RECONSTRUCTION OF A SPACE-TIME-DEPENDENT SOURCE IN SUBDIFFUSION MODELS VIA A PERTURBATION APPROACH*  

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Abstract. In this article we study two inverse problems of recovering a space-time-dependent source component from the lateral boundary observation in a subdiffusion model. The mathematical model involves a Djrbashian–Caputo fractional derivative of order \( \alpha \in (0,1) \) in time, and a second-order elliptic operator with time-dependent coefficients. We establish a well-posedness and a conditional stability result for the inverse problems using a novel perturbation argument and refined regularity estimates of the associated direct problem. Further, we present a numerical algorithm for efficiently and accurately reconstructing the source component, and we provide several two-dimensional numerical results showing the feasibility of the recovery.  

Key words. inverse source problem, subdiffusion, time-dependent coefficient, conditional stability, reconstruction  

AMS subject classifications. 35R30, 35R11, 35B30, 65M32  

DOI. 10.1137/21M1397295  

1. Introduction. This work is concerned with inverse source problems (ISPs) of identifying a space-time-dependent component of the source in the subdiffusion model in a cylindrical domain from the lateral Cauchy data on a part of the boundary. Let \( d \geq 2, \Omega = \omega \times (-\ell,\ell), \omega \subset \mathbb{R}^{d-1} \) be an open bounded domain with a \( C^2 \) boundary, and fix \( T > 0 \) as the final time. For any \( x \in \Omega \), we write \( x = (x',x_d), \) with \( x' \in \omega \) and \( x_d \in (-\ell,\ell) \). For \( m = 0,1 \), we consider the following initial boundary value problem for the function \( u \):  

\[
\begin{cases}
\partial_t^\alpha u + A(t)u = F & \text{in } \Omega \times (0,T), \\
u(x,0) = 0 & \text{in } \Omega, \\
\partial_x^m u(x',\ell,t) = 0 & \text{on } \omega \times (0,T), \\
\partial_{x_d} u(x',-\ell,t) = 0 & \text{on } \omega \times (0,T), \\
u(x,t) = 0 & \text{on } \partial \omega \times (-\ell,\ell) \times (0,T). 
\end{cases}
\]

In the model (1.1), the order \( \alpha \in (0,1) \) is fixed, and the notation \( \partial_t^\alpha u \) denotes the so-called Djrbashian–Caputo fractional derivative of order \( \alpha \) in time, which, for \( \alpha \in (0,1) \), is defined by [26, p. 92]  

\[
\partial_t^\alpha u(t) = \frac{1}{\Gamma(1-\alpha)} \int_0^t (t-s)^{-\alpha} u'(s)ds,
\]

*Received by the editors February 5, 2021; accepted for publication (in revised form) May 10, 2021; published electronically August 10, 2021.  
https://doi.org/10.1137/21M1397295  
Funding: The work of the first author was partially supported by UK EPSRC grant EP/T000864/1. The work of the second author was supported by the French National Research Agency ANR (project MutiOnde) through grant ANR-17-CE40-0029. The work of the third author was supported by Hong Kong RGC grant 15304420.  
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where $\Gamma(z) = \int_0^\infty s^{z-1}e^{-s}ds$ for $\Re(z) > 0$ denotes Euler’s Gamma function (the notation $\Re$ denotes taking the real part of a complex number $z$). When the order $\alpha$ approaches $1^-$, the fractional derivative $\partial_t^\alpha u$ recovers the usual first-order derivative $u'(t)$, and, accordingly, the model coincides with the standard diffusion equation. $A(t)$ is a time-dependent second-order strongly elliptic operator, defined by

$$A(t)u(x) = -\sum_{i,j=1}^d \partial_{x_i}(a_{ij}(x,t)\partial_{x_j}u) + q(t)u, \quad x \in \Omega,$$

where $a = [a_{ij}]_{i,j=1}^d \in C(\overline{\Omega} \times [0,T]; \mathbb{R}^{d \times d})$ is a symmetric matrix-valued function and satisfies suitable regularity conditions given in Assumption 2.1 below, and $q \in C^1([0,T]; L^\infty(\Omega))$ is nonnegative.

The model (1.1) has received much attention in recent years, known by the name “subdiffusion” or “time-fractional diffusion,” due to its extraordinary capability for describing anomalously slow diffusion processes arising in a wide range of practical applications in physics, engineering, and biology. At a microscopic level, it can be derived from continuous-time random walk with a heavy-tailed waiting time distribution (with a divergent mean) in the sense that the probability density function of the walker appearing at time $t > b$ distribution (with a divergent mean) in the sense that the probability density function of the walker appearing at time $t > b$.

The ISPs of interest are to determine some information of the source $F$ from the measurement on a subboundary $\omega \times \{t\} \subset \partial \Omega$ of the domain $\Omega$. Note that the boundary measurement is insufficient to uniquely determine a general source $F$ (see, e.g., [23, section 1.3.1]), and additional assumptions have to be imposed on the source $F$ in order to restore unique recovery. Often it is formulated as recovering either a spatial or a temporal component of the source $F(x,t)$. In this work, the source $F$ is assumed to be of the form

$$F(x,t) = f(x',t)R(x,t).$$

Equation (1.2) can be interpreted as the condition that an unknown source $f(x',t)$ depends only on the depth variable $x'$ and $t$ in the case of $d = 2$, which corresponds to a layer structure, and on the planar location $(x_1,x_2)$ and $t$ but not on the depth in the case of $d = 3$, which can be a good approximation if the domain $\Omega$ is very thin in the direction of $x_3. Note that it arises also naturally in linearizing the inverse potential problem, where the potential coefficient $q$ depends on only $x'$ and $t$ [14]. We investigate the following two inverse problems: (i) ISPn is to recover $f(x',t)$ from the boundary observation $u|_{\omega \times \{t\} \times (0,T)}$ for $m = 1$ in (1.1), and (ii) ISPd is to recover $f(x',t)$ from the flux measurement $\partial_{x_d}u|_{\omega \times \{t\} \times (0,T)}$ for $m = 0$ in (1.1) (i.e., “n” and “d” refer to the Neumann and Dirichlet boundary condition, respectively, on the subboundary $\omega \times \{t\} \times (0,T)$ in the direct problem (1.1)).

This work is devoted to the theoretical analysis and numerical reconstruction of ISPn and ISPd. In Theorem 3.4, we prove a well-posedness result for ISPn in $L^2(0,T; L^2(\omega))$. This is achieved by combining the technique developed in [24], improved regularity estimates, and a novel perturbation argument from [19]. Further, in
Theorem 4.3, we establish a conditional stability result under an additional regularity condition on \( f(x', t) \) for ISPd. To the best of our knowledge, this is the first work rigorously analyzing ISPs of recovering a space-time-dependent source component in a subdiffusion model with time-dependent coefficients. The main technical challenges in the study include the nonlocality of the time-fractional derivative \( \partial_t^\alpha u \) and the time-dependence of the operator \( \mathcal{A}(t) \). The nonlocality essentially limits the solution regularity pickup (see, e.g., [38] and [17, Chapter 6]), and thus sharp regularity estimates for incompatible problem data are needed, which is especially delicate due to limited smoothness of the domain \( \Omega \). This is achieved in Proposition 4.1 by using a refined regularity pickup from [9, Lemma 2.4], exploiting the cylindrical structure of the domain \( \Omega \). The time dependence of the elliptic operator \( \mathcal{A}(t) \) precludes the application of the standard separation of variable technique that has been predominant in existing studies. This challenge is overcome by a perturbation argument and maximal \( L^p \) regularity for time-fractional problems, which plays an important role in the analysis of ISPd. In section 5, we derive the adjoint problem for computing the gradient of a quadratic misfit functional and analyze the regularity of the adjoint variable. Further, we describe the conjugate gradient algorithm for recovering \( f(x', t) \) and provide extensive numerical experiments to illustrate the feasibility of the recovery. The well-posedness and conditional stability results and the reconstruction algorithm represent the main contributions of this work.

Last we situate this work in the existing literature. ISPs of recovering partial information of the source \( F \) in a subdiffusion model from the lateral or terminal data represent an important class of applied inverse problems and have been extensively studied in the past decade. Most of the works devoted to this problem have been stated for sources \( F(x, t) = p(t)q(x) \) and can be divided into three groups: (i) inverse \( t \)-source problem of recovering \( p(t) \) [38, 41, 8, 29], (ii) inverse \( x \)-source problem of recovering \( q(x) \) [39, 43, 16, 36], and (iii) simultaneous inversion of spatial and temporal components [23, 37, 29, 27]. Within group (i), for example, using the decay property of the Mittag--Leffler function \( E_{\alpha,\beta}(z) \), a two-sided stability result of recovering \( p(t) \) was shown in [38] if the observation \( u(x_0, t) \) satisfies \( x_0 \in \text{supp}(q) \). Within group (ii), the unique recovery of the spatial component \( q(x) \) by interior observation was proved in [16] using Duhamel's principle and unique continuation principle, which also gave an iterative reconstruction algorithm. All these works in groups (i) and (ii) are concerned with recovering only either \( p(t) \) or \( q(x) \). The works in group (iii) are close to the current work. The work [23] showed the simultaneous recovery of \( p \) and \( q \) under suitable assumptions. For a two-dimensional heat equation, Rundell and Zhang [37] proved the unique recovery of both \( p \) and \( q \) in a semidiscrete setting (i.e., the temporal component \( p(t) \) is piecewise constant) from sparse observation on the boundary \( \partial \Omega \times (0, T) \). Li and Zhang [29] extended the analysis to the time-fractional model in two dimensions and established the uniqueness of recovering the unknown spatial component \( q(x) \), the time mesh, and the fractional order \( \alpha \) simultaneously from sparse data on the boundary \( \partial \Omega \times (0, T) \). We refer interested readers to the reviews [22, 31] for further pointers to theoretical and numerical results. See also the work of [27] for the unique recovery of a general source \( F \) from the full knowledge of the solution of problem (1.1), with \( \mathcal{A} \) independent of \( t \), on \( \Omega \times (T_1, T) \), with \( T_1 \in (0, T) \).

Kian and Yamamoto [24] proved the first uniqueness and stability results for the ISPs of recovering \( f(x', t) \) of the subdiffusion model in a cylindrical domain. (See also Isakov [14] for relevant results for the standard parabolic problem in the half space.) The analysis in [24] relies on some representation of solutions by means of \( E_{\alpha,\beta}(z) \) which are unavailable for elliptic operators with time-dependent coefficients.
This work extends the results in [24] to the case of the time-dependent diffusion coefficients, and further, by exploiting the maximal $L^p$ regularity, we substantially relax the regularity requirement on $f(x', t)$ for conditional stability.

Inverse problems for subdiffusion with time-dependent coefficients have been scarcely studied so far, due to a lack of mathematical tools, when compared with the time-independent counterpart. The only work we are aware of on an ISP with a time-dependent elliptic operator is [40], which showed the unique recovery of a spatial component from terminal measurement using an energy argument, which seems nontrivial to extend to the case $f(x', t)$. See also the works [44] for recovering a time-dependent factor in the diffusion coefficient $a(t)$, where the special structure does allow applying the established separation of variables technique. Thus, the theoretical analysis for ISPs in the case of time-dependent coefficients remains challenging.

This work extends the results in [24] to the case of the time-dependent diffusion coefficients, and further, by exploiting the maximal regularity, we substantially relax the regularity requirement on $f(x', t)$ for conditional stability.

The rest of the paper is organized as follows. In section 2, we state the assumptions and preliminary estimates. Then in sections 3 and 4 we prove the well-posedness of ISPn and conditional stability of ISPd, respectively. In section 5, we describe a numerical algorithm for recovering $f(x', t)$ for both ISPs and provide several numerical experiments to showcase the feasibility of the recovery. Throughout, the notation $c$ denotes a generic constant which may change at each occurrence, but it is always independent of the unknown source $f(x', t)$ or the associated solution $u$. For a bivariate function $g(x, t)$ or $g(x', t)$, we often abbreviate it to $g(t)$ as a vector-valued function by suppressing the dependence on the spatial variable.

2. Preliminaries: Assumptions and basic estimates. Now we collect several preliminary results. For $m = 0, 1$, we define two realizations $A(t)$ and $\hat{A}(t)$ in $L^2(\Omega)$ of the elliptic operator $A(t)$, with their domains given, respectively, by

\[
D(A(t)) = \{ v \in H^1_0(\Omega) : A(t)v \in L^2(\Omega) \},
\]

\[
D(\hat{A}(t)) = \{ v \in H^1(\Omega) : v|_{\partial \Omega \times (-\ell, \ell)} = 0, A(t)v \in L^2(\Omega), \partial_{x}^m v|_{x = \ell} = 0, \partial_{x}^m v|_{x = -\ell} = 0 \},
\]

and let $A_s = A(t_s)$ and $\hat{A}_s = \hat{A}(t_s)$ for any $t_s \in [0, T]$. Note that we abuse the notation $\hat{A}(t)$ for both $m = 0$ and $m = 1$, which will be clear from the context. For any $s \geq 0$, $A_s$ and $\hat{A}_s$ denote the fractional power of $A_s$ and $\hat{A}_s$ via spectral decomposition, and the associated graph norms by $\| \cdot \|_{D(A_s^\gamma)}$ and $\| \cdot \|_{D(\hat{A}_s^\gamma)}$, respectively. Let $E_s(t)$ and $\hat{E}_s(t)$ be the solution operators corresponding to the source $F$, associated with the elliptic operators $A_s$ and $\hat{A}_s$, respectively, defined by [21, section 3.1]

\[
E_s(t) := \frac{1}{2\pi i} \int_{\Gamma_{\theta, \delta}} e^{z(t^\alpha + A_s)}^{-1} dz \quad \text{and} \quad \hat{E}_s(t) := \frac{1}{2\pi i} \int_{\Gamma_{\theta, \delta}} e^{z(t^\alpha + \hat{A}_s)}^{-1} dz,
\]

with the contour $\Gamma_{\theta, \delta} \subset \mathbb{C}$ (oriented with an increasing imaginary part) given by

\[
\Gamma_{\theta, \delta} = \{ z \in \mathbb{C} : |\arg z| \leq \theta \} \cup \{ z \in \mathbb{C} : z = \rho e^{\pm i\theta}, \rho \geq \delta \}.
\]

Throughout, we fix $\theta \in (\pi/2, \pi)$ so that $z^\alpha \in \Sigma_{\alpha \theta}$ for $z \in \Sigma_{\theta} := \{ z \in \mathbb{C} \setminus \{0\} : |\arg(z)| \leq \theta \}$. Further, we employ the operator $\hat{S}_s(t)$ (corresponding to the initial data) defined by

\[
\hat{S}_s(t) := \frac{1}{2\pi i} \int_{\Gamma_{\theta, \delta}} e^{z(t^\alpha - 1)(z^\alpha + \hat{A}_s)}^{-1} dz.
\]
Then it is known that [21, equation (3.8)]

\[
\frac{d}{dt} \tilde{S}_s(t) = -\tilde{A}_s \tilde{E}_s(t).
\]

The next lemma summarizes the smoothing properties of \( E_s(t), \tilde{E}_s(t), \) and \( \tilde{S}_s(t). \) The notation \( \| \cdot \| \) denotes the operator norm on \( L^2(\Omega). \)

**Lemma 2.1** (see [21, Lemma 1]). For any \( \beta \in [0, 1], \) there hold for any \( t \in [0, T] \)

\[ t^{1+\alpha(\beta-1)} \| \tilde{A}_s E_s(t) \| \leq c \quad \text{and} \quad t^{1+\alpha(\beta-1)} \| \tilde{A}_s \tilde{E}_s(t) \| + t^{1+\alpha} \| \tilde{A}_s \tilde{E}_s(t) \| + t^{\beta \alpha} \| \tilde{A}_s \tilde{S}_s(t) \| \leq c. \]

Throughout, we make the following assumption on the diffusion coefficient matrix \( a. \) The regularity \( a \in C^1([0, T]; C^1(\Omega; \mathbb{R}^{d \times d})) \cap C([0, T]; C^3(\Omega; \mathbb{R}^{d \times d})) \) is sufficient for Lemma 2.3. (ii) is a structural condition to enable unique recovery. The notation \( \cdot \ast \) denotes the standard Euclidean inner product and norm, respectively, on \( \mathbb{R}^d. \)

**Assumption 2.1.** The coefficient \( q \in C^1([0, T]; L^\infty(\Omega)) \cap L^\infty([0, T]; W^{2, \infty}(\Omega)), \) and the symmetric diffusion coefficient matrix \( a \in C^1([0, T]; C^1(\Omega; \mathbb{R}^{d \times d})) \cap C([0, T]; C^3(\Omega; \mathbb{R}^{d \times d})) \) satisfies the following conditions:

(i) There exists \( \lambda \in (0, 1) \) such that for any \( (x, t) \in \Omega \times [0, T], \)

\[ \lambda |\xi|^2 \leq a(x, t) \xi \cdot \xi \leq \lambda^{-1} |\xi|^2 \quad \forall \xi \in \mathbb{R}^d. \]

(ii) \( a_{jd}(x', \pm \ell, t) = 0, x' \in \omega, \) and \( j = 1, \ldots, d-1, \) and \( \partial_{x_d} a_{ij}(t) = 0 \) for \( i, j = 1, \ldots, d-1. \)

Note that the cylindrical domain \( \Omega = \omega \times (-\ell, \ell) \) is only Lipschitz continuous. Thus, some extra assumptions on the domain and the coefficient matrix \( a \) are needed in order to guarantee high-order Sobolev regularity of the elliptic operator \( \mathcal{A}(t) \) with suitable boundary conditions. In the analysis, we need the following elliptic regularity pickup: (i) and (ii) are sufficient for the analysis in sections 3 and 4, respectively. (i) holds under the assumption that the domain \( \omega \) is convex and

\[ a_{id} = 0, \quad \partial_{x_d} a_{dd} = 0, \quad \partial_{x_d} a_{ij} = 0, \quad i, j \in \{1, \ldots, d-1\}. \]

Indeed, if \( \omega \) is convex, then \( \Omega \) is convex and the desired assertion follows from [10, Theorems 3.2.1.2 and 3.2.1.3]. This can be verified using a separation of variables argument [9, Lemma 2.4]. Besides the condition \( (2.3) \), if the domain \( \omega \) is of class \( C^4, \) the separation of variable argument similar to [9, Lemma 2.4] implies Assumption \( \tilde{H} \) in Definition 2.2(ii).

**Definition 2.2**.

(i) A tuple \((\Omega, \mathcal{A}(t))\) is said to satisfy Assumption \( \tilde{H}_{mn}, m, n = 0, 1, \) if for any \( t \in [0, T] \) and any \( f \in L^2(\Omega), \) the boundary value problem

\[
\begin{cases}
\mathcal{A}(t)v = f & \text{in } \Omega, \\
v = 0 & \text{on } \partial \omega \times (-\ell, \ell), \\
\partial^n_{x_d} v(x', \ell) = 0 & \text{on } \omega, \\
\partial^n_{x_d} v(x', -\ell) = 0 & \text{on } \omega,
\end{cases}
\]

admits a unique solution \( v \in H^2(\Omega) \) such that

\[ \|v\|_{H^2(\Omega)} \leq c(\mathcal{A}, m, n, \Omega) \|f\|_{L^2(\Omega)}. \]
(ii) A tuple \((\Omega, \mathcal{A}(t))\) is said to satisfy Assumption \(\widetilde{H}\) if for all \(v \in H^{\max(1,s)}(\Omega)\) satisfying \(\mathcal{A}(t)v \in H^s(\Omega), \ s \in [0,2]\), there hold \(v \in H^{2+s}(\Omega)\) and
\[
\|v\|_{H^{2+s}(\Omega)} \leq c(A, s, \Omega)(\|A v\|_{H^s(\Omega)} + \|v\|_{H^s(\Omega)}).
\]

The following perturbation estimates are useful.

**Lemma 2.3.** Under Assumptions 2.1(i) and H00 / H01 / H11, for any \(t, s \in [0,T]\) and \(\beta \in [0,1]\), there hold
\[
\begin{align*}
&\|A^\beta(I - A(t)^{-1}A(s))v\|_{L^2(\Omega)} \leq c|t-s|\|A^\beta v\|_{L^2(\Omega)} \quad \forall v \in D(A^\beta), \\
&\|A^\beta(I - A(t)^{-1}\partial_t A(s))v\|_{L^2(\Omega)} \leq c|t-s|\|A^\beta v\|_{L^2(\Omega)} \quad \forall v \in D(A^\beta).
\end{align*}
\]

**Proof.** For the operator \(A(t)\), the case \(\beta = 0\) is contained in [19, Corollary 3.1]. To show the estimate for \(\beta = 1\), fix \(t, s \in [0,T], v \in D(A_\ast)\). From Assumption H00, we deduce \(D(A_\ast) = H^s_0(\Omega) \cap H^s(\Omega) = D(A(t)) = D(A(s)), \) i.e., \(v \in D(A(t))\) and \(v \in D(A(s))\). Moreover, applying again Assumption H00, we get
\[
\begin{align*}
&\|A_\ast(I - A(t)^{-1}A(s))v\|_{L^2(\Omega)} \leq c(\|I - A(t)^{-1}A(s))v\|_{H^s(\Omega)} \\
&\leq c\|A(t)(I - A(t)^{-1}A(s))v\|_{L^2(\Omega)} = c\|A(t)v - A(s)v\|_{L^2(\Omega)},
\end{align*}
\]
with \(c > 0\) a constant independent of \(t\) and \(s\). Combining this estimate with [19, equation (2.6)] and Assumption H00, we obtain (2.4) for \(\beta = 1\) and \(q \equiv 0\). We can extend this result to \(q \neq 0\), since for \(q \in C^1([0,T]; L^2(\Omega))\), the mean value theorem implies
\[
\|q(t)v - q(s)v\|_{L^2(\Omega)} \leq \|\partial_t q\|_{L^\infty([0,T]; L^\infty(\Omega))}|t-s|\|v\|_{L^2(\Omega)} \leq c|t-s|\|A^\beta v\|_{L^2(\Omega)}.
\]

The case \(\beta \in (0,1)\) follows by interpolation. The proof of estimate (2.5) is identical under Assumption H01 / H11. \(\square\)

Below we need Bochner–Sobolev spaces \(W^{s,p}(0,T;X)\) for a UMD space \(X\) (see [13] for the definition of UMD spaces, which include Sobolev spaces \(W^{s,p}(\Omega)\) with \(1 < p < \infty\)). For any \(s \geq 0\) and \(1 \leq p < \infty\), we denote by \(W^{s,p}(0,T;X)\) the space of functions \(v : (0,T) \to X\), with the norm defined by complex interpolation. Equivalently, the space is equipped with the quotient norm
\[
\|v\|_{W^{s,p}(0,T;X)} := \inf_{\tilde{v}} \|\tilde{v}\|_{W^{s,p}(\BbbR;X)} := \inf_{\tilde{v}} \|F^{-1}[(1 + |\xi|^2)^{s/2}F(\tilde{v})(\xi)]\|_{L^p(\BbbR;X)},
\]
where the infimum is taken over all possible \(\tilde{v}\) that extend \(v\) from \((0,T)\) to \(\BbbR\), and \(F\) denotes the Fourier transform (with \(F^{-1}\) being its inverse). The following norm equivalence result will be used extensively.

**Lemma 2.4.** Let \(\alpha \in (0,1)\) and \(p \in [1,\infty)\) with \(ap > 1\). If \(v(0) = 0\) and \(\partial_t^\alpha v \in L^p(0,T;X)\), then \(v \in W^{\alpha,p}(0,T;X)\) and
\[
\|v\|_{W^{\alpha,p}(0,T;X)} \leq c\|\partial_t^\alpha v\|_{L^p(0,T;X)}.
\]

Meanwhile, if \(v(0) = 0\), \(v \in W^{\alpha,p}(0,T;X)\), then \(\partial_t^\alpha v \in L^p(0,T;X)\) and
\[
\|\partial_t^\alpha v\|_{L^p(0,T;X)} \leq c\|v\|_{W^{\alpha,p}(0,T;X)}.
\]
Then it is direct that

\[ K \]

which satisfies

\[ \text{and Young's convolution inequality imply (cf., e.g., [17, Theorem 2.2])} \]

\[ \| v \|_{L^p(0,T;X)} \leq c \| g \|_{L^p(0,T;X)}. \]

Let \( \tilde{g} \) be the extension of \( g \) from \( L^p(0,T;X) \) to \( L^p(\mathbb{R};X) \) by zero, i.e., \( \tilde{g}(t) = 0 \) for \( t \in (-\infty,0) \cup (T,\infty) \) and \( \tilde{g}(t) = g(t) \) for \( t \in (0,T) \). Then let

\[ \tilde{v}(t) = \frac{1}{\Gamma(\alpha)} \int_{-\infty}^{t} (t-s)^{\alpha-1} \tilde{g}(s) \, ds, \]

which satisfies

\[ \tilde{g}(t) = \frac{1}{\Gamma(1-\alpha)} \frac{d}{dt} \int_{-\infty}^{t} (t-s)^{-\alpha} \tilde{v}(s) \, ds. \]

Then there holds [26, p. 90]

\[ \| v \|_{W^{\alpha,p}(\mathbb{R},X)} = \| F^{-1}[(1 + |\xi|^2)^{\frac{\alpha}{2}} F[\tilde{v}](\xi)] \|_{L^p(\mathbb{R},X)} \]

\[ = \| F^{-1}[K(\xi)(1 + (i\xi)^\alpha) F[\tilde{v}](\xi)] \|_{L^p(\mathbb{R},X)}, \]

with \( K(\xi) = (1 + |\xi|^2)^{\frac{\alpha}{2}} (1 + (i\xi)^\alpha)^{-1} \). Note that

\[ \lim_{|\xi| \to 0^+} |K(\xi)| = 1 \quad \text{and} \quad \lim_{|\xi| \to \infty} |K(\xi)| = 1, \]

so \( K(\xi) \) is uniformly bounded. Similarly,

\[ \frac{\partial}{\partial \xi} K(\xi) = \frac{\alpha |\xi|^2}{1 + |\xi|^2} (1 + |\xi|^2)^{\frac{\alpha}{2}} (1 + (i\xi)^\alpha)^{-1} + \alpha (1 + |\xi|^2) \frac{\alpha}{2} (1 + (i\xi)^\alpha)^{-2}(i\xi)^\alpha \]

is also bounded. Therefore, the vector-valued Mikhlin multiplier theorem (see, e.g., [6] or [45, Proposition 3]) indicates that \( K(\xi) \) is a Fourier multiplier, and hence

\[ \| v \|_{W^{\alpha,p}(0,T;X)} \leq \| \tilde{v} \|_{W^{\alpha,p}(\mathbb{R},X)} \leq c \| F^{-1}[(1 + (i\xi)^\alpha) F[\tilde{v}](\xi)] \|_{L^p(\mathbb{R},X)} \]

\[ \leq c \| \tilde{v} \|_{L^p(\mathbb{R},X)} + c \| g \|_{L^p(\mathbb{R},X)} \leq c \| g \|_{L^p(\mathbb{R},X)} = c \| g \|_{L^p(0,T;X)} = c \| \partial_t^\alpha v \|_{L^p(0,T;X)}. \]

To prove the second assertion, let \( v \in C^\infty([0,T];X) \) with \( v(0) = 0 \), and we extend \( v \) from \( (0,T) \) to a function \( \tilde{v} \in W^{\alpha,p}(\mathbb{R};X) \) satisfying \( \tilde{v}(t) = 0 \) for all \( t \leq 0 \) and

\[ \| \tilde{v} \|_{W^{\alpha,p}(\mathbb{R},X)} \leq c \| v \|_{W^{\alpha,p}(0,T;X)}. \]

Then it is direct that

\[ \partial_t^\alpha v(t) = \frac{1}{\Gamma(\alpha)} \frac{d}{dt} \int_{-\infty}^{t} (t-s)^{-\alpha} \tilde{v}(s) \, ds = \partial_t^\alpha v(t) \quad \forall t \in (0,T) \]

and

\[ \| \partial_t^\alpha v \|_{L^p(0,T;X)} = \| \partial_t^\alpha \tilde{v} \|_{L^p(0,T;X)} \leq \| \partial_t^\alpha \tilde{v} \|_{L^p(\mathbb{R},X)} \]

\[ = \| F^{-1}(i\xi)^\alpha F[\tilde{v}](\xi) \|_{L^p(\mathbb{R},X)} = \| F^{-1} K_2(\xi)(1 + |\xi|^2)^{\frac{\alpha}{2}} F[\tilde{v}](\xi) \|_{L^p(\mathbb{R},X)}, \]

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with \( K_2(\xi) = |\xi|^\alpha (1 + |\xi|^2)^{-\frac{\alpha}{2}} \). Note that both \(|K_2(\xi)|\) and \(|\xi|^\frac{\alpha}{2} K_2(\xi)\) are uniformly bounded, and hence it is a Fourier multiplier. Then we have
\[
\|\partial_t^\alpha v\|_{L^p(0,T;X)} \leq \|F^{-1}((1 + |\xi|^2)^{\frac{\alpha}{2}} \mathcal{F}[v])\|_{L^p(\mathbb{R};X)} \leq c\|v\|_{W^{\alpha,p}(\mathbb{R};X)}.
\]
This together with (2.6) and the density of \( C^\infty([0,T];X) \) in \( W^{\alpha,p}(0,T;X) \) leads to the second assertion.

We need the following Gronwall’s inequality (see, e.g., [42], [12, Exercise 3, p. 190], or [17, Theorem 4.2]).

**Lemma 2.5.** Let \( c, r > 0 \) and \( y, a \in L^1(0,T) \) be nonnegative functions satisfying
\[
y(t) \leq a(t) + c \int_0^t (t-s)^{r-1} y(s) \, ds, \quad t \in (0,T).
\]
Then there exists \( c = c(r,T) > 0 \) such that
\[
y(t) \leq a(t) + c \int_0^t (t-s)^{r-1} a(s) \, ds, \quad t \in (0,T).
\]

3. **Well-posedness for ISPn.** This section is devoted to ISPn, i.e., recovering the source component \( f(x',t) \) in problem (1.1) with \( m = 1 \) from \( u|_{\omega \times (\ell) \times (0,T)} \). The direct problem is given by
\[
\begin{aligned}
\partial_t^\alpha u + Au &= F \quad \text{in } \Omega \times (0,T), \\
u(x,0) &= 0 \quad \text{in } \Omega, \\
\partial_{x\alpha} u(x', \pm \ell, t) &= 0 \quad \text{on } \omega \times (0,T), \\
u(x,t) &= 0 \quad \text{on } \partial \omega \times (-\ell, \ell) \times (0,T).
\end{aligned}
\]
Subdiffusion with time-dependent coefficients has recently been studied in [28, 19, 21], where well-posedness and several regularity estimates have been established. Our description largely follows the approach developed in [19, 21]. Throughout, for the prefactor \( R(x,t) \) in the source \( F \), we make the following assumption.

**Assumption 3.1.** The function \( R \in L^\infty(\Omega \times (0,T)) \) satisfies \( \partial_{x\alpha} R \in L^\infty(\Omega \times (0,T)) \) and that there exists \( c_R > 0 \) such that \( |R(x', \ell, t)| \geq c_R \) for any \((x', \ell, t) \in \omega \times (0,T)\).

Now we give several regularity estimates for the direct problem (3.1). First we derive a representation of the solution \( u \). The key step is to reformulate problem (3.1) into
\[
\partial_t^\alpha u + \tilde{A}_s u = F(t) + (\tilde{A}_s - \tilde{A}(t)) u(t) \quad \forall t \in (0,T].
\]
According to [19, 21], problem (3.1) has a unique solution \( u \) which satisfies
\[
u(t) = \int_0^t \tilde{E}_s(t-s) F(s) \, ds + \int_0^t \tilde{E}_s(t-s) (\tilde{A}_s - \tilde{A}(s)) u(s) \, ds.
\]
By setting \( t \) to \( t_* \), we can use Lemma 2.3 to estimate the second integral, which involves the crucial perturbation term.

The next result collects a priori estimates on the solution \( u \) to problem (3.1). Let Assumption H11 hold. Then the solution \( u \) to problem (3.1) satisfies
\[
\|u(t)\|_{H^1(\Omega)} \leq c \int_0^t (t-s)^{\frac{\alpha}{2} - 1} \|F(s)\|_{L^2(\Omega)} \, ds \quad \forall t \in (0,T]
\]
and also the following maximal $L^p$ regularity:

\begin{equation}
(3.3) \quad \|\partial^m_t u\|_{L^p(0,T;L^2(\Omega))} + \|u\|_{L^p(0,T;D(\mathcal{A}))} \leq c\|f\|_{L^p(0,T;L^2(\Omega))} \quad \forall 1 < p < \infty.
\end{equation}

**Proof.** The estimate (3.2) can be found in [21, Theorem 2] (with $k = 0$), and (3.3) in [19, Theorem 2.1]. \hfill \square

Further, we denote by $u_f$ the solution of problem (3.1) to explicitly indicate its dependence on $f$. First we show that the inverse problem is indeed ill-posed on the space $L^2(0,T;L^2(\omega))$.

**Corollary 3.2.** Under Assumptions 2.1(i), H11, and 3.1, the map $f \mapsto u_f|_{L^2(0,T;L^2(\omega))}$ is linear and compact on $L^2(0,T;L^2(\omega))$.

**Proof.** The linearity is obvious. The compactness is direct from Lemma 3.1. In fact, by the maximal $L^p$ regularity in Lemmas 3.1 and 2.4 and Assumption 3.1, we have

\[ ||u_f||_{W^{\alpha,2}(0,T;L^2(\Omega))} \leq c ||f||_{L^2(0,T;L^2(\omega))}. \]

Thus, $u_f \in W^{\alpha,2}(0,T;L^2(\Omega)) \cap L^2(0,T;H^2(\Omega))$. Meanwhile, by interpolation, the space $W^{\alpha,2}(0,T;L^2(\Omega)) \cap L^2(0,T;H^2(\Omega))$ embeds compactly into $L^2(0,T;H^2(\Omega))$ [4, Theorem 5.2], which, by the trace theorem, embeds continuously into $L^2(0,T;L^2(\omega))$. Thus the map $f \mapsto u_f|_{\omega \times (t) \times (0,T)}$ is compact on $L^2(0,T;L^2(\omega))$. \hfill \square

Let $w = \partial_{x_d}u_f$. Then $w$ satisfies

\begin{equation}
(3.4) \quad \begin{cases}
\partial^d_t w + \mathcal{A}(t)w = -\partial_{x_d}\mathcal{A}(t)u_f(t) + \partial_{x_d}F(t) & \text{in } \Omega \times (0,T), \\
w(0) = 0 & \text{in } \Omega, \\
w = 0 & \text{on } \partial\Omega \times (0,T).
\end{cases}
\end{equation}

By applying the perturbation argument and using the operator $\mathcal{A}(t)$, the solution $w$ to problem (3.4) can be represented by

\begin{equation}
(3.5) \quad w(t) = \int_0^t E_s(t-s)(-\partial_{x_d}\mathcal{A}(s)u_f(s) + \partial_{x_d}F(s))ds + \int_0^t E_s(t-s)(\mathcal{A}_s - \mathcal{A}(s))w(s)ds.
\end{equation}

Noting the definition $w = \partial_{x_d}u_f$ and the condition $\partial_{x_d}a_{ij}(t) = 0$ for $i,j = 1,\ldots,d-1$ from Assumption 2.1(ii), we deduce

\[-(\partial_{x_d}\mathcal{A}(t))u = \partial_{x_d}(\partial_{x_d}a_{dd}(t)\partial_{x_d}u_f) + \sum_{j=1}^{d-1} \left[ \partial_{x_j}(\partial_{x_d}a_{jd}(t)\partial_{x_d}u_f) + \partial_{x_d}(\partial_{x_d}a_{jd}(t)\partial_{x_j}u_f) \right] - \partial_{x_d}q(t)u_f := B_1(t)w + B_2(t)u_f,
\]

where the (time-dependent) operators $B_1(t)$ and $B_2(t)$ are given, respectively, by

\begin{align*}
B_1(t)w &:= \partial_{x_d}a_{dd}(t)\partial_{x_d}w + 2\sum_{j=1}^{d-1} \partial_{x_d}a_{jd}(t)\partial_{x_j}w + \sum_{j=1}^{d-1} \partial_{x_j}\partial_{x_d}a_{jd}(t)w, \\
B_2(t)u &:= \sum_{j=1}^d (\partial^2_{x_d}a_{jd}(t))\partial_{x_j}u - \partial_{x_d}q(t)u.
\end{align*}

Note that Assumption 2.1(ii) allows eliminating the cross terms $\partial_{x_i}(\partial_{x_d}a_{ij}(t)\partial_{x_j}u)$, $i,j = 1,\ldots,d-1$, which plays a central role in the analysis below, and without this the argument does not work.

The next result gives useful bounds on $w := \partial_{x_d}u_f$. 

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Meanwhile, under Assumption 3.1 and estimate (3.2), we deduce
\[ w_{4454}BANGTI JIN, YAVAR KIAN, AND ZHI ZHOU \]
\[ \begin{align*}
\|w(t)\|_{H^\beta(\Omega)} & \leq c \int_0^t (t-s)^{(1-\frac{\alpha}{2})\alpha-1}\|f(s)\|_{L^2(\omega)}ds, \quad t \in (0,T),
\end{align*} \]
where the constant \( c \) depends only on \( R, A, \beta, \) and \( T. \)

**Proof.** By estimate (3.3) with \( p = 2 \), we have \( -\partial_{xx}A(t)u_f, \partial_{xx}F \in L^2(0,T;L^2(\Omega)) \).
Then Lemma 3.1 shows that problem (3.4) has a unique solution \( w \in L^2(0,T;H^2(\Omega)) \) with \( A(t)w(t), \partial_{xx}w \in L^2(0,T;L^2(\Omega)) \).
Next, we prove the \( H^\beta(\Omega) \) bound on \( w(t) \). We define the operators \( K_1 : L^2(0,T;H^1(\Omega)) \rightarrow L^2(0,T;H^1(\Omega)) \) and \( K_2 : L^2(0,T;L^2(\omega)) \rightarrow L^2(0,T;H^1(\Omega)) \), respectively, by
\[ \begin{align*}
K_1v(t) & = \int_0^t E_s(t-s)B_1v(s)ds, \\
K_2f(t) & = \int_0^t E_s(t-s)B_2u_f(s)ds + \int_0^t E_s(t-s)\partial_{xx}R(s)f(s)ds.
\end{align*} \]

By Lemma 2.1, we have
\[ \|K_1v(t_*)\|_{H^\beta(\Omega)} \leq c \int_0^{t_*} (t_*-s)^{(1-\frac{\alpha}{2})\alpha-1}\|B_1v(s)\|_{L^2(\omega)}ds \]
\[ \leq c \int_0^{t_*} (t_*-s)^{(1-\frac{\alpha}{2})\alpha-1}\|v(s)\|_{H^1(\Omega)}ds. \tag{3.6} \]

Similarly, by Lemma 2.1, under Assumption 3.1, we have
\[ \begin{align*}
\|K_2f(t_*)\|_{H^\beta(\Omega)} & \leq c \int_0^{t_*} (t_*-s)^{(1-\frac{\alpha}{2})\alpha-1}\|f(s)\|_{L^2(\omega)}ds \\
& \quad + c \int_0^{t_*} (t_*-s)^{(1-\frac{\alpha}{2})\alpha-1}\|u_f(s)\|_{H^1(\Omega)}ds. \tag{3.7}
\end{align*} \]

Meanwhile, under Assumption 3.1 and estimate (3.2), we deduce
\[ \|u_f(t)\|_{H^1(\Omega)} \leq c \int_0^t (t-s)^{\frac{-\alpha}{2}-1}\|f(s)\|_{L^2(\omega)}ds. \]

Consequently,
\[ \begin{align*}
& \int_0^{t_*} (t_*-s)^{(1-\frac{\alpha}{2})\alpha-1}\|u_f(s)\|_{H^1(\Omega)}ds \\
& \leq c \int_0^{t_*} (t_*-s)^{(1-\frac{\alpha}{2})\alpha-1}\int_0^s (s-\xi)^{\frac{-\alpha}{2}-1}\|f(\xi)\|_{L^2(\omega)}d\xi ds \\
& = c \int_0^{t_*} \|f(\xi)\|_{L^2(\omega)} \int_{\xi}^{t_*} (t_*-s)^{(1-\frac{\alpha}{2})\alpha-1}(s-\xi)^{\frac{-\alpha}{2}-1}dsd\xi \\
& \leq cT^{\frac{-\alpha}{2}} \int_0^{t_*} (t_*-s)^{(1-\frac{\alpha}{2})\alpha-1}\|f(s)\|_{L^2(\omega)}ds.
\end{align*} \]
This and estimate (3.7) imply

\[ (3.8) \quad \|K_2 f(t_\ast)\|_{H^\beta(\Omega)} \leq c_T \int_0^{t_\ast} (t_\ast - s)^{(1-\frac{2}{q})\alpha - 1}\|f(s)\|_{L_2(\omega)} ds. \]

Next, by Lemmas 2.1 and 2.3, (2.4), and Assumption H00, we have

\[ \|\int_0^{t_\ast} E_s(t_\ast - s)(A_s - A(s))w(s)ds\|_{H^\beta(\Omega)} \leq c \int_0^{t_\ast} \|A_s E_s(t_\ast - s)\|\|A^\beta_s (I - A_s^{-1})A(s)w(s)\|_{L_2(\Omega)} ds \]
\[ \leq c \int_0^{t_\ast} (t_\ast - s)^{-1}(t_\ast - s)^{1(1-\frac{2}{q})\alpha - 1}\|w(s)\|_{H^\beta(\Omega)} ds = c \int_0^{t_\ast} \|w(s)\|_{H^\beta(\Omega)} ds. \]

This estimate, (3.6), (3.8), and the solution representation (3.5) lead to

\[ \|w(t_\ast)\|_{H^\beta(\Omega)} \leq c \int_0^{t_\ast} \|w(s)\|_{H^\beta(\Omega)} ds + c \int_0^{t_\ast} (t_\ast - s)^{(1-\frac{2}{q})\alpha - 1}\|f(s)\|_{L_2(\omega)} ds. \]

This and Gronwall's inequality in Lemma 2.5 imply the desired $H^\beta(\Omega)$ bound. This completes the proof. \( \square \)

Now we can state a well-posedness result for ISPn. Note that below we use the notation $L^2(\omega \times (0, T))$ and $L^2(\omega \times (0, T))$ interchangeably since they are isomorphic by the Fubini–Tonelli theorem.

**Theorem 3.4.** Let Assumptions 2.1, H00, H11, and 3.1 be fulfilled. Then for any $f \in L^2(\omega \times (0, T))$, the solution $u$ of problem (3.1) satisfies $u \in H^\beta(-\ell, \ell; L^2(\omega \times (0, T)))$, $\partial_t^\alpha u, A(t)u \in H^\beta(-\ell, \ell; L^2(\omega \times (0, T)))$. Thus, the map

\[ (3.9) \quad h : (x', t) \mapsto \frac{[\partial_t^\alpha u + (A(t) + a_{dd}(t))\partial_{x_d}^2 u](x', \ell, t)}{R(x', \ell, t)} \in L^2(\omega \times (0, T)) \]

is well-defined, and, further, there exists a bounded linear operator $\mathcal{H} : L^2(0, T; L^2(\omega)) \to L^2(0, T; L^2(\omega))$ such that $f$ solves

\[ (3.10) \quad h = f + \mathcal{H}f, \]

which is well-posed on $L^2(\omega \times (0, T))$. Finally, for every pair $(h, f) \in L^2(\omega \times (0, T)) \times L^2(\omega \times (0, T))$, the solution $u$ of problem (3.1) satisfies (3.10).

**Proof.** By Lemmas 3.1 and 3.3, problem (1.1) has a solution $u_f \in L^2(0, T; H^1(\Omega))$, with $A(t)u_f, \partial_t^\alpha u_f \in L^2(0, T; L^2(\Omega))$ and $w = \partial_{x_d} u_f \in L^2(0, T; H^2(\Omega))$, $\partial_t^\alpha \partial_{x_d} u_f$, $A(t)\partial_{x_d} u_f \in L^2(0, T; L^2(\Omega))$. Hence,

\[ x_d \mapsto u_f(\cdot, x_d, \cdot) \in H^\beta(-\ell, \ell; L^2(\omega \times (0, T))) \subset H^1(-\ell, \ell; L^2(\omega \times (0, T))) \]
\[ x_d \mapsto \partial_t^\alpha u_f(\cdot, x_d, \cdot) \in H^1(-\ell, \ell; L^2(\omega \times (0, T))). \]

By the trace theorem, we can restrict $\partial_{x_d} w = \partial_{x_d}^2 u_f, \partial_t^\alpha u_f$, and $A(t)u_f$ to the boundary $x_d = \ell$. Thus the governing equation in problem (3.1) implies that for $(x', t) \in \omega \times (0, T),

\[ a_{dd}(t)\partial_{x_d} w(x', \ell, t) = a_{dd}(t)\partial_{x_d}^2 u_f(x', \ell, t) \]
\[ = [\partial_t^\alpha u_f + (A(t) + a_{dd}(t)\partial_{x_d}^2)u_f](x', \ell, t) - R(x', \ell, t)f(x', t) \]
\[ = R(x', \ell, t)[h(x', t) - f(x', t)], \]
with the function \( h(x', t) \) given by (3.9). Let the operator \( \mathcal{H} : L^2(\omega \times (0, T)) \to L^2(\omega \times (0, T)) \) be defined by

\[
\mathcal{H}\phi(x', t) = \frac{\partial_{x_d}^2 u_\phi(x', \ell, t)}{R(x', \ell, t)},
\]

where \( u_\phi(x', x_d, t) \) denotes the solution to problem (3.1) with \( F = \phi R \). Then it follows from (3.11) that \( f \) is the solution to

\[
h = f + \mathcal{H} f.
\]

Moreover, by Lemma 3.3, trace inequality, and the defining identity \( w = \partial_{x_d} u_f \), we deduce that for any \( \beta \in (\frac{1}{2}, 1) \)

\[
\|f(t)\|_{L^2(\omega)} \leq \|h(t)\|_{L^2(\omega)} + \|\mathcal{H} f\|_{L^2(\omega)} \leq \|h(t)\|_{L^2(\omega)} + c\|\partial_{x_d}^2 u_f(\cdot, \ell, t)\|_{L^2(\omega)}
\]

\[
\leq \|h(t)\|_{L^2(\omega)} + c\|\omega(t)\|_{H^2_\omega(\Omega)} \leq \|h(t)\|_{L^2(\omega)} + c \int_0^T (t-s)^{(1-\beta)\alpha-1} \|f(s)\|_{L^2(\omega)} ds.
\]

This and the standard Gronwall inequality in Lemma 2.5 yield

\[
\|f(t)\|_{L^2(\omega)} \leq \|h(t)\|_{L^2(\omega)} + c \int_0^T (t-s)^{(1-\beta)\alpha-1} \|h(s)\|_{L^2(\omega)} ds,
\]

which together with Young’s inequality directly implies

\[
\|f\|_{L^2(0,T;L^2(\omega))} \leq c\|h\|_{L^2(0,T;L^2(\omega))}.
\]

This shows the well-posedness of (3.10) and the recovery of \( f \) from the data \( h \). Last, fix \( (h_1, f) \in L^2(0, T; L^2(\omega)) \times L^2(0, T; L^2(\omega)) \) satisfying (3.10) with \( h = h_1 \) and consider \( u \in L^2(0, T; H^1(\Omega)) \) solving problem (1.1) with \( F = hR \). The preceding argument shows that one can define \( h \in L^2(0, T; L^2(\omega)) \) given by (3.9) and \( f \) solves (3.10). This implies \( h_1 = f + \mathcal{H} f = h \). Therefore, we have \( h = h_1 \), and this completes the proof of the theorem.

**Remark 3.5.** Theorem 3.4 actually gives a reconstruction algorithm for recovering \( f \) for ISPd if the given data \( g^f(x', t) = u_{g^f}(x', \ell, t) \) is sufficiently accurate so that the derivatives \( \partial_t^\alpha g^f \) and \( A(t) g^f \) in (3.9) can be evaluated accurately. For noisy data \( g^f \), one can proceed in two steps: first suitably mollify the data \( g^\delta \) so that the mollified data is smooth, and then apply the fixed point iteration.

### 4. Conditional stability for ISPd

In this section, we establish a conditional stability result for ISPd, i.e., recovering \( f(x', t) \) in problem (1.1) with \( m = 0 \) from the lateral flux observation \( \partial_{x_d} u|_{\omega \times (\ell) \times (0, T)} \). The direct problem is given by

\[
\begin{align}
\partial_t u + \mathcal{A}(t) u &= fR \quad \text{in } \Omega \times (0, T), \\
u(\cdot, \ell, \cdot) &= 0 \quad \text{on } \partial \omega \times (-\ell, \ell) \times (0, T), \\
\partial_{x_d} u(\cdot, -\ell, \cdot) &= 0 \quad \text{on } \omega \times (0, T), \\
u(0) &= 0 \quad \text{in } \omega \times (-\ell, \ell).
\end{align}
\]

(4.1)

Note that the estimates in Lemma 3.1 remain valid for problem (4.1). The next result gives an improved regularity result under extra regularity and compatibility assumptions on the source \( F \). This result plays a central role in the stability analysis.

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Proposition 4.1. Let Assumptions 2.1(i), H01, and \( \tilde{H} \) be fulfilled, let \( \gamma \in (\frac{1}{2}, 1) \), and let \( F \in W^{1, \frac{1}{1+\alpha-2\gamma}}(0, T; L^2(\Omega)) \cap L^{\frac{1}{1-\gamma}}(0, T; H^{2\gamma}(\Omega)) \) and \( F(0) = 0 \). Then for any \( \beta \in (\frac{1}{2}, \gamma) \), problem (4.1) has a unique weak solution \( u \in L^{\frac{1}{1-\gamma}}(0, T; H^{2\beta}(\Omega)) \) with

\[
\|u\|_{L^\frac{1}{1-\gamma}(0, T; H^{2\beta}(\Omega))} + \|u\|_{W^{\alpha, \frac{1}{1-\gamma}}(0, T; L^2(\Omega))} \leq c (\|F\|_{W^{1, \frac{1}{1+\alpha-2\gamma}}(0, T; L^2(\Omega))} + \|F\|_{L^{\frac{1}{1-\gamma}}(0, T; H^{2\gamma}(\Omega))}).
\]

Proof. By Sobolev embedding, \( F \in L^\infty(0, T; L^2(\Omega)) \), and the existence and uniqueness of a weak solution \( u \in L^q(0, T; D(\tilde{A}_*)) \) for all \( q \in (1, \infty) \) follows directly from Lemma 3.1 with

\[
\|u\|_{L^q(0, T; D(\tilde{A}_*))} \leq c \|f\|_{W^{1, \frac{1}{1+\alpha-2\gamma}}(0, T; L^2(\omega))}.
\]

It suffices to show the claimed regularity. Using the operator \( \tilde{A}(t) \) and the perturbation argument, since \( u(0) = 0 \), the solution \( u \) can be represented by

\[
u(t) = \int_0^t \tilde{E}_*(s)F(t-s)ds + \int_0^t \tilde{E}_*(s)(\tilde{A}_* - \tilde{A}(t-s))u(t-s)ds.
\]

Then applying \( \tilde{A}_* \) to both sides of the identity and using the governing equation give

\[
\tilde{A}_*^\alpha u(t) = -\tilde{A}_* u(t) + F(t) + (\tilde{A}_* - \tilde{A}(t))u(t)
= -\int_0^t \tilde{A}_* \tilde{E}_*(s)F(t-s)ds + \int_0^t \tilde{A}_* \tilde{E}_*(s)\tilde{A}_*(t-s)u(t-s)ds + F(t)
+ (\tilde{A}_* - \tilde{A}(t))u(t).
\]

Now by fixing \( t \) at \( t_* \) in the identity, applying the identity (2.2) and the integration by parts formula to the first integral, and noting the condition \( F(\cdot, 0) = 0 \) and the fact \( \tilde{S}_*(0) = I \) [17, Lemma 6.3], we obtain

\[
\tilde{A}_*^\alpha \tilde{S}_*(t_*)F(t_*) - \int_0^{t_*} \tilde{A}_* \tilde{E}_*(s)F(t_*-s)ds - \int_0^{t_*} \tilde{A}_* \tilde{E}_*(s)(\tilde{A}_* - \tilde{A}(t_*-s))u(t_*-s)ds + F(t_*)
= \int_0^{t_*} \tilde{A}_* \tilde{E}_*(s)(\tilde{A}_* - \tilde{A}(t_*-s))u(t_*-s)ds.
\]

Then it follows from Lemmas 2.1 and 2.3 that

\[
\|\tilde{A}_*^\beta \tilde{S}_*(t_*)\|_{L^2(\Omega)} \leq \int_0^{t_*} \|\tilde{A}_*^\beta \tilde{S}_*(t_*-s)\||F^*(s)||_{L^2(\Omega)}ds
+ \int_0^{t_*} \|\tilde{A}_*^\beta \tilde{S}_*(t_*-s)\||I - \tilde{A}_*^{-1}\tilde{A}(s)||u(s)||_{L^2(\Omega)}ds
\leq c \int_0^{t_*} (t_*-s)^{-\beta_2}\|F^*(s)||_{L^2(\Omega)}ds + c \int_0^{t_*} (t_*-s)^{-\beta_2}(t_*-s)^{-\beta_2-1}(t_*-s)||\tilde{A}_* u(s)||_{L^2(\Omega)}ds
\leq c \int_0^{t_*} (t_*-s)^{-\beta_2}\|F^*(s)||_{L^2(\Omega)}ds + c \int_0^{t_*} (t_*-s)^{-\beta_2}||\tilde{A}_* u(s)||_{L^2(\Omega)}ds.
\]
Note that $t^{-\beta \alpha} \in L^p(0, T)$ for any $p \in (1, \frac{1}{\gamma - 1}) \subset (1, \frac{1}{\beta \alpha})$ by the choice $\beta < \gamma$. Now choosing $p = \frac{1}{\alpha(1 - \gamma)}$, $q = \frac{1}{1 + (1 - 2\beta)\alpha}$ in the Young’s convolution inequality

$$\|f * g\|_{L^r(0, T)} \leq \|f\|_{L^p(0, T)} \|g\|_{L^q(0, T)} \quad \forall p, q, r \geq 1 \text{ with } r^{-1} + 1 = p^{-1} + q^{-1},$$

we deduce

$$\|\tilde{A}_*^{\alpha} \partial_t^\alpha u(t_*)\|_{L^{\frac{1}{1-\gamma}}(0, T; L^2(\Omega))} \leq c\left(\|F'(s)\|_{L^2(0, T; L^2(\Omega))} + \|\tilde{A}_* u(s)\|_{L^2(0, T; L^2(\Omega))}\right).$$

Further, it follows from the representation (4.3) and Lemmas 2.1 and 2.3 that

$$\|\tilde{A}_*^\alpha u(t_*)\|_{L^2(\Omega)} \leq \int_0^{t_*} \|\tilde{A}_* \tilde{E}_*(s)\|\|F(t_* - s)\|_{L^2(\Omega)} ds + \int_0^{t_*} \|\tilde{A}_* \tilde{E}_*(s)(\tilde{A}_* - \tilde{A}_*)^{-1}(t_* - s)u(t_* - s)\|_{L^2(\Omega)} ds
\leq c\int_0^{t_*} s^{(1-\beta)\alpha - 1}\|F(t_* - s)\|_{L^2(\Omega)} ds + c\int_0^{t_*} s^{-1}\|u(t_* - s)\|_{L^2(\Omega)} ds
\leq c\|F\|_{L^\infty(0, T; L^2(\Omega))} t_*^{(1-\beta)\alpha} + c\int_0^{t_*} \|\tilde{A}_* u(s)\|_{L^2(\Omega)} ds.$$

This and Gronwall’s inequality directly imply $\lim_{t \to 0^+} \|\tilde{A}_*^\alpha u(t)\|_{L^2(\Omega)} = 0$. Hence, from Lemma 2.4 and Assumption H01, we deduce $u \in W^{\alpha, \frac{1}{\alpha(1 - \gamma)}}(0, T; D(\tilde{A}_*^\alpha)) \subset W^{\alpha, \frac{1}{\alpha(1 - \gamma)} \ominus 1}(0, T; H^{2\beta}(\Omega))$. Thus, we conclude that for any fixed $t \in (0, T]$, the solution $u$ satisfies

$$\begin{align*}
A(t)u(t) = F(t) - \partial_t^\alpha u(t) & \quad \text{in } \Omega, \\
u(x', \ell, t) = 0 & \quad \text{on } \omega, \\
\partial_{x_\ell} u(x', -\ell, t) = 0 & \quad \text{on } \omega, \\
u(x, t) = 0 & \quad \text{on } \partial \omega \times (-\ell, \ell).
\end{align*}$$

Note that for any $t \in (0, T]$, there holds $A(\cdot)u(\cdot) = F(\cdot) - \partial_t^\alpha u(\cdot) \in L^{\alpha(1-\gamma)}(0, T; H^{2\beta}(\Omega))$. Then by Assumption $\mathbb{H}$, we obtain $u \in L^{\frac{1}{\alpha(1 - \gamma)}}(0, T; H^{2(1+\beta)}(\Omega))$. This completes the proof.

The conditional stability analysis employs the regularity estimates in Proposition 4.1. Let $u$ be the solution to problem (4.1), and let $v = \partial_{x_\ell} u$. Then $v$ satisfies

$$\begin{align*}
\partial_t^\alpha v + A(t)v &= \mathcal{H} + f \partial_{x_\ell} R & \text{in } \Omega \times (0, T], \\
v &= 0 & \text{on } \partial \omega \times (-\ell, \ell) \times (0, T), \\
v(\cdot, \ell, \cdot) &= \partial_{x_\ell} u(\cdot, \ell, \cdot) & \text{on } \omega \times (0, T], \\
v(\cdot, -\ell, \cdot) &= 0 & \text{on } \omega \times (0, T), \\
v(0) &= 0 & \text{in } \omega \times (-\ell, \ell),
\end{align*}$$

with the function $\mathcal{H}$ given by

$$H(x', x_d, t) = -\partial_{x_d} A(t)u = \sum_{i,j=1}^d \partial_{x_i} \partial_{x_j} a_{ij}(t) \partial_{x_j} u - \partial_{x_d} q(t)u.$$
approach to derive the requisite bound on \( v \). For \( r \geq 1 \), the notation \( X_{\alpha,r} \) denotes the space \( W^{\alpha,r}(0,T;L^2(\omega)) \cap L^r(0,T;H^2(\omega)) \) with the norm
\[
\|v\|_{X_{\alpha,r}} = \|v\|_{W^{\alpha,r}(0,T;L^2(\omega))} + \|v\|_{L^r(0,T;H^2(\omega))}.
\]

Assumption 4.1. \( \omega \in C^4 \), \( f \in W^{1,\frac{1}{1+\alpha-\gamma}}(0,T;L^2(\omega)) \cap L^{\frac{1}{1-\gamma}+\gamma}(0,T;H^{2\gamma}(\omega)) \), for some \( \gamma \in (\frac{3}{4},1) \), with \( f(0) = 0 \), \( \partial_t R \in L^\infty(\Omega \times (0,T)) \) and \( R \in C([0,T];W^{2,\infty}(\Omega)) \).

Lemma 4.2. Let Assumptions 2.1, 3.1, H00, H01, and 4.1 be fulfilled, and let \( u \) be the solution of problem (4.1). Then for any \( \beta \in (\frac{3}{4},\gamma) \), the solution \( v \) to problem (4.6) satisfies
\[
\|\partial_{x_d}v(\cdot,\ell,t)\|_{L^2(\omega)} \leq c\|\partial_{x_d}u(\cdot,\ell,\cdot)\|_{X^{\alpha,\beta-1}} + \frac{c}{\alpha^{1-\gamma}} \int_0^t (t-s)^{\alpha(1-\beta)-1}\|f(\cdot,s)\|_{L^2(\omega)}ds
\]
for any \( t \in (0,T) \), where the constant \( c \) depends on \( R, \Omega, \alpha, \beta, \gamma \), and \( \mathcal{A} \).

Proof. Let \( r = \frac{1}{\alpha(1-\gamma)} \). Since \( f \in W^{1,\frac{1}{1+\alpha-\gamma}}(0,T;L^2(\omega)) \) and \( \partial_t R \in L^\infty(\Omega \times (0,T)) \), we deduce \( F = fR \in W^{1,\frac{1}{1+\alpha-\gamma}}(0,T;L^2(\omega)) \). The assumptions \( f \in L^r(0,T;H^{2\gamma}(\omega)) \) and \( R \in C([0,T];W^{2,\infty}(\Omega)) \) imply \( F = \overline{f}R \in L^{\infty}(\Omega \times (0,T));H^{2\gamma}(\Omega)) \). Further, \( f(\cdot,0) = 0 \) in \( \omega \) indicates \( f(\cdot,0) = 0 \) in \( \Omega \). Thus, \( F = fR \) satisfies the conditions in Proposition 4.1, and since \( \omega \in C^4 \), Assumption H holds. By Proposition 4.1, \( u \in W^{\alpha,r}(0,T;H^{2\beta}(\Omega)) \cap L^r(0,T;H^{2\beta+1}(\Omega)) \) for any \( \beta \in (\frac{3}{4},\gamma) \), and by the trace theorem, there hold
\[
(x',t) \mapsto \partial_{x_d}u(x',\ell,t) \in W^{\alpha,r}([0,T];H^{\frac{3}{2}+\beta}(\omega)) \cap L^r([0,T];H^{\frac{3}{2}+\beta}(\omega))
\]
and \( \partial_{x_d}u(\cdot,-\ell,t) = 0 \). Next we split the solution \( v \) to problem (4.6) into \( v = v_1 + v_2 \), with the functions \( v_1 \) and \( v_2 \), respectively, solving
\[
\begin{cases}
\partial_{x_d}^2v_1 + A(t)v_1 = 0 & \text{in } \Omega \times (0,T), \\
v_1(\cdot,\ell,\cdot) = \partial_{x_d}u(\cdot,\ell,\cdot) & \text{on } \omega \times (0,T), \\
v_1(\cdot,-\ell,\cdot) = 0 & \text{on } \omega \times (0,T), \\
v_1(\cdot,0) = 0 & \text{on } \Omega,
\end{cases}
\]
and
\[
\begin{cases}
\partial_{x_d}^2v_2 + A(t)v_2 = H + f\partial_{x_d}R & \text{in } \Omega \times (0,T), \\
v_2(\cdot,\ell,\cdot) = 0 & \text{on } \partial\Omega \times (0,T), \\
v_2(\cdot,0) = 0 & \text{on } \partial\Omega.
\end{cases}
\]
Next we bound \( v_1 \) and \( v_2 \). To bound \( v_1 \), we first extend \( \partial_{x_d}u(\cdot,\ell,\cdot) \) from \( \omega \times (0,T) \) to \( \Omega \times (0,T) \). Indeed, by the regularity estimate (4.7) and using the classical lifting theorem for Sobolev spaces [30, Chapter 1, Theorem 9.4] and Assumption H00, we deduce that there exists a function \( G \in L^r(0,T;H^2(\Omega)) \) satisfying
\[
-\Delta_x G(x,t) = 0, \quad (x,t) \in Q,
\]
\[
G(x',x_d,t) = \begin{cases} 
\partial_{x_d}u(x',\ell,t), & x_d = \ell, (x',t) \in \omega \times (0,T), \\
0, & x_d = -\ell, (x',t) \in \omega \times (0,T), \\
0, & t = 0, (x',x_d) \in \omega \times (-\ell, \ell).
\end{cases}
\]
Clearly, $G$ satisfies the following estimate:

$$
\|G\|_{L^r(0,T;H^2(\Omega))} \leq c \|\partial_x u(\cdot, \cdot, \cdot)\|_{L^r(0,T;H^2(\omega))}.
$$

Moreover, by interpreting $G$ as the solution to (4.8)–(4.9) in the transposition sense (see, e.g., [30, Chapter 2, Theorem 6.3] or [5]), $G \in W^{\alpha,r}(0,T;L^2(\Omega))$ satisfies

$$
\|G\|_{W^{\alpha,r}(0,T;L^2(\Omega))} \leq c \|\partial_x u(\cdot, \cdot, \cdot)\|_{W^{\alpha,r}(0,T;L^2(\omega))}.
$$

Consequently, we have

$$
\|G\|_{W^{\alpha,r}(0,T;L^2(\Omega))} + \|G\|_{L^r(0,T;H^2(\Omega))} \leq \|\partial_x u(\cdot, \cdot, \cdot)\|_{X_{\alpha,r}}.
$$

Then we can decompose $v_1$ into $v_1 = G + w_1$, with the function $w_1$ solving

$$
\begin{cases}
\partial_t^\alpha w_1 + A(t)w_1 = F_1 & \text{in } \Omega \times (0,T), \\
w_1 = 0 & \text{on } \partial\Omega \times (0,T), \\
w_1(0) = 0 & \text{in } \omega \times (-\ell, \ell),
\end{cases}
$$

with $F_1 = -\partial_t^\alpha G - A(t)G$. Since $\partial_x u(x', \ell, 0) = 0$ for $x' \in \omega$, the uniqueness of the solution of problem (4.8)–(4.9) implies $G(\cdot, 0) = 0$. Then direct computation with Lemma 2.4 gives

$$
\|F_1\|_{L^r(0,T;L^2(\Omega))} \leq \|\partial_t^\alpha G\|_{L^r(0,T;L^2(\Omega))} + \|A(t)G\|_{L^r(0,T;L^2(\Omega))}
$$

(4.11)

$$
\leq c(\|G\|_{W^{\alpha,r}(0,T;L^2(\Omega))} + \|G\|_{L^r(0,T;H^2(\Omega))}).
$$

Thus using the operator $A(t)$ and the perturbation argument, we have

$$
w_1(t_*) = \int_0^{t_*} E_*(t_* - s)F_1(s)ds + \int_0^{t_*} E_*(t_* - s)(A_* - A(s))w_1(s)ds.
$$

By Lemmas 2.1 and 2.3,

$$
\|A_\beta^\alpha w_1(t_*)\|_{L^2(\Omega)} \leq \int_0^{t_*} \|A_\beta^\alpha E_*(t_* - s)\| F_1(s)\|_{L^2(\Omega)}ds
$$

$$
+ \int_0^{t_*} \|A_* E_*(t_* - s)\| \|A_\beta^\alpha (I - A_*^{-1})A(s)w_1(s)\|_{L^2(\Omega)}ds
$$

$$
\leq c \int_0^{t_*} (t_* - s)^{(1-\beta)\alpha - 1}\|F_1(s)\|_{L^2(\Omega)}ds + c \int_0^{t_*} (t_* - s)^{-1}(t_* - s)^{-1}\|A_\beta^\alpha w_1(s)\|_{L^2(\Omega)}ds
$$

$$
\leq c\|F_1\|_{L^r(0,T;L^2(\Omega))} + c \int_0^{t_*} \|A_\beta^\alpha w_1(s)\|_{L^2(\Omega)}ds.
$$

It follows from this estimate and Gronwall’s inequality that $w_1 \in L^\infty(0,T;D(A_\beta^\alpha))$ with

$$
\|w_1\|_{L^\infty(0,T;D(A_\beta^\alpha))} \leq c_T\|F_1\|_{L^r(0,T;L^2(\Omega))}.
$$

Then by the triangle inequality, (4.11), and Assumption H00,

$$
\|v_1\|_{L^\infty(0,T;D(A_\beta^\alpha))} \leq \|w_1\|_{L^\infty(0,T;D(A_\beta^\alpha))} + \|G\|_{L^\infty(0,T;D(A_\beta^\alpha))}
$$

$$
\leq c\|F_1\|_{L^r(0,T;L^2(\Omega))} + \|G\|_{L^\infty(0,T;H^2(\Omega))}.
$$
Meanwhile, the condition \( \beta \in (\frac{4}{3}, \gamma) \) and [4, Theorem 5.2] imply the following embedding inequality:

\[
\|w\|_{L^\infty(0,T;H^{2\beta}(\Omega))} \leq c(\|w\|_{W^{\infty,r}(0,T;L^2(\Omega))} + \|w\|_{L^r(0,T;H^2(\Omega))}).
\]

The last two estimates together give

\[
\|v_1\|_{L^\infty(0,T;D(A^\alpha_u))} \leq c(\|G\|_{W^{\infty,r}(0,T;L^2(\Omega))} + \|G\|_{L^r(0,T;H^2(\Omega))}).
\]

This and estimate (4.10) imply

\[
(4.12) \quad \|v_1\|_{L^\infty(0,T;D(A^\alpha_u))} \leq c\|\partial_{x_d}u(\cdot, \ell, \cdot)\|_{X_{\alpha, r}}.
\]

Moreover, by Assumption H00 and the trace inequality

\[
\|\partial_{x_d}v_1(\cdot, \ell, \cdot)\|_{L^\infty(0,T;L^2(\omega))} \leq c\|v_1\|_{L^\infty(0,T;D(A^\alpha_u))},
\]

we get

\[
(4.13) \quad \|\partial_{x_d}v_1(\cdot, \ell, \cdot)\|_{L^\infty(0,T;L^2(\omega))} \leq c\|\partial_{x_d}u(\cdot, \ell, \cdot)\|_{X_{\alpha, r}}.
\]

Next we bound \( v_2 \). Note that the solution \( v_2(t) \) can be represented by

\[
v_2(t) = \int_0^t E_\ast(t-s)[H(s) + \partial_{x_d}F(s)]ds + \int_0^t E_\ast(t-s)(A_\ast - A(s))v_2(s)ds.
\]

Thus, by Lemmas 2.1 and 2.3 and Assumption 3.1, we get

\[
\|A^\alpha_u v_2(t)\|_{L^2(\Omega)} \leq c\int_0^t (t-s)^{-\alpha(1-\beta)-1}[\|f(s)\|_{L^2(\omega)} + \|H(s)\|_{L^2(\Omega)}]ds
\]

\[
+ c\int_0^t \|A^\beta v_2(s)\|_{L^2(\Omega)}ds.
\]

In light of Assumption 2.1(ii) and the definition \( v = \partial_{x_d}u \), we have

\[
H(t) = -\partial_{x_d}A(t)u
\]

\[
= \partial_{x_d}a_{dd}(t)\partial^2_{x_d}u + 2\sum_{j=1}^{d-1} \partial_{x_d}a_{jd}(t)\partial_{x_j}\partial_{x_d}u + \sum_{j=1}^{d} \partial_{x_j}\partial_{x_d}a_{jd}(t)\partial_{x_d}u
\]

\[
+ \sum_{j=1}^{d-1} \partial^2_{x_d}a_{jd}(t)\partial_{x_j}u - \partial_{x_d}q(t)u
\]

\[
= \partial_{x_d}a_{dd}(t)\partial_{x_d}v + 2\sum_{j=1}^{d-1} \partial_{x_d}a_{jd}(t)\partial_{x_j}v + \left(\sum_{j=1}^{d} \partial_{x_j}\partial_{x_d}a_{jd}(t)\right)v
\]

\[
+ \sum_{j=1}^{d-1} \partial^2_{x_d}a_{jd}(t)\partial_{x_j}u - \partial_{x_d}q(t)u,
\]

from which it directly follows that

\[
\|H(t)\|_{L^2(\Omega)} \leq c(\|v(t)\|_{H^1(\Omega)} + \|u(t)\|_{H^1(\Omega)}), \quad t \in (0,T].
\]
By Lemma 3.1 with $\beta = \frac{1}{2}$ (which holds also for problem (4.1)) and Assumption 3.1, we have
\[ \|u(t)\|_{H^1(\Omega)} \leq c \int_0^t (t - s)^{\frac{3}{2} - 1}\|f(s)\|_{L^2(\omega)} ds. \]
This and (4.12) lead to
\[ \|H(t)\|_{L^2(\omega)} \leq c \int_0^t (t - s)^{\frac{3}{2} - 1}\|f(s)\|_{L^2(\omega)} ds + c(\|\partial_x u(\cdot, \ell, \cdot)\|_{X_{a,r}} + \|v_2(t)\|_{H^1(\Omega)}), \]
which together with (4.14) yields
\[ \|A^2_v v_2(t)\|_{L^2(\Omega)} \leq c \int_0^t (t - s)^{(1 - \beta)\alpha - 1}\|f(s)\|_{L^2(\omega)} ds + c(\|\partial_x u(\cdot, \ell, \cdot)\|_{X_{a,r}} + \|v_2(s)\|_{L^2(\Omega)}) ds. \]
This estimate and Gronwall’s inequality in Lemma 2.5 then imply
\[ \|A^2_v v_2(t)\|_{L^2(\Omega)} \leq c \int_0^t (t - s)^{(1 - \beta)\alpha - 1}\|f(s)\|_{L^2(\omega)} ds + c(\|\partial_x u(\cdot, \ell, \cdot)\|_{X_{a,r}}. \]
It follows from this estimate, Assumption H00, and the trace inequality that
\[ \|\partial_x v_2(\cdot, \ell, t)\|_{L^2(\omega)} \leq c \int_0^t (t - s)^{(1 - \beta)\alpha - 1}\|f(s)\|_{L^2(\omega)} ds + c(\|\partial_x u(\cdot, \ell, \cdot)\|_{X_{a,r}}. \]
Finally, combining this bound with estimate (4.13) yields the desired assertion.

Now we can state a conditional stability result for ISPd.

**Theorem 4.3.** Let Assumptions 2.1, 3.1, H00, H01, and 4.1 be fulfilled, and let $u$ be the solution of problem (4.1). Then there exists a constant $c$ depending on $R, \Omega, T, \alpha, \gamma,$ and $A$ such that
\[ \|f\|_{L^\infty(0,T;L^2(\omega))} \leq c \left( \|\partial_x u(\cdot, \ell, \cdot)\|_{L^\infty(0,T;H^\frac{3}{2}(\omega))} + \|\partial_x u(\cdot, \ell, \cdot)\|_{W^{\frac{1}{2},\frac{3}{2}}(0,T;L^2(\omega))} \right). \]

**Proof.** First projecting the governing equation in (4.1) onto the lateral boundary $\omega \times \{\ell\} \times (0,T)$ and then using the fact that, for all $(x', t) \in \omega \times (0,T)$, we have $u(x', t, 0) = 0$, we thus obtain, for any $(x', t) \in \omega \times (0,T)$,
\[ f(x', t) R(x', \ell, t) = - \left[ a_{dd}(t) \partial^2_{x_d} u + \sum_{j=1}^{d-1} a_{jd}(t) \partial_x \partial_{x_j} u + \sum_{j=1}^d \partial_x a_{jd}(t) \partial_{x_d} u \right](x', \ell, t). \]
This, Assumption 3.1, and the definition $v = \partial_x u$ imply that for all $t \in (0,T)$, there holds
\[ \|f(t)\|_{L^2(\omega)} \leq c_{R}^{-1} c \left( \|\partial_x u(\cdot, \ell, t)\|_{L^2(\omega)} + \|\partial_x u(\cdot, \ell, t)\|_{H^1(\omega)} \right). \]
Under the condition $\gamma \in \left(\frac{3}{2}, 1\right)$, the choice $r = \frac{1}{\alpha(1-\gamma)}$ [4, Theorem 5.2] implies
\[ \|\partial_x u(\cdot, \ell, t)\|_{L^\infty(0,T;H^\frac{3}{2}(\omega))} \leq c \left( \|\partial_x u(\cdot, \ell, t)\|_{W^{\alpha,\gamma}(0,T;L^2(\omega))} + \|\partial_x u(\cdot, \ell, t)\|_{L^\infty(0,T;H^\frac{3}{2}(\omega))} \right). \]

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The last two estimates and Lemma 4.2 imply
\[
\|f(t)\|_{L^2(\omega)} \leq c\|\partial_x u(\cdot, \ell, \cdot)\|_{X_{\alpha,r}} + c \int_0^t (t-s)^{(1-\beta)-1}\|f(s)\|_{L^2(\omega)} ds.
\]

Then Gronwall’s inequality in Lemma 2.5 implies the desired assertion, completing the proof of the theorem.

Remark 4.4. Theorem 4.3 shows the influence of the fractional order $\alpha$ on the stability: the larger the order $\alpha$, the stronger the temporal regularity $\|\partial_x u(\cdot, \ell, \cdot)\|_{W^{\alpha,\infty}(0,T;L^2(\omega))}$ on the data $u(\omega \times (t) \times (0,T)$ the stability needs. This agrees with the smoothing property of the solution operator, and shows also the beneficial influence of anomalous diffusion. Theorem 4.3 improves the corresponding result in [24, Theorem 1.4] (with $\delta > \frac{1}{2}$):
\[
\|f\|_{L^{\infty}(0,T;L^2(\omega))} \leq c(\|\partial_x u(\cdot, \ell, \cdot)\|_{L^{\infty}(0,T;\dot{H}^{\frac{1}{2}}(\omega))} + \|\partial_x u(\cdot, \ell, \cdot)\|_{W^{1,\infty}(0,T;\dot{H}^{\frac{1}{2}}(\omega))}),
\]
This improvement is achieved by the maximal $L^p$ regularity and the suitable interpolation inequality in fractional Sobolev spaces.

Remark 4.5. In the spirit of [24, Corollary 1.5], Theorem 4.3 allows one to prove the stable recovery of a class of the zeroth-order coefficient $q$ from the flux data $\partial_x u|_{\omega \times (t) \times (0,T)}$. This analysis requires the existence of a solution to problem (4.1) in $W^{1,\infty}(0,T,W^{1,\infty}(\Omega)) \cap L^{\infty}(0,T,W^{2,\infty}(\Omega))$. The latter can be achieved using the argument of Proposition 4.1, and we leave the details to future investigation.

5. Numerical experiments and discussions. In this section, we present several numerical experiments to illustrate the feasibility of recovering the space-time-dependent $f$ from lateral boundary observation.

5.1. Numerical algorithm. First we describe a numerical algorithm for recovering $f$ for ISPn (and the algorithm for ISPd is similar). We employ an iterative regularization technique, which approximately minimizes

\[
J(f) := \frac{1}{2}\|u_f - fR\|_{L^2(\omega)}^2,
\]

where $u_f$ denotes the solution to the direct problem (3.1) with $F = fR$. By Corollary 3.2, the map $u_f : L^2(0,T;L^2(\omega)) \rightarrow L^2(0,T;L^2(\omega))$ is linear and compact, and thus standard regularization theory [7, 15] can be applied to justify the reconstruction technique. In particular, when equipped with an appropriate stopping criterion, the approximate minimizer obtained by gradient-type methods, e.g., gradient descent and conjugate gradient (CG) methods, will converge to the exact source component $f^1$ as the noise level tends to zero, and further it will converge at a certain rate dependent on the “regularity” of $f^1$ (in the sense of source condition or conditional stability estimates), when equipped with a suitable stopping criterion.

To (approximately) minimize the functional $J(f)$, we employ the CG method [3]. When applying the method, the main computational effort is to compute the gradient, which can be done efficiently using the adjoint technique. Specifically, let $v$ be the
solution to the following adjoint problem:

\[
\begin{aligned}
& \frac{d}{dt}I_T^{1-\alpha}v(x, t) = 0 \quad \text{in} \Omega, \\
& \partial_{x_d} v(x', \ell, t) = u_f - g^\delta \quad \text{on} \omega \times (0, T), \\
& \partial_{x_d} v(x', -\ell, t) = 0 \quad \text{on} \omega \times (0, T), \\
& v(x, t) = 0 \quad \text{on} \partial \omega \times (-\ell, \ell) \times (0, T),
\end{aligned}
\]

(5.2)

where the notation \( I_T^{1-\alpha}v(t) \) and \( d\partial_T^\alpha v \) denotes the right-sided Riemann–Liouville fractional integral and derivative of \( v \), defined respectively by \cite{26}

\[
\begin{aligned}
& iI_T^{1-\alpha}v(t) = \frac{1}{\Gamma(1-\alpha)} \int_t^T (s-t)^{-\alpha} v(s) \, ds \quad \text{and} \\
& R\partial_T^\alpha v(t) = -\frac{1}{\Gamma(1-\alpha)} \frac{d}{dt} \int_t^T (s-t)^{-\alpha} u(s) \, ds.
\end{aligned}
\]

Then we have the following representation of the gradient \( J'(f) \) of \( J(f) \).

**PROPOSITION 5.1.** The gradient \( J'(f) \) of the functional \( J(f) \) is given by

\[
J'(f) = \int_{-\ell}^\ell Rv \, dx_d,
\]

where \( v \) is the solution to the adjoint problem (5.2).

**Proof.** The derivation follows a standard procedure. The directional derivative \( J'(f)[h] \) of the functional \( J \) with respect to \( f \) in the direction \( h \in L^2(0, T; L^2(\omega)) \) is given by

\[
J'(f)[h] = (u_h, u_f - g^\delta)_{L^2(0, T; L^2(\omega))},
\]

where \( u_h \) is the solution to problem (3.1) with \( h \) in place of \( f \) (or the source \( F = hR) \). Multiplying the equation for \( u_h \) with a test function \( \phi(x, t) \) and then applying integration by parts yields

\[
\int_0^T \int_\Omega (\phi \partial_t^\alpha u_h + a \nabla u_h \cdot \nabla \phi) \, dx \, dt = \int_0^T \int_\Omega Rh \phi \, dx \, dt.
\]

(5.4)

Meanwhile, the weak formulation for the adjoint solution \( v \) is given by

\[
\int_0^T \int_\Omega (\phi R\partial_T^\alpha v + a \nabla v \cdot \nabla \phi) \, dx \, dt = \int_0^T \int_\omega (u_f - g^\delta) \phi \, dx' \, dt.
\]

(5.5)

Then taking \( \phi = v \) in (5.4) and \( \phi = u_h \) in (5.5), appealing to the identity (\cite[p. 76, Lemma 2.7]{26} or \cite[Lemma 2.6]{17})

\[
\int_0^T \int_\Omega v \partial_t^\alpha u_h \, dx \, dt = \int_0^T \int_\Omega u_hR\partial_T^\alpha v \, dx \, dt
\]

(in view of the zero initial/terminal conditions) and subtracting the two identities gives

\[
\int_0^T \int_\Omega Rh \, dx \, dt = \int_0^T \int_\omega (u_f - g^\delta) u_h \, dx' \, dt.
\]

This and the definition of the derivative \( J'(f) \) show the desired assertion. \( \square \)
The next result gives the regularity of the adjoint variable $v$.

**Theorem 5.2.** Let $g^\delta \in L^2(0,T;L^2(\omega))$, and let Assumption H11 be fulfilled. Then there exists a unique solution $v \in L^2(0,T;H^{\frac{1}{2}-\epsilon}(\Omega)) \cap W^{\frac{\epsilon}{2}-\epsilon,2}(0,T;L^2(\Omega))$ for any $\epsilon > 0$ to the adjoint problem (5.2).

**Proof.** For any fixed $t \in [0,T]$, let $N(t)$ be the Neumann map defined by $\phi = N(t)\psi$, with $\phi$ solving

$$
\begin{cases}
  \mathcal{A}(t)\phi = 0 & \text{in } \Omega, \\
  \partial_{\nu}\phi(x',-\ell) = 0 & \text{on } \omega, \\
  \partial_{\nu}\phi(x',\ell) = \psi & \text{on } \omega, \\
  \phi(x) = 0 & \text{on } \partial\omega \times (-\ell,\ell).
\end{cases}
$$

(5.6)

It is known that $\|N(t)\psi\|_{H^{\frac{1}{2}-\epsilon}(\Omega)} \leq c\|\psi\|_{L^2(\Omega)}$ with a range $\mathcal{R}(N(t)) = D(\tilde{A}^{\frac{1}{2}-\epsilon})$, for any small $\epsilon > 0$ [1, Proposition 2.12]. Below we first analyze the case of time-independent coefficients, and then the case of time-dependent coefficients. Let $\psi = u_f - g^\delta \in L^2(0,T;L^2(\omega))$.

**Case (i):** $\mathcal{A}(t) = \mathcal{A}_*$ and $N(t) \equiv N_*$. Note that the solution $v$ to problem (5.2) can be represented by using the operators $E(t)$ and $N(t)$ [25, section 2.2]:

$$
v(t) = \int_t^T \mathcal{A}_* E_* (s-t) N_* \psi(s) \, ds.
$$

(5.7)

Thus, by Lemma 2.1 and Young’s inequality, for any $\theta \in [0,\frac{3}{4})$ and $\epsilon \in (0,2-2\theta - \frac{1}{2})$, there holds

$$
\|v\|_{L^2(0,T;D(\mathcal{A}_*^\theta))} \leq \int_0^T \|\mathcal{A}_*^{\theta+\frac{1}{2}+\frac{\epsilon}{2}} E_* (t)\|\, dt \left( \int_0^T \|\tilde{A}^{\frac{1}{2}-\epsilon} N_* \psi(t)\|_{L^2(\Omega)}^2 \, dt \right)^{\frac{1}{2}}
\leq c_* \|\psi\|_{L^2(0,T;L^2(\omega))}.
$$

It follows from (5.2) that

$$
\frac{\partial}{\partial t} \mathcal{A}_* v + \mathcal{A}_* (v - N_* \psi) = 0,
$$

which implies $\frac{\partial}{\partial t} \mathcal{A}_* v \in L^2(0,T;D(\mathcal{A}_*^{\theta-1}))$. Then by an argument similar to that in the proof of Lemma 2.4, we deduce $v \in W^{\alpha,2}(0,T;D(\mathcal{A}_*^{\theta-1}))$ (see also [20, Theorem 2.1]). Then by interpolation, we derive $v \in W^{\frac{\alpha}{2}-\epsilon,2}(0,T;L^2(\Omega))$ for any $\epsilon > 0$ [4, Theorem 5.2]. Further, in view of the identity

$$
t I^{\frac{1}{2}-\alpha}_T v(t) = \int_t^T \mathcal{A}_* \tilde{S}_*(s-t) N_* \psi(s) \, ds
$$

and Young’s inequality, we deduce for any $\theta \in (1-\frac{1}{2\alpha},\frac{3}{4})$

$$
\|t I^{\frac{1}{2}-\alpha}_T v(t)\|_{L^2(\Omega)} \leq \int_t^T \|\tilde{A}_*^{\frac{1}{2}-\alpha} \tilde{S}_*(s-t)\| \|\tilde{A}_*^\theta N_* \psi(s)\|_{L^2(\Omega)} \, ds
\leq c \int_t^T (s-t)^{-(1-\theta)\alpha} \|\psi(s)\|_{L^2(\omega)} \, ds \leq c \left( \int_t^T (s-t)^{-(1-\theta)\alpha} \, ds \right)^{\frac{1}{2}} \|\psi\|_{L^2(0,T;L^2(\omega))}
\leq c (T-t)^{\frac{1}{2}-(1-\theta)\alpha} \|\psi\|_{L^2(0,T;L^2(\omega))}.
$$

Therefore the terminal condition $t I^{\frac{1}{2}-\alpha}_T v(T) = 0$ holds.
Case (ii): time-dependent elliptic operator $A(t)$. We rewrite the adjoint problem (5.2) as

$$
\partial_t^\alpha v(t) + \tilde{A}_* (v - \psi(t)) = (\tilde{A}_* - \tilde{A}(t))(v - \psi(t)),
$$

with $\partial_T^{1-\alpha} v(T) = 0$. Then the solution $v(t)$ can be represented as

$$
v(t) = \int_t^T \tilde{A}_* \tilde{E}_*(s-t)N(t)s) \, ds + \int_t^T \tilde{A}_* \tilde{E}_*(s-t)((\tilde{A}_* - \tilde{A}(s))(v(s) - N(s)\psi(s)) \, ds.
$$

By Lemmas 2.1 and 2.3, we deduce that for any $\theta \in [0, \frac{3}{4})$ and $\epsilon \in (0, 2 - 2\theta - \frac{1}{2})$,

$$
\|v(t_*)\|_{D(\tilde{A}_*)} \leq c \int_{t_*}^T (s - t_*)^{-(\theta + \frac{1}{2} + \frac{1}{2})}\|\psi(s)\|_{L^2(\omega)} \, ds + c \int_{t_*}^T \|\psi(s)\|_{L^2(\omega)} \, ds.
$$

Then squaring both sides of (5.10) and integrating over $[t_0, T]$ leads to

$$
\|v(t)\|_{L^2(t_0, T; D(\tilde{A}_*)))}^2 \leq c_\epsilon \|\psi\|_{L^2(0, T; L^2(\omega))}^2 + c \int_{t_0}^T \|v\|_{L^2(t; D(\tilde{A}_*)))}^2 \, dt.
$$

This together with Gronwall’s inequality implies that for any $\theta \in [0, \frac{3}{4})$

$$
\|v\|_{L^2(0, T; D(\tilde{A}_*)))} \leq c\|\psi\|_{L^2(0, T; L^2(\omega))}.
$$

By (5.8), $\partial_\alpha \partial^2 v \in L^2(0, T; D(\tilde{A}_*^{-1}))$, and by Lemma 2.4, $v \in W^{\alpha, 2}(0, T; D(\tilde{A}_*^{-1}))$. Then by interpolation, we derive $u \in W^{\frac{3}{2} - \epsilon, 2}(0, T; L^2(\Omega))$ for any $\epsilon > 0$ [4, Theorem 5.2].

Remark 5.3. It follows from Theorem 5.2 that for data $g^\delta \in L^2(0, T; L^2(\omega))$, the gradient $J'(f)$ belongs to $L^2(0, T; H^{\frac{3}{2} - \epsilon}(\omega)) \cap W^{\frac{3}{2} - \epsilon, 2}(0, T; L^2(\omega))$ for any small $\epsilon > 0$ if the factor $R$ is smooth, and, further, $\partial_T^{1-\alpha} J'(f)(x', t) = 0$ for $x' \in \omega$ and $J'(f)(x', t) = 0$ for $(x', t) \in \partial \omega \times (0, T)$. These conditions will impact the convergence behavior of the CG method, dependent on the regularity of $f^\dagger$.

Now we can describe the conjugate gradient method [3] for minimizing $J$. The complete procedure is listed in Algorithm 5.1. In the algorithm, steps 6–7 compute the conjugate descent direction, and step 8 computes the optimal step size using the sensitivity problem. In general, the algorithm converges within tens of iterations; see the numerical experiments below. At each iteration, the algorithm involves three forward solves (direct problem, adjoint problem, and sensitivity problem), which represent the main computational effort. For the stopping criterion at step 11, we employ the discrepancy principle [34, 7, 15], i.e.,

$$
k^* = \arg \min \{k \in \mathbb{N} : \|u_{fs} - g^\delta\|_{L^2(0, T; L^2(\omega))} \leq c\delta\},
$$

where $c > 1$ and $\delta = \|g^1 - g^\delta\|_{L^2(0, T; L^2(\omega))}$ is the noise level of the data $g^\delta$.

Algorithm 5.1 can also be applied to ISPd by viewing the zero Dirichlet data on $\omega \times \{t\}$ as the measurement, and then the measurement $\partial_{x_4} u|_{\omega \times \{t\} \times (0, T)}$ as the Neumann data on $\omega \times \{t\} \times (0, T)$ for problem (3.1). However, the discrepancy principle (5.11) cannot be applied directly, due to a lack of the noise level for the Dirichlet boundary data.
Algorithm 5.1. Conjugate gradient method for minimizing the functional $J$ in (5.1).

1: Initialize $f^0$, and set $k = 0$.
2: for $k = 0, \ldots, K$ do
3:   Solve for $u^k$ from problem (3.1) with $F = f^kR$, and compute the residual $r^k = u^k|_{\Omega \times (0,T)} - g^\delta$.
4:   Solve for $u^k$ from problem (5.2) with $r^k$.
5:   Compute the gradient $J'(f^k)$ by (5.3).
6:   Compute the conjugate coefficient $\gamma^k$ by
   
   $\gamma^k = \begin{cases} 0, & k = 0, \\ \frac{\|J'(f^{k-1})\|_{L^2(0,T;L^2(\Omega))}^2}{\|J'(f^{k-1})\|_{L^2(0,T;L^2(\Omega))}^2}, & k \geq 1. \end{cases}$
7:   Compute the conjugate direction $d^k$ by $d^k = -J'(f^k) + \gamma^k d^{k-1}$.
8:   Solve for $u_{db}$ from problem (3.1) with $F = d^kR$.
9:   Compute the step size $s^k$ by
   
   $s^k = -\frac{(u_{db}, r^k)_{L^2(0,T;L^2(\Omega))}}{\|u_{db}\|_{L^2(0,T;L^2(\Omega))}^2}$.
10: Update the source component $f^{k+1} = f^k + s^kd^k$.
11: Check the stopping criterion.
12: end for

5.2. Numerical results and discussions. Now we present several examples to illustrate the feasibility of recovering $f$. The domain $\Omega$ is taken to be the unit square $\Omega = (-\frac{1}{2}, \frac{1}{2})^2$, with $\omega = (-\frac{1}{2}, \frac{1}{2})$, $q \equiv 0$, and the final time $T = 1$. The direct and adjoint problems are all discretized by the standard continuous piecewise linear Galerkin method in space and backward Euler convolution quadrature in time; see [19] for the error analysis for relevant direct problems and the review [18] for various numerical schemes. The domain $\Omega$ is first divided into $M^2$ small squares each of width $1/M$, and then further divided into triangles by connecting the upper right vertex with the lower left vertex of each small square to obtain a uniform triangulation. For the inversion step, we take $M = 100$ and $N = 1000$. The same spatial and temporal mesh is used for approximating $f$. The factor $R(x,t)$ is fixed at $R \equiv 1$. The exact data $g^\dagger$ on the lateral boundary $\omega \times \{\ell\} \times (0, T)$ is obtained by solving the direct problem (1.1) with the exact $f^\dagger$ on a finer mesh. The noisy boundary data $g^\delta$ is generated from the exact data $g^\dagger$ by

$g^\delta(x', t) = g^\dagger(x', t) + \varepsilon\|g^\dagger\|_{L^\infty(\omega \times (0,T))}\xi(x', t) \quad \forall (x', t) \in \omega \times (0, T),$

where $\xi(x', t)$ follows the standard normal distribution, and $\varepsilon$ denotes the relative noise level. In Algorithm 5.1, the maximum number of CG iterations is fixed at 50, and the constant $c$ in (5.11) is taken to be $c = 1.01$. Throughout, we measure the accuracy of a reconstruction $\hat{f}$ by the $L^2$ error $e(\hat{f})$ defined by

$e(\hat{f}) = \|\hat{f} - f^\dagger\|_{L^2(0,T;L^2(\Omega))}.$

5.2.1. Numerical results for ISPn. First we illustrate the case of time-independent coefficients.
Example 5.1. \(a(x_1, x_2) = 1 + \sin(\pi x_1) x_2 (1 - x_2)\) and \(f^1(x_1, t) = (\frac{1}{2} - x_1^2) t (T - t) e^t\).

The numerical results for Example 5.1 are presented in Table 1, where the numbers in parentheses denote the stopping index determined by the discrepancy principle (5.11). For noisy data \(g^\delta\), the method reaches convergence within ten iterations, and thus it is fairly efficient. It is observed that as the relative noise level \(\epsilon\) increases from zero to 5e-2, the error \(e(f)\) also increases, whereas the required number of CG iterations decreases. For a fixed noise level \(\epsilon\), the reconstruction error \(e\) tends to decrease with the order \(\alpha\), and all the reconstructions are fairly accurate; see Figure 1 for typical reconstructions and the associated pointwise errors \(e = f - f^1\) (which slightly abuses the notation \(e\)). These results clearly show the feasibility of recovering \(f\) from the lateral boundary data, corroborating the theoretical results in [24].

**Table 1**
The reconstruction errors \(e\) for Example 5.1.

<table>
<thead>
<tr>
<th>(\alpha) (\backslash) (\epsilon)</th>
<th>0</th>
<th>1e-3</th>
<th>5e-3</th>
<th>1e-2</th>
<th>5e-2</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.25</td>
<td>8.61e-5 (50)</td>
<td>3.87e-4 (13)</td>
<td>7.36e-4 (10)</td>
<td>1.26e-3 (7)</td>
<td>2.33e-3 (4)</td>
</tr>
<tr>
<td>0.50</td>
<td>4.27e-5 (50)</td>
<td>3.91e-4 (10)</td>
<td>6.84e-4 (8)</td>
<td>1.29e-3 (6)</td>
<td>2.19e-3 (3)</td>
</tr>
<tr>
<td>0.75</td>
<td>8.61e-5 (50)</td>
<td>3.62e-4 (16)</td>
<td>5.93e-4 (11)</td>
<td>8.84e-4 (9)</td>
<td>1.71e-3 (4)</td>
</tr>
</tbody>
</table>

![Fig. 1. Reconstructions and the pointwise errors for Example 5.1 with \(\epsilon = 1e-2\) and \(\epsilon = 5e-2\).](image)

Now we give two examples with time-dependent coefficients. The notation \(\chi_S\) denotes the characteristic function of a set \(S\).

Example 5.2. The diffusion coefficient \(a\) is given by \(a(x_1, x_2, t) = (1 + \sin(\pi x_1) x_2 (1 - x_2))(1 + \sin t)\), and consider two different source components:

(i) \(f^1(x_1, t) = \sin(x_1 + \frac{1}{2}) t \sin(T - t) e^t\).

(ii) \(f^1(x_1, t) = \sin(x_1 + \frac{1}{2}) t \sin(T - t) e^t \chi_{[0,0.7]}(t)\).

In case (i), \(f\) is smooth in time, but it is discontinuous for case (ii). The results for Example 5.2 are shown in Table 2. The results for case (i) are largely comparable with that for Example 5.1, and all the observations remain valid; see also Figure 2.
The behavior of ISP_n is largely independent of the fractional order \( \alpha \), due to the good regularity and compatibility of \( f^\dagger \). In sharp contrast, the results for case (ii) exhibit a different trend: for a fixed noise level \( \varepsilon \), the reconstruction error \( e \) increases with the order \( \alpha \), and also it takes more CG iterations to reach the convergence (see also Figure 3). This is attributed to the discontinuity in time of \( f^\dagger \) and the regularity of the adjoint \( v \) in problem (5.2): the temporal regularity of the adjoint \( v \) increases steadily with \( \alpha \); cf. Theorem 5.2, which makes it increasingly harder to approximate a discontinuous \( f^\dagger \). This is clearly visible from the error plots in Figure 4, where the errors around the discontinuity dominate. This is especially pronounced for \( \alpha = 0.50 \) and \( \alpha = 0.75 \).

<table>
<thead>
<tr>
<th>Case</th>
<th>( \alpha ) ( \setminus ) ( \varepsilon )</th>
<th>0</th>
<th>1e-3</th>
<th>5e-3</th>
<th>1e-2</th>
<th>5e-2</th>
</tr>
</thead>
<tbody>
<tr>
<td>(i)</td>
<td>0.25</td>
<td>2.89e-5 (50)</td>
<td>2.98e-4 (13)</td>
<td>1.21e-3 (9)</td>
<td>1.97e-3 (8)</td>
<td>6.08e-3 (5)</td>
</tr>
<tr>
<td></td>
<td>0.50</td>
<td>2.93e-5 (50)</td>
<td>3.07e-4 (12)</td>
<td>1.18e-3 (9)</td>
<td>2.09e-3 (8)</td>
<td>6.14e-3 (5)</td>
</tr>
<tr>
<td></td>
<td>0.75</td>
<td>3.49e-5 (50)</td>
<td>2.61e-4 (13)</td>
<td>8.50e-4 (9)</td>
<td>1.44e-3 (8)</td>
<td>4.24e-3 (5)</td>
</tr>
<tr>
<td>(ii)</td>
<td>0.25</td>
<td>3.69e-4 (50)</td>
<td>4.51e-4 (13)</td>
<td>1.12e-3 (9)</td>
<td>1.84e-3 (8)</td>
<td>5.71e-3 (4)</td>
</tr>
<tr>
<td></td>
<td>0.50</td>
<td>1.66e-3 (50)</td>
<td>1.68e-3 (13)</td>
<td>2.09e-3 (10)</td>
<td>2.61e-3 (9)</td>
<td>6.33e-3 (5)</td>
</tr>
<tr>
<td></td>
<td>0.75</td>
<td>2.99e-3 (50)</td>
<td>3.38e-3 (25)</td>
<td>4.49e-3 (14)</td>
<td>5.34e-3 (11)</td>
<td>8.49e-3 (6)</td>
</tr>
</tbody>
</table>

Fig. 2. Reconstructions and the pointwise errors for Example 5.2(i) with \( \varepsilon = 1e\cdot2 \) and \( \varepsilon = 5e\cdot2 \).

**5.2.2. Numerical results for ISPd.** Now we present two examples for ISPd, with the setting similar to that of Example 5.2.

**Example 5.3.** The diffusion coefficient \( a \) is given by \( a(x_1, x_2, t) = (1 + \sin(\pi(x_1 + \frac{1}{2}))(\frac{1}{4} - x_2^2))(1 + \sin t) \), and consider two different source components:

(i) \( f^\dagger(x_1, t) = \sin(x_1 + \frac{1}{2})\pi t(T - t)e^t \).

(ii) \( f^\dagger(x_1, t) = \sin(x_1 + \frac{1}{2})\pi t(T - t)e^t\chi[0,0.7](t) \).
Fig. 3. The convergence of the error for Example 5.2(ii), where the red dots indicate the stopping index determined by the discrepancy principle (5.11). (Color available online.)

Fig. 4. Reconstructions and the pointwise errors for Example 5.2(ii) with $\varepsilon = 1 e^{-2}$.

Table 3
The reconstruction errors $e$ for Example 5.3.

<table>
<thead>
<tr>
<th>Case</th>
<th>$\alpha \backslash \varepsilon$</th>
<th>$0$</th>
<th>$1e^{-3}$</th>
<th>$5e^{-3}$</th>
<th>$1e^{-2}$</th>
<th>$5e^{-2}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>(i)</td>
<td>0.25</td>
<td>4.51e-3 (50)</td>
<td>4.52e-3 (42)</td>
<td>4.61e-3 (20)</td>
<td>4.84e-3 (18)</td>
<td>6.28e-3 (5)</td>
</tr>
<tr>
<td></td>
<td>0.50</td>
<td>4.50e-3 (50)</td>
<td>4.51e-3 (41)</td>
<td>4.61e-3 (20)</td>
<td>4.86e-3 (18)</td>
<td>6.11e-3 (4)</td>
</tr>
<tr>
<td></td>
<td>0.75</td>
<td>4.47e-3 (50)</td>
<td>4.49e-3 (37)</td>
<td>4.56e-3 (18)</td>
<td>4.70e-3 (13)</td>
<td>5.84e-3 (6)</td>
</tr>
<tr>
<td>(ii)</td>
<td>0.25</td>
<td>3.87e-3 (50)</td>
<td>3.88e-3 (36)</td>
<td>3.97e-3 (20)</td>
<td>4.20e-3 (17)</td>
<td>5.47e-3 (4)</td>
</tr>
<tr>
<td></td>
<td>0.50</td>
<td>3.99e-3 (50)</td>
<td>4.00e-3 (40)</td>
<td>4.10e-3 (22)</td>
<td>4.37e-3 (17)</td>
<td>5.95e-3 (6)</td>
</tr>
<tr>
<td></td>
<td>0.75</td>
<td>4.35e-3 (50)</td>
<td>4.36e-3 (50)</td>
<td>4.60e-3 (33)</td>
<td>4.97e-3 (24)</td>
<td>7.07e-3 (11)</td>
</tr>
</tbody>
</table>

Note that case (ii) does not satisfy the condition of Theorem 4.3. The numerical results for Example 5.3 are shown in Table 3, where the stopping index is taken so that the reconstruction error $e$ is smallest (since the discrepancy principle (5.11) does not apply directly). The observations from Examples 5.1 and 5.2 are still valid, except the algorithm takes more iterations to reach convergence. This might be due to the fact that the approximation of the exact flux data (for the direct problem) is less accurate, which also limits the attainable accuracy of the reconstruction for data.
with low noise level. The results for case (ii) show that for a fixed noise level $\epsilon$, the error $e$ increases with $\alpha$, and also it takes more CG iterations to reach convergence, due to the mismatch between the temporal regularity of $f$ and the gradient $J'(f)$. This is also clear from the error plots in Figures 5 and 6, where the errors around the discontinuity become increasingly dominating as $\alpha$ increases.

These numerical results indicate that indeed it is feasible to recover a space-time-dependent source from the lateral boundary observation in a cylindrical domain for both time-independent and time-dependent diffusion coefficients, and standard regularization techniques, e.g., the conjugate gradient method (when equipped with the discrepancy principle (5.11)), can deliver accurate reconstructions for both exact and noisy data. This provides numerical evidence to the theoretical results in Theorems 3.4 and 4.3.
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