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Published in: Inverse Problems in Science and Engineering

DOI: 10.1080/17415977.2011.590897

2011


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RESEARCH ARTICLE

Fast reconstruction of harmonic functions from Cauchy data
using the Dirichlet-to-Neumann map and integral equations

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(September 2010)

We propose and investigate a method for the stable determination of a harmonic function from knowledge of its value and its normal derivative on a part of the boundary of the (bounded) solution domain (Cauchy problem). We reformulate the Cauchy problem as an operator equation on the boundary using the Dirichlet-to-Neumann map. To discretize the obtained operator, we modify and employ a method denoted as "Classic II" given in [15, Section 3], which is based on Fredholm integral equations and Nyström discretization schemes. Then, for stability reasons, to solve the discretized integral equation we use the method of smoothing projection introduced in [17, Section 7], which makes it possible to solve the discretized operator equation in a stable way with minor computational cost and high accuracy. With this approach, for sufficiently smooth Cauchy data, also the normal derivative can be accurately computed on the part of the boundary where no data is initially given.

Keywords: Alternating method; Cauchy problem; Dirichlet-to-Neumann map; Laplace equation; Second kind boundary integral equation.

AMS Subject Classification: 35R25; 65N20; 65N35; 31A10; 31A05

1. Introduction

The stable reconstruction of a harmonic function from given Cauchy data is a problem of fundamental importance in many engineering applications in fluid and heat flow, such as in non-destructive testing and tomography, see, for example, [9, 14, 24, 31, 32]. The governing model is the Laplace equation with overspecified data given on a part (arc) of the boundary of the solution domain in the form of the solution and its normal derivative; the solution $u$ satisfies

\[
\begin{align*}
\Delta u &= 0 \quad \text{in } \Omega, \\
\frac{\partial u}{\partial \nu} &= g_C \quad \text{on } \Gamma_C, \\
u &= f_C \quad \text{on } \Gamma_C.
\end{align*}
\]

We assume here that $\Omega$ is a planar bounded Lipschitz domain in $\mathbb{R}^2$ with $\Gamma_C$ an (open) arc of the boundary $\Gamma = \partial \Omega$, and define $\Gamma_U = \Gamma \setminus \Gamma_C$. The element $\nu$ is the outward unit normal to the boundary $\Gamma$. On the boundary part $\Gamma_U$, the solution and its normal derivative are unknown and have to be reconstructed. We assume

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that the given Cauchy data $f_C$ and $g_C$ are sufficiently smooth and compatible such that there exists a solution $u$. Note that uniqueness of the solution is well established, see, for example [5, 7]. It is well-known that the Cauchy problem (1) is ill-posed and thus measurement errors in the data can completely destroy the reconstructions unless regularizing methods are employed.

Reconstruction of $u$ in (1) is a classical problem and there are therefore many different numerical methods in the literature for its solution, some are listed in [6]. In 1989, Kozlov and Maz’ya [22, 23] proposed the alternating method, which is an iterative procedure for the reconstruction of the solution $u$. This method preserves the governing Laplace operator and the regularizing character is achieved by appropriate change of the boundary conditions. The alternating method has successfully been employed to several applied problems, see, for example, [2–4, 8, 10, 13, 18–21, 25, 27–29]. However, in most studies it has been reported that this procedure can be time-consuming and that non-accurate reconstructions of the normal derivative are obtained. There are more involved methods that can be more efficient, see [11] and [12].

Recently, in [17], the authors of the present paper investigated ways of implementing the alternating method to speed up convergence and minimize the computational cost. The authors took advantage of the reformulation of the alternating method in terms of an operator equation on the boundary, together with a recent integral equation method [15, 16]. Inspired and encouraged by those results, we reformulate the Cauchy problem as another operator equation on the boundary. In the literature, the most straightforward reformulation based on the Dirichlet-to-Neumann map seems to have been overlooked and we therefore present this reformulation in this paper and shall compare the obtained results with those in [17]. Note that, however, a method in this direction was given in [33], where the Cauchy problem (1) was discretized using the boundary element method, and the corresponding linear system of equations was regularized using various techniques including Tikhonov regularization. Moreover, there is also a recent investigation on an effective way to numerically implement the Dirichlet-to-Neumann map, see [15], making our present approach for the Cauchy problem timely. We point out that our focus is to produce a fast method that is straightforward to implement and has high accuracy. The obtained approximation could perhaps then be used as a priori information in methods like [6, 22].

To discretize the obtained operator, we modify and employ a method denoted as Classic II given in [15, Section 3]. This method was originally given to compute the Dirichlet-to-Neumann map on the boundary of a two-dimensional domain exterior to a single contour, and is based on Fredholm integral equations and Nyström discretization schemes. We outline how to adjust the method to our case. Then, for stability reasons due to the ill-posedness of the Cauchy problem (1), to solve the discretized integral equation we use the method of smoothing projection introduced in [17, Section 7], which makes it possible to solve the discretized operator equation in a stable way with minor computational cost and high accuracy.

For the outline of this paper, in Section 2, we present the reformulation of the Cauchy problem and point out some properties of the Dirichlet-to-Neumann map and its relation to the Poincaré-Steklov operator. In Section 3, geometry and parameters for the numerical investigations are presented. The numerical method for the discretization of the operator from Section 2, based on Fredholm integral equations and Nyström discretization schemes, are given in Section 4. In Section 5, we show how to numerically construct the solution and its normal derivative on $\Gamma_U$, and give some numerical results including noisy data. Conclusions are found in Section 6.
2. Reformulation of the Cauchy problem (1)

2.1. The Dirichlet-to-Neumann map

Let $L^2(\Omega)$ be the standard $L^2$-space with the standard norm. As usual, $H^1(\Omega)$ is the Sobolev space of real-valued functions in $\Omega$ with finite norm given by the relation $\|u\|_{H^1(\Omega)}^2 = \|u\|_{L^2(\Omega)}^2 + \|\nabla u\|_{L^2(\Omega)}^2$, where $\nabla = (\partial_{x_1}, \partial_{x_2})$.

For trace spaces, we recall that the space of traces of functions from $H^1(\Omega)$ on $\Gamma$ is $H^{1/2}(\Gamma)$. Restrictions of elements in $H^{1/2}(\Gamma)\times H^{-1/2}(\Gamma)$ constitute the space $H^{1/2}(\Gamma)\times H^{-1/2}(\Gamma)$.

We then recall some facts about the Dirichlet-to-Neumann map, for the proofs, we refer to [26]. Given $f \in H^{1/2}(\Gamma)$, the Dirichlet problem

$$\begin{cases}
\Delta u = 0 & \text{in } \Omega, \\
u = f & \text{on } \Gamma,
\end{cases}$$

has a unique solution in $H^1(\Omega)$. Moreover, the normal derivative $g = \partial u/\partial \nu$ of $u$ on $\Gamma$ is well-defined as an element in the dual space $H^{-1/2}(\Gamma)$. The operator $D$ that maps $f$ to $g$ is a bounded operator denoting the Dirichlet-to-Neumann map. Furthermore,

$$(Df, f_1)_{H^{-1/2}(\Gamma)\times H^{1/2}(\Gamma)} = (f, Df_1)_{H^{1/2}(\Gamma)\times H^{-1/2}(\Gamma)},$$

where $(\cdot, \cdot)_{H^{-1/2}(\Gamma)\times H^{1/2}(\Gamma)}$ is the duality pairing between the space $H^{1/2}(\Gamma)$ and $H^{-1/2}(\Gamma)$ induced by the scalar product in $L^2(\Gamma)$. For connections between the Dirichlet-to-Neumann map and the Poincaré-Steklov operator, see [1] and [30].

2.2. Reformulation of (1)

To reformulate problem (1), we use restrictions of the Dirichlet-to-Neumann map to the respective boundary part. First, let the operator $A_{CU}$ be defined such that $A_{CU}f_U$ is the normal derivative on $\Gamma_C$ of the solution to

$$\begin{cases}
\Delta u = 0 & \text{in } \Omega, \\
u = f_U & \text{on } \Gamma_U, \\
u = 0 & \text{on } \Gamma_C.
\end{cases}$$

Similarly, let $A_{CC}$ be defined such that $A_{CC}f_C$ is the normal derivative on $\Gamma_C$ of the solution to

$$\begin{cases}
\Delta u = 0 & \text{in } \Omega, \\
u = 0 & \text{on } \Gamma_U, \\
u = f_C & \text{on } \Gamma_C.
\end{cases}$$

Then the Cauchy problem (1) is equivalent to solving the following operator equation on the boundary

$$A_{CU}f_U = g_C - A_{CC}f_C.$$
Note that in both these problems (4) and (5), we possibly have discontinuous Dirichlet data. Thus, suitable spaces for the data are then the corresponding $L^2$-spaces on the respective boundary part, and the corresponding solution will then be in $L^2(\Omega)$. The normal derivative on $\Gamma_C$ in (4) has meaning since local regularity results for elliptic equations clearly imply that $u$ has, locally, derivatives of second order near $\Gamma_C$ due to the zero Dirichlet condition imposed on $\Gamma_C$. Similarly, since we assumed that Cauchy data in (1) is sufficiently smooth, the normal derivative on $\Gamma_U$ exists in (5).

Now, it is straightforward, using the pairing (3), to obtain the following.

**Theorem 2.1:** The adjoint operator $A^*_{CU}$ of $A_{CU}$ is defined by

$$A^*_{CU} f_C = \frac{\partial u}{\partial \nu} |_{\Gamma_U},$$

where $u$ is the solution to (5).

With knowledge of the adjoint operator, we can then, for example, employ Tikhonov regularization to (6) and obtain a stable approximation $f_{U,\lambda}$ from

$$(A^*_{CU} A_{CU} + \lambda I) f_{U,\lambda} = A^*_{CU} (g_C - A_{CC} f_C).$$

Alternatively, iterative methods, such as the Landweber-Fridman or conjugate gradient methods can be employed to solve (6). However, since our focus is to develop a fast method that is straightforward to implement and has high accuracy, to solve (6) we shall instead employ a recent method based on smoothing projections that was introduced in [17, Section 7].

Provided data is smooth enough, the normal derivative, $g_U$, also exists on $\Gamma_U$ of the solution to the Cauchy problem (1). To construct it, let $A_{UU} f_U$ be the normal derivative of the solution to (4) on $\Gamma_U$, and let $A_{UC} f_C$ be the normal derivative of the solution to (5) on $\Gamma_U$. Then

$$g_U = A_{UU} f_U + A_{UC} f_C.$$  \hfill (7)

Note that from Theorem 2.1, we have $A_{UC} = A^*_{CU}$.

### 3. Configuration for the numerics

To compare results, we shall use the same configuration as in [17] and recall its definition below. We make no distinction between points in the real plane $\mathbb{R}^2$ and points in the complex plane $C$, thus all points are denoted $z$ or $\tau$. The solution domain that we use for the numerical experiments is a bounded domain enclosed by a curve with the parameterization

$$\tau(t) = (1 + 0.1 \cos 5t)e^{it}, \quad -\pi < t \leq \pi,$$  \hfill (8)

see Figure 1. Note that this geometry is not trivial since its curvature is varying. The two arcs $\Gamma_U$ and $\Gamma_C$ are defined by

$$\tau(t) \in \Gamma_U, \quad -\pi < t < -\frac{\pi}{2}, \quad \text{and} \quad \tau(t) \in \Gamma_C, \quad \frac{\pi}{2} < t < \pi.$$  \hfill (9)
and the closure of these parts have two points in common, \( \gamma_1 = \tau(\pi) \) and \( \gamma_2 = \tau(-\pi/2) \). The Cauchy data is generated from the harmonic function

\[
u(z) = \Re\left\{ \frac{1}{z - S_1} \right\},
\]

where \( S_1 = 1.4 + 1.4i \), see Figure 1.

All numerical experiments presented are performed in MATLAB version 7.9 and executed on an ordinary workstation equipped with an Intel Core2 Duo E8400 CPU at 3.00 GHz.

![Figure 1. The solution domain \( \Omega \) with boundary \( \Gamma = \Gamma_U \cup \Gamma_C \) given by (8) and (9). The arcs \( \Gamma_U \) and \( \Gamma_C \) meet at the two points \( \gamma_1 \) and \( \gamma_2 \). A total of 256 discretization points are constructed on \( \Gamma \), 64 of which are located on \( \Gamma_U \). A source \( S_1 \), for the generation of Cauchy data via (10), is marked by ‘∗’.](image.png)

4. Discretization

We discretize problem (1) via the reformulation in Section 2.2 using a variant of an integral equation scheme originally designed for the fast and accurate computation of the Dirichlet-to-Neumann map on the boundary of a two-dimensional domain exterior to a single contour. The original scheme is denoted as Classic II in [15, Section 3], and we now briefly review the main steps of our variant of that method.

The solution \( u(z) \) is represented in terms of an unknown layer density \( \rho(z) \) on \( \Gamma \) via a double-layer potential. Enforcing Dirichlet boundary conditions \( f(z) \) on \( \Gamma \) leads to a Fredholm second kind integral equation

\[
(I + k) \rho(z) = f(z), \quad z \in \Gamma,
\]
where the action of the compact integral operator $k$ on $\rho(z)$ is given by

$$k\rho(z) = \frac{1}{\pi} \int_{\Gamma} \rho(\tau) \Im \left\{ \frac{d\tau}{\tau - z} \right\}. \quad (12)$$

Once (11) is solved for $\rho(z)$, the normal derivative of $u(z)$ at $\Gamma$ can be computed by applying a Cauchy-singular integro-differential operator $K$ to $\rho(z)$

$$\frac{\partial u}{\partial \nu}(z) = K\rho(z), \quad z \in \Gamma. \quad (13)$$

The action of $K$ on $\rho(z)$ is given by

$$K\rho(z) = \Im \left\{ \frac{\nu(z)}{\pi} \int_{\Gamma} \frac{\rho'(\tau) d\tau}{\tau - z} \right\}, \quad (14)$$

where the differentiation $\rho'(z) = d\rho(z)/dz$ is along the tangent to $\Gamma$ and $\nu(z)$ is the outward unit normal to $\Gamma$ at $z$.

We discretize the operators $k$ and $K$ using a Nyström scheme based on the composite trapezoidal quadrature rule. We use 256 discretization points on $\Gamma$, of which the 64 first points are located on $\Gamma_U$ and the remaining 192 points are located on $\Gamma_C$. The Cauchy-singular integral in (14) is to be interpreted in the principal value sense and we use the method denoted *global regularization* in [16] to achieve this. The Fast Fourier Transform, carried out with MATLAB’s built-in functions `fft` and `ifft`, is used for differentiation. The discretization results in two square matrices $k$ and $K$, both of dimension $256 \times 256$.

Now define the matrix $A$ as the composition

$$A = K(I + k)^{-1}, \quad (15)$$

where $I$ is the identity matrix. Clearly, $A$ is a discretization of the Dirichlet-to-Neumann map given in Section 2.1, and if we let $f$ and $g$ be column vectors containing the values of $u(z)$ and $\partial u(z)/\partial \nu$ at the discretization points we have

$$Af = g. \quad (16)$$

On partitioned form, where we have separated points on $\Gamma_C$ from points on $\Gamma_U$, this relation reads

$$\begin{bmatrix} A_{UU} & A_{UC} \\ A_{CU} & A_{CC} \end{bmatrix} \begin{bmatrix} f_U \\ f_C \end{bmatrix} = \begin{bmatrix} g_U \\ g_C \end{bmatrix}. \quad (17)$$

Thus, in order to get the discretizations of the operators $A_{UU}$, $A_{UC}$, $A_{CU}$, and $A_{CC}$ defined in Section 2.2, one only has to pick the appropriate blocks from the matrix $A$ in (16) via (17). Computing all the entries of $A$ takes about 0.03 seconds on the workstation mentioned in Section 3.

5. Solving for $f_U$ and $g_U$

Now that we have access to the discretized operators $A_{UU}$, $A_{UC}$, $A_{CU}$, $A_{CC}$, as well as the known data $f_C$ and $g_C$, it is easy to solve the discretized counterpart of (6) for $f_U$ and subsequently use the discretized counterpart of (7) for $g_U$. We
concentrate on the setup detailed in Section 3. The reference solutions, that is, the correct analytical values for $f_U$ and $g_U$ are shown in Figure 2.

Rather than using (6) as it stands we shall, for stability reasons, use the method of smoothing projections introduced in [17, Section 7]. We represent $f_U$ in terms of $n$ coefficients $\hat{f}_U$ in a monomial basis on the canonical interval $[-1, 1]$

$$f_U = V_n \hat{f}_U,$$

where $V_n$ is a $64 \times n$ Vandermonde matrix. Thus, we first solve (in the least squares sense) the $192 \times n$ overdetermined linear system

$$(A_C V_n) \hat{f}_U = g_C - A_C f_C$$

for the unknown $\hat{f}_U$. Then we use (18) to obtain $f_U$. Finally, we compute $g_U$ from

$$g_U = A_U f_U + A_U f_C.$$

Doing this for $n = 1, 2, \ldots, 32$, takes an additional 0.01 seconds on the workstation mentioned in Section 3. The excellent quality of the reconstruction is shown in Figure 3. Comparison with Figure 7 in [17] shows an improvement in achievable accuracy with between one and two digits. To further illustrate this and to make it easier to compare, the results from Figure 7 in [17] are also included in Figure 3.

It is also of interest to add noise to the Cauchy data to verify the stability of the method. Naturally, the more noise that is added, the less accurate the reconstruction will be. Furthermore, the quality of the reconstruction varies between realizations and if a very large number of points are sampled on $\Gamma_C$, one could try filtering the Cauchy data to reduce the noise. Here, we shall ignore such issues and simply add Gaussian noise with mean zero and standard deviation $

$$\sigma = 0.01 \cdot \max |f_C|$$

to $f_C$. The element $f_C$ and the corresponding noisy data are shown in Figure 4(a); no noise is added to $g_C$. As it turns out, a low degree basis with $n = 2$ for $f_U$ most often gives the best reconstruction at this high level of noise. A typical example is
shown in Figure 4(b). The reconstruction of the derivative is more inaccurate as expected.

6. Conclusion

We have proposed and investigated a method for the stable reconstruction of a harmonic function from Cauchy data. The aim was to produce a fast method that
is straightforward to implement and has high accuracy. To achieve this, the Cauchy problem was rewritten as an operator equation on the boundary using the Dirichlet-to-Neumann map. To discretize the obtained operator, we modified and employed a method denoted as Classic II in [14, Section 3]. This method was originally given to compute the Dirichlet-to-Neumann map on the boundary of a two-dimensional domain exterior to a single contour, and is based on Fredholm integral equations and Nyström discretization schemes. For stability reasons, to solve the discretized integral equation, we used the method of smoothing projection introduced in [16, Section 7], which makes it possible to solve the discretized operator equation in a stable way with minor computational cost and high accuracy. A numerical example was investigated in a bounded domain having a non-trivial boundary (curvature is varying). Compared with the numerical results in [16], using the proposed approach, we obtain a higher accuracy in the reconstructed function values and normal derivatives, of between one and two digits. Stability against noise in the data was also investigated, showing that a stable solution can be obtained with increasing accuracy as the noise is decreasing. Moreover, the present approach is more efficient and faster, and also more straightforward to implement requiring a small amount of computer code.

Acknowledgment The first author was supported by the Swedish Research Council under the contract 621-2007-6234.

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