1 Introduction

Consider a physical domain $\Omega \subset \mathbb{R}^d$, $d \in \mathbb{N}^+$ be bounded with the boundary Γ and donate I := (0, T), $Q := \Omega \times I$ and $S := \Gamma \times I$ with T > 0 given.

Consider the heat equation

$$\frac{\partial u}{\partial t} - \sum_{i,j=1}^{d} \frac{\partial}{\partial x_j} \left(a_{ji}(x,t) \frac{\partial u}{\partial x_i} \right) = F(x,t), \quad (x,t) \in Q, \tag{1.1}$$

with the initial and Dirichlet conditions, respectively

$$u(x,0) = u_0(x), \quad x \in \Omega, \tag{1.2}$$

$$u(x,t) = \varphi(x,t), \quad (x,t) \in S \tag{1.3}$$

where

$$a_{ij} \in L^{\infty}(Q), \ a_{ij} = a_{ji}, \ \forall i, j \in \{1, 2, ..., d\},$$

$$\lambda_1 \|\xi\|^2 \le \sum_{i,j=1}^d a_{ij} \xi_i \xi_j \le \lambda_2 \|\xi\|^2, \ \forall \xi \in \mathbb{R}^d,$$

$$\varphi \in L^2(S), \ u_0 \in L^2(\Omega), \ F \in L^2(I; H^{-1}(\Omega)),$$

with λ_1 v λ_2 are positive constants.

The direct problem is to determine u when the coefficients of the questions (1.1) and the data u_0, φ and F are given. The inverse problem is to identify a missed data such as the right hand side F when some additional observation on the solution u are available.

Consider the structure F(x,t) = f(.)q(x,t) + g(x,t), with f(.) is either f(x,t), f(x) or f(t). The problem here is to find f(.) when $\ell u(f) = u(x,t) = \omega(x,t)$ is given on Q.

To solve this problem, we need to minimize the least square functional

$$J_{\gamma}(f) = \frac{1}{2} \|\ell u(f) - \omega\|_{L^{2}(Q)}^{2}.$$

However, this minimization problem is unstable and there might be many minimizers to it. Therefore, we minimize the Tikhonov functional instead

$$J_{\gamma}(f) = \frac{1}{2} \|u - \omega\|_{L^{2}(Q)}^{2} + \frac{\gamma}{2} \|f - f^{*}\|_{L^{2}(*)}^{2},$$

with $\gamma > 0$ being a regularization parameter, f^* is an a prior estimation of f and $\|.\|_{L^2(*)}$ depends on f(*) appropriately.

2 Variational problem

To introduce the concept of weak form, we use the standard Sobolev spaces $H^1(\Omega)$, $H^1_0(\Omega)$, $H^{1,0}(Q)$ and $H^{1,1}(Q)$. Further, for a Banach space B, we define

$$L^{2}(0,T;B) = \left\{ u : u(t) \in B \text{ a.e } t \in (0,T) \text{ and } \|u\|_{L^{2}(0,T;B)} < \infty \right\},$$

with the norm

$$||u||_{L^2(0,T;B)} = \int_0^T ||u(t)||_B^2 dt.$$

In the sequel, we shall use the space W(0,T) define as

$$W(0,T) = \left\{ u : u \in L^2(0,T; H^1(\Omega)), u_t \in L^2(0,T; (H^1(\Omega))') \right\}$$

Suppose that $F \in L^2(Q)$, a week solution in W(0,T) of the problem (1.1) - (1.3) is a function $u(x,t) \in W(0,T)$ satisfying the identity

$$\int_{Q} \left[\frac{\partial u}{\partial t} v + \sum_{i,j=1}^{d} a_{ji} \frac{\partial u}{\partial x_{i}} \frac{\partial v}{\partial x_{j}} \right] dxdt = \int_{Q} Fv dxdt + \int_{S} \varphi v dsdt.$$
 (2.1)

and

$$u(x,0) = u_0, x \in \Omega. \tag{2.2}$$

$$||u||_{L^{2}(0,T;B)} \le c_{d} \left(||F||_{L^{2}(Q)} + ||u_{0}||_{L^{2}(\Omega)} + ||\varphi||_{L^{2}(S)} \right)$$
(2.3)

We suppose that F has the form F(x,t) = f(x,t)q(x,t) + g(x,t) with $f \in L^2(Q)$, $q \in L^\infty(Q)$ and $g \in L^2(Q)$. We hope to recover f(x,t) from the observation. Since the solution u(x,t) depends on the function f(x,t), so we denote it by u(x,t,f) or u(f). To identify f(x,t), we minimize the Tikhonov functional

$$J_{\gamma}(f) = \frac{1}{2} \|\ell u(f) - \omega\|_{L^{2}(Q)}^{2} + \frac{\gamma}{2} \|f - f^{*}\|_{L^{2}(Q)}^{2}.$$
 (2.4)

We will prove that J_{γ} is Frechet differentiable and drive a formula for its gradient. Want to do so, we need the adjoint problem

$$\begin{cases}
-\frac{\partial p}{\partial t} - \sum_{i,j=1}^{d} \frac{\partial}{\partial x_j} \left(a_{ji}(x,t) \frac{\partial p}{\partial x_i} \right) = \ell u(f) - \omega, & (x,t) \in Q, \\
u(x,t) = 0, & (x,t) \in S \\
p(x,T) = 0, & x \in \Omega.
\end{cases}$$
(2.5)

By changing the time direction, meaning $\overline{p}(x,t) = p(x,T-t)$, we will get a Neumann problem for parabolic equations.

Theorem 2.1. The functional J_{γ} is Frechet differentiable and its gradient ∇J_{γ} at f has the form

$$\nabla J_{\gamma}(f) = q(x,t)p(x,t) + \gamma (f(x,t) - f^{*}(x,t))$$
(2.6)

Proof. By taking a small variation $\delta f \in L^2(Q)$ of f and denoting $\delta u(f) = u(f + \delta f) - u(f)$, we have

$$J_{0}(f + \delta f) - J_{0}(f) = \frac{1}{2} \|\ell u(f + \delta f) - \omega\|_{L^{2}(Q)}^{2} - \frac{1}{2} \|\ell u(f) - \omega\|_{L^{2}(Q)}^{2}$$

$$= \frac{1}{2} \|\ell \delta u(f) + \ell u(f) - \omega\|_{L^{2}(Q)}^{2} - \frac{1}{2} \|\ell u(f) - \omega\|_{L^{2}(Q)}^{2}$$

$$= \frac{1}{2} \|\ell \delta u(f)\|_{L^{2}(Q)}^{2} + \langle \ell \delta u(f), \ell u(f) - \omega \rangle_{L^{2}(Q)}$$

where $\delta u(f)$ is the solution to the problem

$$\begin{cases}
\frac{\partial \delta u}{\partial t} - \sum_{i,j=1}^{d} \frac{\partial}{\partial x_j} \left(a_{ji}(x,t) \frac{\partial \delta u}{\partial x_i} \right) = q(x,t) \delta f, & (x,t) \in Q, \\
\delta u(x,t) = 0, & (x,t) \in S, \\
\delta u(x,0) = 0, & x \in \Omega.
\end{cases}$$
(2.7)

Because the priori estimate (2.3) for the direct problem, we have

$$\|\ell \delta u(f)\|_{L^2(Q)}^2 = o\left(\|\delta f\|_{L^2(Q)}\right) \text{ when } \|\delta f\|_{L^2(Q)} \to 0.$$

What is more, applying the Green formula for (2.5) and (2.7), we get

$$\int_{Q} \delta u \left(\ell u(f) - \omega \right) dx dt = \int_{Q} p(x, t) q(x, t) \delta f(x, t) dx dt$$

Therefore,

$$J_{0}(f + \delta f) - J_{0}(f) = \int_{Q} \delta u \left(\ell u - \omega\right) ds + o\left(\|\delta f\|_{L^{2}(Q)}\right)$$
$$= \int_{Q} qp \delta f dx dt + o\left(\|\delta f\|_{L^{2}(I)}\right)$$
$$= \langle qp, \delta f \rangle_{L^{2}(Q)} + o\left(\|\delta f\|_{L^{2}(Q)}^{2}\right).$$

It is easy to get this

$$J_{\gamma}(f+\delta f) - J_{\gamma}(f) = \langle qp, \delta f \rangle_{L^{2}(Q)} + \gamma \langle f - f^{*}, \delta f \rangle_{L^{2}(Q)} + o\left(\left\|\delta f\right\|_{L(Q)}^{2}\right).$$

Hence the functional J_{γ} is Frechet differentiable and its gradient ∇J_{γ} at f has the form (2.5). The theorem is proved.

Remark 2.1. In this theorem, we write the Tikhonov functional for F(x,t) = f(x,t)q(x,t) + g(x,t). But when F has another structure, the penalty term should be modified

• F(x,t) = f(x)q(x,t) + g(x,t): the penalty functional is $||f - f^*||_{L^2(\Omega)}$ and

$$\nabla J_0(f) = \int_0^T q(x,t)p(x,t)dt.$$

• F(x,t) = f(t)q(x,t) + g(x,t): the penalty functional is $||f - f^*||_{L^2(0,T)}$ and

$$\nabla J_0(f) = \int_{\Omega} q(x,t)p(x,t)dt.$$

To find f satisfied (2.4), we use the conjugate gradient method (CG). It proceeds as follows: assume that at the kth iteration, we have f^k . Then the next iteration is

$$f^{k+1} = f^k + \alpha_k d^k.$$

with

$$d^{k} = \begin{cases} -\nabla J_{\gamma}(f^{k}), & k = 0, \\ -\nabla J_{\gamma}(f^{k}) + \beta_{k}d^{k-1}, & k > 0, \end{cases}$$

$$\beta_k = \frac{\left\| \nabla J_{\gamma}(f^k) \right\|_{L^2(I)}^2}{\left\| \nabla J_{\gamma}(f^{k-1}) \right\|_{L^2(I)}^2},$$

and

$$\alpha_k = \operatorname*{arg\,min}_{\alpha \geq 0} J_{\gamma}(f^k + \alpha d^k).$$

To identify α_k , we consider two problems

Problem 2.1. Denote the solution of this problem is u[f]

$$\begin{cases} \frac{\partial u}{\partial t} - \sum_{i,j=1}^{d} \frac{\partial}{\partial x_j} \left(a_{ji}(x,t) \frac{\partial u}{\partial x_i} \right) = f(t)h(x,t), & (x,t) \in Q, \\ \frac{\partial u}{\partial \mathcal{N}}(x,t) = 0, & (x,t) \in S, \\ u(x,0) = 0, & x \in \Omega. \end{cases}$$

Problem 2.2. Denote the solution of this problem is $u(u_0,\varphi)$

$$\begin{cases} \frac{\partial u}{\partial t} - \sum_{i,j=1}^{d} \frac{\partial}{\partial x_j} \left(a_{ji}(x,t) \frac{\partial u}{\partial x_i} \right) = g(x,t), & (x,t) \in Q, \\ \frac{\partial u}{\partial \mathcal{N}}(x,t) = \varphi(x,t), & (x,t) \in S, \\ u(x,0) = u_0(x), & x \in \Omega. \end{cases}$$

If so, the observation operators have the form $\ell u(f) = \ell u[f] + \ell u(u_0, \varphi) = Af + \ell u(u_0, \varphi)$, with A being bounded linear operators from $L^2(Q)$ to $L^2(Q)$.

We have

$$J_{\gamma}(f^{k} + \alpha d^{k}) = \frac{1}{2} \left\| \ell u(f^{k} + \alpha d^{k}) - \omega \right\|_{L^{2}(S_{1})}^{2} + \frac{\gamma}{2} \left\| f^{k} + \alpha d^{k} - f^{*} \right\|_{L^{2}(I)}^{2}$$

$$= \frac{1}{2} \left\| \alpha A d^{k} + A f^{k} + \ell u(u_{0}, \varphi) - \omega \right\|_{L^{2}(S_{1})}^{2} + \frac{\gamma}{2} \left\| f^{k} + \alpha d^{k} - f^{*} \right\|_{L^{2}(I)}^{2}$$

$$= \frac{1}{2} \left\| \alpha A d^{k} + \ell u(f^{k}) - \omega \right\|_{L^{2}(S_{1})}^{2} + \frac{\gamma}{2} \left\| f^{k} + \alpha d^{k} - f^{*} \right\|_{L^{2}(I)}^{2}.$$

Differentiating $J_{\gamma}(f^k + \alpha d^k)$ with respect to α , we get

$$\frac{\partial J_{\gamma}(f^k + \alpha d^k)}{\partial \alpha} = \alpha \left\| A d^k \right\|_{L^2(S_1)}^2 + \left\langle A d^k, \ell u(f^k) - \omega \right\rangle_{L^2(S_1)}^2$$
$$+ \gamma \alpha \left\| d^k \right\|_{L^2(I)}^2 + \gamma \left\langle d^k, f^k - f^* \right\rangle_{L^2(I)}^2.$$

Putting
$$\frac{\partial J_{\gamma}(f^k + \alpha d^k)}{\partial \alpha} = 0$$
, we obtain

$$\begin{split} \alpha_k &= -\frac{\left\langle Ad^k, \ell u(f^k) - \omega \right\rangle_{L^2(S_1)} + \gamma \left\langle d^k, f^k - f^* \right\rangle_{L^2(I)}}{\|Ad^k\|_{L^2(S_1)}^2 + \gamma \|d^k\|_{L^2(I)}^2} \\ &= -\frac{\left\langle d^k, A^* \left(\ell u(f^k) - \omega \right) \right\rangle_{L^2(I)} + \gamma \left\langle d^k, f^k - f^* \right\rangle_{L^2(I)}}{\|Ad^k\|_{L^2(S_1)}^2 + \gamma \|d^k\|_{L^2(I)}^2} \\ &= -\frac{\left\langle d^k, A^* \left(\ell u(f^k) - \omega \right) + \gamma (f^k - f^*) \right\rangle_{L^2(I)}}{\|Ad^k\|_{L^2(S_1)}^2 + \gamma \|d^k\|_{L^2(I)}^2} \\ &= -\frac{\left\langle d^k, \nabla J_\gamma(f^k) \right\rangle_{L^2(I)}}{\|Ad^k\|_{L^2(S_1)}^2 + \gamma \|d^k\|_{L^2(I)}^2}. \end{split}$$

Because of $d^k = r^k + \beta_k d^{k-1}$, $r^k = -\nabla J_{\gamma}(f^k)$ and $\langle r^k, d^{k-1} \rangle_{L^2(I)} = 0$, we get

$$\alpha_k = \frac{\left\| r^k \right\|_{L^2(I)}^2}{\left\| A d^k \right\|_{L^2(S_1)}^2 + \gamma \left\| d^k \right\|_{L^2(I)}^2}.$$

CG algorithm

- 1. Set k = 0, initiate f^0 .
- 2. For k = 0, 1, 2, ... Calculate

$$r^k = -\nabla J_{\gamma}(f^k).$$

Update

$$d^{k} = \begin{cases} r^{k}, & k = 0, \\ r^{k} + \beta_{k} d^{k-1}, & k > 0, \end{cases}$$

$$\beta_k = \frac{\left\| r^k \right\|_{L^2(Q)}^2}{\left\| r^{k-1} \right\|_{L^2(Q)}^2}.$$

3. Calculate

$$\alpha_k = \frac{\left\| r^k \right\|_{L^2(Q)}^2}{\left\| A d^k \right\|_{L^2(S_1)}^2 + \gamma \left\| d^k \right\|_{L^2(Q)}^2}.$$

Update

$$f^{k+1} = f^k + \alpha_k d^k.$$

3 Finite element method

We rewrite the Tikhonov functional

$$J_{\gamma}(f) = \frac{1}{2} \|\ell u[f] + \ell u(u_0, \varphi) - \omega\|_{L^2(S_1)}^2 + \frac{\gamma}{2} \|f - f^*\|_{L^2(I)}^2$$

$$= \frac{1}{2} \|Af + \ell u(u_0, \varphi) - \omega\|_{L^2(S_1)}^2 + \frac{\gamma}{2} \|f - f^*\|_{L^2(I)}^2$$

$$= \frac{1}{2} \|Af - \hat{\omega}\|_{L^2(S_1)}^2 + \frac{\gamma}{2} \|f - f^*\|_{L^2(I)}^2,$$

with $\hat{\omega} = \omega - \ell u(u_0, \varphi)$.

The solution f^{γ} of the minimization problem (2.4) is characterized by the first-order optimality condition

$$\nabla J_{\gamma}(f^{\gamma}) = A^*(Af^{\gamma} - \hat{\omega}) + \gamma(f^{\gamma} - f^*) = 0, \tag{3.1}$$

with $A^*: L^2(Q) \to L^2(Q)$ is the adjoint operator of A defined by $A^*\phi = p$ where $\phi = \ell u(f) - \omega$ and p is the solution of the adjoint problem (2.5).

3.1 Finite element approximate of A, A^*

We will approximate (3.1) by using the space-time finite element method with finite spaces $X_h \subset Y_h$. For the space-time domain $Q = \Omega \times I \subset \mathbb{R}^{d+1}$, we consider a sequence of admissible decompositions Q_h into shape regular simplicity finite element q_l

$$Q_h = \cup_{l=1}^N \bar{q}_l.$$

Denote $\{(x_k, t_k)\}_{k=1}^M$ is a set of nodes $(x_k, t_k) \in \mathbb{R}^{d+1}$. We introduce a reference element $q \in \mathbb{R}^{d+1}$ which any element q_l can maple to q by using

$$\begin{pmatrix} x \\ t \end{pmatrix} = \begin{pmatrix} x_k \\ t_k \end{pmatrix} + J_l \begin{pmatrix} \xi \\ \tau \end{pmatrix}, \begin{pmatrix} \xi \\ \tau \end{pmatrix} \in q.$$

with Δ_l is the volume of q_l

$$\Delta_l = \int_{q_l} dx dt = \det J_l \int_q d\xi d\tau = |q| \det J_l,$$

and the local mesh width

$$h_l = \Delta_l^{\frac{1}{d+1}}, \ h := \max_{l=1,\dots,N} h_l.$$

Note that

$$|q| = \begin{cases} \frac{1}{2}, & d = 1, \\ \frac{1}{6}, & d = 2. \end{cases}$$

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$$\int_{Q} \left[\frac{\partial u_{h}}{t} v_{h} + \sum_{i,j=1}^{d} a_{ji} \frac{\partial u_{h}}{\partial x_{i}} \frac{\partial v_{h}}{\partial x_{j}} \right] dxdt = \int_{Q} F v_{h} dxdt + \int_{S} \varphi v_{h} dsdt, \forall v_{h} \in Y_{h}.$$
(3.2)

and

$$u_h(x,0) = u_0, x \in \Omega. \tag{3.3}$$

The discrete variational problem (3.2) admits a unique solution $u_h \in X_h$. Hence, the discrete version of the optimal control problem will be

$$J_{\gamma,h}(f) = \frac{1}{2} \|A_h f - \hat{\omega}_h\|_{L^2(S_1)}^2 + \frac{\gamma}{2} \|f - f^*\|_{L^2(I)}^2.$$

Let f_h^{γ} be the solution of this problem is characterized by the variational equation

$$\nabla J_{\gamma,h}(f_h^{\gamma}) = A_h^* (A_h f^{\gamma} - \hat{\omega}_h) + \gamma (f_h^{\gamma} - f^*) = 0, \tag{3.4}$$

where A_h^* is the adjoint operator of A_h . But it is hardly to find A_h^* from A_h in practice. So we define a proximate \hat{A}_h^* of A^* instead. In deed, we have $\hat{A}_h^*\phi = p_h$, with p_h is the approximate solution of adjoint problem (2.5). Therefore, the equation above will be

$$\nabla J_{\gamma,h}(f_h^{\gamma}) \simeq \nabla J_{\gamma,h}(\hat{f}_h^{\gamma}) = \hat{A}_h^*(A_h \hat{f}^{\gamma} - \hat{\omega}_h) + \gamma(\hat{f}_h^{\gamma} - f^*) = 0, \tag{3.5}$$

Moreover, the observation will have noise in practice, so instead of $\omega(x,t)$, we only get $\omega^{\delta}(x,t)$ satisfy

$$\left\|\omega - \omega^{\delta}\right\|_{L^{2}(S_{1})} \leq \delta.$$

Therefore, instead of getting \hat{f}_h^{γ} satisfies the equation (3.5), we will get $\hat{f}_h^{\gamma,\delta}$ satisfies

$$\nabla J_{\gamma,h}\left(\hat{f}_h^{\gamma,\delta}\right) = \hat{A}_h^*(A_h\hat{f}_h^{\gamma,\delta} - \hat{\omega}_h^{\delta}) + \gamma(\hat{f}_h^{\gamma,\delta} - f^*) = 0, \tag{3.6}$$

with $\hat{\omega}_h^{\delta} = \omega^{\delta} - \ell u_h(u_0, \varphi)$.

3.2 Convergence results

Theorem 3.1. Let u(x,t) be the solution of variational problem (2.1) - (2.2) and $u_h(x,t)$ be the solution for (3.2) - (3.3) Then there holds the error estimate

$$||u - u_h||_{L^2(Q)} \le ||u - u_h||_{L^2(I; H^1(\Omega))} \le c_2 h_{xt} |u|_{H^2(\Omega)}.$$
(3.7)

What is more,

$$\left\| \left(A^* - \hat{A}_h^* \right) \phi \right\|_{L^2(Q)}^2 = \int_Q (p - p_h)^2 dx dt = \|p - p_h\|_{L^2(Q)}^2$$

$$\Rightarrow \| (A^* - A_h^*) q \|_{L^2(Q)} \le c_5 h. \tag{3.8}$$

Let $u_h[f]$ v $u_h(u_0, \varphi)$ are the approximate solutions of **Problems** 2.1 and **Problems** 2.2 by using space-time finite element method. We define A_h of A is $A_h f = \ell u_h[f]$ and $\hat{\omega}_h = \omega - \ell u_h(u_0, \varphi)$. We have

$$\|(A - A_h) f\|_{L^2(Q)} = \|\ell u[f] - \ell u_h[f]\|_{L^2(\Omega)} \le \|u[f] - u_h[f]\|_{L^2(Q)} \le c_3 h, \tag{3.9}$$

and

$$\|\hat{\omega} - \hat{\omega}_h\|_{L^2(Q)} = \|u(u_0, \varphi) - u_h(u_0, \varphi)\|_{L^2(Q)} \le c_4 h. \tag{3.10}$$

Theorem 3.2. Let f^{γ} and \hat{f}_h^{γ} are the solution of variational problems (3.1) and (3.5), respectively. Then there hold a error estimate

$$\left\| f^{\gamma} - \hat{f}_h^{\gamma} \right\|_{L^2(Q)} \le c_6 h.$$
 (3.11)

Proof. From equations (3.1) and (3.5), we will have

$$\gamma \left(f^{\gamma} - \hat{f}_{h}^{\gamma} \right) = \hat{A}_{h}^{*} \left(A_{h} \hat{f}_{h}^{\gamma} - \hat{\omega}_{h} \right) - A^{*} \left(A f^{\gamma} - \hat{\omega} \right)$$

$$= \left(\hat{A}_{h}^{*} - A^{*} \right) \left(A_{h} \hat{f}_{h}^{\gamma} - \hat{\omega}_{h} \right) + A^{*} A_{h} \left(\hat{f}_{h}^{\gamma} - f^{\gamma} \right)$$

$$+ A^{*} \left(A_{h} - A \right) f^{\gamma} + A^{*} \left(\hat{\omega} - \hat{\omega}_{h} \right)$$

According to (3.8), (3.9) and (3.10), we have

$$\left\| \left(\hat{A}_{h}^{*} - A^{*} \right) \left(A_{h} \hat{f}_{h}^{\gamma} - \hat{\omega}_{h} \right) \right\|_{L^{2}(I)} \leq c_{7} h_{xt},$$

$$\left\| A^{*} \left(A_{h} - A \right) f^{\gamma} \right\|_{L^{2}(I)} \leq c_{8} h_{xt},$$

$$\left\| A^{*} \left(\hat{\omega} - \hat{\omega}_{h} \right) \right\|_{L^{2}(I)} \leq c_{9} h_{xt}.$$

We take apart this

$$A^*A_h\left(\hat{f}_h^{\gamma}-f^{\gamma}\right)=A^*\left(A_h-A\right)\left(\hat{f}_h^{\gamma}-f^{\gamma}\right)+A^*A\left(\hat{f}_h^{\gamma}-f^{\gamma}\right).$$

Moreover, we have

$$\left\langle A^* \left(A_h - A \right) \left(\hat{f}_h^{\gamma} - f^{\gamma} \right), f^{\gamma} - \hat{f}_h^{\gamma} \right\rangle_{L^2(I)} \le c_{10} h_{xt}^2 \left\| f^{\gamma} - \hat{f}_h^{\gamma} \right\|_{L^2(I)}^2,$$

$$\left\langle A^* A \left(\hat{f}_h^{\gamma} - f^{\gamma} \right), f^{\gamma} - \hat{f}_h^{\gamma} \right\rangle_{L^2(I)} = - \left\| A \left(f^{\gamma} - \hat{f}_h^{\gamma} \right) \right\|_{L(S_t)}^2 < 0.$$

The theorem is proved.

Remark 3.1. Let f^{γ} and \hat{f}_h^{γ} are the solution of variational problems (3.1) and (3.5), respectively. Then there hold a error estimate

$$\|f^{\gamma} - \hat{f}_h^{\gamma,\delta}\|_{L^2(I)} \le c_{11}(h+\delta).$$
 (3.12)

4 Numerical results

Let $\Omega = [0,1]^2$, T=1 and $a(x,t)=1+x^2+y^2+t^2$ and we want to re-simulate the process of conduction heat transfer and identifying the heat source

$$u(x,t) = e^t \sin(\pi x) \sin(\pi y).$$

The form of the heat source

$$F(x,t) = f(x,t)h(x,t) + g(x,t),$$

where

$$h(x,t) = 2 + x^2 + y^2 + t^2$$
.

Set $f^* = 0$, $\gamma = 10^{-6}$.

f(.):

$$f(x,t) = \phi(t)\sin(\pi x)\sin(\pi y)$$

where

- Example 1: $\phi(t) = \sin(\pi t)$
- Example 2: $\phi(t) = \begin{cases} 2t, & t \in [0, 0.5], \\ 2(1-t), & t \in [0.5, 1], \end{cases}$
- Example 3: $\phi(t) = \begin{cases} 1, & t \in [0.25, 0.75], \\ 0, & t \notin [0.25, 0.75], \end{cases}$

5 Conclusion

References