi_1, \nabla\phi_n) & \dots & (\nabla\phi_n, \nabla\phi_n) \end{bmatrix}}_{\text{Stiffness matrix } \mathbf{A}} \underbrace{\begin{bmatrix} u_1 \\ \vdots \\ u_n \end{bmatrix}}_{\mathbf{u}} = \underbrace{\begin{bmatrix} (f, \phi_1) \\ \vdots \\ (f, \phi_n) \end{bmatrix}}_{\text{Load vector } \mathbf{F}} \quad (29)$$

The entries are:

$$A_{ij} = (\nabla\phi_j, \nabla\phi_i) := \int_{\Omega} \nabla\phi_j \cdot \nabla\phi_i \, dx \quad (30)$$

$$F_i = (f, \phi_i) := \int_{\Omega} f\phi_i \, dx \quad (31)$$

Our goal is to **solve** this system for the unknown vector of nodal values \mathbf{u} .

Algorithm 1: Global Assembly

Input : Mesh \mathcal{T}_h , Source f

```
1 Initialize  $\mathbf{A} = 0$  (sparse  $n \times n$  matrix)
2 Initialize  $\mathbf{F} = 0$  ( $n \times 1$  vector)
3 foreach element  $e$  in mesh  $\mathcal{T}_h$  do
4   Initialize  $\mathbf{A}^e = 0$  ( $3 \times 3$  local matrix)
5   Initialize  $\mathbf{F}^e = 0$  ( $3 \times 1$  local vector)
6   Let  $\phi_1, \phi_2, \phi_3$  be the local shape functions on  $e$ 
7   for  $i = 1$  to 3 do
8      $F_i^e = \int_e f \phi_i \, dx$ 
9     for  $j = 1$  to 3 do
10       $A_{ij}^e = \int_e \nabla \phi_j \cdot \nabla \phi_i \, dx$ 
11    end
12  end
13  for  $i = 1$  to 3 do
14     $F[g_i] += F_i^e$ 
15    for  $j = 1$  to 3 do
16       $A[g_i, g_j] += A_{ij}^e$ 
17    end
18  end
19 end
20 ApplyDirichletBC( $\mathbf{A}, \mathbf{F}$ )
21 return  $\mathbf{A}, \mathbf{F}$ 
```

Structure of the Matrix A

- A is a large, **sparse matrix**.
- The entry A_{ij} is non-zero **only if** nodes i and j share an element.
- It is also **symmetric** ($A_{ij} = A_{ji}$) and **positive-definite**.

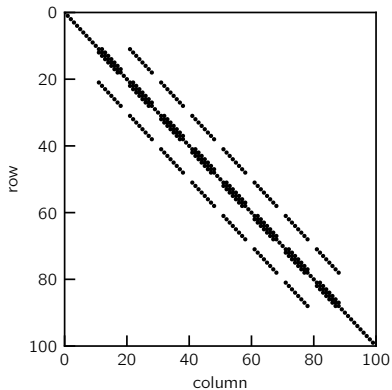


Figure: Sparsity pattern of A .

Solving the System

Because \mathbf{A} is sparse, symmetric, and positive-definite, we don't use direct solvers (like \mathbf{A}^{-1}). We use efficient iterative solvers like the Conjugate Gradient (CG) Method.

Algorithm 2: Conjugate Gradient Method

Input : Matrix \mathbf{A} , vector \mathbf{b} , initial guess \mathbf{x}_0

Output: Solution \mathbf{x}

```
1  $\mathbf{r}_0 = \mathbf{b} - \mathbf{A}\mathbf{x}_0$ 
2  $\mathbf{p}_0 = \mathbf{r}_0$ 
3 for  $k = 0$  to  $maxit - 1$  do
4    $\alpha_k = (\mathbf{r}_k^T \mathbf{r}_k) / (\mathbf{p}_k^T \mathbf{A} \mathbf{p}_k)$ 
5    $\mathbf{x}_{k+1} = \mathbf{x}_k + \alpha_k \mathbf{p}_k$ 
6    $\mathbf{r}_{k+1} = \mathbf{r}_k - \alpha_k \mathbf{A} \mathbf{p}_k$ 
7   if  $\|\mathbf{r}_{k+1}\|_2 \leq tol$  then
8     return  $\mathbf{x}_{k+1}$ 
9    $\beta_k = (\mathbf{r}_{k+1}^T \mathbf{r}_{k+1}) / (\mathbf{r}_k^T \mathbf{r}_k)$ 
10   $\mathbf{p}_{k+1} = \mathbf{r}_{k+1} + \beta_k \mathbf{p}_k$ 
11 end
12 return  $\mathbf{x}_{k+1}$ 
```

Solution of the Poisson Equation

After solving $\mathbf{A}\mathbf{u} = \mathbf{F}$ for the vector \mathbf{u} , we have the value u_j at each node. We can then reconstruct the full solution $u_h = \sum u_j \phi_j$ for visualization.

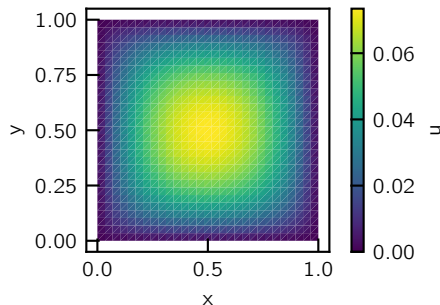


Figure: The computed discrete solution u_h for a source $f = 1$ on the unit square.

- [1] T Wick. *Numerical Methods for Partial Differential Equations*. en. Hannover : Institutionelles Repositorium der Leibniz Universität Hannover, 2020. DOI: 10.15488/9248.
- [2] OC Zienkiewicz and RL Taylor. *The finite element method. Volume 1: The basis*. 5th ed. Oxford: Butterworth-Heinemann, 2002. 689 pp. URL: https://www.meil.pw.edu.pl/content/download/58297/306302/file/FEM_Zienkiewicz%20Vol1.pdf (visited on 10/28/2025).