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# A Trotter type result for the stochastic porous media equations

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## Abstract

This paper proves the continuous dependence with respect to diffusivity of the solutions to the stochastic porous media equations with noncoercive monotone diffusivity function and multiplicative noise.

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*Key words and phrases:* stochastic porous media equations, maximal monotone graphs, Yosida approximation, Wiener process.

## 1 Introduction

Let  $\mathcal{O}$  be an open bounded domain of  $\mathbb{R}^d$  ( $1 \leq d \leq 3$ ) with smooth boundary  $\partial\mathcal{O}$ . We also consider the stochastic partial differential equations

$$(1) \quad \begin{cases} dX(t) - \Delta\Psi(X(t)) dt \ni \sigma(X(t)) dW(t), & \text{in } (0, T) \times \mathcal{O} \\ \Psi(X(t)) \ni 0, & \text{on } (0, T) \times \partial\mathcal{O} \\ X(0) = x, & \text{in } \mathcal{O} \end{cases}$$

where  $x$  is the initial data and  $\Psi : \mathbb{R} \rightarrow 2^{\mathbb{R}}$  is a maximal monotone (possibly multivalued) graph with polynomial growth and  $\sigma(X)$  is defined by

$$(2) \quad \sigma(x)h = \sum_{k=1}^{\infty} \mu_k(h, e_k) x e_k, \quad \forall x \in H^{-1}(\mathcal{O}), \quad \forall h \in L^2(\mathcal{O}),$$

where  $(\cdot, \cdot)$  is the scalar product in  $L^2(\mathcal{O})$ .

We note that

$$\sigma(X) dW = \sum_{k=1}^{\infty} \mu_k X d\beta_k e_k, \quad \forall t \geq 0,$$

which is linear in  $X$ . Here  $\{e_k\}$  is an orthonormal basis in  $L^2(\mathcal{O})$ ,  $\{\mu_k\}$  is a sequence of positive numbers and  $\{\beta_k\}$  a sequence of independent standard Brownian motion on a filtered probability space  $(\Omega, \mathcal{F}, \{\mathcal{F}_t\}_{t \geq 0}, \mathbb{P})$ .

In this work we shall suppose that the sequence  $\{\mu_k\}$  is such that

$$(3) \quad \sum_{k=1}^{\infty} \mu_k^2 \lambda_k^2 = C < \infty,$$

where  $\lambda_k$  are the eigenvalues of the Laplace operator  $-\Delta$  in  $\mathcal{O}$  with Dirichlet boundary conditions.

Recall that the operator  $A : D(A) \subset H^{-1}(\mathcal{O}) \rightarrow H^{-1}(\mathcal{O})$  is defined by  $Ax = -\Delta\Psi(x)$  where

$$D(A) = \{x \in H^{-1}(\mathcal{O}) \cap L^1(\mathcal{O}) : \Psi(x) \in H_0^1(\mathcal{O})\}.$$

The Sobolev space  $H^{-1}(\mathcal{O})$  (the dual of  $H_0^1(\mathcal{O})$ ) is endowed with the norm

$$|x|_{H^{-1}(\mathcal{O})} = |x|_{-1} = \left| (-\Delta)^{-1} x \right|_{H_0^1(\mathcal{O})}.$$

(Here  $(-\Delta)^{-1} x = y$  is the solution to Dirichlet problem  $-\Delta y = x$  in  $\mathcal{O}$ ,  $y \in H_0^1(\mathcal{O})$ ).

The scalar product in  $H^{-1}(\mathcal{O})$  is given by

$$\langle x, z \rangle_{-1} = \int_{\mathcal{O}} (-\Delta)^{-1} x z d\xi, \quad \forall x, z \in H_0^1(\mathcal{O}).$$

We note that since  $d \leq 3$  we have by Sobolev embedding theorem

$$|e_k|_{\infty} \leq C |e_k|_{H^2(\mathcal{O})} \leq C |\Delta e_k|_{L^2(\mathcal{O})} \leq C \lambda_k$$

and for some constant  $c_1 > 0$

$$\sum_{k=1}^{\infty} \mu_k^2 |x e_k|_{-1}^2 \leq c_1 \sum_{k=1}^{\infty} \mu_k^2 \lambda_k^2 |x|_{-1}^2 \leq C_1 |x|_{-1}^2, \quad \forall x \in H^{-1}(\mathcal{O}).$$

We obtain that  $\sigma(x)$  is a Hilbert Schmidt from  $L^2(\mathcal{O})$  to  $H^{-1}(\mathcal{O})$ . Note that since  $\sigma$  is linear we have that  $x \rightarrow \sigma(x)$  is Lipschitz from  $H^{-1}(\mathcal{O})$  to  $L_2(L^2(\mathcal{O}), H^{-1}(\mathcal{O}))$ .

Recall from [9] the following definition:

**Definition 1** Let  $x \in H^{-1}(\mathcal{O})$ . An  $H^{-1}(\mathcal{O})$  valued continuous  $\mathcal{F}_t$ -adapted process  $X = X(t, x)$  is called a solution to (1) on  $[0, T]$  if

$$X \in L^p(\Omega \times (0, T) \times \mathcal{O}) \cap L^2(0, T; L^2(\Omega, H^{-1}(\mathcal{O})))$$

and there exists  $\eta \in L^{p/m}(\Omega \times (0, T) \times \mathcal{O})$  such that  $\mathbb{P}$ -a.s.

(4)

$$\begin{aligned} \langle X(t), e_j \rangle_2 &= \langle x, e_j \rangle_2 + \int_0^t \int_{\mathcal{O}} \eta(s, \xi) \Delta e_j(\xi) d\xi ds \\ &+ \sum_{k=1}^{\infty} \mu_k \int_0^t \langle X(s) e_k, e_j \rangle_2 d\beta_k(s), \quad \forall j \in \mathbb{N}, \quad t \in [0, T], \end{aligned}$$

and

$$\eta \in \Psi(X), \quad \text{a.e. in } \Omega \times (0, T) \times \mathcal{O}.$$

Here  $m$  is the exponent arising in the assumption (6) and  $\{e_k\}$  is the above orthonormal basis in  $L^2(\mathcal{O})$ . Taking into account that  $-\Delta e_k = \lambda e_k$  in  $\mathcal{O}$  we may equivalently write (4) as follows

$$\begin{aligned} \langle X(t), e_j \rangle_{-1} &= \langle x, e_j \rangle_2 - \int_0^t \int_{\mathcal{O}} \eta(s, \xi) e_j(\xi) d\xi ds \\ &\quad + \sum_{k=1}^{\infty} \mu_k \int_0^t \langle X(s) e_k, e_j \rangle_{-1} d\beta_k(s), \quad \forall j \in \mathbb{N}, \quad t \in [0, T]. \end{aligned}$$

We know also from [9] that for  $\Psi$  a maximal monotone multivalued function from  $\mathbb{R}$  into  $2^{\mathbb{R}}$  such that  $0 \in \Psi(0)$  and

$$\sup\{|\theta| : \theta \in \Psi(r)\} \leq C(1 + |r|^m), \quad \forall r \in \mathbb{R}$$

under condition (3), for each  $x \in L^p(\mathcal{O})$ ,  $p \geq \max\{2m, 4\}$  there is a unique nonnegative solution  $X \in L^\infty(0, T; L^p(\Omega; L^p(\mathcal{O})))$  to the equation (1).

In this work we are interested in the continuous dependence of the solution as function of  $\Psi$  for the stochastic porous media equation (1). This problem is relevant in asymptotic analysis and approximation of stochastic porous media equations.

To this propose we consider a family of maximal monotone graphs  $\{\Psi^\alpha\}_{\alpha>0}$ ,  $\Psi$  and denote  $A^\alpha = -\Delta\Psi^\alpha(x)$ , with

$$D(A^\alpha) = \{x \in H^{-1}(\mathcal{O}) \cap L^1(\mathcal{O}) : \Psi^\alpha(x) \in H_0^1(\mathcal{O})\}.$$

Suppose that the following assumptions are satisfied:

**H<sub>1</sub>** There exist some constants  $m \geq 1$  and  $C$  independent of  $\alpha$  such that

$$(5) \quad \sup\{|\theta| : \theta \in \Psi^\alpha(r)\} \leq C(1 + |r|^m), \quad \forall r \in \mathbb{R}$$

and

$$(6) \quad \sup\{|\theta| : \theta \in \Psi(r)\} \leq C(1 + |r|^m), \quad \forall r \in \mathbb{R}.$$

**H<sub>2</sub>** For all  $\alpha > 0$  we have  $0 \in \Psi^\alpha(0)$  and  $0 \in \Psi(0)$ .

**H<sub>3</sub>** We have  $\Psi^\alpha \rightarrow \Psi$  as  $\alpha \rightarrow 0$  in the graph sense, i. e.,

$$(1 + \lambda\Psi^\alpha)^{-1}x \longrightarrow (1 + \lambda\Psi)^{-1}x, \quad \forall \lambda > 0, \quad \forall x \in \mathbb{R}$$

for  $\alpha \rightarrow 0$ .

The main result is stated and proved in Section 2 and some examples are given in Section 3.

The following notations will be used throughout this paper.

$L^p(\mathcal{O})$ ,  $p \geq 1$ , is the usual space of  $p$ -integrable functions with norm denoted by  $|\cdot|_p$ . The scalar product in  $L^2(\mathcal{O})$  and the duality induced by the space  $L^2(\mathcal{O})$  will be denoted by  $\langle \cdot, \cdot \rangle_2$ .

For  $p, q \in [1, +\infty]$  by  $L_W^q(0, T; L^p(\Omega; H))$  ( $H$  a Hilbert space) we shall denote the space of all  $q$ -integrable processes  $u : [0, T] \rightarrow L^p(\Omega; H)$  which are adapted to the filtration  $\{\mathcal{F}_t\}_{t \geq 0}$ .

By  $C_W([0, T]; L^2(\Omega; H))$  we shall denote the space of all  $H$ -valued adapted processes which are mean square continuous (see [12], [13]).

This space is endowed with the norm

$$\|X\|_{C_W([0, T]; L^2(\Omega, \mathcal{F}, \mathbb{P}; H))}^2 = \sup_{t \in [0, T]} \mathbb{E} |X(t)|_H^2.$$

The main result (Theorem 2 below) amounts to saying that if  $\Psi^\alpha \rightarrow \Psi$ , for  $\alpha \rightarrow 0$ , then the solution  $X^\alpha$  to (7) is convergent to the solution  $X$  to (1) and this may be seen as a Trotter type result for equation (1) (see e. g. [1], [3], [11] for corresponding deterministic results).

The Theorem 2 below is the main result of this paper.

## 2 The main result

**Theorem 2** Assume that  $H_1, H_2, H_3$  and (3) hold. For each  $\alpha$  consider the corresponding equations

$$(7) \quad \begin{cases} dX^\alpha(t) - \Delta \Psi^\alpha(X^\alpha(t)) dt \ni \sigma(X^\alpha(t)) dW(t), & \text{in } (0, T) \times \mathcal{O} \\ \Psi(X(t)) \ni 0, & \text{on } (0, T) \times \partial\mathcal{O} \\ X(0) = x, & \text{in } \mathcal{O} \end{cases}.$$

Then for each  $x \in L^p(\mathcal{O})$ , the corresponding solution  $X^\alpha$  to (7) is convergent in

$$C_W([0, T]; L^2(\Omega, \mathcal{F}, \mathbb{P}; H^{-1}(\mathcal{O})))$$

for  $\alpha \rightarrow 0$  to the solution  $X$  to (1), i. e.,

$$\lim_{\alpha \rightarrow 0} \mathbb{E} |X^\alpha(t) - X(t)|_{H^{-1}(\mathcal{O})}^2 = 0$$

uniformly on  $[0, T]$ .

**Proof.** Let  $X_\lambda$  be the solution to approximating equation

$$(8) \quad \begin{cases} dX_\lambda(t) - \Delta(\Psi_\lambda(X_\lambda(t)) + \lambda X_\lambda(t)) dt = \sigma(X_\lambda(t)) dW(t), & \text{in } (0, T) \times \mathcal{O} \\ X_\lambda(0) = x, & \text{in } \mathcal{O} \end{cases}$$

where  $\Psi_\lambda$  is the Yosida approximation of  $\Psi$ , i. e.,

$$(9) \quad \Psi_\lambda(x) = \frac{1}{\lambda}(x - J_\lambda(x)) \in \Psi\left((1 + \lambda\Psi)^{-1}(x)\right), \quad \lambda > 0, x \in \mathbb{R},$$

and  $J_\lambda(x) = (1 + \lambda\Psi)^{-1}(x)$ . Note that  $x \mapsto \Psi_\lambda(x) + \lambda x$  is strictly monotonically increasing.

Denote

$$\begin{cases} A_\lambda x = -\Delta(\Psi_\lambda(x) + \lambda x); \\ D(A_\lambda) = \{x \in H^{-1}(\mathcal{O}) \cap L^1(\mathcal{O}) : \Psi_\lambda(x) + \lambda x \in H_0^1(\mathcal{O})\}. \end{cases}$$

Consider also  $X_\lambda^\alpha$  the corresponding solution to equation

$$(10) \quad \begin{cases} dX_\lambda^\alpha(t) - \Delta(\Psi_\lambda^\alpha(X_\lambda^\alpha(t)) + \lambda X_\lambda^\alpha(t)) dt = \sigma(X_\lambda^\alpha(t)) dW(t), & \text{in } (0, T) \times \mathcal{O} \\ X_\lambda^\alpha(0) = x, & \text{in } \mathcal{O} \end{cases}$$

where  $\Psi_\lambda^\alpha$  is the Yosida approximation of  $\Psi^\alpha$  for each  $\alpha$ .

Denote

$$\begin{cases} A_\lambda^\alpha x = -\Delta(\Psi_\lambda^\alpha(x) + \lambda x); \\ D(A_\lambda^\alpha) = \{x \in H^{-1}(\mathcal{O}) \cap L^1(\mathcal{O}) : \Psi_\lambda^\alpha(x) + \lambda x \in H_0^1(\mathcal{O})\}. \end{cases}$$

We have

$$\begin{aligned} \mathbb{E} |X^\alpha(t) - X(t)|_{-1}^2 &\leq 3 \left( \mathbb{E} |X^\alpha(t) - X_\lambda^\alpha(t)|_{-1}^2 + \mathbb{E} |X_\lambda^\alpha(t) - X_\lambda(t)|_{-1}^2 \right. \\ &\quad \left. + \mathbb{E} |X_\lambda(t) - X(t)|_{-1}^2 \right). \end{aligned}$$

By (6) we know from [[9], (3.14)] that for  $\lambda \rightarrow 0$  we have

$$(11) \quad (X_\lambda - X) \rightarrow 0 \quad \text{strongly in } L^2(\Omega; C([0, T]; H^{-1}(\mathcal{O}))).$$

We shall prove now that as  $\lambda \rightarrow 0$  we have

$$(12) \quad (X_\lambda^\alpha - X^\alpha) \rightarrow 0 \quad \text{strongly in } L^2(\Omega; C([0, T]; H^{-1}(\mathcal{O})))$$

uniformly in  $\alpha > 0$ .

Consider the section

$$\eta^\alpha \in \Psi^\alpha(X^\alpha), \quad a.e. \quad \text{in } \Omega \times (0, T) \times \mathcal{O}$$

which arises in [7].

Using Ito's formula for equation

$$d(X^\alpha(t) - X_\lambda^\alpha(t)) - \Delta(\eta^\alpha(t) - \Psi_\lambda^\alpha(X_\lambda^\alpha(t)) - \lambda X_\lambda^\alpha(t)) dt = \sigma(X^\alpha(t) - X_\lambda^\alpha(t)) dW(t)$$

with  $\varphi(t, x) = |x|_{-1}^2 e^{-\varepsilon t}$ , we get that

$$\begin{aligned} & \frac{1}{2} |X^\alpha(t) - X_\lambda^\alpha(t)|_{-1}^2 e^{-\varepsilon t} + \int_0^t \int_{\mathcal{O}} [\eta^\alpha(s) - \Psi_\lambda^\alpha(X_\lambda^\alpha(s)) - \lambda X_\lambda^\alpha(s)] (X^\alpha(s) - X_\lambda^\alpha(s)) e^{-\varepsilon s} d\xi ds \\ & \leq \int_0^t e^{-\varepsilon s} \langle X^\alpha(s) - X_\lambda^\alpha(s), \sigma(X^\alpha(s) - X_\lambda^\alpha(s)) dW(s) \rangle_{-1} \\ & \quad + \left( -\frac{1}{2} \varepsilon \right) \left( \int_0^t |X^\alpha(s) - X_\lambda^\alpha(s)|_{-1}^2 e^{-\varepsilon s} ds \right) \\ & \quad + c \sum_{k=1}^{\infty} \mu_k^2 \lambda_k^2 \int_0^t |X^\alpha(s) - X_\lambda^\alpha(s)|_{-1}^2 e^{-\varepsilon s} ds, \quad \mathbb{P}\text{-a.s.} \end{aligned}$$

By (9) we have  $x = \lambda \Psi_\lambda^\alpha(x) + (1 + \lambda \Psi^\alpha)^{-1}(x)$  and this yields

$$\begin{aligned} & [\eta^\alpha(s) - \Psi_\lambda^\alpha(X_\lambda^\alpha(s)) - \lambda X_\lambda^\alpha(s)] (X^\alpha(s) - X_\lambda^\alpha(s)) \\ & = \left( \eta^\alpha(s) - \Psi^\alpha \left( (1 + \lambda \Psi^\alpha)^{-1} X_\lambda^\alpha(s) \right) \right) \left( X^\alpha(s) - (1 + \lambda \Psi^\alpha)^{-1} X_\lambda^\alpha(s) \right) \\ & \quad - \lambda (\eta^\alpha(s) - \Psi_\lambda^\alpha(X_\lambda^\alpha(s))) \Psi_\lambda^\alpha(X_\lambda^\alpha(s)) - \lambda X_\lambda^\alpha(s) (X^\alpha(s) - X_\lambda^\alpha(s)) \\ & \geq \lambda \left( |\Psi_\lambda^\alpha(X_\lambda^\alpha(s))|^2 - \eta^\alpha(s) \Psi_\lambda^\alpha(X_\lambda^\alpha(s)) \right) + \lambda \left( |X_\lambda^\alpha(s)|^2 - X_\lambda^\alpha(s) X^\alpha(s) \right) \\ & \geq -\frac{\lambda}{4} |\eta^\alpha(s)|^2 - \frac{\lambda}{4} |X^\alpha(s)|^2, \quad \mathbb{P}\text{-a.s.} \end{aligned}$$

using the monotonicity of  $\Psi^\alpha$  and  $\Psi_\lambda^\alpha(x) \in \Psi^\alpha \left( (1 + \lambda \Psi^\alpha)^{-1}(x) \right)$  for all  $x \in \mathbb{R}$ .

Hence for  $\varepsilon > 0$  large enough we obtain for all  $\lambda \in (0, 1)$  and  $t \in [0, T]$

$$(13) \quad \begin{aligned} \frac{1}{2} |X^\alpha(t) - X_\lambda^\alpha(t)|_{-1}^2 e^{-\varepsilon t} & \leq \frac{\lambda}{4} \int_0^t \int_{\mathcal{O}} \left( |\eta^\alpha(s)|^2 + |X^\alpha(s)|^2 \right) d\xi ds \\ & \quad + \int_0^t e^{-\varepsilon s} \langle X^\alpha(s) - X_\lambda^\alpha(s), \sigma(X^\alpha(s) - X_\lambda^\alpha(s)) dW(s) \rangle_{-1}. \end{aligned}$$

We get for  $\varepsilon > 0$ , for all  $\lambda \in (0, 1)$ , and  $r \in [0, T]$  that

$$(14) \quad \begin{aligned} \frac{1}{4} \mathbb{E} \sup_{t \in [0, r]} |X^\alpha(s) - X_\lambda^\alpha(s)|_{-1}^2 e^{-\varepsilon t} & \leq \frac{\lambda}{4} \mathbb{E} \int_0^r \int_{\mathcal{O}} \left( |\eta^\alpha(s)|^2 + |X^\alpha(s)|^2 \right) d\xi ds \\ & \quad + c \mathbb{E} \left( \int_0^r |X^\alpha(s) - X_\lambda^\alpha(s)|_{-1}^2 e^{-\varepsilon s} ds \right) \end{aligned}$$

since by the Burkholder-Davis-Gundy inequality, we have

$$\begin{aligned}
& \int_0^t e^{-\varepsilon s} \langle X^\alpha(s) - X_\lambda^\alpha(s), \sigma(X^\alpha(s) - X_\lambda^\alpha(s)) dW(s) \rangle_{-1} \\
& \leq \mathbb{E} \left( c \int_0^r |X^\alpha(s) - X_\lambda^\alpha(s)|_{-1}^4 e^{-2\varepsilon s} ds \right)^{\frac{1}{2}} \\
& \leq \mathbb{E} \sup_{s \in [0, r]} |X^\alpha(s) - X_\lambda^\alpha(s)|_{-1} e^{-\varepsilon s/2} \left( c \int_0^r |X^\alpha(s) - X_\lambda^\alpha(s)|_{-1}^2 e^{-\varepsilon s} ds \right)^{1/2} \\
& \leq \frac{1}{4} \mathbb{E} \sup_{s \in [0, r]} |X^\alpha(s) - X_\lambda^\alpha(s)|_{-1}^2 e^{-\varepsilon s} + c \mathbb{E} \left( \int_0^r |X^\alpha(s) - X_\lambda^\alpha(s)|_{-1}^2 e^{-\varepsilon s} ds \right).
\end{aligned}$$

By the hypothesis  $\mathbf{H}_1$  we have for all  $x \in \mathbb{R}$  and all  $\eta^\alpha \in \Psi^\alpha(x)$

$$|\eta^\alpha| \leq C(1 + |x|^m).$$

Consequently for  $\eta^\alpha \in \Psi^\alpha(X^\alpha)$ , *a.e.* in  $\Omega \times (0, T) \times \mathcal{O}$  we get that

$$\begin{aligned}
(15) \quad \frac{\lambda}{4} \mathbb{E} \int_0^t \int_{\mathcal{O}} (|\eta^\alpha(s)|^2 + |X^\alpha(s)|^2) d\xi ds & \leq \frac{\lambda}{4} C \mathbb{E} \int_0^t \int_{\mathcal{O}} (|(1 + |X^\alpha(s)|^m)|^2 + |X^\alpha(s)|^2) d\xi ds \\
& \leq \frac{\lambda}{4} C \left( 1 + \mathbb{E} \int_0^t \int_{\mathcal{O}} |X^\alpha(s)|^p d\xi ds \right)
\end{aligned}$$

since  $p \geq \max\{2m, 2\}$  and  $C$  is independent of  $\lambda$  and  $\alpha$ .

We prove that

$$(16) \quad \operatorname{ess\,sup}_{t \in [0, T]} \mathbb{E} |X_\lambda^\alpha(t, x)|_p^p \leq \exp\left(c \frac{p-1}{2}\right) |x|_p^p \quad \forall \lambda > 0, \alpha > 0,$$

where  $c > 0$  is independent of  $t, x, \lambda$  and  $\alpha$ .

Note that relation (16) is similar to Lemma 3.1 from [9], but in the present paper we are interested to get  $c$  independent of  $\alpha$ .

Indeed, for  $A_\lambda^\alpha x = -\Delta(\Psi_\lambda^\alpha(x) + \lambda x)$ , we take  $(A_\lambda^\alpha)_\varepsilon$  the Yosida approximation of  $A_\lambda^\alpha$ ,

$$(A_\lambda^\alpha)_\varepsilon = \frac{1}{\varepsilon} \left( I - (I + \varepsilon A_\lambda^\alpha)^{-1} \right), \quad \varepsilon > 0$$

and we apply the Ito formula to

$$(17) \quad d(X_\lambda^\alpha)_\varepsilon(t) + (A_\lambda^\alpha)_\varepsilon(X_\lambda^\alpha)_\varepsilon(t) dt = \sigma((X_\lambda^\alpha)_\varepsilon(t)) dW(t)$$

for the function  $\varphi(x) = \frac{1}{p} |x|_p^p$ . (More precisely we first apply Ito's formula to (17) for the function  $\varphi_\gamma(x) = \frac{1}{p} |(1 - \gamma \Delta)^{-1} x|_p^p$ ,  $\gamma > 0$ , and then we let  $\gamma \rightarrow 0$ . For more details see [[6], Lemma 3.5]).

We get

$$\begin{aligned}
(18) \quad \mathbb{E} \varphi((X_\lambda^\alpha)_\varepsilon(t)) + \mathbb{E} \int_0^t \left\langle (A_\lambda^\alpha)_\varepsilon((X_\lambda^\alpha)_\varepsilon(s)), |(X_\lambda^\alpha)_\varepsilon(s)|^{p-2} (X_\lambda^\alpha)_\varepsilon(s) \right\rangle_2 ds \\
& = \varphi(x) + \frac{p-1}{2} \sum_{k=1}^{\infty} \mu_k^2 \mathbb{E} \int_0^t \int_{\mathcal{O}} |(X_\lambda^\alpha)_\varepsilon(s)|^{p-2} |(X_\lambda^\alpha)_\varepsilon(s) e_k|^2 d\xi ds \\
& \leq \varphi(x) + \frac{p-1}{2} c \mathbb{E} \int_0^t \int_{\mathcal{O}} |(X_\lambda^\alpha)_\varepsilon(s)|^p d\xi ds.
\end{aligned}$$

By [[6], (3.25)] , we have  $|(Y_\lambda^\alpha)_\varepsilon|_p \leq |(X_\lambda^\alpha)_\varepsilon|_p$  and this leads to

$$\left\langle (A_\lambda^\alpha)_\varepsilon (X_\lambda^\alpha)_\varepsilon, |(X_\lambda^\alpha)_\varepsilon|^{p-2} (X_\lambda^\alpha)_\varepsilon \right\rangle_2 = \frac{1}{\varepsilon} \left\langle (X_\lambda^\alpha)_\varepsilon - (Y_\lambda^\alpha)_\varepsilon, |(X_\lambda^\alpha)_\varepsilon|^{p-2} (X_\lambda^\alpha)_\varepsilon \right\rangle_2 \geq 0$$

where  $(Y_\lambda^\alpha)_\varepsilon = (I + \varepsilon A_\lambda^\alpha)^{-1} (X_\lambda^\alpha)_\varepsilon$ .

On another hand we have from [[6], Lemma 3.4]

$$\begin{aligned} (X_\lambda^\alpha)_\varepsilon &\rightarrow X_\lambda^\alpha \quad \text{strongly in } L_W^\infty(0, T; L^2(\Omega; H^{-1}(\mathcal{O}))), \\ (X_\lambda^\alpha)_\varepsilon &\rightarrow X_\lambda^\alpha \quad \text{weak}^* \text{ in } L_W^\infty(0, T; L^p(\Omega; L^p(\mathcal{O}))). \end{aligned}$$

Using Gronwall's lemma in (18) and letting  $\varepsilon$  tend to 0, we obtain (16) with  $c > 0$  is independent of  $t, x, \lambda$  and  $\alpha$ .

From [[9], (3.8)] we have for  $\lambda \rightarrow 0$

$$X_\lambda^\alpha \rightarrow X^\alpha \quad \text{weak}^* \text{ in } L^\infty(0, T; L^p(\Omega; L^p(\mathcal{O}))).$$

Using [[10], Proposition III.12.] this yields

$$\begin{aligned} \operatorname{ess\,sup}_{t \in [0, T]} \mathbb{E} |X^\alpha(t, x)|_p^p &\leq \liminf_\lambda \left( \operatorname{ess\,sup}_{t \in [0, T]} \mathbb{E} |X_\lambda^\alpha(t, x)|_p^p \right) \\ &\leq \exp\left(c \frac{p-1}{2}\right) |x|_p^p \leq C_1 |x|_p^p \end{aligned}$$

with  $C_1 > 0$  is independent of  $t, x, \lambda$  and  $\alpha$ .

Coming back to (15) we get that

$$\begin{aligned} \frac{\lambda}{4} \mathbb{E} \int_0^t \int_{\mathcal{O}} (|\eta^\alpha(s)|^2 + |X^\alpha(s)|^2) d\xi ds &\leq \frac{\lambda}{4} C \left( 1 + \mathbb{E} \int_0^t \int_{\mathcal{O}} |X^\alpha(s)|^p d\xi ds \right) \\ &\leq \frac{\lambda}{4} C_2 \left( 1 + \operatorname{ess\,sup}_{t \in [0, T]} \mathbb{E} |X^\alpha(t, x)|_p^p \right) \\ &\leq \frac{\lambda}{4} C_3 \left( 1 + |x|_p^p \right) \end{aligned}$$

with  $C_3 > 0$  is independent of  $t, x, \lambda$  and  $\alpha$ .

Using Gronwall's lemma in (14) we get that

$$(X_\lambda^\alpha - X^\alpha) \rightarrow 0 \quad \text{strongly in } L^2(\Omega; C([0, T]; H^{-1}(\mathcal{O})))$$

for  $\lambda \rightarrow 0$  uniformly in  $\alpha > 0$ .

In order to complete the proof it suffices to show that

$$(X_\lambda^\alpha - X_\lambda) \rightarrow 0 \quad \text{strongly in } L^2(\Omega; C([0, T]; H^{-1}(\mathcal{O}))), \quad \forall \lambda > 0.$$

as  $\alpha \rightarrow 0$ .

Applying Ito's formula in equation

$$\begin{aligned} d(X_\lambda^\alpha(t) - X_\lambda(t)) - \Delta(\Psi_\lambda^\alpha(X_\lambda^\alpha(t)) + \lambda X_\lambda^\alpha(t) - \Psi_\lambda(X_\lambda(t)) - \lambda X_\lambda(t)) dt \\ = \sigma(X_\lambda^\alpha(t) - X_\lambda(t)) dW(t) \end{aligned}$$



with  $\varphi(t, x) = |x|_{-1}^2 e^{-\varepsilon t}$  we have

$$\begin{aligned}
& \frac{1}{2} |X_\lambda^\alpha(t) - X_\lambda(t)|_{-1}^2 e^{-\varepsilon t} + \int_0^t \int_{\mathcal{O}} [\Psi_\lambda^\alpha(X_\lambda^\alpha(s)) - \Psi_\lambda(X_\lambda(s))] (X_\lambda^\alpha(s) - X_\lambda(s)) e^{-\varepsilon s} d\xi ds \\
& + \lambda \int_0^t \int_{\mathcal{O}} |X_\lambda^\alpha(s) - X_\lambda(s)|^2 e^{-\varepsilon s} d\xi ds \\
& \leq \int_0^t e^{-\varepsilon s} \langle X_\lambda^\alpha(s) - X_\lambda(s), \sigma(X_\lambda^\alpha(s) - X_\lambda(s)) dW(s) \rangle_{-1} \\
& + \left(-\frac{1}{2}\varepsilon\right) \left(\int_0^t |X_\lambda^\alpha(s) - X_\lambda(s)|_{-1}^2 e^{-\varepsilon s} ds\right) + c \sum_{k=1}^{\infty} \mu_k^2 \lambda_k^2 \int_0^t |X_\lambda^\alpha(s) - X_\lambda(s)|_{-1}^2 e^{-\varepsilon s} ds,
\end{aligned}$$

and for  $\varepsilon > 0$ , large enough, we get after some calculation involving the Burkholder-Davis-Gundy inequality, that

$$\begin{aligned}
& \frac{1}{4} \mathbb{E} \sup_{t \in [0, r]} |X_\lambda^\alpha(t) - X_\lambda(t)|_{-1}^2 e^{-\varepsilon t} \\
& + \mathbb{E} \int_0^r \int_{\mathcal{O}} [\Psi_\lambda^\alpha(X_\lambda^\alpha(s)) - \Psi_\lambda(X_\lambda(s))] (X_\lambda^\alpha(s) - X_\lambda(s)) e^{-\varepsilon s} d\xi ds \\
& \leq c \mathbb{E} \left( \int_0^r |X_\lambda^\alpha(s) - X_\lambda(s)|_{-1}^2 e^{-\varepsilon s} ds \right).
\end{aligned}$$

It is easily seen that

$$[\Psi_\lambda^\alpha(X_\lambda^\alpha(s)) - \Psi_\lambda(X_\lambda(s))] (X_\lambda^\alpha(s) - X_\lambda(s)) \geq [\Psi_\lambda^\alpha(X_\lambda(s)) - \Psi_\lambda(X_\lambda(s))] (X_\lambda^\alpha(s) - X_\lambda(s)),$$

$\mathbb{P}$ -a.s., since by the monotonicity of  $\Psi_\lambda$  we have that

$$(\Psi_\lambda^\alpha(X_\lambda^\alpha(s)) - \Psi_\lambda^\alpha(X_\lambda(s))) (X_\lambda^\alpha(s) - X_\lambda(s)) \geq 0.$$

We obtain that

$$(19) \quad \frac{1}{4} \mathbb{E} \sup_{t \in [0, r]} |X_\lambda^\alpha(t) - X_\lambda(t)|_{-1}^2 e^{-\varepsilon t}$$

$$\begin{aligned}
(20) \quad & \leq c \mathbb{E} \int_0^r |X_\lambda^\alpha(s) - X_\lambda(s)|_{-1}^2 e^{-\varepsilon s} ds \\
& + \mathbb{E} \int_0^r \int_{\mathcal{O}} [\Psi_\lambda^\alpha(X_\lambda(s)) - \Psi_\lambda(X_\lambda(s))] (X_\lambda(s) - X_\lambda^\alpha(s)) e^{-\varepsilon s} d\xi ds.
\end{aligned}$$

We have also that

$$\begin{aligned}
& \mathbb{E} \int_0^r \int_{\mathcal{O}} [\Psi_\lambda^\alpha(X_\lambda(s)) - \Psi_\lambda(X_\lambda(s))] (X_\lambda(s) - X_\lambda^\alpha(s)) e^{-\varepsilon s} d\xi ds \\
& = \langle \Psi_\lambda^\alpha(X_\lambda(s)) - \Psi_\lambda(X_\lambda(s)), (X_\lambda(s) - X_\lambda^\alpha(s)) e^{-\varepsilon s} \rangle_{L^2(\Omega \times [0, r] \times \mathcal{O})} \\
& \leq |\Psi_\lambda^\alpha(X_\lambda(s)) - \Psi_\lambda(X_\lambda(s))|_{L^2(\Omega \times [0, r] \times \mathcal{O})} |(X_\lambda(s) - X_\lambda^\alpha(s)) e^{-\varepsilon s}|_{L^2(\Omega \times [0, r] \times \mathcal{O})}.
\end{aligned}$$

Since  $p > 2$  we have that

$$\begin{aligned}
& |(X_\lambda(s) - X_\lambda^\alpha(s)) e^{-\varepsilon s}|_{L^2(\Omega \times [0, r] \times \mathcal{O})} \\
& \leq C |X_\lambda|_{L^p(\Omega \times [0, r] \times \mathcal{O})} + C |X_\lambda^\alpha|_{L^p(\Omega \times [0, r] \times \mathcal{O})} \\
& \leq C \left( \int_0^r \mathbb{E} |X_\lambda(s)|_{L^p(\mathcal{O})}^p ds \right)^{1/p} + C \left( \int_0^r \mathbb{E} |X_\lambda^\alpha(s)|_{L^p(\mathcal{O})}^p ds \right)^{1/p}
\end{aligned}$$

and by [[9], Lemma 3.1] and (16) we have

$$|(X_\lambda(s) - X_\lambda^\alpha(s)) e^{-\varepsilon s}|_{L^2(\Omega \times [0,r] \times \mathcal{O})} \leq C_4 \left(1 + |x|_p^p\right)^{1/p},$$

where  $C_4$  is independent of  $x, t, \lambda$  and  $\alpha$ .

On the other hand by  $\mathbf{H}_1$  and [[9], Lemma 3.1] we have

$$\begin{aligned} (21) \quad & |\Psi_\lambda^\alpha(X_\lambda(s)) - \Psi_\lambda(X_\lambda(s))|_{L^2(\Omega \times [0,r] \times \mathcal{O})} \\ & \leq \left( \mathbb{E} \int_0^t \int_{\mathcal{O}} (|\Psi_\lambda^\alpha(X_\lambda(s))|^2) d\xi ds \right)^{1/2} + \left( \mathbb{E} \int_0^t \int_{\mathcal{O}} (|\Psi_\lambda(X_\lambda(s))|^2) d\xi ds \right)^{1/2} \\ & \leq C_5 \left( \mathbb{E} \int_0^t \int_{\mathcal{O}} (|1 + |X_\lambda(s)|^m|^2) d\xi ds \right)^{1/2} \\ & \leq C_6 \left(1 + |x|_p^p\right)^{1/2}. \end{aligned}$$

with  $C_6$  independent of  $x, t, \lambda$  and  $\alpha$ .

Using  $\mathbf{H}_3$ , and

$$(\Psi_\lambda(X_\lambda(s)) - \Psi_\lambda^\alpha(X_\lambda(s))) = \frac{1}{\lambda} \left( (1 + \lambda\Psi)^{-1} X_\lambda(s) - (1 + \lambda\Psi^\alpha)^{-1} X_\lambda(s) \right)$$

we get

$$(22) \quad \Psi_\lambda^\alpha(X_\lambda) \rightarrow \Psi_\lambda(X_\lambda) \text{ as } \alpha \rightarrow 0, \text{ a. e. on } \Omega \times [0, r] \times \mathcal{O}.$$

We obtain from (21) and (22) via the Lebesgue dominated convergence theorem that

$$|\Psi_\lambda^\alpha(X_\lambda(s)) - \Psi_\lambda(X_\lambda(s))|_{L^2(\Omega \times [0,r] \times \mathcal{O})} \rightarrow 0 \text{ as } \alpha \rightarrow 0.$$

Gronwall's lemma applied to (19) leads to

$$\begin{aligned} & \frac{1}{4} \mathbb{E} \sup_{t \in [0,r]} |X_\lambda^\alpha(t) - X_\lambda(t)|_{-1}^2 \\ & \leq |\Psi_\lambda^\alpha(X_\lambda(s)) - \Psi_\lambda(X_\lambda(s))|_{L^2(\Omega \times [0,r] \times \mathcal{O})} |(X_\lambda(s) - X_\lambda^\alpha(s)) e^{-\varepsilon s}|_{L^2(\Omega \times [0,r] \times \mathcal{O})} \end{aligned}$$

and finally we get that

$$\mathbb{E} \sup_{t \in [0,r]} |X_\lambda^\alpha(t) - X_\lambda(t)|_{-1}^2 \rightarrow 0, \text{ as } \alpha \rightarrow 0, \quad \forall \lambda > 0.$$

We can now come back to

$$\begin{aligned} \mathbb{E} |X^\alpha(t) - X(t)|_{-1}^2 & \leq 3 \left( \mathbb{E} |X^\alpha(t) - X_\lambda^\alpha(t)|_{-1}^2 + \mathbb{E} |X_\lambda^\alpha(t) - X_\lambda(t)|_{-1}^2 \right. \\ & \quad \left. + \mathbb{E} |X_\lambda(t) - X(t)|_{-1}^2 \right). \end{aligned}$$

Given  $\varepsilon > 0$  we first choose  $\lambda$ , independent of  $\alpha$ , such that the first and the tierd terms are less then  $\frac{\varepsilon}{3}$ . Having fixed  $\lambda$  this way we can choose  $\alpha$  such that the second term is less then  $\frac{\varepsilon}{3}$  and finally we obtain

$$\mathbb{E} |X^\alpha(t) - X(t)|_{-1}^2 \leq \varepsilon \text{ uniformly on } [0, T].$$

The proof of the main result is now complete.  $\blacksquare$

### 3 Examples

1° Let  $\Psi : \mathbb{R} \rightarrow 2^{\mathbb{R}}$  defined by

$$\Psi^\alpha (X) = |X|^\alpha \text{sign}X, \quad 0 \leq \alpha < 1.$$

Equation (7) is called in this case the stochastic fast diffusion equation and is relevant in plasma physics (see [2]).

The case  $\alpha = 0$  is relevant in stochastic models for self-organized criticality. The existence and longtime behaviors of solutions for equation were studied in [6], [7], [9], [14].

The extinction in finite time of solution for  $0 < \alpha < 1$  was studied in [8].

As a consequence of Theorem 2 we obtain:

**Corollary 3** Consider the solution  $X^\alpha$  to equation

$$(23) \quad \begin{cases} dX^\alpha (t) - \Delta (|X^\alpha (t)|^\alpha \text{sign}X^\alpha (t)) dt \ni \sigma (X^\alpha (t)) dW (t), & \text{in } (0, T) \times \mathcal{O} \\ |X^\alpha (t)|^\alpha \text{sign}X^\alpha (t) \ni 0, & \text{on } (0, T) \times \partial\mathcal{O} \\ X (0) = x, & \text{in } \mathcal{O} \end{cases} .$$

Then for each  $x \in L^p (\mathcal{O})$  and  $\alpha \rightarrow 0$  the corresponding solution  $X^\alpha$  to equation (23) is convergent in

$$C_W ([0, T]; L^2 (\Omega, \mathcal{F}, \mathbb{P}; H^{-1} (\mathcal{O})))$$

to the solution  $X$  to equations

$$(24) \quad \begin{cases} dX (t) - \Delta (\text{sign}X (t)) dt \ni \sigma (X (t)) dW (t), & \text{in } (0, T) \times \mathcal{O} \\ \text{sign}X (t) \ni 0, & \text{on } (0, T) \times \partial\mathcal{O} \\ X (0) = x, & \text{in } \mathcal{O} \end{cases}$$

i. e.,

$$\mathbb{E} |X^\alpha (t) - X (t)|_{H^{-1}(\mathcal{O})}^2 \rightarrow 0 \text{ uniformly on } [0, T] \text{ as } \alpha \rightarrow 0.$$

**Proof.**

It is easily seen that  $\Psi, \Psi^\alpha : \mathbb{R} \rightarrow 2^{\mathbb{R}}$  are maximal monotone graphs.

Since  $\Psi^\alpha (X) = |X|^\alpha \text{sign}X = |X|^{\alpha-1} X$  and  $\alpha < 1 \leq m$  we have

$$\begin{aligned} \sup \{|\theta| : \theta \in \Psi^\alpha (X)\} &= \sup \{|\theta| : \theta \in |X|^\alpha \text{sign}X\} \\ &\leq C (1 + |X|^m), \quad \forall X \in \mathbb{R}. \end{aligned}$$

We also have that

$$(1 + \lambda \Psi^\alpha)^{-1} x \longrightarrow (1 + \lambda \Psi)^{-1} x, \quad \forall \lambda > 0, \quad \forall x \in \mathbb{R}$$

(for details see [1]).

The proof of the Corollary is now complete. ■

**Remark 4** The limit equation (24) is related to the model of self-organized criticality under stochastic perturbation (see [9]).

2° The diffusivity function  $\Psi : \mathbb{R} \rightarrow 2^{\mathbb{R}}$  of stochastic fast diffusion equation can also be written as

$$\Psi^\alpha (X) = |X|^{1-\alpha} \text{sign}X, \quad 0 < \alpha \leq 1.$$

In case  $\alpha$  is near 0, the corresponding equation can be regarded as a perturbation of stochastic heat equation.

By Theorem 2 we have that for each  $x \in L^p(\mathcal{O})$  and  $\alpha \rightarrow 0$  the solution  $X^\alpha$  to equation

$$(25) \quad \begin{cases} dX^\alpha(t) - \Delta \left( |X^\alpha(t)|^{1-\alpha} \text{sign} X^\alpha(t) \right) dt \ni \sigma(X^\alpha(t)) dW(t), & \text{in } (0, T) \times \mathcal{O} \\ |X^\alpha(t)|^{1-\alpha} \text{sign} X^\alpha(t) \ni 0, & \text{on } (0, T) \times \partial\mathcal{O} \\ X(0) = x, & \text{in } \mathcal{O} \end{cases} ,$$

is convergent in  $C_W([0, T]; L^2(\Omega, \mathcal{F}, \mathbb{P}; H^{-1}(\mathcal{O})))$  to the solution  $X$  to the linear stochastic heat equations

$$\begin{cases} dX(t) - \Delta X(t) dt = \sigma(X(t)) dW(t), & \text{in } (0, T) \times \mathcal{O} \\ X(t) = 0, & \text{on } (0, T) \times \partial\mathcal{O} \\ X(0) = x, & \text{in } \mathcal{O} \end{cases} ,$$

i. e.,  $\mathbb{E} |X^\alpha(t) - X(t)|_{H^{-1}(\mathcal{O})}^2 \rightarrow 0$  uniformly on  $[0, T]$  as  $\alpha \rightarrow 0$ .

To conclude the second example, we just have to repeat the proof of the Corollary 3.

**3°** Let  $\Psi : \mathbb{R} \rightarrow 2^{\mathbb{R}}$  be a maximal monotone graph of the form

$$\Psi(X) = \begin{cases} \Psi_1(X), & \text{if } X < a \\ [\Psi_1(a), \Psi_2(a)], & \text{if } X = a \\ \Psi_2(X), & \text{if } X > a \end{cases} ,$$

where  $\Psi_1$  and  $\Psi_2$  are continuous and monotone functions satisfying the assumption (6).

We define the approximation

$$\Psi^\alpha(X) = \begin{cases} \Psi_1(X), & \text{if } X < a - \alpha \\ \Psi_1(a - \alpha) \frac{a + \alpha - X}{2\alpha} + \Psi_2(a + \alpha) \frac{a - \alpha - X}{-2\alpha}, & \text{if } a - \alpha \leq X \leq a + \alpha \\ \Psi_2(X), & \text{if } a + \alpha < X \end{cases} .$$

Note that we have the approximation of a maximal monotone graph by continuous and monotone functions.

Using Theorem 2 we can prove the following corollary.

**Corollary 5** For each  $x \in L^p(\mathcal{O})$  and  $\alpha \rightarrow 0$  the corresponding solution  $X^\alpha$  to equation

$$(26) \quad \begin{cases} dX^\alpha(t) - \Delta \Psi^\alpha(X^\alpha(t)) dt = \sigma(X^\alpha(t)) dW(t), & \text{in } (0, T) \times \mathcal{O} \\ \Psi(X^\alpha(t)) = 0, & \text{on } (0, T) \times \partial\mathcal{O} \\ X(0) = x, & \text{in } \mathcal{O} \end{cases} ,$$

is convergent in  $C_W([0, T]; L^2(\Omega, \mathcal{F}, \mathbb{P}; H^{-1}(\mathcal{O})))$  to the solution  $X$  of equations

$$\begin{cases} dX(t) - \Delta \Psi(X(t)) dt \ni \sigma(X(t)) dW(t), & \text{in } (0, T) \times \mathcal{O} \\ \Psi(X(t)) \ni 0, & \text{on } (0, T) \times \partial\mathcal{O} \\ X(0) = x, & \text{in } \mathcal{O} \end{cases} ,$$

i. e.,  $\mathbb{E} |X^\alpha(t) - X(t)|_{H^{-1}(\mathcal{O})}^2 \rightarrow 0$  uniformly on  $[0, T]$  as  $\alpha \rightarrow 0$ .

**Proof.** Since  $\Psi_1$  and  $\Psi_2$  satisfies assumption (6) of  $\mathbf{H}_1$  and for all  $x \in \mathbb{R}$  we have  $\lim_{\alpha \rightarrow 0} \Psi^\alpha(x) = \Psi(x)$ , the proof of the Corollary 5 is immediate. ■

As a particular case we have the Heavside step function

$$H(x) = \begin{cases} 0, & \text{if } x < 0 \\ [0, 1], & \text{if } X = 0 \\ 1, & \text{if } X > 0 \end{cases},$$

which is relevant in the anomalous (singular) diffusion equation of the type

$$dX(t) = \Delta(H(X(t) - x_c)) dt + \sigma(X(t)) dW(t),$$

with  $x_c$  the critical value (see [9]).

Another particular case is

$$\Psi(X) = \text{sign}X = \begin{cases} \frac{X}{|X|}, & \text{if } X \neq 0 \\ [-1, 1], & \text{if } X = 0 \end{cases},$$

which as mentioned above is relevant in the stochastic models for self-organized criticality.

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