

# Solving Harmonic Function using SVD

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## 1 Introduction

I will try to solve the Harmonic function using the SVD technique before going to more complicated PDEs.

## 2 Harmonic function

The harmonic function is given by the equation:

$$f(x, y) = 2\pi^2 \sin(\pi x) \sin(\pi y)$$

Find  $U : \Omega \rightarrow \mathbb{R}$  such that

$$\begin{aligned} -\Delta u &= f \text{ in } \Omega, \\ u &= 0 \text{ on } \partial\Omega, \\ \Omega &= (0, 1) \times (0, 1) \end{aligned}$$

## 3 Finite difference method

$$-\Delta u = - \left( \frac{\partial^2}{\partial x^2} + \frac{\partial^2}{\partial y^2} \right) \quad (1)$$

The differential equation can be approximated as

$$-\Delta u = - \left( \frac{u_{i+1,j} - 2u_{i,j} + u_{i-1,j}}{dx^2} + \frac{u_{i,j+1} - 2u_{i,j} + u_{i,j-1}}{dy^2} \right) \quad (2)$$

By assuming  $dx = dy = h$ , we can rewrite the Eq. (2) to be

$$-\Delta u = - \left( \frac{u_{i+1,j} + u_{i-1,j} + u_{i,j+1} + u_{i,j-1} - 4u_{i,j}}{h^2} \right) \quad (3)$$

At the boundary node, the equation will be

$$u_0 = 0 \quad (4)$$

For node inside the boundary, the equation will be

$$-\frac{1}{h^2} (u_{i+1,j} + u_{i-1,j} + u_{i,j+1} + u_{i,j-1} - 4u_{i,j}) = f(x, y) \quad (5)$$

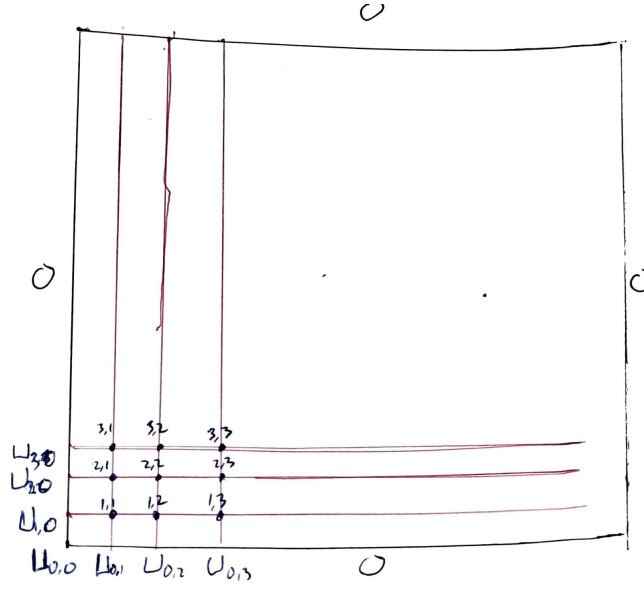


Figure 1: Discretization of the domain

## 4 Constructing the matrix

We will have a set of linear equations based on the size of  $h$ .

$$-\frac{1}{h^2} \begin{bmatrix} 1 & \cdots & & & & & & & 0 \\ \vdots & \ddots & & & & & & & \\ & 1 & \cdots & 1 & -4 & 1 & \cdots & 1 & \cdots \\ & & & & & & \ddots & & \\ 0 & & & & & & \cdots & & 1 \end{bmatrix} \begin{bmatrix} u_{00} \\ u_{01} \\ u_{02} \\ u_{03} \\ \vdots \\ u_{10} \\ u_{11} \\ u_{12} \\ u_{13} \\ \vdots \\ u_{nn} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ \vdots \\ 0 \\ f(x, y) \\ f(x, y) \\ f(x, y) \\ \vdots \\ 0 \end{bmatrix} \quad (6)$$

This matrix has a diagonally dominant property.

## 5 Result

### 5.0.1 Finite difference method

The finite difference method gives the fastest result with CPU time = 36.96. But the back draw of this method is the error cannot be reduced any further.

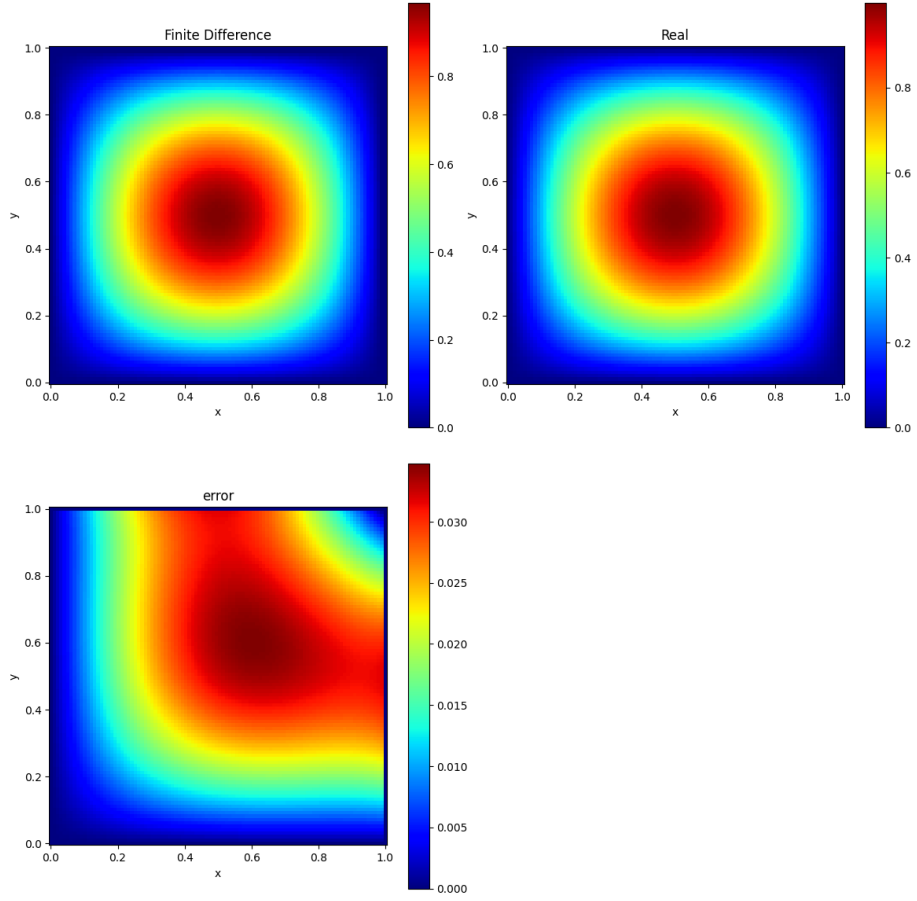


Figure 2: Finite Difference method result

The result is shown in Fig. 2.

### 5.0.2 Classic matrix calculation

Classic matrix calculation is done with CPU time = 103.9. It is done using the following formula

$$Ax = B \quad (7)$$

$$x = \text{inv}(A)B \quad (8)$$

Matrix calculation gives a better accuracy. The result is shown in Fig. 3.

### 5.0.3 Matrix calculation using SVD decomposition

Matrix decomposition using the SVD technique is done by numpy library. To decompose the matrix A, it needs CPU time = 1005.03 and the calculation using

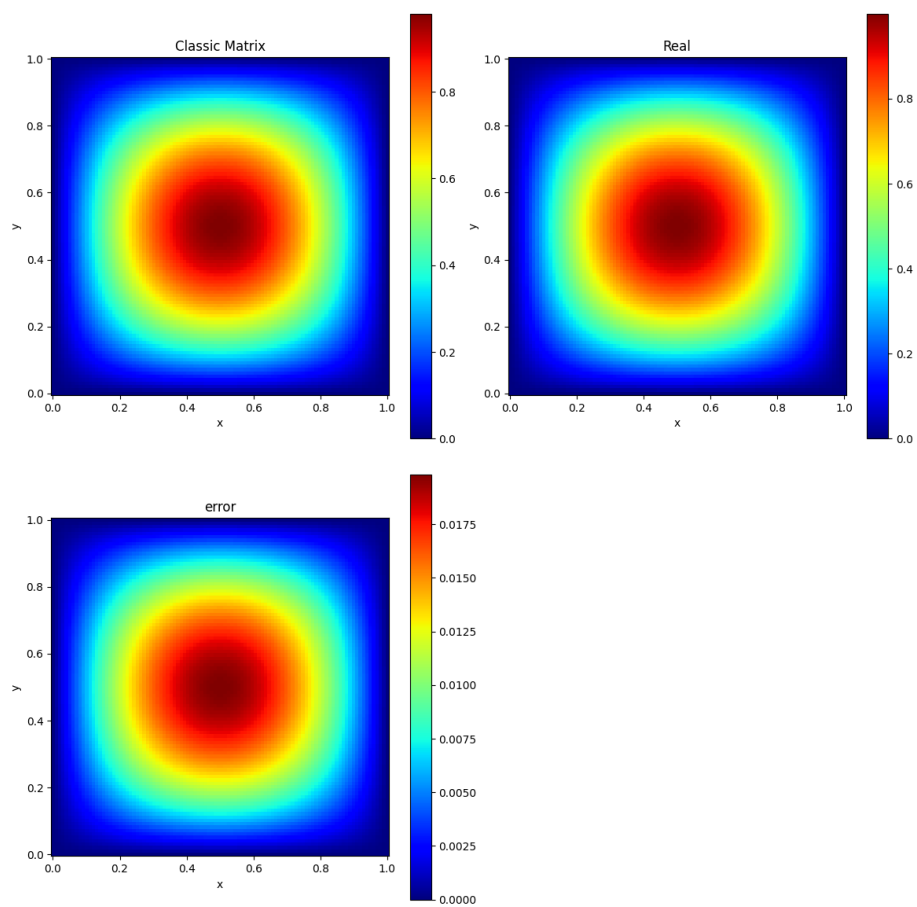


Figure 3: Classic matrix calculation result

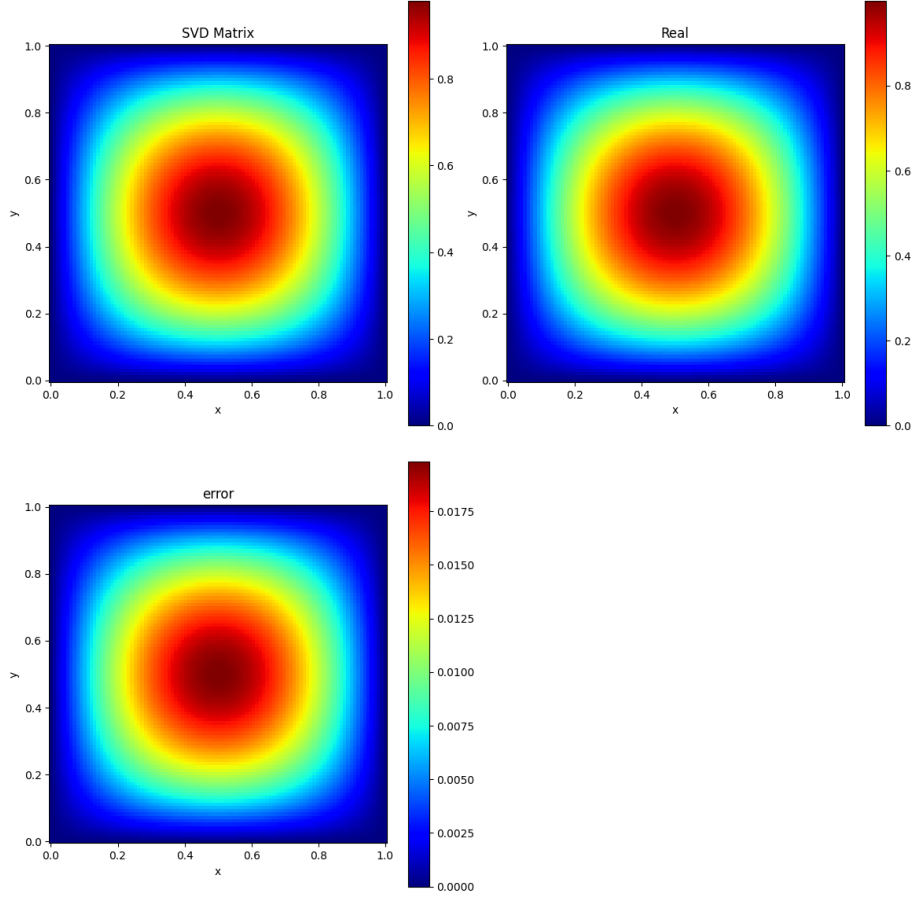


Figure 4: SVD matrix decomposition calculation result

decomposed matrix needs CPU time = 414.24. It is done by continuing the Eq. (7) and (8)

$$Ax = B$$

$$x = \text{inv}(A)B$$

$$A = U\Sigma V^T \quad (9)$$

$$x = \text{inv}(U\Sigma V^T)B \quad (10)$$

$$x = \text{inv}(V^T)\text{inv}(\Sigma)\text{inv}(U)B \quad (11)$$

Total CPU time using SVD Decomposition is the highest. The result is shown in Fig. 4.

## 6 Conclusion

Method	CPU time
Finite Difference	36.96
$Ax = B$	103.9
SVD decomposition	1005.03
$\text{inv}(V^T)\text{inv}(\Sigma)\text{inv}(U)B$	414.24

Table 1: CPU time summary