TREFFTZ METHOD FOR AN INVERSE GEOMETRY PROBLEM IN STEADY-STATE HEAT CONDUCTION

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Abstract. The paper addresses a boundary identification problem in two-dimensional steady-state heat conduction. The proposed approach based on the Trefftz method allows one to reconstruct the unknown part of a regular domain boundary from the given temperature measurements on it, provided that both the temperature and heat flux on the remaining part of the boundary are known. The reconstruction of an unknown boundary is done through successive approximations with a polynomial or a truncated Fourier series. The proposed solution method, whose merit lies in the avoidance of large systems of nonlinear equations, is fast converging, accurate and numerically stable, as demonstrated in the included numerical examples.

Keywords: inverse geometry problem, boundary identification, Trefftz method

1. Introduction

An important category of problems considered in science and engineering are inverse problems. When the system geometry is not fully known but is being investigated, such problems are referred to as inverse geometry problems (IGPs). Some typical examples could be determining a moving interface between liquid and solid phases or nondestructive detection of internal cracks or cavities in materials. According to the classification proposed in [1], inverse analysis concerning IGPs can be applied to problems related to (i) shape and design optimization, (ii) identification of defects and (iii) identification of an unknown part of the boundary. The paper focuses attention on the latter.

Most of the studies concerning this task are devoted to problems described by Laplace’s equation. As known, Laplace’s equation characterizes a large group of physical problems. The present study refers specifically to steady-state isotropic heat conduction.

Numerical reconstruction of the unknown boundary can be achieved by a variety of computational approaches, e.g. the radial basis functions method [2], the boundary element regularization method [3], the smoothed grid finite element method [4]
or the level set method \[5\] - to name a few. As the present study employs the Trefftz method, we focus particular attention to the approaches applying this method or related techniques. The modified collocation Trefftz method for a boundary detection problem was used in \[6, 7\] and the same technique was applied for an inverse geometric problem governed by the biharmonic equation \[8\] and the modified Helmholtz equation \[9\]. The regularized collocation Trefftz approach was successfully tested on different shapes in a two-dimensional void detection problem \[10\]. The method of fundamental solutions, which may be viewed as one of the Trefftz methods, also belongs to computational techniques effective in IGPs, see e.g. \[11, 12\].

The cited references concerning the solution of a boundary identification problem by the (modified) collocation Trefftz method propose a discrete reconstruction of an unknown boundary. In contrast, the present study uses a polynomial or a Fourier expansion of the unknown boundary. The computational procedure first outputs the solution of Laplace’s equation by the Trefftz method. The further step consists in specification of the coefficients of a polynomial or a Fourier representation of the unknown boundary so as to best fit the calculated temperatures to their known values. The latter requires solving systems of nonlinear algebraic equations but with relatively few unknowns which is the specific value of the approach.

2. Boundary identification problem

We consider a problem in a finite two-dimensional region \( \Omega \) whose boundary \( \partial \Omega \) is composed of two disjoint curves \( \Gamma \) and \( \gamma \):

\[
\partial \Omega = \Gamma \cup \gamma
\]

such that \( \Gamma \) is known and \( \gamma \) unknown. Highly irregular shapes of \( \Omega \) will not be addressed in this paper as they usually require a division into subdomains and local solutions, while the present approach uses global Trefftz-type solutions which could fail in complex geometries. We assume that \( \gamma \) is smooth and can be described by a function \( y = y(x) \) or \( r = r(\theta) \), depending on the type of coordinates. In spite of these constraints on the topology of \( \Omega \), the proposed algorithm allows a great variety of shapes occurring in real-world applications.

The discussed problem is that of stationary heat flow with no heat sources or sinks, so the governing equation is

\[
\Delta T = 0 \quad \text{in} \; \Omega \quad (\Delta \text{ - Laplace operator})
\]

which we solve subject to

\[
T = T_\gamma \quad \text{on} \; \Gamma
\]
\[ \frac{\partial T}{\partial n} = q_f \quad \text{on } \Gamma \]  

(4)

where \( \partial/\partial n \) denotes differentiation along the outward normal \( n \). Equations (3) and (4) represent an overspecified condition on \( \Gamma \) (known temperature and heat flux) in order to compensate for the missing information concerning geometric characteristics of the problem domain. On the remaining (unknown) part of the boundary, we impose a Dirichlet condition:

\[ T = T_f \quad \text{on } \gamma \]  

(5)

with \( T_f \) not necessarily constant. Since \( \gamma \) is being sought, condition (5) should be understood as a given temperature at the points \((x, y)\) or \((r, \theta)\) whose first coordinates are known while second coordinates are to be found.

3. Trefftz method

The Trefftz method, developed at the end of the 1920s, has recently become very popular. It approximates the solution of a differential equation with a linear combination of certain basis functions (named T-complete functions) satisfying the given differential equation. For Laplace’s equation in \((x, y)\)-coordinates, the proper basis is composed of the functions commonly known as harmonic polynomials:

\[ \left\{ 1, \text{Re} (x + iy)^k, \text{Im} (x + iy)^k, \quad k = 1, 2, \ldots \right\} \]  

(6)

In polar coordinates the corresponding basis functions are

\[ \left\{ 1, r^k \cos(k\theta), r^k \sin(k\theta), \quad k = 1, 2, \ldots \right\} \]  

(7a)

unless for to a problem in a multiple domain, when we use the bases:

\[ \left\{ 1, \ln(r), r^k \cos(k\theta), r^k \sin(k\theta), r^{-k} \cos(k\theta), r^{-k} \sin(k\theta) \quad k = 1, 2, \ldots \right\} \]  

(7b)

According to the logic of the Trefftz method, the approximate solution of (2)-(5), denoted \( T_{\text{app}} \), can be written as a linear combination of the form

\[ T_{\text{app}} = \sum_{k=0}^{K} c_k U_k \]  

(8)

with \( U_k \) denoting the T-complete functions of type (6), (7a) or (7b) and \( c_k \) - the coefficients to be found. In order to specify the coefficients \( c_k \) we minimize a functional
\( \Phi(c_0, c_1, ..., c_K) = \left[ \sum_{k=0}^{K} c_k U_k \right] - T_f \right] d\Gamma + \left[ \sum_{k=0}^{K} c_k \frac{\partial U_k}{\partial c_k} \right] - q_f \right] d\Gamma \) \hspace{1cm} (9)

which leads to a system of linear equations:

\( \frac{\partial \Phi}{\partial c_k} = 0, \quad k = 0, 1, ..., K \) \hspace{1cm} (10)

4. Boundary identification algorithm

The problem of identification of an unknown boundary \( \gamma \) can be set up in terms of searching for either \((x, y)\)- or \((r, \theta)\) -coordinates of the boundary points. Regardless of chosen coordinates, the proposed solution scheme assumes reconstruction of a boundary through its approximation by a function whose parameters need to be specified. The advantage of such an approach lies in reducing computational effort as we are able to recover infinitely many boundary points at the cost of finding only a few parameters of an approximating curve.

The boundary identification algorithm further described in this section can be summarized as follows:

STEP 1. Approximate the temperature \( T \) of a considered body with \( T_{\text{app}} \), using the Trefftz method.

STEP 2. Choose the form of an approximation to \( \gamma \): with a polynomial \( y = P_N(x) \) or a finite trigonometric series \( r = Q_N(\theta) \).

STEP 3. Choose \( \delta \) - the demanded level of accuracy in approximation of the given \( T \) on \( \gamma \) by \( T_{\text{app}} \) on the graph of \( P_N(x) \)/or \( Q_N(\theta) \).

STEP 4. Specify the coefficients of \( P_N(x) \)/or \( Q_N(\theta) \) for \( N = 1 \). If the required accuracy \( \delta \) is achieved with \( P_1(x) \)/or \( Q_1(\theta) \), stop. Otherwise go to STEP 5.

STEP 5. Increase \( N \) by 1. Then specify the coefficients of \( P_N(x) \)/or \( Q_N(\theta) \). If the required accuracy \( \delta \) is achieved with \( P_N(x) \)/or \( Q_N(\theta) \), stop. Otherwise repeat STEP 5.

4.1. Implementation in \((x, y)\)-coordinates

In the considered problem we assume that \( \gamma \) is a smooth curve \( y = f(x) \) which we can approximate with an \( N^{th} \) degree polynomial:

\[ P_N(x) = \sum_{n=0}^{N} a_n x^n, \quad x \in [a, b] \] \hspace{1cm} (11)
By construction, \( P_N(x) \) has to agree with \( \gamma \) at both ends (corresponding to \( x = a \) and \( x = b \)), hence to determine \( N + 1 \) coefficients of the polynomial \( P_N(x) \), we do not need to solve \( N + 1 \) equations but only \( N - 1 \).

Condition (5), when expressed in terms of approximate functions, gives

\[
T_{app}(x, P_N(x)) = T_\gamma(x)
\]  

which holds in the whole interval \([a, b]\) or at \( M \) selected points \( x_1, x_2, \ldots, x_M \).

In general, such a condition cannot be met in the strict sense, therefore we claim fulfillment in a variational sense, i.e. seeking the minimum of a functional

\[
\Phi(P_N(x)) = \left\| T_{app}(x, \sum_0^N a_n x^n) - T_\gamma(x) \right\|^2
\]  

where \( \| \| \) denotes \( L^2 \)-norm of a function or an \( M \)-dimensional vector. Minimization of \( \Phi(P_N(x)) \) leads to a system of nonlinear algebraic equations

\[
\frac{\partial \Phi}{\partial a_n} = 0, \quad n = 0, 1, \ldots, N
\]  

which we solve numerically. It is worth noting that the Trefftz method itself can approximate temperature but not shape. However, it is involved in polynomial boundary reconstruction through delivering the temperature \( T_{app} \), which appears in (13) and consequently in (14).

Once \( T_{app} \) is found, the polynomial approximations of \( \gamma \) will be determined in order of an increasing degree. Before further description of the procedure, for each polynomial \( P_N(x) \) we define a nonnegative number, \( E^T(N) \), which measures the relative error between the exact and calculated temperature on the unknown boundary:

\[
E^T(N) = \frac{\left\| T_{app}[x \rightarrow P_N(x)] - T_\gamma \right\|}{\left\| T_\gamma \right\|},
\]  

provided \( T_\gamma \neq 0 \). At the \( N^{th} \) step, \( E^T(N) \) will be compared with the maximum admissible error, \( \delta \), chosen prior to computation. The first reconstruction of \( \gamma \) is with a polynomial of the lowest degree (0 or 1). We check whether this initial approximation, say \( P_1(x) \), is sufficiently accurate, i.e. whether \( E^T(1) < \delta \). If so, we have the solution. Otherwise, we search for a polynomial of degree 2, whose coefficients minimize (13) and unless \( E^T(2) < \delta \), we construct (in the same
manner) the next polynomial approximations of degree $N = 3, 4, \ldots$, until we achieve $E^T(N) < \delta$ for some $N$. Note that we derive the coefficients of $P_N(x)$’s from nonlinear equations (14). Numerical procedures for solving nonlinear equations usually need a certain initial guess. If $N = 2$, we plot $\Phi(P_2(x))$ versus $a_2$ in order to suggest initial value for $a_2$, while $a_0$ and $a_1$ can be expressed as functions of $a_2$. For $N > 2$, one can take the $(N - 1)^{th}$ polynomial as an initial guess for finding the coefficients of the $N^{th}$ polynomial.

4.2. Implementation in $(r, \phi)$-coordinates

Described in polar coordinates, the unknown boundary is assumed to be the graph of a function $r = f(\theta)$. A reasonable approximation of $f(\theta)$ seems to be that by a finite series of trigonometric functions:

$$Q_N(\theta) = a_0 + \sum_{n=1}^{N} (a_n \cos(n\theta) + b_n \sin(n\theta)), \quad \theta \in [\theta_0, \theta_1]$$

(16)

in which we admit $\theta$ from 0 to $2\pi$. The adjustment of $a_n$’s and $b_n$’s runs similarly to the $(x,y)$ case, i.e. through minimization of a functional $\Phi(Q_N(\theta))$ analogous to (13). Finding $Q_N(\theta)$ requires more computation in comparison to finding a polynomial $P_N(x)$ as we have to specify $2N + 1$ parameters of the $N^{th}$ order approximation, whereas in (11) only $N + 1$.

The start-up approximation of $\gamma$ is of $0^{th}$ or $1^{st}$ order. When $f(\theta_1) = f(\theta_1)$, we first take $Q_0(\theta) = a_0$. Otherwise, we put $Q_1(\theta) = a_0 + a_1 \cos\theta + b_1 \sin\theta$ and specify e.g. $b_1$ through plotting $\Phi(Q_1(\theta))$ versus $b_1$ (while $a_0$ and $a_1$ can be expressed in terms of $b_1$). Further proceeding runs similar to that described in Section 4.1.

5. Numerical results

In this section, we present a couple of numerical examples. The computations were performed on numerically simulated (synthetic) temperature, as in the reference papers [6-11]. Such an approach allows convenient and accurate evaluation of the results because the exact solution is known.

For better assessment of the solution method, each example was based on a different analytical solution to Laplace’s equation. The measurements on $\gamma$ were simulated at $M$ points uniformly distributed on the given range of $x$ or $\theta$. We assume the algorithm stops when $E^T(N)$ attains $\sim 10^{-6}$. For assessing the accuracy of boundary reconstruction in measurable terms, we introduce the error $\varepsilon(N)$, defined by
\[ \varepsilon(N) = \frac{\| \gamma_N^{\text{app}} - \gamma \|}{\| \gamma \|} \]  

(17)

*Example 1.*

Let the problem domain be described by inequalities

\[
\Omega: \begin{cases} 
0 \leq x \leq 1 \\
0 \leq y \leq f(x)
\end{cases}
\]  

(18)

where

\[ f(x) = \frac{1 + 2x + \sin x + \cos(2\pi)}{x^3 + 4} \]  

(19)

will be the unknown boundary \( \gamma \) to be found. The exact temperature is \( T(x, y) = e^x \cos x + x + 1 \). We take \( T \)-complete functions of the type (6) with \( K = 14 \) in the expansion (8) and \( M = 30 \). The problem domain is presented in Figure 1(a) and a polynomial reconstruction of \( \gamma \) in Figure 1(b).

![Fig. 1. Problem domain \( \Omega \) (a) and boundary identification by a polynomial of degree \( N = 8 \) (b)](image)

Further evidence concerning accuracy and rate of convergence is given in Table 1.

<table>
<thead>
<tr>
<th>( N )</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>5</th>
<th>6</th>
<th>7</th>
<th>8</th>
</tr>
</thead>
<tbody>
<tr>
<td>( E^2(N) ), (%)</td>
<td>3.56</td>
<td>3.49</td>
<td>0.37</td>
<td>0.35</td>
<td>0.021</td>
<td>0.020</td>
<td>0.0008</td>
</tr>
<tr>
<td>( \varepsilon(N) ), (%)</td>
<td>12.14</td>
<td>11.80</td>
<td>1.26</td>
<td>1.17</td>
<td>0.073</td>
<td>0.066</td>
<td>0.0027</td>
</tr>
</tbody>
</table>
The errors listed in Table 1 show high accuracy and fast convergence of the solution procedure as only few iterations were sufficient for very accurate reconstruction of the unknown boundary.

Example 2.
This example refers to a region defined (in polar coordinates) by inequalities

\[ \begin{aligned}
\Omega: & \quad 0 \leq \theta < 2\pi \\
& \quad 0 \leq r \leq f(\theta)
\end{aligned} \tag{20} \]

with

\[ f(\theta) = \frac{3 + \cos(\theta - \pi/3)\sin 4\theta}{5 + \sin 2\theta} \tag{21} \]

When constrained to \( \theta \in [\pi, 2\pi] \), the function \( f(\theta) \) will be the unknown boundary to be identified. In this example we take \( T(x, y) = \cosh x \sin y + y + 4 \). The considered case requires T-complete functions (7a). We take \( K = 16 \) in (8) and \( M = 60 \). Figure 2a shows the domain boundaries (known and unknown) and Figure 2b compares the exact \( \gamma \) with its approximation corresponding to \( N = 8 \).

Table 2 provides with more information concerning convergence properties and accuracy of the algorithm.

<table>
<thead>
<tr>
<th>( N )</th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>5</th>
<th>6</th>
<th>7</th>
<th>8</th>
</tr>
</thead>
<tbody>
<tr>
<td>( E^T(N) ), %</td>
<td>4.21</td>
<td>3.87</td>
<td>1.45</td>
<td>0.24</td>
<td>0.027</td>
<td>0.007</td>
<td>0.002</td>
<td>0.0002</td>
</tr>
<tr>
<td>( \varepsilon(N) ), %</td>
<td>18.05</td>
<td>15.21</td>
<td>8.59</td>
<td>1.92</td>
<td>0.29</td>
<td>0.076</td>
<td>0.023</td>
<td>0.003</td>
</tr>
</tbody>
</table>
As in the previous example, we achieved very good accuracy of the boundary identification. Moreover, the small number of iterations before terminating the procedure indicates fast convergence.

**Example 3.**
We consider a doubly connected domain described in polar coordinates by

\[
\Omega: \begin{cases} 0 \leq \theta < 2\pi \\ f(\theta) \leq r \leq g(\theta) \end{cases}
\]  

where

\[
g(\theta) = \frac{4 - \cos \theta \sin 2\theta}{6 \sin 4\theta}
\]  

is its outer boundary where the temperature was known at \(M = 60\) points and

\[
f(\theta) = \frac{4 + \cos 3\theta}{12 + \cos^2 \theta \sin 2\theta}
\]  

is the unknown inner boundary. We generated the measurements from \(T(x, y) = e^x \cos x + 10xy + 3\). The suitable Trefftz-bases for this example are those of type (7b) and we take \(K = 37\). Figure 3 presents the boundaries of \(\Omega\) and the approximation of \(\gamma\) found by the algorithm.

![Fig. 3. Problem domain \(\Omega\) (a) and boundary identification by expansion (16) with \(N = 9\) (b)](image)

Table 3 includes information for assessing the obtained approximations.
Errors for stopping criterion, \( E^T(N) \), and errors of boundary detection, \( \varepsilon(N) \)

<table>
<thead>
<tr>
<th>( N )</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>5</th>
<th>6</th>
<th>7</th>
<th>8</th>
<th>9</th>
</tr>
</thead>
<tbody>
<tr>
<td>( E^T(N) ), %</td>
<td>2.94</td>
<td>0.30</td>
<td>0.073</td>
<td>0.034</td>
<td>0.031</td>
<td>0.003</td>
<td>0.0012</td>
<td>0.0004</td>
</tr>
<tr>
<td>( \varepsilon(N) ), %</td>
<td>20.09</td>
<td>1.68</td>
<td>0.466</td>
<td>0.235</td>
<td>0.236</td>
<td>0.023</td>
<td>0.014</td>
<td>0.0042</td>
</tr>
</tbody>
</table>

The present example, like the two already discussed, demonstrated very high accuracy of the final-order solution and fast convergence of the algorithm.

6. Sensitivity to input errors

The calculation was repeated using the input data disturbed by random errors. For that purpose we generated \( M \) normally distributed random numbers from the range between \(-3\%\) and \(3\%\). Such a level of noise was intended to create realistic test data. The evaluation of the influence of those noisy inputs on the final boundary reconstruction is through calculating a relative error \( \delta \) given by

\[
\delta = \frac{\left\| \gamma_{N}^{app} - \gamma_{N}^{app} \right\|}{\gamma_{N}^{app}}
\]  

(25)

where \( \gamma_{N}^{app} \) denotes the \( N \)-th-order approximation of \( \gamma \) obtained from noisy data. It seems crucial to analyze the highest order reconstruction provided by the algorithm in case the errors accumulated from iteration to iteration. Figure 4 presents the results of boundary identification from the noisy data. The depicted approximate profiles of \( \gamma \) are those obtained for \( N = 8 \) in Examples 1-2 and \( N = 9 \) in Example 3.

![Fig. 4. Results of boundary identification from data containing 3% noise](image-url)
Visual inspection of Figure 4 shows a fairly good agreement between exact and approximate positions of the corresponding boundary points. It indicates small impact of the measurement errors on the predicted locations of the unknown boundary points. In support of this conclusion, we present the errors (δ) between the reconstructions of γ from clean and noisy data (see Table 4).

<table>
<thead>
<tr>
<th>Example 1 (N = 8)</th>
<th>Example 2 (N = 8)</th>
<th>Example 3 (N = 9)</th>
</tr>
</thead>
<tbody>
<tr>
<td>δ [%]</td>
<td>2.06</td>
<td>3.01</td>
</tr>
<tr>
<td></td>
<td></td>
<td>4.06</td>
</tr>
</tbody>
</table>

The percentage changes in boundary approximations - when compared to the measurement errors - are of the same order of magnitude. In conclusion, at the assumed level of noise, the algorithm turned out to be stable.

7. Conclusions and final remarks

We proposed an algorithm for identification of the unknown part of a domain boundary. Unlike the existing approaches based on the Trefftz method, the present algorithm offers a continuous solution which accurately recovers the unknown shape after a few iterations, as demonstrated in the included examples. Small random errors introduced to the input data resulted in comparably small disturbances in the final boundary reconstructions, which proved numerical stability of the solutions. Another advantage of the method is that it separates specification of the coefficients: those referring to temperature approximation are obtained from a linear system of equations while the coefficients of a shape approximation - from a distinct nonlinear system of equations with relatively few unknowns. That way, large (and numerically problematic) systems of nonlinear equations can be avoided. Finally, the proposed method can be extended without changes on boundary identification problems governed by other linear differential equations for which the T-complete functions are known or can be somehow generated.

References


