

Neural networks for boundary value problems

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- Methods of fundamental solution(MFS) with two approaches by neural network
- Convergence analysis of the MFS
- Numerical results
- Future works

Problem

$$\begin{cases} Lu = 0 & \text{on } \Omega \\ u|_{\Gamma_1} = g & \text{and } \frac{\partial u}{\partial n}|_{\Gamma_2} = f. \end{cases} \quad (1)$$

where $\Omega \subset \mathbb{R}^n$ is an open set and $\partial\Omega$ consists of two disjoint parts, Γ_1 and Γ_2 , such that $\partial\Omega = \Gamma_1 \cup \Gamma_2$.

Definition

A fundamental solution to a linear differential operator L is a distribution E such that $L(E) = \delta$

Fundamental solution of Laplace operator

$$\phi_y(x) = \begin{cases} -\frac{1}{2\pi} \log(|x - y|) & \text{for } d = 2 \\ \frac{1}{(d-2)\omega_n \|x-y\|^{d-2}} & \text{for } d \geq 3 \end{cases} \cdot$$

Method of fundamental solution

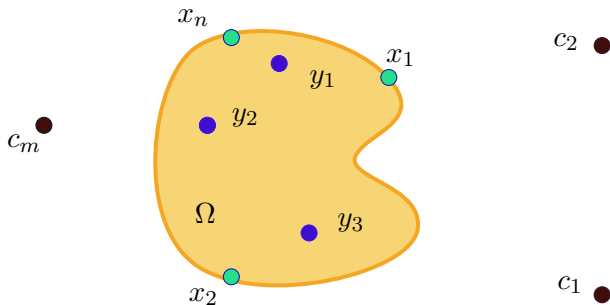


Figure: Inner, Boundary, Outer points

According to the single layer representation,

$$u(x) = \int_{\partial\Omega} \phi_y(x) \mathcal{G}(y) \, dy. \quad (2)$$

If x are boundary point, $\phi_y(x)$ was not well-defined (singular integrals).

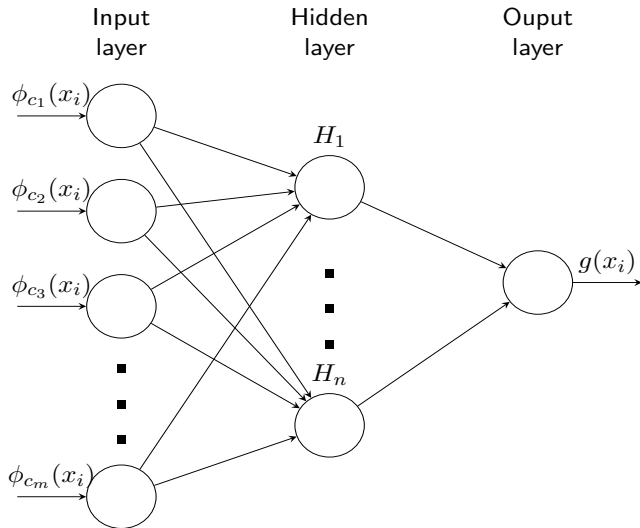
Hence auxiliary boundary Ω' was introduced

$$u(x) = \int_{\partial\Omega'} \phi_y(x) \mathcal{G}'(y) \, dy$$

Discretizing the integrals, we get

$$u(x) \approx \sum_{j=1}^m a_j \phi_{c_j}(x) \quad (3)$$

Method of fundamental solution



The first approach

- Dirichlet typed input:
 $(\phi_{c_1}(x_i), \dots, \phi_{c_m}(x_i))$
where $x_i \in \Gamma_1$
Neumann typed input :
 $(\frac{\partial}{\partial \mathbf{n}} \phi_{c_1}(x_i), \dots, \frac{\partial}{\partial \mathbf{n}} \phi_{c_m}(x_i))$
where $x_i \in \Gamma_2$
- Dirichlet typed output :
 $g(x_i)$ where $x_i \in \Gamma_1$
Neumann typed output:
 $f_{\mathbf{n}}(x_i)$ where $x_i \in \Gamma_2$

The second approach

- Input

$$(\phi_{c_j}(x_1), \dots, \phi_{c_j}(x_n), \\ \frac{\partial}{\partial \mathbf{n}} \phi_{c_j}(x'_1), \dots, \frac{\partial}{\partial \mathbf{n}} \phi_{c_j}(x'_n))$$

- Output $\phi_{c_j}(y)$

The first approach

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The second approach

- Input
 $(\phi_{c_j}(x_1), \dots, \phi_{c_j}(x_n),$
 $\frac{\partial}{\partial \mathbf{n}} \phi_{c_j}(x'_1), \dots, \frac{\partial}{\partial \mathbf{n}} \phi_{c_j}(x'_n))$
- Output $\phi_{c_j}(y)$

Convergence analysis of the MFS

(*Katsurada and Okamoto ('96), and Fairweather and Karageorghis ('98)*) For the Dirichlet problem for the Laplace equation on the circle $(0, \rho)$, boundary function is analytical, solution is analytically harmonic continuable to the whole plane:

$$\|u - u_M\|_{L^\infty(\Omega)} \leq C \left(\frac{\rho}{R}\right)^M$$

(*Kitagawa ('88, '91)*) u is not analytically continuable to the whole plane, but rather only up to an extension $B(0; r_0)$

$$\|u - u_M\|_{L^\infty(\Omega)} \leq \|u\|_{L^\infty(\partial B(0, r_0))} \left(\frac{2}{1 - \frac{\rho}{R}}\right) \left[(1 + A(R, \rho)) \left(\frac{\rho}{r_0}\right)^{M/3} + 4 \left(\frac{\rho}{R}\right)^{M/3} \right]$$

where $A(R, \rho)$ is some constant between 1 and 2.

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where $A(R, \rho)$ is some constant between 1 and 2.

- What is the optimal **distribution** of the boundary and outer points?
- What is the optimal **distance** of the outer points from the boundary?
- What is the optimal number (or rather: the **ratio**) of the boundary and outer points?
- Does the two approaches deliver **similar accuracy**?
- With an optimal choice of all parameters, which **convergence rate** can be achieved?
- What is the setup and the parameters in a **neural network** used in the computations?

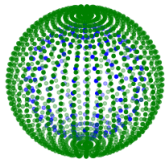
Example

$$\begin{cases} \Delta^2 u - 3u = 0 & \text{on } \Omega \\ u(x, y, z) = \sinh(x + y + z) & \text{on } \partial\Omega \end{cases}$$

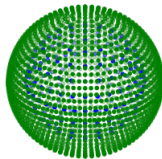
where $\Omega = \{(x, y, z) | x^2 + y^2 + z^2 < 1\}$.

Nodal distributions

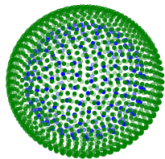
● Boundary points
● Outer points



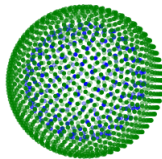
● Boundary points
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● Boundary points
● Outer points



Nodal distributions

| Collocation points | Error(dis.1) | Error(dis.2) | Error(dis.3) | Error(dis.4) |
|--------------------|--------------|--------------|--------------|--------------|
| (0.6, 0.6, 0.1) | -0.007686 | 0.002031 | 0.001083 | -1.168*e-5 |
| (0.3, 0.3, 0.3) | -0.021422 | 0.005856 | 0.002753 | -0.000319 |
| (0, 0, 0) | -0.028143 | 0.007756 | 0.003630 | -0.000422 |
| (-.6, .7, -.3) | -0.000238 | 0.000237 | -0.000335 | -0.000581 |

Table: Different nodal distribution leads to different error distribution

Example

Let us consider the Laplace equation on the Amoeba-like domain where the boundary points have the position of

$$\left(e^{\sin \theta} \sin^2(2\theta) + e^{\cos \theta} \cos^2(2\theta) \right) (\cos \theta, \sin \theta).$$

The boundary conditions are given by the Dirichlet boundary condition if $0 \leq \theta < \pi$ and the Neumann boundary condition if that $0 \leq \theta < 2\pi$ such that the analytical solution is

$$u(x, y) = \cos(x) \cosh(y) + \sin(x) \sinh(y)$$

Irregular domains

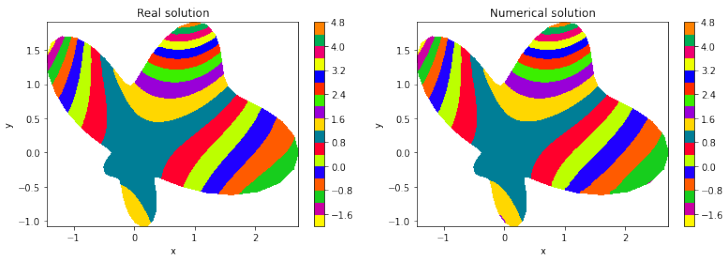


Figure: The analytical and numerical solution on Amoeba-like domain

| Collocation points | $(0, 0)$ | $(1, 1)$ | $(2, 0)$ | $(-1, 1)$ |
|---------------------|-----------|----------|-----------|-----------|
| Analytical solution | 1 | 1.822627 | -0.416146 | -0.155167 |
| Numerical solution | 1.0056722 | 1.823955 | -0.437396 | -0.15932 |

Table: The numerical solution of some points on the mixed boundary problem using the second approach.

Example

In a unit square Ω let us consider Laplace equation with Dirichlet boundary

$$\begin{cases} u(x, 0) = 0, & u(x, 1) = \sin(\pi x) & \text{for } 0 < x < 1 \\ u(0, y) = 0, & u(1, y) = 0 & \text{for } 0 < y < 1. \end{cases} \quad (4)$$

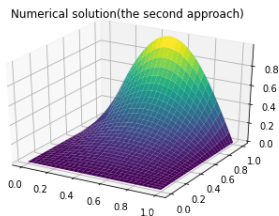
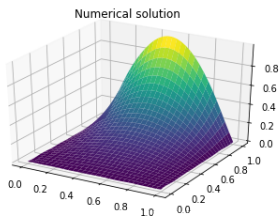


Figure: Numerical solutions are solved by the first approach (left) and the second approach (right) of example on the square

Distance between two surfaces

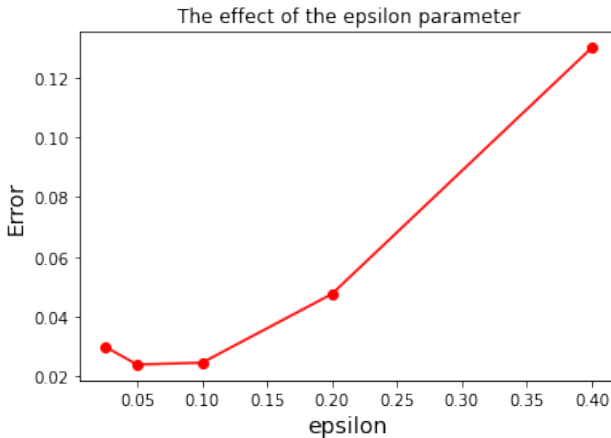


Figure: The effect of ϵ (the first approach)

The number of boundary and outer points

| The number of boundary and outer points | Error (Frobenius norm) |
|---|------------------------|
| $n = 16, m = 64$ | 0.2461622809658327 |
| $n = 32, m = 64$ | 0.3182853004347026 |
| $n = 32, m = 128$ | 0.21974143048835057 |
| $n = 16, m = 128$ | 0.18014076148115063 |
| $n = 16, m = 512$ | 0.10919759728510069 |
| $n = 16, m = 2048$ | 0.07990843843706859 |
| $n = 32, m = 8192$ | 0.04762830527750495 |

Table: The effect of the number of boundary points and outer points (the first approach)

The number of boundary and outer points

| Collocation points | True solution | (100,100) | (400, 400) | (1000, 100) | (100, 1000) |
|--------------------|---------------|-----------|------------|-------------|-------------|
| (0.2, 0.2) | 0.034124 | 0.031156 | 0.033824 | 0.0342969 | 0.0341972 |
| (0.5, 0.5) | 0.199268 | 0.200808 | 0.200867 | 0.1984138 | 0.1958812 |
| (0.7, 0.7) | 0.311947 | 0.314457 | 0.311565 | 0.311674 | 0.306193 |
| (0.2, 0.7) | 0.226643 | 0.224749 | 0.223706 | 0.2267211 | 0.222510 |
| (0.7, 0.2) | 0.046969 | 0.045743 | 0.046496 | 0.0466224 | 0.0464201 |

Table: The different choices of the number of boundary points and outer points, $\epsilon = 0.15$ (the second approach)

Neural Networks Structures

| Neural network | Error |
|--|----------|
| Linear Regression, no hidden layer, epochs = 1000 | 0.048206 |
| Neural network, 1 hidden layer of size 100, epochs = 1000 | 0.026957 |
| Neural network, 1 hidden layer of size 500, epochs = 1000 | 0.026286 |
| Neural network, 2 hidden layers of size 100, epochs = 1000 | 0.022334 |
| CNN including convolution of 2 FC layers, epochs = 10000 | 0.017556 |

Table: Different neural network structures for example on the square.(the first approach)

- If the fundamental solution is not continuous? (E.g. The wave equation)
- If the equations do not have fundamental solutions?
- Error estimation on Sobolev space.
- Application on the acoustic problem.
- Application on moving boundary problem.

THANK YOU FOR YOUR ATTENTION!