FOURIER TRUNCATION METHOD FOR AN INVERSE SOURCE PROBLEM FOR SPACE-TIME FRACTIONAL DIFFUSION EQUATION

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ABSTRACT. In this article, we study an inverse problem to determine an unknown source term in a space time fractional diffusion equation, whereby the data are obtained at a certain time. In general, this problem is ill-posed in the sense of Hadamard, so the Fourier truncation method is proposed to solve the problem. In the theoretical results, we propose a priori and a posteriori parameter choice rules and analyze them.

1. Introduction

In this work, we consider the inverse problem of finding the source function f in the problem

$$\frac{\partial^{\beta}}{\partial t^{\beta}}u(x,t) = -r^{\beta}(-\Delta)^{\frac{\alpha}{2}}u(x,t) + h(t)f(x), \quad (x,t) \in \Omega_{T},$$

$$u(-1,t) = u(1,t) = 0, \quad 0 < t < T,$$

$$u(x,0) = 0, \quad x \in \Omega,$$

$$u(x,T) = g(x), \quad x \in \Omega,$$
(1.1)

where $\Omega_T = (-1,1) \times (0,T)$, r > 0 is a parameter, $h \in C[0,T]$ is a given function, $\beta \in (0,1)$, $\alpha \in (1,2)$ are fractional order of the time and the space derivatives, respectively, and T > 0 is a final time. The function u = u(x,t) denotes a concentration of contaminant at a position x and time t. The symbol $\frac{\partial^{\beta} u}{\partial t^{\beta}}$ is the Caputo fractional derivative of order β for differentiable function u; it writes

$$\frac{\partial^{\beta}}{\partial t^{\beta}}u(t) = \frac{1}{\Gamma(1-\alpha)} \int_{0}^{t} \frac{u'(s)}{(t-s)^{\beta}} ds,$$

and $\Gamma(.)$ denotes the standard Gamma function. Note that if the fractional order β tends to unity, the fractional derivative $\frac{\partial^{\beta}}{\partial t^{\beta}}u$ converges to the first-order derivative $\frac{du}{dt}$ [6], and thus the problem reproduces the diffusion model. See, e.g., [6, 12] for the definition and properties of Caputo's derivative.

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It is known that the inverse source problem mentioned above is ill-posed in general, i.e., a solution does not always exist, and in the case of existence of a solution, it does not depend continuously on the given data. In fact, from a small noise of a physical measurement, for example (h,g) is noised by observation data $(h^{\varepsilon}, g^{\varepsilon})$ with order of $\varepsilon > 0$

$$||h^{\varepsilon} - h||_{C([0,T])} + ||g^{\varepsilon} - g||_{L^{2}(-1,1)} \le \varepsilon$$
 (1.2)

where we denote $\|\theta\|_{C([0,T])} = \sup_{0 \le t \le T} |\theta(t)|$ for any $\theta \in C([0,T])$. It is well-known that if ε is small, the sought solution f may have a large error. An example for illustrating this is given in Theorem 2.13. Hence some regularization method are required for stable computation of a sought solution.

The inverse source problem attracted many authors and its physical background can be found in [18]. Wei et al [20, 19, 21] studied an inverse source problem in a spatial fractional diffusion equation by quasi-boundary value and truncation methods. Recently, Kirane et al [7, 6] studied conditional well-posedness to determine a space dependent source in one-dimensional and two-dimensional time-fractional diffusion equations. Rundell et al [4, 13] considered an inverse problem for a onedimensional time-fractional diffusion problem. However, the inverse source problem for both time and space fractional is limited. Recently, Tatar et al [17] considered Problem (1.1) with a general source h(t,x)f(x). They show that the inverse source problem is well-posed in the sense of Hadamard except for a finite set of r > 0. However the source function is also unstable in L^2 norm (See Theorem 2.14 below). The topic in this paper is to finding approximate solution. Hence, our purpose is different and not contradict with the results in [17]. Motivated by above reasons, in this study, we apply the Fourier regularization method to establish a regularized solution. We estimate a convergence rate under an a priori bound assumption of the sought solution and a priori parameter choice rule. Because the a priori bound is difficult to obtain in practical application, so we also estimate a convergence rate under the a posteriori parameter choice rule which is independent on the a priori bound.

This article is organized as follows. In Section 2, we give a formula of the source function f and establish some lemmas and theorems which are useful to the next results. Moreover, the ill-posedness of the inverse source problem is also given in this section. In Section 2, we propose Fourier regularization method and give two convergence estimates under an a priori assumption for the exact solution and two regularization parameter choice rules.

2. Inverse source problem

2.1. Formula of the source function. First, we introduce a few properties of the eigenvalues of the operator $(-\Delta)^{\alpha/2}$, see [5, 12].

Theorem 2.1 (Eigenvalues of the fractional Laplacian operator). 1. Each eigenvalues of $(-\Delta)^{\alpha/2}$ is real. The family of eigenvalues $\{\overline{\lambda_k}\}_{k=1}^{\infty}$ satisfy $0 \le \overline{\lambda_1} \le \overline{\lambda_2} \le \overline{\lambda_3} \le \ldots$, and $\overline{\lambda_k} \to \infty$ as $k \to \infty$.

2. We take $\{\overline{\lambda_k}, \phi_k\}$ the eigenvalues and corresponding eigenvectors of the fractional Laplacian operator in Ω with Dirichlet boundary conditions on $\partial\Omega$:

$$-\Delta \phi_k(x) = \overline{\lambda_k} \phi_k(x), \quad x \in \Omega,$$

$$\phi_k(x) = 0, \quad on \ \partial \Omega,$$
 (2.1)

for k = 1, 2, ...

Then we define the operator $(-\Delta)^{\frac{\alpha}{2}}$ by

$$(-\Delta)^{\alpha/2}u := \sum_{k=0}^{\infty} c_k \phi_k(x) = -\sum_{k=0}^{\infty} c_k \overline{\lambda}_k^{\alpha/2} \phi_k(x),$$

which maps $H_0^{\alpha}(\Omega)$ into $L^2(\Omega)$. Let $0 \neq \gamma < \infty$. By $H^{\gamma}(\Omega)$ we denote the space of all functions $g \in L^2(\Omega)$ with the property

$$\sum_{k=1}^{\infty} (1 + \overline{\lambda_k})^{2\gamma} |g_k|^2 < \infty, \tag{2.2}$$

where $g_k = \int_{\Omega} g(x)\phi_k(x)dx$. Then we also define

$$||g||_{H^{\gamma}(\Omega)} = \sqrt{\sum_{k=1}^{\infty} (1 + \overline{\lambda_k})^{2\gamma} |g_k|^2}$$
. If $\gamma = 0$ then $H^{\gamma}(\Omega)$ is $L^2(\Omega)$.

Now we use the separation of variables to yield the solution of (1.1). Suppose that the solution of (1.1) is defined by Fourier series

$$u(x,t) = \sum_{k=1}^{\infty} u_k(t)\phi_k(x), \quad \text{with } u_k(t) = \langle u(.,t), \phi_k(x) \rangle.$$
 (2.3)

Then the eigenfunction expansions can be defined by the Fourier method. That is, we multiply both sides of (1.1) by $\phi_k(x)$ and integrate the equation with respect to x. Using the Green formular and $\phi_k|_{\partial\Omega}=0$, we obtain an uncouple system of initial value problem for the fractional differential equations for the unknown Fourier coefficient $u_k(t)$

$$\frac{\partial^{\beta}}{\partial t^{\beta}} u_k(t) = -r^{\beta} (-\Delta)^{\frac{\alpha}{2}} u_k(t) + h(t) f_k(x), \quad (x, t) \in \Omega \times (0, T),$$

$$u_k(0) = \langle u(x, 0), \varphi_k(x) \rangle. \tag{2.4}$$

From the result in [17], the formula of solution corresponding to the initial value problem for (2.4) is obtained as follows, from u(x, 0) = 0.

$$u(x,t) = \sum_{k=1}^{\infty} \left(\int_0^t \tau^{\beta-1} E_{\beta,\beta} \left(-\left(\frac{k\pi}{2}\right)^{\alpha} r^{\beta} \tau^{\beta} \right) \langle f(x)h(t-\tau), \phi_k(x) \rangle d\tau \right) \phi_k(x). \tag{2.5}$$

By a change variable in the integral, we can rewrite

$$u(x,t) = \sum_{k=1}^{\infty} \left(\int_0^t (t-\tau)^{\beta-1} E_{\beta,\beta} \left(-\left(\frac{k\pi}{2}\right)^{\alpha} r^{\beta} (t-\tau)^{\beta} \right) h(\tau) d\tau \right) \langle f(x), \phi_k(x) \rangle \phi_k(x).$$
(2.6)

Letting t = T in the latter equality, we obtain

$$f(x) = \sum_{k=1}^{\infty} \frac{\langle g(x), \phi_k(x) \rangle \phi_k(x)}{\int_0^T (T-\tau)^{\beta-1} E_{\beta,\beta} \left(-\left(\frac{k\pi}{2}\right)^{\alpha} r^{\beta} (T-\tau)^{\beta}\right) h(\tau) d\tau}.$$
 (2.7)

to abbreviate notation, we set

$$\Phi_{\beta}(\lambda_k^{\alpha}, \tau, r) = (T - \tau)^{\beta - 1} E_{\beta, \beta} \left(-\left(\frac{k\pi}{2}\right)^{\alpha} r^{\beta} (T - \tau)^{\beta} \right),$$

then the source function f is rewritten as

$$f(x) = \sum_{k=1}^{\infty} \frac{\langle g(x), \phi_k(x) \rangle \phi_k(x)}{\int_0^T \Phi_{\beta}(\lambda_k^{\alpha}, \tau, r) h(\tau) d\tau}.$$
 (2.8)

Remark 2.2. Applying [17, Theorem 2.1], we obtain the existence and uniqueness of problem (1.1) such that $u \in L^2(0,T;H^{\alpha}(\Omega))$. The regularity estimate for u as in (2.6) is mentioned in [17] and so, we omit it here.

2.2. **Preliminary results.** Now, we consider the following definition and lemmas which are useful for our main results.

Definition 2.3 ([12]). The Mittag-Leffler function is

$$E_{\alpha,\beta}(z) = \sum_{k=0}^{\infty} \frac{z^k}{\Gamma(\alpha k + \beta)}, \quad z \in \mathbb{C}$$

where $\alpha > 0$ and $\beta \in \mathbb{R}$ are arbitrary constants.

Lemma 2.4 ([14]). For $\lambda > 0$ and $0 < \beta < 1$, we have

$$\frac{d}{dt}E_{\beta,1}(-\lambda t^{\beta}) = -\lambda t^{\beta-1}E_{\beta,\beta}(-\lambda t^{\beta}), \quad t > 0.$$
(2.9)

Lemma 2.5 ([12]). For $\alpha > 0$ and $\beta \in \mathbb{R}$, we have

$$E_{\alpha,\beta}(z) = zE_{\alpha,\alpha+\beta}(z) + \frac{1}{\Gamma(\beta)}, \quad z \in \mathbb{C}.$$

Lemma 2.6 ([14]). The following equality holds for $\lambda > 0$, $\alpha > 0$ and $m \in \mathbb{N}$

$$\frac{d^m}{dt^m}E_{\alpha,1}(-\lambda t^\alpha) = -\lambda t^{\alpha-m}E_{\alpha,\alpha-m+1}(-\lambda t^\alpha), \quad t > 0.$$
 (2.10)

Lemma 2.7 ([17]). If $\alpha \leq 2$, β is arbitrary real number, μ is such that $\frac{\pi\alpha}{2} < \mu < \min\{\pi\alpha, \pi\}$, $\mu \leq |arg(z)| \leq \pi$ then there exists two constants $A_0 > 0$ and $A_1 > 0$ such that

$$\frac{A_0}{1+|z|} \le |E_{\alpha,\beta}(z)| \le \frac{A_1}{1+|z|}. (2.11)$$

Lemma 2.8 ([14]). Let $E_{\beta,\beta}(-\eta) \ge 0$, $0 < \beta < 1$, we have

$$\int_{0}^{M} \left| t^{\beta - 1} E_{\beta,\beta}(-\overline{\lambda_{k}} t^{\beta}) \right| dt = \int_{0}^{M} t^{\beta - 1} E_{\beta,\beta}(-\overline{\lambda_{k}} t^{\beta}) dt$$

$$= -\frac{1}{\overline{\lambda_{k}}} \int_{0}^{M} \frac{d}{dt} E_{\beta,1}(-\overline{\lambda_{k}} t^{\beta}) dt$$

$$= \frac{1}{\overline{\lambda_{k}}} \left(1 - E_{\beta,1}(-\overline{\lambda_{k}} M^{\beta}) \right). \tag{2.12}$$

Lemma 2.9 ([17]). For any $\lambda_k^{\alpha} = (\frac{k\pi}{2})^{\alpha}$ satisfying $\lambda_k^{\alpha} \geq \lambda_1^{\alpha}$ there exists positive constant C depending on $\{\beta, T, \frac{\pi}{2}\}$ such that

$$\frac{C}{r^{\beta}T^{\beta}\lambda_{k}^{\alpha}} \le E_{\beta,\beta+1}(-\lambda_{k}^{\alpha}r^{\beta}T^{\beta}) \le \frac{1}{r^{\beta}T^{\beta}\lambda_{k}^{\alpha}}.$$
(2.13)

Lemma 2.10. Let $h:[0,T]\to\mathbb{R}^+$ be a continuous function such that $|\mathcal{I}(h)|=\inf_{t\in[0,T]}|h(t)|>0$. Set $\|h\|_{C[0,T]}=\sup_{t\in[0,T]}|h(t)|$. Then we have

$$\frac{|\mathcal{I}(h)|(1 - E_{\beta,1}(-\lambda_1^{\alpha} r^{\beta} T^{\beta}))}{\lambda_{\mu}^{\alpha} r^{\beta}} \le \int_0^T \Phi_{\beta}(\lambda_k^{\alpha}, \tau, r) h(\tau) d\tau \le \frac{\|h\|_{C[0,T]}}{\lambda_{\mu}^{\alpha} r^{\beta}}.$$
 (2.14)

A proof of the above lemma can be found in [17].

Theorem 2.11. Let $g \in H^{\alpha}(\Omega)$. Then the inverse source problem has the solution $f \in L^{2}(\Omega)$.

Proof. The solution f exists if and only if the series in the right-hand side of (2.8) converges. Hence, we show this point. Indeed, using Lemma 2.10 and noting that $g \in H^{\alpha}(\Omega)$, we obtain

$$\sum_{k=1}^{\infty} \left| \frac{\langle g(x), \phi_k(x) \rangle}{\int_0^T \Phi_{\beta}(\lambda_k^{\alpha}, \tau, r) h(\tau) d\tau} \right|^2 \ge \sum_{k=1}^{\infty} \frac{\lambda_k^{2\alpha} r^{2\beta} \langle g(x), \phi_k(x) \rangle^2}{\|h\|_{C[0, T]}^2} \\
= \frac{r^{2\beta}}{\|h\|_{C[0, T]}^2} \|g\|_{H^{\alpha}(\Omega)}^2, \tag{2.15}$$

and

$$\sum_{k=1}^{\infty} \left| \frac{\langle g(x), \phi_k(x) \rangle}{\int_0^T \Phi_{\beta}(\lambda_k^{\alpha}, \tau, r) h(\tau) d\tau} \right|^2 \leq \sum_{k=1}^{\infty} \frac{\lambda_k^{2\alpha} r^{2\beta} \langle g(x), \phi_k(x) \rangle^2}{|\mathcal{I}(h)|^2 (1 - E_{\beta, 1}(-\lambda_1^{\alpha} r^{\beta} T^{\beta}))^2}$$

$$= \frac{r^{2\beta}}{|\mathcal{I}(h)|^2 (1 - E_{\beta, 1}(-\lambda_1^{\alpha} r^{\beta} T^{\beta}))^2} \|g\|_{H^{\alpha}(\Omega)}^2.$$
(2.16)

From two latter inequality, we conclude that the series $\sum_{k=1}^{\infty} \frac{\langle g(x), \phi_k(x) \rangle \phi_k(x)}{\int_0^T \Phi_{\beta}(\lambda_k^{\alpha}, \tau, r) h(\tau) d\tau}$ is convergent. The proof is complete.

Theorem 2.12. Let $R:[0,T]\to\mathbb{R}$ be as in Lemma 2.10, then the solution (u(x,t),f(x)) of Problem (1) is unique.

Proof. Let f_1 and f_2 be the source functions corresponding to the final values g_1 and g_2 respectively. Suppose that $g_1 = g_2$ then we prove that $f_1 = f_2$. In fact, it is well-known that $E_{\beta,\beta}(-(\frac{k\pi}{2})^{\alpha}r^{\beta}(t-\tau)^{\beta}) \geq 0$ for $\tau \leq t$. Since $||h||_{C[0,T]} \geq |\mathcal{I}(h)| > 0$ for $t \in [0,T]$, we have

$$\int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h(\tau) d\tau$$

$$\geq |\mathcal{I}(h)| \int_{0}^{T} (T - \tau)^{\beta - 1} E_{\beta, \beta} \left(-\left(\frac{k\pi}{2}\right)^{\alpha} r^{\beta} (T - \tau)^{\beta} \right) d\tau$$

$$= |\mathcal{I}(h)| T^{\beta} E_{\beta, \beta + 1} \left(-\left(\frac{k\pi}{2}\right)^{\alpha} (rT)^{\beta} \right) > 0.$$
(2.17)

We have the estimate

$$f_1(x) - f_2(x) = \sum_{k=1}^{\infty} \frac{\langle g_1(x) - g_2(x), \phi_k(x) \rangle \phi_k(x)}{\int_0^T \Phi_{\beta}(\lambda_k^{\alpha}, \tau, r) h(\tau) d\tau} = 0.$$
 (2.18)

The proof is complete.

Theorem 2.13. The inverse source problem of finding the function f is ill-posed in the Hadamard sense in the L^2 norm.

Proof. Let us Define a linear operator $\mathcal{P}: L^2(\Omega) \to L^2(\Omega)$ as follows

$$\mathcal{P}f(x) = \int_{\Omega} p(x,\omega)f(\omega)d\omega$$

$$= \sum_{k=1}^{\infty} \left[\int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r)h(\tau)d\tau \right] \langle f(x), \phi_{k}(x) \rangle \phi_{k}(x)$$
(2.19)

where

$$p(x,\omega) = \sum_{k=1}^{\infty} \left[\int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h(\tau) d\tau \right] \phi_{k}(x) \phi_{k}(\omega).$$

Because $p(x,\omega) = p(\omega,x)$ we know that \mathcal{K} is self-adjoint operator. Next, we prove its compactness. Defining the finite rank operators \mathcal{K}_N as follows

$$\mathcal{P}_N f(x) = \sum_{k=1}^N \left[\int_0^T \Phi_\beta(\lambda_k^\alpha, \tau, r) h(\tau) d\tau \right] \langle f(x), \phi_k(x) \rangle \phi_k(x). \tag{2.20}$$

Then, from (2.19) and (2.20), we have

$$\|\mathcal{P}_{N}f - \mathcal{P}f\|_{L^{2}(\Omega)}^{2} = \sum_{k=N+1}^{\infty} \left[\int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h(\tau) d\tau \right]^{2} |\langle f(x), \phi_{k}(x) \rangle|^{2}$$

$$\leq \sum_{k=N+1}^{\infty} \frac{\|h\|_{C[0,T]}^{2}}{\lambda_{k}^{2\alpha}} |\langle f(x), \phi_{k}(x) \rangle|^{2}$$

$$\leq \frac{\|h\|_{C[0,T]}^{2}}{\lambda_{N}^{2\alpha}} \sum_{k=N+1}^{\infty} |\langle f(x), \phi_{k}(x) \rangle|^{2}.$$
(2.21)

This implies

$$\|\mathcal{P}_N f - \mathcal{P} f\|_{L^2(\Omega)} \le \left(\frac{\|h\|_{C[0,T]}^2}{\lambda_N^{2\alpha}} \|f\|_{L^2(\Omega)}^2\right)^{1/2} = \frac{\|h\|_{C[0,T]}}{\lambda_N^{\alpha}} \|f\|_{L^2(\Omega)}. \tag{2.22}$$

Therefore, $\|\mathcal{P}_N - \mathcal{P}\| \to 0$ in the sense of operator norm in $L(L^2(\Omega); L^2(\Omega))$ as $N \to \infty$. Also, \mathcal{P} is a compact operator. Next, the singular values for the linear self-adjoint compact operator \mathcal{P} are

$$\psi_{k} = \int_{0}^{T} (T - \tau)^{\beta - 1} E_{\beta, \beta} \left(-\left(\frac{k\pi}{2}\right)^{\alpha} r^{\beta} (T - \tau)^{\beta} \right) h(\tau) d\tau$$

$$= \int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h(\tau) d\tau,$$
(2.23)

and corresponding eigenvectors is ϕ_k which is known as an orthonormal basis in $L^2(\Omega)$. From (2.19), the inverse source problem we introduced above can be formulated as an operator equation.

$$\mathcal{P}f(x) = g(x) \tag{2.24}$$

and by Kirsch [8], we conclude that it is ill-posed. To illustrate an ill-posed problem, we present an example. Let us choose the input final data $g^m(x) = \frac{\phi_m(x)}{\sqrt{r^{2\beta}\lambda_m^{\alpha}}}$. Following (2.9), we know $\lambda_m^{\alpha} = (\frac{m\pi}{2})^{\alpha}$. By (2.8), the source term corresponding to g^m is

$$f^{m}(x) = \sum_{k=1}^{\infty} \frac{\langle g^{m}(x), \phi_{k}(x) \rangle \phi_{k}(x)}{\int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h(\tau) d\tau}$$

$$= \sum_{k=1}^{\infty} \frac{\langle \frac{\phi_{m}(x)}{\sqrt{r^{2\beta}(\frac{m\pi}{2})^{\alpha}}}, \phi_{k}(x) \rangle \phi_{k}(x)}{\int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h(\tau) d\tau}$$

$$= \frac{\phi_{m}(x)}{\sqrt{r^{2\beta}(\frac{m\pi}{2})^{\alpha}} \int_{0}^{T} \Phi_{\beta}(\lambda_{m}^{\alpha}, \tau, r) h(\tau) d\tau}.$$
(2.25)

Let us choose another input final data g = 0. By (2.7), the source term corresponding to g is f = 0. An error in L^2 norm between two input final data is

$$||g^{m} - g||_{L^{2}(\Omega)} = ||\frac{\phi_{m}(x)}{|r^{\beta}|\sqrt{(\frac{m\pi}{2})^{\alpha}}}||_{L^{2}(\Omega)} = \frac{1}{|r^{\beta}|\sqrt{(\frac{m\pi}{2})^{\alpha}}},$$
 (2.26)

where $\alpha \in (1,2)$. Therefore

$$\lim_{m \to +\infty} \|g^m - g\|_{L^2(\Omega)} = \lim_{m \to +\infty} \frac{1}{|r^{\beta}| \sqrt{(\frac{m\pi}{2})^{\alpha}}} = 0.$$
 (2.27)

And an error in $L^2(-1,1)$ norm between two corresponding source term is

$$||f^{m} - f||_{L^{2}(\Omega)} = ||\frac{\phi_{m}(x)}{|r^{\beta}|\sqrt{(\frac{m\pi}{2})^{\alpha}} \int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r)h(\tau)d\tau}||_{L^{2}(\Omega)}$$

$$= \frac{1}{|r^{\beta}|\sqrt{(\frac{m\pi}{2})^{\alpha}} \int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r)h(\tau)d\tau},$$
(2.28)

which we note that $\beta \in (0,1)$ and r is positive number. From (2.28) and using the inequality as in Lemma 2.10, we obtain

$$||f^m - f||_{L^2(\Omega)} \ge \frac{\sqrt{(\frac{m\pi}{2})^{\alpha}}}{||h||_{C[0,T]}}.$$
 (2.29)

This leads to

$$\lim_{m \to +\infty} \|f^m - f\|_{L^2(\Omega)} > \lim_{m \to +\infty} \frac{\sqrt{(\frac{m\pi}{2})^{\alpha}}}{\|h\|_{C[0,T]}} = +\infty.$$
 (2.30)

Combining (2.27) with (2.30), we conclude that the inverse source problem is illposed.

Theorem 2.14 (A conditional stability estimate). Assume that there exists $\gamma > 0$ such that $||f||_{H^{\alpha\gamma}(\Omega)} \leq M$ for M > 0. Then

$$||f||_{L^{2}(\Omega)} \le \mathcal{K}_{\alpha,\beta}(h,r,T)M^{\frac{1}{\gamma+1}} ||g||_{L^{2}(\Omega)}^{\frac{\gamma}{\gamma+1}}.$$
 (2.31)

where

$$\mathcal{K}_{\alpha,\beta}(h,r,T) = \frac{(r^{\beta})^{\frac{\gamma}{\gamma+1}}}{|\mathcal{I}(h)|^{\frac{\gamma}{\gamma+1}} big(1 - E_{\alpha,1}(-\lambda_1^{\alpha}r^{\beta}T^{\beta}))^{\frac{\gamma}{\gamma+1}}}.$$
 (2.32)

Proof. According (2.8), by Hölder's inequality, we have

$$||f||_{L^{2}(\Omega)}^{2} = \sum_{k=1}^{\infty} \left| \frac{\langle g(x), \phi_{k}(x) \rangle \phi_{k}(x)}{\int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h(\tau) d\tau} \right|^{2}$$

$$\leq \sum_{k=1}^{\infty} \frac{|\langle g(x), \phi_{k}(x) \rangle|^{\frac{2}{\gamma+1}} |\langle g(x), \phi_{k}(x) \rangle|^{\frac{2\gamma}{\gamma+1}}}{\left| \int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h(\tau) d\tau \right|^{2}}$$

$$\leq \left(\sum_{k=1}^{\infty} \frac{|\langle g(x), \phi_{k}(x) \rangle|^{2}}{\left| \int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h(\tau) d\tau \right|^{2(\gamma+1)}} \right)^{\frac{1}{\gamma+1}}$$

$$\times \left(\sum_{k=1}^{\infty} |\langle g(x), \phi_{k}(x) \rangle|^{2} \right)^{\frac{\gamma}{\gamma+1}}$$

$$\leq \left(\sum_{k=1}^{\infty} \frac{|\langle f(x), \phi_{k}(x) \rangle|^{2}}{\left| \int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h(\tau) d\tau \right|^{2\gamma}} \right)^{\frac{1}{\gamma+1}} ||g||_{L^{2}(\Omega)}^{\frac{2\gamma}{\gamma+1}}.$$

Here we have used the fact that

$$\langle g(x), \phi_k(x) \rangle = \langle f(x), \phi_k(x) \rangle \Big| \int_0^T \Phi_\beta(\lambda_k^\alpha, \tau, r) h(\tau) d\tau \Big|^2.$$

Using Lemma 2.10, we have

$$\left| \int_0^T \Phi_{\beta}(\lambda_k^{\alpha}, \tau, r) h(\tau) d\tau \right|^{2\gamma} \ge \frac{|\mathcal{I}(h)|^{2\gamma}}{\lambda_k^{2\alpha\gamma} r^{2\beta\gamma}} \left(1 - E_{\beta, 1}(-\lambda_1^{\alpha} r^{\beta} T^{\beta}) \right)^{2\gamma} \tag{2.34}$$

and this inequality leads to

$$\sum_{k=1}^{\infty} \frac{|\langle f(x), \phi_{k}(x) \rangle|^{2}}{\left| \int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h(\tau) d\tau \right|^{2\gamma}} \\
\leq \sum_{k=1}^{\infty} \frac{r^{2\beta\gamma} \lambda_{k}^{2\alpha\gamma} |\langle f(x), \phi_{k}(x) \rangle|^{2}}{|\mathcal{I}(h)|^{2\gamma} \left(1 - E_{\beta,1}(-\lambda_{1}^{\alpha} r^{\beta} T^{\beta})\right)^{2\gamma}} \\
\leq \frac{r^{2\beta\gamma}}{|\mathcal{I}(h)|^{2\gamma} \left(1 - E_{\beta,1}(-\lambda_{1}^{\alpha} r^{\beta} T^{\beta})\right)^{2\gamma}} \sum_{k=1}^{\infty} \lambda_{k}^{2\alpha\gamma} |\langle f(x), \phi_{k}(x) \rangle|^{2} \\
= \frac{r^{2\beta\gamma} ||f||_{H^{\alpha\gamma}(\Omega)}^{2}}{|\mathcal{I}(h)|^{2\gamma} \left(1 - E_{\beta,1}(-\lambda_{1}^{\alpha} r^{\beta} T^{\beta})\right)^{2\gamma}}.$$
(2.35)

Combining (2.33) with (2.35), we obtain

$$||f||_{L^{2}(\Omega)}^{2} \leq \frac{r^{\frac{2\beta\gamma}{\gamma+1}} ||f||_{H^{\alpha\gamma}(\Omega)}^{\frac{2}{\gamma+1}}}{|\mathcal{I}(h)|^{\frac{2\gamma}{\gamma+1}} (1 - E_{\beta,1}(-\lambda_{1}^{\alpha} r^{\beta} T^{\beta}))^{\frac{2\gamma}{\gamma+1}}} ||g||_{L^{2}(\Omega)}^{\frac{2\gamma}{\gamma+1}}$$

$$\leq |\mathcal{K}_{\alpha,\beta}(h,r,T)|^{2} M^{\frac{2}{\gamma+1}} ||g||_{L^{2}(\Omega)}^{\frac{2\gamma}{\gamma+1}}.$$
(2.36)

3. Fourier truncation regularization and error estimate

In this section, we eliminate all the components of large k from the solution and define the truncation regularized solution as follows:

$$f^{\epsilon,N}(x) = \sum_{k=1}^{N} \frac{\langle g^{\epsilon}(x), \phi_k(x) \rangle \phi_k(x)}{\int_0^T \Phi_{\beta}(\lambda_k^{\alpha}, \tau, r) h^{\epsilon}(\tau) d\tau}$$
(3.1)

where the positive integer N plays the role of regularization parameter. Next, we consider an a *a-priori* and an a *a-posteriori* choice to find the regularization parameter. Under each choice of the regularization parameter, the error estimates between the exact solution f given by (2.7) and the regularized approximation solution $f^{\epsilon,N}$ given by (3.1) can be obtained.

3.1. An a priori parameter choice. Afterwards, we will give an error estimation for $||f(x) - f^{\epsilon,N}(x)||_{L^2(\Omega)}$ and show convergence rate under a suitable choice for the regularization parameter.

Theorem 3.1. Let $f^{\epsilon,N}$ be the regularized solution for problem (1.1) with noisy data g^{ϵ} and f(x) be the exact solution for problem (1.1). Let us choose parameter regularization $N = [\mu]$, where $[\mu]$ denotes the largest integer less than or equal to μ . Then we have the following:

- If $0 < \gamma \le 1$ then choose $\mu = \frac{2}{\pi} \left(\frac{M}{\epsilon}\right)^{\frac{1}{\alpha(\gamma+1)}}$, we have the estimate $\|f(x) f^{\epsilon,N}(x)\|_{L^{2}(\Omega)} \le \varepsilon^{\frac{\gamma}{\gamma+1}} M^{\frac{1}{\gamma+1}} \mathcal{D}_{\alpha,\beta}(f,h,h^{\varepsilon},r,T). \tag{3.2}$
- If $\gamma > 1$, choose $\mu = \frac{2}{\pi} \left(\frac{M}{\epsilon} \right)^{\frac{1}{2\alpha}}$, we obtain the error estimate

$$||f(x) - f^{\epsilon, N}(x)||_{L^2(\Omega)} \le \varepsilon^{\frac{1}{2}} M^{\frac{1}{2}} \mathcal{D}_{\alpha, \beta}(f, h, h^{\varepsilon}, r, T), \tag{3.3}$$

where

$$\mathcal{D}_{\alpha,\beta}(f,h,h^{\varepsilon},r,T) = \left[1 + \max\left\{\frac{r^{\beta}}{|\mathcal{I}(h)|(1 - E_{\beta,1}(-\lambda_1^{\alpha}r^{\beta}T^{\beta}))}, \frac{\|f\|_{L^2(\Omega)}}{|\mathcal{I}(h^{\varepsilon})|}\right\}\right]. \quad (3.4)$$

Remark 3.2. If the function h depends on x and t, i.e. h = h(t, x) then we can not represent f as the Fourier series as (2.8). Hence, we can not use some usual regularization methods. The regularized problem is open and difficult when h depends on x and t.

Proof of Theorem 3.1. Using (2.7) and (3.1) and the triangle inequality, we have

$$f(x) - f^{\epsilon,N}(x) = \sum_{k=1}^{\infty} \frac{\langle g(x), \phi_k(x) \rangle \phi_k(x)}{\int_0^T \Phi_{\beta}(\lambda_k^{\alpha}, \tau, r) h(\tau) d\tau} - \sum_{k=1}^N \frac{\langle g^{\epsilon}(x), \phi_k(x) \rangle \phi_k(x)}{\int_0^T \Phi_{\beta}(\lambda_k^{\alpha}, \tau, r) h^{\epsilon}(\tau) d\tau}$$

$$= \sum_{k=1}^{\infty} \frac{\langle g(x), \phi_k(x) \rangle \phi_k(x)}{\int_0^T \Phi_{\beta}(\lambda_k^{\alpha}, \tau, r) h(\tau) d\tau} - \sum_{k=1}^N \frac{\langle g(x), \phi_k(x) \rangle \phi_k(x)}{\int_0^T \Phi_{\beta}(\lambda_k^{\alpha}, \tau, r) h(\tau) d\tau}$$

$$+ \sum_{k=1}^N \frac{\langle g(x), \phi_k(x) \rangle \phi_k(x)}{\int_0^T \Phi_{\beta}(\lambda_k^{\alpha}, \tau, r) h(\tau) d\tau} - \sum_{k=1}^N \frac{\langle g^{\epsilon}(x), \phi_k(x) \rangle \phi_k(x)}{\int_0^T \Phi_{\beta}(\lambda_k^{\alpha}, \tau, r) h^{\epsilon}(\tau) d\tau}.$$
(3.5)

Hence

$$f(x) - f^{\epsilon,N}(x) = \underbrace{\sum_{k=N+1}^{\infty} \frac{\langle g(x), \phi_k(x) \rangle \phi_k(x)}{\int_0^T \Phi_{\beta}(\lambda_k^{\alpha}, \tau, r) h(\tau) d\tau}}_{Q_1} + \underbrace{\sum_{k=1}^N \frac{\langle g(x) - g^{\epsilon}(x), \phi_k(x) \rangle \phi_k(x)}{\int_0^T \Phi_{\beta}(\lambda_k^{\alpha}, \tau, r) h^{\epsilon}(\tau) d\tau}}_{Q_2}$$

$$+ \underbrace{\sum_{k=1}^N \frac{\langle g(x), \phi_k(x) \rangle \phi_k(x)}{\int_0^T \Phi_{\beta}(\lambda_k^{\alpha}, \tau, r) h^{\epsilon}(\tau) d\tau}}_{Q_2} \times \underbrace{\sum_{k=1}^N \frac{\int_0^T \Phi_{\beta}(\lambda_k^{\alpha}, \tau, r) h^{\epsilon}(\tau) d\tau}{\int_0^T \Phi_{\beta}(\lambda_k^{\alpha}, \tau, r) h^{\epsilon}(\tau) d\tau}}_{Q_2}.$$

$$(3.6)$$

First, we have the following estimate

$$\|\mathcal{Q}_{1}\|_{L^{2}(\Omega)}^{2} = \sum_{k=N+1}^{\infty} \frac{|\langle g(x), \phi_{k}(x) \rangle|^{2}}{\left|\int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h(\tau) d\tau\right|^{2}}$$

$$= |\langle f(x), \phi_{k}(x) \rangle|^{2}$$

$$\leq \sum_{k=N+1}^{\infty} (1 + \lambda_{k})^{-2\alpha\gamma} (1 + \lambda_{k})^{2\alpha\gamma} |\langle f(x), \phi_{k}(x) \rangle|^{2}$$

$$\leq (1 + \lambda_{N})^{-2\alpha\gamma} M^{2}.$$
(3.7)

Hence, we obtain

$$\|\mathcal{Q}_1\|_{L^2(\Omega)} \le (1+\lambda_N)^{-\alpha\gamma} M. \tag{3.8}$$

Second, the term $\|Q_2\|_{L^2(\Omega)}$ is bounded by

$$\|\mathcal{Q}_{2}\|_{L^{2}(\Omega)}^{2} \leq \sum_{k=1}^{N} \frac{|\langle g(x) - g^{\epsilon}(x), \phi_{k}(x) \rangle|^{2}}{\left|\int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h^{\epsilon}(\tau) d\tau\right|^{2}}$$

$$\leq \sum_{k=1}^{N} \frac{\lambda_{k}^{2\alpha} r^{2\beta} |\langle g(x) - g^{\epsilon}(x), \phi_{k}(x) \rangle|^{2}}{|\mathcal{I}(h^{\epsilon})|^{2} \left(1 - E_{\beta, 1}(-\lambda_{1}^{\alpha} r^{\beta} T^{\beta})\right)^{2}}$$

$$\leq \sup_{1 \leq k \leq N} \frac{\lambda_{k}^{2\alpha} r^{2\beta}}{|\mathcal{I}(h^{\epsilon})|^{2} \left(1 - E_{\beta, 1}(-\lambda_{1}^{\alpha} r^{\beta} T^{\beta})\right)^{2}}$$

$$\times \sum_{k=1}^{N} |\langle g(x) - g^{\epsilon}(x), \phi_{k}(x) \rangle|^{2}$$

$$\leq \frac{\lambda_{N}^{2\alpha} r^{2\beta}}{|\mathcal{I}(h^{\epsilon})|^{2} \left(1 - E_{\beta, 1}(-\lambda_{1}^{2} r^{\beta} T^{\beta})\right)^{2}} \|g^{\epsilon} - g\|_{L^{2}(\Omega)}^{2}$$

$$\leq \frac{\lambda_{N}^{2\alpha} r^{2\beta}}{|\mathcal{I}(h^{\epsilon})|^{2} \left(1 - E_{\beta, 1}(-\lambda_{1}^{2} r^{\beta} T^{\beta})\right)^{2}} \epsilon^{2}.$$
(3.9)

Hence

$$\|\mathcal{Q}_2\|_{L^2(\Omega)} \le \frac{\lambda_N^{\alpha} r^{\beta}}{|\mathcal{I}(h^{\varepsilon})|(1 - E_{\beta,1}(-\lambda_1^2 r^{\beta} T^{\beta}))} \epsilon. \tag{3.10}$$

Finally, the term $\|Q_3\|_{L^2(\Omega)}$ can be estimated as follows

$$\|Q_3\|_{L^2(\Omega)}^2$$

$$\leq \left[\sum_{k=1}^{N} \left| \frac{\langle g(x), \phi_{k}(x) \rangle \phi_{k}(x)}{\int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h(\tau) d\tau} \right|^{2} \right] \left[\sum_{k=1}^{N} \left| \frac{\int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h^{\epsilon}(\tau) - h(\tau) d\tau}{\int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h^{\epsilon}(\tau) d\tau} \right|^{2} \right] \\
\leq \left[\sum_{k=1}^{N} \frac{\left| \int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) (h^{\varepsilon}(\tau) - h(\tau)) d\tau \right|^{2}}{\left| \int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h^{\varepsilon}(\tau) d\tau} \right|^{2}} \right] \left[\sum_{k=1}^{N} \frac{\left| \langle g(x), \phi_{k}(x) \rangle \right|^{2}}{\left| \int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h(\tau) d\tau \right|^{2}} \right] \\
\leq \frac{\|h^{\varepsilon} - h\|_{C[0,T]}^{2}}{|\mathcal{I}(h^{\varepsilon})|^{2}} \sum_{k=1}^{\infty} \frac{\left| \langle g(x), \phi_{k}(x) \rangle \right|^{2}}{\left| \int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h(\tau) d\tau \right|^{2}}$$

$$= \frac{\|h^{\varepsilon} - h\|_{C[0,T]}^{2}}{|\mathcal{I}(h^{\varepsilon})|^{2}} \|f\|_{L^{2}(\Omega)}^{2}$$

$$\leq \frac{\varepsilon^{2} \|f\|_{L^{2}(\Omega)}^{2}}{|\mathcal{I}(h^{\varepsilon})|^{2}}.$$

$$(3.11)$$

Hence

$$\|\mathcal{Q}_3\|_{L^2(\Omega)} \le \frac{\varepsilon \|f\|_{L^2(\Omega)}}{|\mathcal{I}(h^{\varepsilon})|}.$$
(3.12)

Combining (3.7), (3.9) and (3.11), it yields

$$||f(x) - f^{\epsilon, N}(x)||_{L^{2}(\Omega)} \le (1 + \lambda_{N})^{-\alpha \gamma} M + \varepsilon \frac{||f||_{L^{2}(\Omega)}}{|\mathcal{I}(h^{\varepsilon})|} + \varepsilon \frac{\lambda_{N}^{\alpha} r^{\beta}}{|\mathcal{I}(h^{\varepsilon})|(1 - E_{\beta, 1}(-\lambda_{1}^{2} r^{\beta} T^{\beta}))}.$$

$$(3.13)$$

This and the fact that $N \leq \mu \leq N+1$ give

$$\begin{split} &\|f(x) - f^{\epsilon,N}(x)\|_{L^{2}(\Omega)} \\ &\leq \left(\frac{\mu\pi}{2}\right)^{-\gamma\alpha} M + \varepsilon \left(\frac{\mu\pi}{2}\right)^{\alpha} \max\left\{\frac{r^{\beta}}{|\mathcal{I}(h^{\varepsilon})|\left(1 - E_{\beta,1}(-\lambda_{1}^{\alpha}r^{\beta}T^{\beta})\right)}, \frac{\|f\|_{L^{2}(\Omega)}}{|\mathcal{I}(h^{\varepsilon})|}\right\} \\ &\leq \varepsilon^{\frac{\gamma}{\gamma+1}} M^{\frac{1}{\gamma+1}} \left[1 + \max\left\{\frac{r^{\beta}}{|\mathcal{I}(h^{\varepsilon})|\left(1 - E_{\beta,1}(-\lambda_{1}^{\alpha}r^{\beta}T^{\beta})\right)}, \frac{\|f\|_{L^{2}(\Omega)}}{|\mathcal{I}(h^{\varepsilon})|}\right\}\right]. \end{split}$$

3.2. An a posteriori parameter choice. In this subsection, we consider an a posteriori regularization parameter choice by the discrepancy principle. Define

$$F_N g^{\epsilon} = \sum_{k=1}^{N} \langle g(x), \phi_k(x) \rangle \phi_k(x). \tag{3.14}$$

By the discrepancy principle, we take $K = K(\varepsilon, g^{\varepsilon})$ as the solution of

$$\|(I - F_N)g^{\varepsilon}\|_{L^2(\Omega)} \le m\epsilon \le \|(I - F_{N-1})g^{\varepsilon}\|_{L^2(\Omega)}, \quad m > 1.$$
 (3.15)

Lemma 3.3. We have

$$N \le \frac{2}{\pi} \left(\frac{\|h\|_{C[0,T]} M}{r^{\beta} (m-1)\epsilon} \right)^{\frac{1}{\alpha(\gamma+1)}}.$$
 (3.16)

Proof. From $||g^{\varepsilon} - g||_{L^{2}(\Omega)} \leq \varepsilon$ and (3.15), we have

$$||F_{N-1}g - g||_{L^{2}(\Omega)} = ||(F_{N-1} - I)g^{\varepsilon} - (I - F_{N-1})(g - g^{\varepsilon})||_{L^{2}(\Omega)}$$

$$\geq ||(F_{N-1} - I)g^{\varepsilon}||_{L^{2}(\Omega)} - ||(I - F_{N-1})(g - g^{\varepsilon})||_{L^{2}(\Omega)} \quad (3.17)$$

$$\geq (m - 1)\varepsilon.$$

On the other hand, for $k \geq N$, we obtain

$$\left| \int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h(\tau) d\tau \right| \leq \|h\|_{C[0,T]} \int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) d\tau$$

$$= \|h\|_{C[0,T]} \frac{\left(1 - E_{\beta,1}(-\lambda_{k}^{\alpha} r^{\beta} T^{\beta})\right)}{\lambda_{k}^{\alpha} r^{\beta}}$$

$$\leq \frac{\|h\|_{C[0,T]}}{\lambda_{N}^{\alpha} r^{\beta}}.$$
(3.18)

This implies

$$||F_{N-1}g - g||_{L^{2}(\Omega)}^{2}|$$

$$= \sum_{k=N}^{\infty} |\langle g(x), \phi_{k}(x) \rangle|^{2}$$

$$= \sum_{k=N}^{\infty} \left| \int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h(\tau) d\tau \langle f(x), \phi_{k}(x) \rangle \right|^{2}$$

$$\leq \frac{||h||_{C[0,T]}^{2}}{\lambda_{N}^{2\alpha} r^{2\beta}} \sum_{k=N}^{\infty} |\langle f(x), \phi_{k}(x) \rangle|^{2}$$

$$\leq \frac{||h||_{C[0,T]}^{2}}{\lambda_{N}^{2\alpha} r^{2\beta}} \sum_{k=N}^{\infty} (1 + \lambda_{k})^{-2\alpha\gamma} (1 + \lambda_{k})^{2\alpha\gamma} |\langle f(x), \phi_{k}(x) \rangle|^{2}$$

$$\leq \frac{||h||_{C[0,T]}^{2}}{\lambda_{N}^{2\alpha} r^{2\beta} \lambda_{N}^{2\alpha\gamma}} \sum_{k=N}^{\infty} (1 + \lambda_{k})^{2\alpha\gamma} |\langle f(x), \phi_{k}(x) \rangle|^{2}$$

$$\leq \frac{||h||_{C[0,T]}^{2}}{\lambda_{N}^{2\alpha} r^{2\beta} \lambda_{N}^{2\alpha\gamma}} ||f||_{H^{\alpha\gamma}(\Omega)}^{2}$$

$$\leq ||h||_{C[0,T]}^{2} \frac{M^{2}}{r^{2\beta}} \frac{1}{\lambda_{N}^{2\alpha}(\gamma+1)}.$$
(3.19)

Hence,

$$||F_{N-1}g - g||_{L^2(\Omega)} \le \frac{M}{r^\beta} \frac{||h||_{C[0,T]}}{\lambda_N^{\alpha(\gamma+1)}}.$$
 (3.20)

From (3.17) and (3.20), we have

$$(m-1)\epsilon \le \frac{M}{r^{\beta}} \frac{\|h\|_{C[0,T]}}{\lambda_N^{\alpha(\gamma+1)}}.$$
(3.21)

It follows from $\lambda_k^{\alpha} = (\frac{k\pi}{2})^{\alpha}$ and (3.21) that

$$N \le \frac{2}{\pi} \left(\frac{\|h\|_{C[0,T]} M}{r^{\beta} (m-1)\epsilon} \right)^{\frac{1}{\alpha(\gamma+1)}}.$$
 (3.22)

Next we present an error estimate for the approximate solution of problem (1.1).

Theorem 3.4. Let $f^{\epsilon,N}$ and f be as in Theorem 3.1. Then we have

$$||f(x) - f^{\varepsilon,N}(x)||_{L^{2}(\Omega)} \le \varepsilon^{\frac{\gamma}{\gamma+1}} M^{\frac{1}{\gamma+1}} \left[\mathcal{L}_{\beta}(f, h^{\varepsilon}, h, r, m, T) + \mathcal{K}_{\alpha,\beta}(h, r, T)(m+1)^{\frac{\gamma}{\gamma+1}} \right].$$
(3.23)

where

$$\mathcal{L}_{\beta}(f, h^{\varepsilon}, h, r, m, T) = \left(\frac{\|h\|_{C[0,T]}}{r^{\beta}(m-1)|\mathcal{I}(h^{\varepsilon})|^{\gamma+1}}\right)^{\frac{1}{\gamma+1}} \max\left\{\|f\|_{L^{2}(\Omega)}, \frac{r^{\beta}}{\left(1 - E_{\beta,1}(-\lambda_{1}^{\alpha}r^{\beta}T^{\beta})\right)}\right\},$$

$$\mathcal{K}_{\alpha,\beta}(h, r, T) = \frac{(r^{\beta})^{\frac{\gamma}{\gamma+1}}}{|\mathcal{I}(h)|^{\frac{\gamma}{\gamma+1}}\left(1 - E_{\alpha,1}(-\lambda_{1}^{\alpha}r^{\beta}T^{\beta})\right)^{\frac{\gamma}{\gamma+1}}}.$$

Proof. Using the triangle inequality,

$$||f(x) - f^{\varepsilon,N}(x)||_{L^2(\Omega)} \le ||f(x) - f^N(x)||_{L^2(\Omega)} + ||f(x) - f^{\epsilon,N}(x)||_{L^2(\Omega)}.$$
(3.24)

We split the proof into three steps.

Step 1: Estimate $||f(\cdot) - f^N(\cdot)||_{L^2(\Omega)}$.

$$||f(x) - f^{N}(x)||_{H^{\gamma}(\Omega)} \leq ||\sum_{k=N+1}^{\infty} \langle f(x), \phi_{k}(x) \rangle \phi_{k}(x)||$$

$$= \left(\sum_{k=N+1}^{\infty} \left(1 + \lambda_{k}^{\alpha}\right)^{2\gamma} |\langle f(x), \phi_{k}(x) \rangle|^{2}\right)^{1/2} \leq M.$$
(3.25)

By triangle inequality and (3.15),

$$||Af(x) - Af^{N}(x)||_{L^{2}(\Omega)} \leq ||(I - F_{N})g||$$

$$\leq ||(I - F_{N})g^{\epsilon} + (I - F_{N})(g - g^{\epsilon})||$$

$$\leq ||(I - F_{N})g^{\epsilon}|| + ||(I - F_{N})(g - g^{\epsilon})||$$

$$\leq (m + 1)\varepsilon.$$
(3.26)

Therefore, by the conditional stability (2.31), we have

$$||f(\cdot) - f^{N}(\cdot)||_{L^{2}(\Omega)} \le \mathcal{K}_{\alpha,\beta}(h,r,T)((m+1)\varepsilon)^{\frac{\gamma}{\gamma+1}}.$$
 (3.27)

Next, we obtain

$$f^{N}(x) - f^{\epsilon,N}(x)$$

$$= \sum_{k=1}^{N} \frac{\langle g(x), \phi_{k}(x) \rangle \phi_{k}(x)}{\int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h(\tau) d\tau} - \sum_{k=1}^{N} \frac{\langle g^{\epsilon}(x), \phi_{k}(x) \rangle \phi_{k}(x)}{\int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h^{\epsilon}(\tau) d\tau}$$

$$\leq \sum_{k=1}^{N} \frac{\langle g(x), \phi_{k}(x) \rangle \phi_{k}(x)}{\int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h(\tau) d\tau} - \sum_{k=1}^{N} \frac{\langle g(x), \phi_{k}(x) \rangle \phi_{k}(x)}{\int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h^{\epsilon}(\tau) d\tau}$$

$$+ \sum_{k=1}^{N} \frac{\langle g(x) - g^{\epsilon}(x), \phi_{k}(x) \rangle \phi_{k}(x)}{\int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h^{\epsilon}(\tau) d\tau}$$

$$\leq \sum_{k=1}^{N} \frac{\int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h(\tau) - h^{\epsilon}(\tau) d\tau}{\int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h(\tau) d\tau} \sum_{i=Q_{3}}^{N} \frac{\langle g(x), \phi_{k}(x) \rangle \phi_{k}(x)}{\int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h^{\epsilon}(\tau) d\tau}.$$

$$(3.28)$$

$$+ \sum_{k=1}^{N} \frac{\langle g(x) - g^{\epsilon}(x), \phi_{k}(x) \rangle \phi_{k}(x)}{\int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h^{\epsilon}(\tau) d\tau}.$$

$$:= Q_{4}$$

Using (3.12), we obtain

$$\|\mathcal{Q}_3\|_{L^2(\Omega)} \le \varepsilon \frac{\|f\|_{L^2(\Omega)}}{|\mathcal{I}(h^{\varepsilon})|}.$$
(3.29)

We now estimate the norm of Q_4 . Using Lemma 2.8, we have

$$\|\mathcal{Q}_{4}\|_{L^{2}(\Omega)}^{2} = \sum_{k=1}^{N} \left| \frac{\langle g(x) - g^{\varepsilon}(x), \phi_{k}(x) \rangle}{\int_{0}^{T} \Phi_{\beta}(\lambda_{k}^{\alpha}, \tau, r) h^{\varepsilon}(\tau) d\tau} \right|^{2}$$

$$\leq \sum_{k=1}^{N} \frac{\langle g(x) - g^{\varepsilon}(x), \phi_{k}(x) \rangle^{2}}{\frac{|\mathcal{I}(h_{\varepsilon})|^{2} \left(1 - E_{\beta,1}(-\lambda_{1}^{\alpha} r^{\beta} T^{\beta})\right)^{2}}{\lambda_{k}^{2\alpha} r^{2\beta}}}$$

$$\leq \frac{\lambda_{N}^{2\alpha} r^{2\beta}}{|\mathcal{I}(h_{\varepsilon})|^{2} \left(1 - E_{\beta,1}(-\lambda_{1}^{\alpha} r^{\beta} T^{\beta})\right)^{2}} \sum_{k=1}^{N} \langle g(x) - g^{\varepsilon}(x), \phi_{k}(x) \rangle^{2}$$

$$\leq \frac{\varepsilon^{2} \lambda_{N}^{2\alpha} r^{2\beta}}{|\mathcal{I}(h_{\varepsilon})|^{2} \left(1 - E_{\beta,1}(-\lambda_{1}^{\alpha} r^{\beta} T^{\beta})\right)^{2}}.$$

$$(3.30)$$

Hence

$$\|\mathcal{Q}_4\|_{L^2(\Omega)} \le \left(\frac{N\pi}{2}\right)^{\alpha} \frac{\epsilon}{|\mathcal{I}(h^{\varepsilon})|} \frac{r^{\beta}}{\left(1 - E_{\beta,1}(-\lambda_1^{\alpha} r^{\beta} T^{\beta})\right)}.$$
 (3.31)

From above observations, we deduce that

$$||f^{N}(x) - f^{\epsilon,N}(x)||_{L^{2}(\Omega)} \le \left(\frac{N\pi}{2}\right)^{\alpha} \frac{\varepsilon}{|\mathcal{I}(h^{\varepsilon})|} \max\left\{||f||_{L^{2}(\Omega)}, \frac{r^{\beta}}{(1 - E_{\beta,1}(-\lambda_{1}^{\alpha}r^{\beta}T^{\beta}))}\right\}.$$

$$(3.32)$$

Substituting (3.22) in (3.32), we obtain

$$||f^{N}(x) - f^{\epsilon,N}(x)||_{L^{2}(\Omega)} \le \varepsilon^{\frac{\gamma}{\gamma+1}} M^{\frac{1}{\gamma+1}} \mathcal{L}_{\beta}(f, h^{\varepsilon}, h, r, m, T).$$
 (3.33)

Combining (3.27) with (3.32), we obtain the final estimate as follows:

$$||f(x) - f^{\epsilon,N}(x)||_{L^{2}(\Omega)} \le \varepsilon^{\frac{\gamma}{\gamma+1}} M^{\frac{1}{\gamma+1}} \left[\mathcal{L}_{\beta}(f, h^{\varepsilon}, h, r, m, T) + \mathcal{K}_{\alpha,\beta}(h, r, T)(m+1)^{\frac{\gamma}{\gamma+1}} \right].$$
(3.34)

hereby

$$\mathcal{L}_{\beta}(f, h^{\varepsilon}, h, r, m, T) = \left(\frac{\|h\|_{C[0,T]}}{r^{\beta}(m-1)|\mathcal{I}(h^{\varepsilon})|^{\gamma+1}}\right)^{\frac{1}{\gamma+1}} \max\left\{\|f\|_{L^{2}(\Omega)}, \frac{r^{\beta}}{\left(1 - E_{\beta,1}(-\lambda_{1}^{\alpha}r^{\beta}T^{\beta})\right)}\right\},$$

$$\mathcal{K}_{\alpha,\beta}(h, r, T) = \frac{\left(r^{\beta}\right)^{\frac{\gamma}{\gamma+1}}}{|\mathcal{I}(h)|^{\frac{\gamma}{\gamma+1}}\left(1 - E_{\alpha,1}(-\lambda_{1}^{\alpha}r^{\beta}T^{\beta})\right)^{\frac{\gamma}{\gamma+1}}}.$$

This completes the proof.

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