

Large deviation principles for the stochastic quasi-geostrophic equations

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Abstract

In this paper we establish the large deviation principle for the stochastic quasi-geostrophic equation with small multiplicative noise in the subcritical case. The proof is mainly based on the weak convergence approach. Some analogous results are also obtained for the small time asymptotics of the stochastic quasi-geostrophic equation.

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1 Introduction

The main aim of this work is to establish large deviation principles for the stochastic quasi-geostrophic equation, which is an important model in geophysical fluid dynamics. We consider the following two dimensional (2D) stochastic quasi-geostrophic equation in the periodic domain $\mathbb{T}^2 = \mathbb{R}^2 / (2\pi\mathbb{Z})^2$:

$$\frac{\partial\theta(t, x)}{\partial t} = -u(t, x) \cdot \nabla\theta(t, x) - \kappa(-\Delta)^\alpha\theta(t, x) + (G(\theta)\xi)(t, x) \quad (1.1)$$

with initial condition

$$\theta(0, x) = \theta_0(x). \quad (1.2)$$

Here $0 < \alpha < 1, \kappa > 0$ are real numbers, $\theta(t, x)$ (representing the potential temperature) is a real-valued function of t and x , $\xi(t, x)$ is a Gaussian random field, white noise in time and subject to the restrictions imposed below, u (representing the fluid velocity) is determined by θ via the following relation:

$$u = (u_1, u_2) = (-R_2\theta, R_1\theta) = R^\perp\theta, \quad (1.3)$$

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where R_j is the j -th periodic Riesz transform. The case $\alpha = \frac{1}{2}$ is called the critical case, the case $\alpha > \frac{1}{2}$ subcritical and the case $\alpha < \frac{1}{2}$ supercritical.

Equation (1.1) is used to describe models arising in meteorology and oceanography. In the deterministic case ($G = 0$) such equations are important models in geophysical fluid dynamics. Indeed, they are special cases of general quasi-geostrophic approximations for atmospheric and oceanic fluid flows with small Rossby and Ekman numbers. These models arise under the assumptions of fast rotation, uniform stratification and uniform potential vorticity. The case $\alpha = 1/2$ exhibits similar features (singularities) as the 3D Navier-Stokes equations and can therefore serve as a model case for the latter. For more details about the geophysical background, see for instance [7, 24]. In the deterministic case, this equation has been already intensively investigated because of both its intrinsic mathematical importance and its applications in geophysical fluid dynamics (see e.g. [5, 8, 17, 18, 19, 27] and the references therein). For example, the global existence of weak solutions has been obtained in [27] and one very remarkable result in [5] proved the existence of a classical solution for $\alpha = \frac{1}{2}$ and the other in [19] proved that solutions for $\alpha = \frac{1}{2}$ with periodic C^∞ data remain C^∞ for all time.

Recently, in [28] the two last named authors and Rongchan Zhu have studied the 2D stochastic quasi-geostrophic equation on \mathbb{T}^2 for general parameter $\alpha \in (0, 1)$ and for both additive as well as multiplicative noise. For the subcritical case $\alpha > \frac{1}{2}$ the authors obtained a (probabilistically strong) solution. In this paper, we want to establish the large deviation principles for stochastic quasi-geostrophic equation both for small noise and for short time in the subcritical case.

The large deviation theory concerns the asymptotic behavior of a family of random variables $\{\theta_\varepsilon\}$ and we refer to the monographs [9, 31] for many historical remarks and extensive references. It asserts that for some tail or extreme event A , $P(\theta_\varepsilon \in A)$ converges to zero exponentially fast as $\varepsilon \rightarrow 0$ and the exact rate of convergence is given by the so-called rate function. The large deviation principle was first established by Varadhan in [34] and he also studied the small time asymptotics of finite dimensional diffusion processes in [35]. Since then, many important results concerning the large deviation principle have been established. For results on the large deviation principle for stochastic differential equations in finite dimensional case we refer to [15]. For the extensions to infinite dimensional diffusions or SPDE, we refer the readers to [3, 6, 12, 21, 22, 26, 30, 32, 36] and the references therein.

The large deviation principle for the stochastic quasi-geostrophic equation with small multiplicative noise is proved in Section 3 and the small time large deviations for this equation in Section 4 in the subcritical case (i.e. $\alpha > \frac{1}{2}$). The proof of small noise LDP is mainly based on the weak convergence approach from [2]. Compared to some recent works on LDP for SPDE (cf. [6, 21, 26]), the main difficulty here lies in dealing with the nonlinear term in (1.1) since the solution to the stochastic quasi-geostrophic equation is not as regular as in the case of SPDE within the variational framework (see [6, 21, 26] for many examples). For example, for 2D Navier-Stokes equation, the solution lies in the first order Sobolev space by which the nonlinear term can be dominated. Compared with this, the solution of the stochastic quasi-geostrophic equation only lies in H^α (see definition below) and the nonlinear term cannot be handled as for 2D Navier-Stokes equation. Here we use the regularity of solutions of the deterministic equation to control the nonlinear term. Indeed, the solution of the deterministic quasi-geostrophic equation will be in H^δ if the initial value lies in H^δ (see Theorem A.1). Our main result on small noise large deviations for equation (1.1) is formulated in Theorem 3.9. The small time

large deviation principle describes the behavior of the temperature of the fluid when time is very small. The proof is mainly inspired by the approach from [36]. We first establish the large deviation principle on $L^\infty([0, T], H)$ if the initial value is smooth (see Theorem 4.1). However, since the solution to the stochastic quasi-geostrophic equation is very irregular, we cannot approximate the initial value similarly as in [36] for the 2D Navier-Stokes equation to obtain the result for more general initial value. In order to overcome this difficulty, we establish the small time large deviation principle with general initial value on a larger state space (see Theorem 4.2). Here we use the L^p -norm estimate to control the nonlinear term. But these L^p -norm estimates we cannot prove by Galerkin approximation, instead we use another approximation which can be seen as a piecewise linear equation on small subintervals (see (4.11)).

2 Notations and preliminaries

In the following, we will restrict ourselves to flows which have zero average on the torus, i.e.

$$\int_{\mathbb{T}^2} \theta dx = 0.$$

Thus (1.3) can be restated as

$$u = \left(-\frac{\partial\psi}{\partial x_2}, \frac{\partial\psi}{\partial x_1}\right) \quad \text{and} \quad (-\Delta)^{1/2}\psi = -\theta.$$

Set

$$H = \left\{f \in L^2(\mathbb{T}^2) : \int_{\mathbb{T}^2} f dx = 0\right\}$$

and let $|\cdot|$ and $\langle \cdot, \cdot \rangle$ denote the usual norm and inner product in H respectively. On the periodic domain \mathbb{T}^2 , it is well known that

$$\{\sin(kx) | k \in \mathbb{Z}_+^2\} \cup \{\cos(kx) | k \in \mathbb{Z}_-^2\}$$

form an eigenbasis (we denote it by $\{e_k\}$) of $-\Delta$ and the corresponding eigenvalues are $|k|^2$. Here

$$\mathbb{Z}_+^2 = \{(k_1, k_2) \in \mathbb{Z}^2 | k_2 > 0\} \cup \{(k_1, 0) \in \mathbb{Z}^2 | k_1 > 0\}, \quad \mathbb{Z}_-^2 = \{(k_1, k_2) \in \mathbb{Z}^2 | (-k_1, -k_2) \in \mathbb{Z}_+^2\}.$$

Now we define

$$\|f\|_{H^s}^2 = \sum_k |k|^{2s} \langle f, e_k \rangle^2$$

and let H^s denote the (Sobolev) space of all f such that $\|f\|_{H^s}$ is finite.

Set $\Lambda = (-\Delta)^{1/2}$, then we have

$$\|f\|_{H^s} = |\Lambda^s f|.$$

By the singular integral theory of Calderón and Zygmund (cf.[29, Chapter 3]), for any $p \in (1, \infty)$, there exists a constant $C(p)$ such that

$$\|u\|_{L^p} \leq C(p) \|\theta\|_{L^p}. \tag{2.1}$$

For fixed $\alpha \in (0, 1)$, we define the linear operator

$$A_\alpha : D(A_\alpha) = H^{2\alpha}(\mathbb{T}^2) \subset H \rightarrow H, \quad A_\alpha u = \kappa(-\Delta)^\alpha u.$$

It is well known that A_α is positive definite and self-adjoint with the same eigenbasis as that of $-\Delta$ mentioned above. We denote the eigenvalues of A_α by $0 < \lambda_1 \leq \lambda_2 \leq \dots$ and renumber the above eigenbasis correspondingly as e_1, e_2, \dots .

We first recall the following product estimate (cf.[27, Lemma A.4]).

Lemma 2.1 Suppose that $s > 0$ and $p \in (1, \infty)$. If $f, g \in C^\infty(\mathbb{T}^2)$, then

$$\|\Lambda^s(fg)\|_{L^p} \leq C (\|f\|_{L^{p_1}} \|\Lambda^s g\|_{L^{p_2}} + \|g\|_{L^{p_3}} \|\Lambda^s f\|_{L^{p_4}}), \quad (2.2)$$

where $p_i \in (1, \infty), i = 1, \dots, 4$ satisfy that

$$\frac{1}{p} = \frac{1}{p_1} + \frac{1}{p_2} = \frac{1}{p_3} + \frac{1}{p_4}.$$

For the reader's convenience we also recall the following standard Sobolev inequality (cf.[29, Chapter V]):

Lemma 2.2 Suppose that $q > 1, p \in [q, \infty)$ and

$$\frac{1}{p} + \frac{\sigma}{2} = \frac{1}{q}.$$

If $\Lambda^\sigma f \in L^q$, then we have $f \in L^p$ and there is a constant $C \geq 0$ (independent of f) such that

$$\|f\|_{L^p} \leq C \|\Lambda^\sigma f\|_{L^q}.$$

3 Freidlin-Wentzell's large deviations in the subcritical case

In this section, we consider the large deviation principle for the stochastic quasi-geostrophic equation with small multiplicative noise. Here we will use the weak convergence approach introduced by Budhiraja and Dupuis in [2]. Let us first recall some standard definitions and results from large deviation theory (cf.[11]).

Let $\{X^\varepsilon\}$ be a family of random variables defined on a probability space (Ω, \mathcal{F}, P) taking values in some Polish space E .

Definition 3.1 (Rate function) A function $I : E \rightarrow [0, \infty]$ is called a rate function if I is lower semicontinuous. A rate function I is called a good rate function if the level set $\{x \in E : I(x) \leq M\}$ is compact for each $M < \infty$.

Definition 3.2 (I)(Large deviation principle) The sequence $\{X^\varepsilon\}$ is said to satisfy the large deviation principle with rate function I if for each Borel subset A of E

$$-\inf_{x \in A^\circ} I(x) \leq \liminf_{\varepsilon \rightarrow 0} \varepsilon \log P(X^\varepsilon \in A) \leq \limsup_{\varepsilon \rightarrow 0} \varepsilon \log P(X^\varepsilon \in A) \leq -\inf_{x \in \bar{A}} I(x),$$

where A° and \bar{A} denote the interior and closure of A in E respectively.

(II)(Laplace principle) The sequence $\{X^\varepsilon\}$ is said to satisfy the Laplace principle with rate function I if for each bounded continuous real-valued function h defined on E

$$\lim_{\varepsilon \rightarrow 0} \varepsilon \log E\{\exp[-\frac{1}{\varepsilon}h(X^\varepsilon)]\} = - \inf_{x \in E} \{h(x) + I(x)\}.$$

It is well known that the large deviation principle and the Laplace principle are equivalent if E is a Polish space and the rate function is good. The equivalence is essentially a consequence of Varadhan's lemma and Bryc's converse theorem (cf.[11]).

Suppose $W(t)$ is a cylindrical Wiener process on a Hilbert space U (with inner product $\langle \cdot, \cdot \rangle_U$ and norm $|\cdot|_U$) defined on a probability space $(\Omega, \mathcal{F}, \mathcal{F}_t, P)$ (i.e. the paths of W take values in $C([0, T], Y)$, where Y is another Hilbert space such that the embedding $U \subset Y$ is Hilbert-Schmidt). Now we define

$$\begin{aligned} \mathcal{A} &= \left\{ \phi : \phi \text{ is a } U\text{-valued } \{\mathcal{F}_t\}\text{-predictable process s.t. } \int_0^T |\phi(s)|_U^2 ds < \infty \text{ a.s.} \right\}; \\ S_M &= \left\{ v \in L^2([0, T], U) : \int_0^T |v(s)|_U^2 ds \leq M \right\}; \\ \mathcal{A}_M &= \{ \phi \in \mathcal{A} : \phi(\omega) \in S_M, P\text{-a.s.} \}. \end{aligned}$$

Here we remark that we will always refer to the weak topology on the set S_M in this paper.

Suppose $g^\varepsilon : C([0, T], Y) \rightarrow E$ is a measurable map and $X^\varepsilon = g^\varepsilon(W)$. Now we formulate the following sufficient conditions for the Laplace principle (equivalently, large deviation principle) of X^ε as $\varepsilon \rightarrow 0$.

Hypothesis 3.3 There exists a measurable map $g^0 : C([0, T], Y) \rightarrow E$ such that the following conditions hold:

- 1) Let $\{v^\varepsilon : \varepsilon > 0\} \subset \mathcal{A}_M$ for some $M < \infty$. If v^ε converge to v as S_M -valued random elements in distribution, then $g^\varepsilon(W(\cdot) + \frac{1}{\sqrt{\varepsilon}} \int_0^\cdot v^\varepsilon(s) ds)$ converge in distribution to $g^0(\int_0^\cdot v(s) ds)$.
- 2) For every $M < \infty$, the set $K_M = \{g^0(\int_0^\cdot v(s) ds) : v \in S_M\}$ is a compact subset of E .

The following crucial result was proven in [2] (see also [1] for finite dimensional case).

Theorem 3.4 ([2, Theorem 4.4]) If $\{g^\varepsilon\}$ satisfies Hypothesis 3.3, then $\{X^\varepsilon\}$ satisfies the Laplace principle (hence large deviation principle) on E with the good rate function I given by

$$I(f) = \inf_{\{v \in L^2([0, T], U) : f = g^0(\int_0^\cdot v(s) ds)\}} \left\{ \frac{1}{2} \int_0^T |v(s)|_U^2 ds \right\}. \quad (3.1)$$

Now we reformulate (1.1)-(1.3) in the following form of an abstract stochastic evolution equation:

$$\begin{cases} d\theta(t) + A_\alpha \theta(t) dt + u(t) \cdot \nabla \theta(t) dt = G(\theta) dW(t), \\ \theta(0) = \theta_0 \in H, \end{cases} \quad (3.2)$$

where u satisfies (1.3).

We first need to impose some assumptions on G such that (3.2) has a unique solution. Let $L_2(U, H)$ be the space of all Hilbert-Schmidt operators from U to H and $\{f_n\}$ be an ONB of U . Recall that we only consider the subcritical case (i.e. $\alpha > \frac{1}{2}$) in this work. Let $\beta > 3$ be some fixed constant.

Hypothesis 3.5 Suppose that G satisfies the following conditions:

i) There exist some positive real numbers C_1, C_2, C_3 and $\rho_1 < 2\kappa$ such that

$$\|G(\theta)\|_{L_2(U,H)}^2 \leq C_1|\theta|^2 + \rho_1|\Lambda^\alpha\theta|^2 + C_2, \theta \in H^\alpha;$$

$$\|G(\theta)\|_{L_2(U,H^{-\beta})}^2 \leq C_3(|\theta|^2 + 1), \theta \in H^\alpha.$$

ii) If $\theta_n, \theta \in H^\alpha$ and $\theta_n \rightarrow \theta$ in H , then for all $v \in C^\infty(\mathbb{T}^2)$,

$$\lim_{n \rightarrow \infty} |G(\theta_n)^*(v) - G(\theta)^*(v)|_U = 0,$$

where the asterisk denotes the adjoint operator.

iii) For some p with $0 < 1/p < \alpha - \frac{1}{2}$, there exists some constant C such that

$$\int_{\mathbb{T}^2} \left(\sum_j |G(\theta)(f_j)|^2 \right)^{p/2} dx \leq C \left(\int_{\mathbb{T}^2} |\theta|^p dx + 1 \right), \theta \in H^\alpha \cap L^p(\mathbb{T}^2); \quad (3.3)$$

iv) There exist some constants C and $\beta_1 < 2\kappa$ such that

$$\|\Lambda^{-1/2}(G(\theta_1) - G(\theta_2))\|_{L_2(U,H)}^2 \leq C|\Lambda^{-1/2}(\theta_1 - \theta_2)|^2 + \beta_1|\Lambda^{\alpha-\frac{1}{2}}(\theta_1 - \theta_2)|^2, \theta_1, \theta_2 \in H^\alpha. \quad (3.4)$$

Now we give the definition of the (probabilistically) strong solution to (3.2).

Definition 3.6 We say that there exists a (probabilistically) strong solution to (3.2) on $[0, T]$ if for every probability space $(\Omega, \mathcal{F}, \{\mathcal{F}_t\}_{t \in [0, T]}, P)$ with an \mathcal{F}_t -cylindrical Wiener process W , there exists an \mathcal{F}_t -adapted process $\theta : [0, T] \times \Omega \rightarrow H$ such that for P -a.s. $\omega \in \Omega$

$$\theta(\cdot, \omega) \in L^\infty([0, T]; H) \cap L^2([0, T]; H^\alpha) \cap C([0, T]; H^{-\beta})$$

and P -a.s.

$$\langle \theta(t), \varphi \rangle + \int_0^t \langle A_\alpha^{1/2} \theta(s), A_\alpha^{1/2} \varphi \rangle ds - \int_0^t \langle u(s) \cdot \nabla \varphi, \theta(s) \rangle ds = \langle \theta_0, \varphi \rangle + \left\langle \int_0^t G(\theta(s)) dW(s), \varphi \right\rangle$$

for all $t \in [0, T]$ and all $\varphi \in C^1(\mathbb{T}^2)$.

Remark Note that $\operatorname{div} u = 0$, so for regular functions θ and φ we have

$$\langle u(s) \cdot \nabla(\theta(s) + \varphi), \theta(s) + \varphi \rangle = 0.$$

Hence,

$$\langle u(s) \cdot \nabla \theta(s), \varphi \rangle = -\langle u(s) \cdot \nabla \varphi, \theta(s) \rangle.$$

This relation justifies the integral equation in Definition 3.6.

We recall the following existence and uniqueness result from [28].

Theorem 3.7 ([28, Theorem 4.3]) Assume $\alpha > \frac{1}{2}$ and Hypothesis 3.5 hold. Then for each initial condition $\theta_0 \in L^p$, there exists a pathwise unique probabilistically strong solution θ of equation (3.2) on $[0, T]$ with initial condition $\theta(0) = \theta_0$ such that

$$E \sup_{t \in [0, T]} |\Lambda^{-1/2} \theta(t)|^2 < \infty.$$

Moreover, the solution θ satisfies

$$E \sup_{t \in [0, T]} \|\theta(t)\|_{L^p}^p + E \int_0^T |\Lambda^\alpha \theta(t)|^2 dt < \infty.$$

Now we consider the stochastic quasi-geostrophic equation driven by small multiplicative noise:

$$d\theta^\varepsilon(t) + A_\alpha \theta^\varepsilon(t) dt + u^\varepsilon(t) \cdot \nabla \theta^\varepsilon(t) dt = \sqrt{\varepsilon} G(\theta^\varepsilon) dW(t) \quad (3.5)$$

with $\theta^\varepsilon(0) = \theta_0 \in L^p$. Here u^ε satisfies (1.3) with θ replaced by θ^ε . By Theorem 3.7, under Hypothesis 3.5, there exists a pathwise unique strong solution of (3.5) in $L^\infty([0, T], H) \cap L^2([0, T], H^\alpha) \cap C([0, T], H^{-\beta})$. Therefore, there exist Borel-measurable functions

$$g^\varepsilon : C([0, T], Y) \rightarrow L^\infty([0, T], H) \cap L^2([0, T], H^\alpha) \cap C([0, T], H^{-\beta})$$

such that $\theta^\varepsilon(\cdot) = g^\varepsilon(W(\cdot))$.

Now the aim is to prove the large deviation principle for θ^ε . For this purpose we need to impose some further assumptions on G .

Hypothesis 3.8 Assume G satisfies the following conditions:

i) $G(\theta)$ is a bounded operator from U to H^δ for some $\delta > 2 - 2\alpha$ such that

$$\|G(\theta)\|_{L(U, H^\delta)} \leq C(\|\theta\|_{H^{\delta+\alpha}} + 1), \quad \theta \in H^{\delta+\alpha} \quad (3.6)$$

and for $r := (2 - 2\alpha) \vee \alpha$

$$\|G(\theta)\|_{L(U, H^r)} \leq C(\|\theta\|_{H^{\delta+\alpha}} + 1), \quad \theta \in H^{\delta+\alpha}. \quad (3.7)$$

ii)

$$\|G(\theta_1) - G(\theta_2)\|_{L(U, H)} \leq C\|\theta_1 - \theta_2\|_{H^\alpha}, \quad \theta_1, \theta_2 \in H^\alpha.$$

Remark (i) (3.6) can also be replaced by the following condition:

$$\|G(\theta)\|_{L(U, H^{\delta-\alpha})} \leq C(\|\theta\|_{H^\delta} + 1).$$

(ii) Typical examples for G satisfying Hypothesis 3.5 and 3.8 have the following form: for $\theta \in H^\alpha$

$$G(\theta)y = \sum_{k=1}^{\infty} b_k \langle y, f_k \rangle_U g(\theta), \quad y \in U,$$

where $g \in C_b^1(\mathbb{R})$ and b_k are C^∞ functions on \mathbb{T}^2 satisfying

$$\sum_{k=1}^{\infty} b_k^2(\xi) \leq C, \quad \sum_{k=1}^{\infty} |\Lambda^{\delta \vee r} b_k|^2 \leq C.$$

For $v \in L^2([0, T], U)$, we consider the following skeleton equation

$$\frac{d\theta_v(t)}{dt} = -A_\alpha \theta_v(t) - u_v(t) \cdot \nabla \theta_v(t) + G(\theta_v)v(t) \quad (3.8)$$

with $\theta_v(0) = \theta_0 \in H^\delta \cap L^p$. Here u_v satisfies (1.3) with θ replaced by θ_v . Then by Hypothesis 3.5 and 3.8 we have

$$\|G(\theta)v\|_{L^p} \leq C|v|_U(\|\theta\|_{L^p} + 1); \quad (\text{A.1})$$

$$\|G(\theta)v\|_{H^\delta} \leq C|v|_U(\|\theta\|_{H^{\delta+\alpha}} + 1); \quad (\text{A.2})$$

$$|\Lambda^{-1/2}(G(\theta_1) - G(\theta_2))v| \leq |v|_U(C|\Lambda^{-1/2}(\theta_1 - \theta_2)| + \sqrt{\beta_1}|\Lambda^{\alpha-\frac{1}{2}}(\theta_1 - \theta_2)|). \quad (\text{A.3})$$

By a similar argument as in [27, Theorems 3.5 and 3.7], we know that (3.8) has a unique solution $\theta_v \in L^\infty([0, T], H^\delta \cap L^p) \cap L^2([0, T], H^{\delta+\alpha}) \cap C([0, T], H^{-\beta})$. For the completeness we include the proof of this result in the Appendix.

Remark Here we want to emphasize that although by Theorem A.1 in Appendix if $\theta_0 \in H^\delta \cap L^p$, then we have $\theta_v \in L^\infty([0, T], H^\delta \cap L^p) \cap L^2([0, T], H^{\delta+\alpha}) \cap C([0, T], H^{-\beta})$. However, this might be not true for θ^ε . This is the reason why we establish the large deviation principle for θ^ε on $L^\infty([0, T], H) \cap L^2([0, T], H^\alpha) \cap C([0, T], H^{-\beta})$ (which is the state space of θ^ε) instead of $L^\infty([0, T], H^\delta) \cap L^2([0, T], H^{\delta+\alpha}) \cap C([0, T], H^{-\beta})$.

Define $g^0 : C([0, T], Y) \rightarrow L^\infty([0, T], H) \cap L^2([0, T], H^\alpha) \cap C([0, T], H^{-\beta})$ by

$$g^0(h) = \begin{cases} \theta_v, & \text{if } h = \int_0^\cdot v(s)ds \text{ for some } v \in L^2([0, T], U), \\ 0, & \text{otherwise.} \end{cases}$$

Now we formulate the main result concerning the large deviation principle for θ^ε .

Theorem 3.9 Suppose that Hypothesis 3.5 and Hypothesis 3.8 hold. Then for any $\theta_0 \in H^\delta \cap L^p$ with p in Hypothesis 3.5 iii), $\{\theta^\varepsilon\}$ satisfies the Laplace principle (hence large deviation principle) on $L^\infty([0, T], H) \cap L^2([0, T], H^\alpha) \cap C([0, T], H^{-\beta})$ with a good rate function given by (3.1).

Proof To prove the theorem it suffices to verify the two conditions in Hypothesis 3.3 so that Theorem 3.4 is applicable to obtain the large deviation principle for θ^ε .

[Step 1] First we show that the set $K_M = \{g^0(\int_0^\cdot v(s)ds) : v \in S_M\}$ is a compact subset of $L^\infty([0, T], H) \cap L^2([0, T], H^\alpha) \cap C([0, T], H^{-\beta})$.

Let $\{\theta_n\}$ be a sequence in K_M where θ_n corresponds to the solution of (3.8) with $v_n \in S_M$ in place of v . By the weak compactness of S_M in $L^2([0, T], U)$, there exists a subsequence (which we still denote it by $\{v_n\}$) converging to a limit v weakly in $L^2([0, T], U)$.

Let $w_n = \theta_n - \theta_v$, it suffices to show that $w_n \rightarrow 0$ (in fact, a subsequence is enough) in $L^\infty([0, T], H) \cap L^2([0, T], H^\alpha) \cap C([0, T], H^{-\beta})$ as $n \rightarrow \infty$.

Note that $u_n \cdot \nabla w_n \in H^{-\alpha}$, where u_n satisfies (1.3) with θ replaced by θ_n . In fact, we have uniform L^p norm bound for θ_n, w_n by Theorem A.1. And we also have

$$\begin{aligned} |{}_{H^{-\alpha}}\langle u_n \cdot \nabla w_n, \varphi \rangle_{H^\alpha}| &= |{}_{H^{-\alpha}}\langle \nabla \cdot (u_n w_n), \varphi \rangle_{H^\alpha}| \leq |\Lambda^\alpha \varphi| |\Lambda^{1-\alpha}(u_n \cdot w_n)| \\ &\leq |\Lambda^\alpha \varphi| (|\Lambda^{1-\alpha+\sigma} w_n| \|\theta_n\|_{L^p} + |\Lambda^{1-\alpha+\sigma} \theta_n| \|w_n\|_{L^p}), \end{aligned}$$

where $\sigma = \frac{2}{p} < 2\alpha - 1$ and we use $\operatorname{div} u_n = 0$ in the first equality and Lemmas 2.1, 2.2 and (2.1) in the last inequality. Thus by $\operatorname{div} u_n = 0$, we obtain

$${}_{H^{-\alpha}}\langle u_n \cdot \nabla w_n, w_n \rangle_{H^\alpha} = 0. \quad (\text{3.9})$$

If $\delta < 1$ we get

$$\begin{aligned}
|\langle (u_n - u_v) \cdot \nabla \theta_v, w_n \rangle| &= |\langle \nabla \cdot ((u_n - u_v)\theta_v), w_n \rangle| \leq |\Lambda^\alpha w_n| |\Lambda^{1-\alpha}((u_n - u_v) \cdot \theta_v)| \\
&\leq C |\Lambda^\alpha w_n| (|\Lambda^{2-\alpha-\delta} w_n| |\Lambda^\delta \theta_v| + |\Lambda^{1-\alpha+\delta-(1-\alpha)} \theta_v| |\Lambda^{2-\alpha-\delta} w_n|) \\
&\leq C |\Lambda^\alpha w_n| |\Lambda^\alpha w_n|^\gamma |w_n|^{1-\gamma} |\Lambda^\delta \theta_v| \\
&\leq \frac{\kappa}{4} |\Lambda^\alpha w_n|^2 + C |\Lambda^\delta \theta_v|^N |w_n|^2,
\end{aligned} \tag{3.10}$$

where $\gamma = \frac{2-\alpha-\delta}{\alpha}$, $N = \frac{2\alpha}{2\alpha-2+\delta}$ and we use $\operatorname{div}(u_n - u_v) = 0$ in the first equality, Lemmas 2.1, 2.2 and (2.1) in the second inequality, the interpolation inequality and $\delta > 2 - 2\alpha$ in the third inequality and Young's inequality in the last inequality.

Similarly, if $\delta \geq 1$ we get

$$\begin{aligned}
|\langle (u_n - u_v) \cdot \nabla \theta_v, w_n \rangle| &= |\langle \nabla \cdot ((u_n - u_v)\theta_v), w_n \rangle| \leq |\Lambda^\alpha w_n| |\Lambda^{1-\alpha}((u_n - u_v) \cdot \theta_v)| \\
&\leq C |\Lambda^\alpha w_n| |\Lambda^{1-\alpha+\sigma_1} w_n| |\Lambda^\delta \theta_v| \\
&\leq C |\Lambda^\alpha w_n| |\Lambda^\alpha w_n|^{\gamma_1} |w_n|^{1-\gamma_1} |\Lambda^\delta \theta_v| \\
&\leq \frac{\kappa}{4} |\Lambda^\alpha w_n|^2 + C |\Lambda^\delta \theta_v|^{N_1} |w_n|^2,
\end{aligned}$$

where $0 < \sigma_1 < 2\alpha - 1$, $\gamma_1 = \frac{1-\alpha+\sigma_1}{\alpha}$, $N_1 = \frac{2\alpha}{2\alpha-1-\sigma_1}$ and we use $\operatorname{div}(u_n - u_v) = 0$ in the first equality, Lemmas 2.1, 2.2 and (2.1), $\delta \geq 1$ in the second inequality, the interpolation inequality in the third inequality and Young's inequality in the last inequality.

In the following we only prove the result for $\delta < 1$ and the argument for $\delta \geq 1$ is similar.

By (3.8) we have

$$\begin{aligned}
&|w_n(t)|^2 + 2\kappa \int_0^t |\Lambda^\alpha w_n|^2 ds \\
&= 2 \int_0^t (-\langle u_n \cdot \nabla \theta_n, w_n \rangle + \langle u_v \cdot \nabla \theta_v, w_n \rangle) ds \\
&\quad + 2 \int_0^t \langle G(\theta_n)v_n - G(\theta_v)v, w_n \rangle ds \\
&= -2 \int_0^t \langle (u_n - u_v) \cdot \nabla \theta_v, w_n \rangle ds \\
&\quad + 2 \int_0^t \langle (G(\theta_n) - G(\theta_v))v_n, w_n \rangle ds \\
&\quad + 2 \int_0^t \langle G(\theta_v)(v_n - v), w_n \rangle ds \\
&\leq \int_0^t \left[\kappa |\Lambda^\alpha w_n|^2 + C(|\Lambda^\delta \theta_v|^N + |v_n|_U^2) |w_n|^2 \right. \\
&\quad \left. + 2 \langle G(\theta_v)(v_n - v), w_n \rangle \right] ds,
\end{aligned}$$

where in the second equality we use (3.9) and in the last inequality we use (3.10), Hypothesis 3.8 ii) and Young's inequality.

Let

$$h_n(t) := \int_0^t G(\theta_v)(v_n - v) ds,$$

then we have

$$\begin{aligned} \sup_{t \in [0, T]} \|P_k h_n(t) - h_n(t)\|_{H^r} &\leq \int_0^T \|(P_k - I)G(\theta_v)\|_{L(U, H^r)} \|v_n - v\|_U dt \\ &\leq (2M)^{1/2} \left(\int_0^T \|(P_k - I)G(\theta_v)\|_{L(U, H^r)}^2 dt \right)^{1/2} \rightarrow 0 \text{ as } k \rightarrow \infty. \end{aligned}$$

Here P_k is the orthogonal projection in H onto the space spanned by e_1, \dots, e_k and we use (3.7) and $\theta_v \in L^2([0, T]; H^{\delta+\alpha})$ which follows from Theorem A.1 in the last step.

Since $P_k H^r \subset H^r$ is compact and $v_n \rightharpoonup v$ weakly in $L^2([0, T]; U)$, by (3.7) it is easy to show that $P_k h_n \rightarrow 0$ in $C([0, T], H^r)$ as $n \rightarrow \infty$ (see e.g. [21, Lemma 3.2]) using the Arzèla-Ascoli theorem (since for any subsequence the limit is the same, this convergence holds for the whole sequence). Hence we obtain that $h_n \rightarrow 0$ in $C([0, T], H^r)$ as $n \rightarrow \infty$.

And we also have

$$\begin{aligned} &\int_0^t \langle G(\theta_v)(v_n(s) - v(s)), w_n(s) \rangle ds \\ &= \langle w_n(t), h_n(t) \rangle - \int_0^t \langle w'_n(s), h_n(s) \rangle ds \\ &= \langle w_n(t), h_n(t) \rangle + \int_0^t \langle A_\alpha w_n + u_n \cdot \nabla \theta_n - u_v \cdot \nabla \theta_v, h_n \rangle ds \\ &\quad - \int_0^t \langle G(\theta_n)v_n - G(\theta_v)v, h_n \rangle ds \\ &=: I_1 + I_2 + I_3. \end{aligned} \tag{3.11}$$

Note that

$$I_1 \leq \varepsilon |w_n(t)|^2 + C |h_n(t)|^2;$$

and by Hypothesis 3.5 i) and (A.4)

$$\begin{aligned} I_3 &\leq \sup_{s \in [0, T]} |h_n(s)| \int_0^T (\|G(\theta_n)\|_{L_2(U, H)} |v_n|_U + \|G(\theta_v)\|_{L_2(U, H)} |v|_U) ds \\ &\leq C \sup_{s \in [0, T]} \|h_n(s)\|_{H^r} \left(\int_0^T (|\Lambda^\alpha \theta_v|^2 + |\Lambda^\alpha \theta_n|^2 + C) ds \right)^{1/2} \leq C \sup_{s \in [0, T]} \|h_n(s)\|_{H^r}. \end{aligned}$$

For $\varphi \in H^{2-2\alpha}$, we obtain

$$\begin{aligned} |\langle u_n \cdot \nabla \theta_n - u_v \cdot \nabla \theta_v, \varphi \rangle| &= |\langle \nabla \cdot (u_n \theta_n - u_v \theta_v), \varphi \rangle| \\ &\leq C |\Lambda^{2\alpha-1}(u_n \theta_n - u_v \theta_v)| |\Lambda^{2-2\alpha} \varphi| \\ &\leq C (|\Lambda^\alpha \theta_n|^2 + |\Lambda^\alpha \theta_v|^2) |\Lambda^{2-2\alpha} \varphi|, \end{aligned}$$

where we use $\operatorname{div} u_n = 0$ and $\operatorname{div} u_v = 0$ in the first equality and Lemmas 2.1, 2.2 and (2.1) in the last inequality.

Hence

$$\|u_n \cdot \nabla \theta_n - u_v \cdot \nabla \theta_v\|_{H^{-(2-2\alpha)}} \leq C (|\Lambda^\alpha \theta_n|^2 + |\Lambda^\alpha \theta_v|^2).$$

Therefore,

$$\begin{aligned} I_2 &\leq \int_0^t (\|A_\alpha w_n(s)\|_{H^{-\alpha}} + \|u_n \cdot \nabla \theta_n - u_v \cdot \nabla \theta_v\|_{H^{-(2-2\alpha)}}) \|h_n(s)\|_{H^r} ds \\ &\leq C \sup_{s \in [0, T]} \|h_n(s)\|_{H^r} \int_0^t (\|w_n\|_{H^\alpha} + \|\theta_n\|_{H^\alpha}^2 + \|\theta_v\|_{H^\alpha}^2) ds \\ &\leq C \sup_{s \in [0, T]} \|h_n(s)\|_{H^r}, \end{aligned}$$

where in the last step we use (A.4).

Then the Gronwall lemma and (3.11) yield that

$$\sup_{t \in [0, T]} |w_n(t)|^2 + \frac{\kappa}{2} \int_0^T |\Lambda^\alpha w_n|^2 ds \leq C \sup_{t \in [0, T]} \|h_n(t)\|_{H^r} \left(\exp \left\{ C \int_0^T (|\Lambda^\delta \theta_v|^N + |v_n|_U^2) ds \right\} + 1 \right).$$

Then by (A.4) we have

$$\sup_{t \in [0, T]} |w_n(t)|^2 + \frac{\kappa}{2} \int_0^T |\Lambda^\alpha w_n|^2 ds \rightarrow 0, \quad n \rightarrow \infty.$$

[Step 2] Suppose that $\{v_\varepsilon : \varepsilon > 0\} \subset \mathcal{A}_M$ for some $M < \infty$ and v_ε converge to v as S_M -valued random elements in distribution. Then, by Girsanov's theorem, $\bar{\theta}_{v_\varepsilon} = g^\varepsilon(W(\cdot) + \frac{1}{\sqrt{\varepsilon}} \int_0^\cdot v^\varepsilon(s) ds)$ solves the following equation

$$d\bar{\theta}_{v_\varepsilon}(t) + A_\alpha \bar{\theta}_{v_\varepsilon}(t) dt + u_{\bar{\theta}_{v_\varepsilon}}(t) \cdot \nabla \bar{\theta}_{v_\varepsilon}(t) dt = G(\bar{\theta}_{v_\varepsilon}) v_\varepsilon(t) dt + \sqrt{\varepsilon} G(\bar{\theta}_{v_\varepsilon}) dW(t). \quad (3.12)$$

Here $u_{\bar{\theta}_{v_\varepsilon}}$ satisfies (1.3) with θ replaced by $\bar{\theta}_{v_\varepsilon}$.

Since S_M is a Polish space, by the Skorohod theorem, we can construct processes $(\tilde{v}_\varepsilon, \tilde{v}, \tilde{W}_\varepsilon)$ such that the joint distribution of $(\tilde{v}_\varepsilon, \tilde{W}_\varepsilon)$ is the same as that of (v_ε, W) , the distribution of v coincides with that of \tilde{v} and $\tilde{v}_\varepsilon \rightarrow \tilde{v}$ a.s. as S_M -valued random elements.

Setting $w_\varepsilon := \bar{\theta}_{\tilde{v}_\varepsilon} - \theta_{\tilde{v}}$, it suffices to prove that $w_\varepsilon \rightarrow 0$ in probability in $L^\infty([0, T], H) \cap L^2([0, T], H^\alpha) \cap C([0, T], H^{-\beta})$. For w_ε and $u_{\bar{\theta}_{\tilde{v}_\varepsilon}}$ we also have similar estimates as (3.9) and (3.10). In the following we write $v_\varepsilon = \tilde{v}_\varepsilon, W = \tilde{W}_\varepsilon$ for simplicity.

Itô's formula and (3.9) imply that

$$\begin{aligned}
& |w_\varepsilon(t)|^2 + 2\kappa \int_0^t |\Lambda^\alpha w_\varepsilon|^2 ds \\
&= 2 \int_0^t (-\langle u_{\bar{\theta}_{v_\varepsilon}} \cdot \nabla \bar{\theta}_{v_\varepsilon}, w_\varepsilon \rangle + \langle u_v \cdot \nabla \theta_v, w_\varepsilon \rangle) ds \\
&\quad + 2 \int_0^t \langle G(\bar{\theta}_{v_\varepsilon}(s))v_\varepsilon(s) - G(\theta_v(s))v(s), w_\varepsilon(s) \rangle ds \\
&\quad + 2\sqrt{\varepsilon} \int_0^t \langle w_\varepsilon, G(\bar{\theta}_{v_\varepsilon})dW \rangle + \varepsilon \int_0^t \|G(\bar{\theta}_{v_\varepsilon})\|_{L_2(U,H)}^2 ds \\
&= -2 \int_0^t \langle (u_{\bar{\theta}_{v_\varepsilon}} - u_v) \cdot \nabla \theta_v, w_\varepsilon \rangle ds + 2 \int_0^t \langle (G(\bar{\theta}_{v_\varepsilon}(s)) - G(\theta_v(s)))v_\varepsilon(s), w_\varepsilon(s) \rangle ds \\
&\quad + 2 \int_0^t \langle G(\theta_v(s))(v_\varepsilon(s) - v(s)), w_\varepsilon(s) \rangle ds \\
&\quad + 2\sqrt{\varepsilon} \int_0^t \langle w_\varepsilon, G(\bar{\theta}_{v_\varepsilon})dW \rangle + \varepsilon \int_0^t \|G(\bar{\theta}_{v_\varepsilon})\|_{L_2(U,H)}^2 ds \\
&\leq \int_0^t [\kappa |\Lambda^\alpha w_\varepsilon|^2 + C(|\Lambda^\delta \theta_v|^N + |v_\varepsilon|_U^2)|w_\varepsilon|^2] ds \\
&\quad + 2 \int_0^t \langle G(\theta_v(s))(v_\varepsilon(s) - v(s)), w_\varepsilon(s) \rangle ds \\
&\quad + 2\sqrt{\varepsilon} \int_0^t \langle w_\varepsilon, G(\bar{\theta}_{v_\varepsilon})dW \rangle + \varepsilon \int_0^t \|G(\bar{\theta}_{v_\varepsilon})\|_{L_2(U,H)}^2 ds,
\end{aligned} \tag{3.13}$$

where in the last inequality we use (3.10), Hypothesis 3.8 ii) and Young's inequality.

Similarly we define

$$h_\varepsilon(t) = \int_0^t G(\theta_v(s))(v_\varepsilon(s) - v(s))ds.$$

Then by the same argument as [Step 1] we know $h_\varepsilon(t) \rightarrow 0$ in $C([0, T], H^r)$ a.s. as $\varepsilon \rightarrow 0$.

By Itô's formula and a similar argument as in (3.11) we have

$$\begin{aligned}
& \int_0^t \langle G(\theta_v(s))(v_\varepsilon(s) - v(s)), w_\varepsilon(s) \rangle ds \\
& \leq \varepsilon |w_\varepsilon(t)|^2 + C \left(1 + \int_0^t |\Lambda^\alpha \bar{\theta}_{v_\varepsilon}|^2 ds \right) \sup_{s \in [0, T]} \|h_\varepsilon(s)\|_{H^r} - \sqrt{\varepsilon} \int_0^t \langle h_\varepsilon, G(\bar{\theta}_{v_\varepsilon})dW \rangle.
\end{aligned}$$

Define

$$\tau_{L,\varepsilon} := T \wedge \inf \left\{ t : |\bar{\theta}_{v_\varepsilon}(t)|^2 + \int_0^t |\Lambda^\alpha \bar{\theta}_{v_\varepsilon}(s)|^2 ds > L \right\}.$$

Since $\bar{\theta}_{v_\varepsilon}$ is weakly continuous in H , $\tau_{L,\varepsilon}$ is a stopping time with respect to $\mathcal{F}_{t+} = \bigcap_{s>t} \mathcal{F}_s$ and

$|\bar{\theta}_{v_\varepsilon}(t \wedge \tau_{L,\varepsilon})| \leq L$. By the Burkholder-Davis-Gundy inequality one has

$$\begin{aligned} & \sqrt{\varepsilon} E \sup_{t \in [0, \tau_{L,\varepsilon}]} \left| \int_0^t \langle w_\varepsilon - h_\varepsilon, G(\bar{\theta}_{v_\varepsilon}) dW \rangle \right| \\ & \leq C \sqrt{\varepsilon} E \left(\int_0^{\tau_{L,\varepsilon}} |\bar{\theta}_{v_\varepsilon} - \theta_v - h_\varepsilon|^2 \|G(\bar{\theta}_{v_\varepsilon})\|_{L_2(U,H)}^2 ds \right)^{1/2} \\ & \leq C \sqrt{\varepsilon} E \left(\int_0^{\tau_{L,\varepsilon}} (|\Lambda^\alpha \bar{\theta}_{v_\varepsilon}|^2 + 1) ds \right)^{1/2} \leq C \sqrt{\varepsilon}. \end{aligned}$$

Combining the above estimates with (3.13) and applying Gronwall's lemma we have

$$\begin{aligned} & \sup_{s \in [0, t]} |w_\varepsilon(s)|^2 + \frac{\kappa}{2} \int_0^t |\Lambda^\alpha w_\varepsilon|^2 ds \\ & \leq \left[C(1 + \int_0^t |\Lambda^\alpha \bar{\theta}_{v_\varepsilon}|^2 ds) \sup_{s \in [0, T]} \|h_\varepsilon(s)\|_{H^r} + 2\sqrt{\varepsilon} \sup_{t \in [0, T]} \left| \int_0^t \langle w_\varepsilon - h_\varepsilon, G(\bar{\theta}_{v_\varepsilon}) dW \rangle \right| \right. \\ & \quad \left. + \varepsilon \int_0^t \|G(\bar{\theta}_{v_\varepsilon})\|_{L_2(U,H)}^2 ds \right] \exp \left\{ C \int_0^T (|\Lambda^\delta \theta_v|^N + |v_\varepsilon|_U^2) dr \right\}. \end{aligned}$$

Then we have

$$\sup_{t \in [0, \tau_{L,\varepsilon}]} |w_\varepsilon(t)|^2 + \frac{\kappa}{2} \int_0^{\tau_{L,\varepsilon}} |\Lambda^\alpha w_\varepsilon|^2 ds \rightarrow 0$$

in probability as $\varepsilon \rightarrow 0$.

By Itô's formula and standard argument (cf. [28, Theorem 3.3]) we have

$$\sup_{\varepsilon \in [0, 1]} E \left[\sup_{t \in [0, T]} |\bar{\theta}_{v_\varepsilon}(t)|^2 + \int_0^T |\Lambda^\alpha \bar{\theta}_{v_\varepsilon}(t)|^2 dt \right] < \infty.$$

Let L be fixed. Then for a suitable constant C

$$\sup_{\varepsilon \in [0, 1]} P(\tau_{L,\varepsilon} = T) \geq 1 - \frac{C}{L}.$$

Therefore, we have

$$\sup_{t \in [0, T]} |w_\varepsilon(t)|^2 + \frac{\kappa}{2} \int_0^T |\Lambda^\alpha w_\varepsilon|^2 ds \rightarrow 0$$

in probability as $\varepsilon \rightarrow 0$.

Now the proof of Theorem 3.9 is complete. \square

4 The small time large deviations in the subcritical case

In this section, we establish some small time large deviations results for the stochastic quasi-geostrophic equation. The proof is mainly inspired by the approach used in [36]. We consider the stochastic quasi-geostrophic equation (3.2) again and assume that G satisfies Hypothesis

3.5. Then by Theorem 3.7, for $\theta_0 \in L^p$ there exists a pathwise unique strong solution of (3.2) in $L^\infty([0, T], H) \cap L^2([0, T], H^\alpha) \cap C([0, T], H^{-\beta})$ for $\beta > 3$.

We assume the following additional conditions on G :

S.1) There exists a constant L such that for some $\delta > 0$

$$\|G(\theta)\|_{L_2(U, H^\delta)}^2 \leq L(1 + \|\theta\|_{H^\delta}^2), \quad \theta \in H^\delta;$$

S.2) There exists a constant L_1 such that

$$\|G(\theta_1) - G(\theta_2)\|_{L_2(U, H)}^2 \leq L_1|\theta_1 - \theta_2|^2, \quad \theta_1, \theta_2 \in H.$$

Let $\varepsilon > 0$. By the scaling property of the Wiener process, it is easy to see that $\theta(\varepsilon t)$ coincides in law with the solution of the following equation

$$d\theta^\varepsilon(t) + \varepsilon A_\alpha \theta^\varepsilon(t) dt + \varepsilon u^\varepsilon(t) \cdot \nabla \theta^\varepsilon(t) dt = \sqrt{\varepsilon} G(\theta^\varepsilon) dW(t) \quad (4.1)$$

with $\theta^\varepsilon(0) = \theta_0$. Here u^ε satisfies (1.3) with θ replaced by θ^ε .

Let μ^ε be the law of θ^ε on $L^\infty([0, T], H)$. Now we formulate the small time large deviation principle for (4.1) on $L^\infty([0, T], H)$ for regular initial value θ_0 .

Theorem 4.1 Suppose that S.1) for some $\delta \geq \alpha$ and $\delta > 2 - 2\alpha$, S.2) and Hypothesis 3.5 hold. Then for $\theta_0 \in H^\delta \cap L^p$ with p in Hypothesis 3.5 iii), μ^ε satisfies the large deviation principle on $L^\infty([0, T], H)$ with rate function I given by

$$I(f) = \inf_{\{v \in L^2([0, T], U) : f(t) = \theta_0 + \int_0^t G(f(s))v(s) ds\}} \left\{ \frac{1}{2} \int_0^T |v(s)|_U^2 ds \right\}. \quad (4.2)$$

Proof Let v^ε be the solution of the stochastic equation

$$v^\varepsilon(t) = \theta_0 + \sqrt{\varepsilon} \int_0^t G(v^\varepsilon(s)) dW(s) \quad (4.3)$$

and ν^ε be the law of v^ε on $L^\infty([0, T], H)$. Then by [21] we know that ν^ε satisfies the large deviation principle with rate function I given by (4.2). Now it is sufficient to show that the two families of probability measures μ^ε and ν^ε are exponentially equivalent, i.e. for any $\eta > 0$,

$$\lim_{\varepsilon \rightarrow 0} \varepsilon \log P\left(\sup_{0 \leq t \leq T} |\theta^\varepsilon(t) - v^\varepsilon(t)|^2 > \eta\right) = -\infty. \quad (4.4)$$

Then the conclusion in Theorem 4.1 follows directly from [11, Theorem 4.2.13].

In the following we assume that $\delta < 1$ and for $\delta \geq 1$ the proof is similar.

For $M > 0$, we define the following stopping times:

$$\tau_{\varepsilon, M} = \inf\{t \geq 0 : \|v^\varepsilon(t)\|_{H^\delta}^2 > M\}.$$

Then we have

$$\begin{aligned} & P\left(\sup_{0 \leq t \leq T} |\theta^\varepsilon(t) - v^\varepsilon(t)|^2 > \eta, \sup_{0 \leq t \leq T} \|v^\varepsilon(t)\|_{H^\delta}^2 \leq M\right) \\ & \leq P\left(\sup_{0 \leq t \leq T \wedge \tau_{\varepsilon, M}} |\theta^\varepsilon(t) - v^\varepsilon(t)|^2 > \eta\right). \end{aligned} \quad (4.5)$$

Applying Itô's formula to $|v^\varepsilon(t \wedge \tau_{\varepsilon, M}) - \theta^\varepsilon(t \wedge \tau_{\varepsilon, M})|^2$ we get

$$\begin{aligned}
& |v^\varepsilon(t \wedge \tau_{\varepsilon, M}) - \theta^\varepsilon(t \wedge \tau_{\varepsilon, M})|^2 + 2\varepsilon\kappa \int_0^{t \wedge \tau_{\varepsilon, M}} |\Lambda^\alpha(v^\varepsilon(s) - \theta^\varepsilon(s))|^2 ds \\
&= 2\varepsilon \int_0^{t \wedge \tau_{\varepsilon, M}} \langle A_\alpha v^\varepsilon(s), (v^\varepsilon(s) - \theta^\varepsilon(s)) \rangle ds + 2\varepsilon \int_0^{t \wedge \tau_{\varepsilon, M}} \langle u^\varepsilon \cdot \nabla \theta^\varepsilon, (v^\varepsilon - \theta^\varepsilon) \rangle ds \\
&+ 2\sqrt{\varepsilon} \int_0^{t \wedge \tau_{\varepsilon, M}} \langle v^\varepsilon - \theta^\varepsilon, (G(v^\varepsilon) - G(\theta^\varepsilon)) dW \rangle \\
&+ \varepsilon \int_0^{t \wedge \tau_{\varepsilon, M}} \|G(v^\varepsilon) - G(\theta^\varepsilon)\|_{L_2(U, H)}^2 ds.
\end{aligned}$$

Note that by similar arguments as in (3.9) and (3.10), we have

$$\begin{aligned}
& |\langle u^\varepsilon \cdot \nabla \theta^\varepsilon, v^\varepsilon - \theta^\varepsilon \rangle| \\
&= |\langle u^\varepsilon \cdot \nabla(\theta^\varepsilon - v^\varepsilon), \theta^\varepsilon - v^\varepsilon \rangle| \\
&+ \langle (u^\varepsilon - u_{v^\varepsilon}) \cdot \nabla v^\varepsilon, \theta^\varepsilon - v^\varepsilon \rangle \\
&+ \langle u_{v^\varepsilon} \cdot \nabla v^\varepsilon, \theta^\varepsilon - v^\varepsilon \rangle| \\
&\leq \frac{\kappa}{2} |\Lambda^\alpha(\theta^\varepsilon - v^\varepsilon)|^2 + C|\Lambda^\delta v^\varepsilon|^N |\theta^\varepsilon - v^\varepsilon|^2 + C|\Lambda^\delta v^\varepsilon|^4,
\end{aligned}$$

where u_{v^ε} satisfies (1.3) with θ replaced by v_ε . Here for the last term we use the following estimate:

$$\begin{aligned}
|\langle u_{v^\varepsilon} \cdot \nabla v^\varepsilon, \theta^\varepsilon - v^\varepsilon \rangle| &= |\langle \nabla \cdot (u_{v^\varepsilon} v^\varepsilon), \theta^\varepsilon - v^\varepsilon \rangle| \\
&\leq C|\Lambda^\alpha(\theta^\varepsilon - v^\varepsilon)| |\Lambda^{1-\alpha}(u_{v^\varepsilon} v^\varepsilon)| \\
&\leq |\Lambda^\alpha(\theta_n^\varepsilon - v_n^\varepsilon)| |\Lambda^\delta v^\varepsilon|^2,
\end{aligned}$$

where in the first equality we use $\operatorname{div} u_{v^\varepsilon} = 0$ and in the last inequality we use Lemmas 2.1, 2.2 and $\delta \geq (2 - 2\alpha) \vee \alpha$.

Therefore, by S.2) and Young's inequality we get

$$\begin{aligned}
& |v^\varepsilon(t \wedge \tau_{\varepsilon, M}) - \theta^\varepsilon(t \wedge \tau_{\varepsilon, M})|^2 + 2\varepsilon\kappa \int_0^{t \wedge \tau_{\varepsilon, M}} |\Lambda^\alpha(v^\varepsilon(s) - \theta^\varepsilon(s))|^2 ds \\
&\leq 2\varepsilon \int_0^{t \wedge \tau_{\varepsilon, M}} \left(\frac{\kappa}{2} |\Lambda^\alpha(v^\varepsilon - \theta^\varepsilon)|^2 + C|\Lambda^\alpha v^\varepsilon|^2 \right) ds \\
&+ 2\varepsilon \int_0^{t \wedge \tau_{\varepsilon, M}} \left(\frac{\kappa}{2} |\Lambda^\alpha(\theta^\varepsilon - v^\varepsilon)|^2 + C|\Lambda^\delta v^\varepsilon|^N |\theta^\varepsilon - v^\varepsilon|^2 + C|\Lambda^\delta v^\varepsilon|^4 \right) ds \\
&+ 2\sqrt{\varepsilon} \int_0^{t \wedge \tau_{\varepsilon, M}} \langle v^\varepsilon - \theta^\varepsilon, (G(v^\varepsilon) - G(\theta^\varepsilon)) dW \rangle \\
&+ \varepsilon C \int_0^{t \wedge \tau_{\varepsilon, M}} |v^\varepsilon - \theta^\varepsilon|^2 ds.
\end{aligned}$$

Then by Gronwall's lemma and $\delta \geq \alpha$ we have

$$\begin{aligned}
& |v^\varepsilon(t \wedge \tau_{\varepsilon, M}) - \theta^\varepsilon(t \wedge \tau_{\varepsilon, M})|^2 \leq \left[2\varepsilon \int_0^{t \wedge \tau_{\varepsilon, M}} (C|\Lambda^\delta v^\varepsilon|^2 + C|\Lambda^\delta v^\varepsilon|^4) ds \right. \\
&+ \left. 2\sqrt{\varepsilon} \int_0^{t \wedge \tau_{\varepsilon, M}} \langle v^\varepsilon - \theta^\varepsilon, (G(v^\varepsilon) - G(\theta^\varepsilon)) dW \rangle \right] e^{\varepsilon C \int_0^{t \wedge \tau_{\varepsilon, M}} |\Lambda^\delta v^\varepsilon|^N ds + Ct\varepsilon}.
\end{aligned}$$

To estimate the stochastic integral term, we will use the following result from [4, 10], namely that there exists a universal constant c such that for any $q \geq 2$ and for any continuous martingale M_t with $M_0 = 0$, one has

$$\|M_t^*\|_{L^q} \leq cq^{1/2} \| \langle M \rangle_t^{1/2} \|_{L^q}, \quad (4.6)$$

where $M_t^* = \sup_{0 \leq s \leq t} |M_s|$.

By this result and S.2) we have

$$\begin{aligned} & (E) \sup_{0 \leq s \leq t \wedge \tau_{\varepsilon, M}} |v^\varepsilon(s) - \theta^\varepsilon(s)|^{2/q} \\ & \leq C e^{\varepsilon CM^{N/2}t + Ct\varepsilon} \left[(\varepsilon Mt + \varepsilon M^2 t)^2 + q\varepsilon \left(E \int_0^{t \wedge \tau_{\varepsilon, M}} |v^\varepsilon(r) - \theta^\varepsilon(r)|^4 dr \right)^{q/2} \right]^{2/q} \\ & \leq C e^{\varepsilon CM^{N/2}t + Ct\varepsilon} \left[(\varepsilon Mt + \varepsilon M^2 t)^2 + q\varepsilon \int_0^t (E) \sup_{0 \leq r \leq s \wedge \tau_{\varepsilon, M}} |v^\varepsilon(r) - \theta^\varepsilon(r)|^{2/q} ds \right]. \end{aligned}$$

Then Gronwall's lemma yields that

$$\begin{aligned} & (E) \sup_{0 \leq s \leq T \wedge \tau_{\varepsilon, M}} |v^\varepsilon(s) - \theta^\varepsilon(s)|^{2/q} \\ & \leq C e^{\varepsilon CM^{N/2}T + CT\varepsilon} (\varepsilon MT + \varepsilon M^2 T)^2 \exp \left[CqT\varepsilon e^{\varepsilon CM^{N/2}T + CT\varepsilon} \right]. \end{aligned}$$

Fixing M and taking $q = 2/\varepsilon$ we obtain

$$\begin{aligned} & \varepsilon \log P \left(\sup_{0 \leq t \leq T \wedge \tau_{\varepsilon, M}} |\theta^\varepsilon(t) - v^\varepsilon(t)|^2 > \eta \right) \\ & \leq \varepsilon \log \frac{E[\sup_{0 \leq s \leq T \wedge \tau_{\varepsilon, M}} |v^\varepsilon(s) - \theta^\varepsilon(s)|^{2q}]}{\eta^q} \\ & \leq \log C (\varepsilon MT + \varepsilon M^2 T)^2 - 2 \log \eta + CT e^{\varepsilon CM^{N/2}T + CT\varepsilon} + \varepsilon CM^{N/2}T + CT\varepsilon \\ & \rightarrow -\infty, \text{ as } \varepsilon \rightarrow 0. \end{aligned}$$

Therefore, by (4.5) there exists ε_0 such that for every ε satisfying $0 < \varepsilon \leq \varepsilon_0$,

$$P \left(\sup_{0 \leq t \leq T} |\theta^\varepsilon(t) - v^\varepsilon(t)|^2 > \eta, \sup_{0 \leq t \leq T} \|v^\varepsilon(t)\|_{H^\delta}^2 \leq M \right) \leq e^{-R/\varepsilon}. \quad (4.7)$$

By the same argument as in [36, Lemma 3.2] and S.1) we have

$$\lim_{M \rightarrow \infty} \sup_{0 < \varepsilon \leq 1} \varepsilon \log P \left(\sup_{0 \leq t \leq T} \|v^\varepsilon(t)\|_{H^\delta}^2 > M \right) = -\infty. \quad (4.8)$$

Then for any $R > 0$, there exists a constant M such that for every $\varepsilon \in (0, 1]$ the following inequality holds:

$$P \left(\sup_{0 \leq t \leq T} \|v^\varepsilon(t)\|_{H^\delta}^2 > M \right) \leq e^{-R/\varepsilon}. \quad (4.9)$$

By (4.7) and (4.9), we know that there exists ε_0 such that for every ε satisfying $0 < \varepsilon \leq \varepsilon_0$ we have

$$P \left(\sup_{0 \leq t \leq T} |\theta^\varepsilon(t) - v^\varepsilon(t)|^2 > \eta \right) \leq 2e^{-R/\varepsilon}.$$

Since R is arbitrary, we obtain (4.4).

Hence the proof of Theorem 4.1 is complete. \square

Note that the solution of (4.1) is not as regular as in the case of the 2D stochastic Navier-Stokes equation. In Theorem 4.1 we use the regularity of v^ε to control the nonlinear term, but we can not approximate the initial value in (4.1) to obtain the large deviation principle on $L^\infty([0, T], H)$ for general initial value in L^p as Xu and Zhang did in [36] for the 2D stochastic Navier-Stokes equation since the nonlinear term can not be dominated. To overcome this difficulty, now we enlarge the state space of the solution and use L^p norm estimate to control the nonlinear term. Then we establish the large deviation principle on $L^\infty([0, T], H^{-1/2})$.

We consider the following condition on G .

S.3) There exists a constant L_2 such that

$$\|\Lambda^{-1/2}(G(\theta_1) - G(\theta_2))\|_{L_2(U, H)}^2 \leq L_2 |\Lambda^{-1/2}(\theta_1 - \theta_2)|^2, \theta_1, \theta_2 \in H^\alpha.$$

Remark Typical examples for G satisfying Hypothesis 3.5 and S.1)-S.3) have the following form: for $\theta \in H^\alpha$

$$G(\theta)y = \sum_{k=1}^{\infty} b_k \langle y, f_k \rangle_U \theta, y \in U,$$

where b_k are C^∞ functions on \mathbb{T}^2 satisfying $\sum_{k=1}^{\infty} b_k^2(\xi) \leq M$ and $\sum_{k=1}^{\infty} |\Lambda^{1+\varepsilon} b_k|^2 \leq M$ for some $\varepsilon > 0$.

Let $\bar{\mu}^\varepsilon$ be the law of θ^ε on $L^\infty([0, T], H^{-1/2})$. Now we formulate our main result about the small time large deviation principle for (4.1).

Theorem 4.2 Suppose that S.1) for $\delta \geq (\frac{3}{4} - \frac{\alpha}{2}) \vee (\alpha - \frac{1}{2})$ and Hypothesis 3.5, S.3) hold. Then for $\theta_0 \in L^p$, $\bar{\mu}^\varepsilon$ satisfies the large deviation principle on $L^\infty([0, T], H^{-1/2})$ with rate function I given by (4.2).

It is sufficient to show that the two families of probability measures $\bar{\mu}^\varepsilon$ and ν^ε (for simplicity we still use the same notation) are exponentially equivalent, i.e. for any $\eta > 0$,

$$\lim_{\varepsilon \rightarrow 0} \varepsilon \log P\left(\sup_{0 \leq t \leq T} |\Lambda^{-1/2}(\theta^\varepsilon(t) - v^\varepsilon(t))|^2 > \eta\right) = -\infty. \quad (4.10)$$

Then the conclusion in Theorem 4.2 follows directly from [11, Theorem 4.2.13].

In order to show (4.10) we prove a few lemmas in below.

Lemma 4.3

$$\lim_{M \rightarrow \infty} \sup_{0 < \varepsilon \leq 1} \varepsilon \log P\left(\sup_{0 \leq t \leq T} \|\theta^\varepsilon(t)\|_{L^p}^p > M\right) = -\infty.$$

Proof We consider the same approximation $\theta^{\varepsilon, n}$ to θ^ε as in [28, Theorem 3.3]. We pick a smooth function $\phi \geq 0$ such that $\text{supp } \phi \subset [1, 2]$ and $\int_0^\infty \phi = 1$. Then for $\sigma > 0$ we define

$$U_\sigma[\theta](t) := \int_0^\infty \phi(\tau) (k_\sigma * R^\perp \theta)(t - \sigma\tau) d\tau,$$

where k_σ is the periodic Poisson kernel in \mathbb{T}^2 given by $\widehat{k}_\sigma(\zeta) = e^{-\sigma|\zeta|}$, $\zeta \in \mathbb{Z}^2$, and we set $\theta(t) = 0$ for $t < 0$.

We take a sequence δ_n converging to 0 and consider the following equation:

$$d\theta^{\varepsilon,n}(t) + \varepsilon A_\alpha \theta^{\varepsilon,n}(t) dt + \varepsilon u^{\varepsilon,n}(t) \cdot \nabla \theta^{\varepsilon,n}(t) dt = \sqrt{\varepsilon} k_{\delta_n} * G(\theta^{\varepsilon,n}) dW(t) \quad (4.11)$$

with initial data $\theta^{\varepsilon,n}(0) = k_{\delta_n} * \theta_0$ and $u^{\varepsilon,n} = U_{\delta_n}[\theta^{\varepsilon,n}]$. For a fixed n , this is a linear equation in $\theta^{\varepsilon,n}$ on each subinterval $[t_k^n, t_{k+1}^n]$ with $t_k^n = k\delta_n$, since $u^{\varepsilon,n}$ is determined by the values of $\theta^{\varepsilon,n}$ on the two previous subintervals.

Then by [28, Theorem 3.3, Step 2], there exists a weak solution to (4.11) which converges in distribution to θ^ε in $L^2([0, T], H) \cap C([0, T], H^{-\beta})$.

By [20, Lemma 5.1] we have (here we write $\theta(t) = \theta^{\varepsilon,n}(t)$, $u(t) = u^{\varepsilon,n}(t)$ to simplify the notation)

$$\begin{aligned} \|\theta(t)\|_{L^p}^p &= \|k_{\delta_n} * \theta_0\|_{L^p}^p + \varepsilon \int_0^t \left[-p \int_{\mathbb{T}^2} |\theta(s)|^{p-2} \theta(s) (\Lambda^{2\alpha} \theta(s) + u(s) \cdot \nabla \theta(s)) dx \right. \\ &\quad \left. + \frac{1}{2} p(p-1) \varepsilon \int_{\mathbb{T}^2} |\theta(s)|^{p-2} \left(\sum_j |k_{\delta_n} * G(\theta(s))(f_j)|^2 \right) dx \right] ds \\ &\quad + p\sqrt{\varepsilon} \int_0^t \int_{\mathbb{T}^2} |\theta(s)|^{p-2} \theta(s) k_{\delta_n} * G(\theta(s)) dx dW(s) \\ &\leq \|\theta_0\|_{L^p}^p + \frac{1}{2} p(p-1) \varepsilon \int_0^t \int_{\mathbb{T}^2} |\theta(s)|^{p-2} \left(\sum_j |k_{\delta_n} * G(\theta(s))(f_j)|^2 \right) dx ds \\ &\quad + p\sqrt{\varepsilon} \int_0^t \int_{\mathbb{T}^2} |\theta(s)|^{p-2} \theta(s) k_{\delta_n} * G(\theta(s)) dx dW(s) \\ &\leq \|\theta_0\|_{L^p}^p + \varepsilon \int_0^t \left(\int_{\mathbb{T}^2} |\theta(s)|^p dx + C \int \left(\sum_j |k_{\delta_n} * G(\theta(s))(f_j)|^2 \right)^{p/2} dx \right) ds \\ &\quad + p\sqrt{\varepsilon} \int_0^t \int_{\mathbb{T}^2} |\theta(s)|^{p-2} \theta(s) k_{\delta_n} * G(\theta(s)) dx dW(s), \end{aligned}$$

where in the first inequality we used $\operatorname{div} u = 0$ and $\int |\theta|^{p-2} \theta \Lambda^{2\alpha} \theta \geq 0$ (cf. [27, Lemma 3.2]) as well as Young's inequality in the second inequality.

Then by Hypothesis 3.5 (iii) we have

$$\begin{aligned} \sup_{t \in [0, T]} \|\theta(t)\|_{L^p}^p &\leq \|\theta_0\|_{L^p}^p + \varepsilon CT + C\varepsilon \int_0^T \sup_{t \in [0, s]} \|\theta(t)\|_{L^p}^p ds \\ &\quad + p\sqrt{\varepsilon} \sup_{0 \leq t \leq T} \left| \int_0^t \int_{\mathbb{T}^2} |\theta(s)|^{p-2} \theta(s) k_{\delta_n} * G(\theta(s)) dx dW(s) \right|. \end{aligned}$$

Therefore, for $q \geq 2$ we obtain

$$\begin{aligned} (E(\sup_{t \in [0, T]} \|\theta(t)\|_{L^p}^{pq}))^{1/q} &\leq \|\theta_0\|_{L^p}^p + \varepsilon CT + C\varepsilon (E(\int_0^T \sup_{t \in [0, s]} \|\theta(t)\|_{L^p}^p ds)^q)^{1/q} \\ &\quad + p\sqrt{\varepsilon} (E \sup_{0 \leq t \leq T} \left| \int_0^t \int_{\mathbb{T}^2} |\theta(s)|^{p-2} \theta(s) k_{\delta_n} * G(\theta(s)) dx dW(s) \right|^q)^{1/q}. \end{aligned}$$

Using (4.6) and Minkowski's inequality we have

$$\begin{aligned}
& p\sqrt{\varepsilon}(E \sup_{0 \leq t \leq T} |\int_0^t \int_{\mathbb{T}^2} |\theta(s)|^{p-2} \theta(s) k_{\delta_n} * G(\theta(s)) dx dW(s)|^q)^{1/q} \\
& \leq pc\sqrt{q\varepsilon}(E(\int_0^T (\int_{\mathbb{T}^2} |\theta(s)|^{p-1} (\sum_j |k_{\delta_n} * G(\theta(s))(f_j)|^2)^{1/2} dx)^2 ds)^{q/2})^{1/q} \\
& \leq pc\sqrt{q\varepsilon}(E(\sup_{s \in [0, T]} \|\theta(s)\|_{L^p}^{p-1} (\int_0^T (\int_{\mathbb{T}^2} (\sum_j |k_{\delta_n} * G(\theta(s))(f_j)|^2)^{p/2} dx)^{2/p} ds)^{1/2})^q)^{1/q} \\
& \leq pc\sqrt{q\varepsilon}(E(\sup_{s \in [0, T]} \|\theta(s)\|_{L^p}^{p-1} (\int_0^T (\int_{\mathbb{T}^2} (\sum_j |k_{\delta_n} * G(\theta(s))(f_j)|^2)^{p/2} dx) ds)^{1/p})^q)^{1/q} \\
& \leq \frac{1}{2}(E \sup_{s \in [0, T]} \|\theta(s)\|_{L^p}^{pq})^{1/q} + c(p)(q\varepsilon)^{p/2}(E(\int_0^T (\int_{\mathbb{T}^2} (\sum_j |k_{\delta_n} * G(\theta(s))(f_j)|^2)^{p/2} dx) ds)^q)^{1/q} \\
& \leq \frac{1}{2}(E \sup_{s \in [0, T]} \|\theta(s)\|_{L^p}^{pq})^{1/q} + c(p)(q\varepsilon)^{p/2} \left[\int_0^T (1 + (E\|\theta(s)\|_{L^p}^{pq})^{1/q}) ds \right],
\end{aligned}$$

where in the last inequality we use Hypothesis 3.5 iii) and Jensen's inequality.

Hence,

$$\begin{aligned}
(E(\sup_{t \in [0, T]} \|\theta(t)\|_{L^p}^{pq}))^{1/q} & \leq 2\|\theta_0\|_{L^p}^p + \varepsilon CT + C\varepsilon \int_0^T (E \sup_{t \in [0, s]} \|\theta(t)\|_{L^p}^{pq})^{1/q} ds \\
& \quad + c(p)(q\varepsilon)^{p/2} \left[\int_0^T (1 + (E\|\theta(s)\|_{L^p}^{pq})^{1/q}) ds \right].
\end{aligned}$$

Applying Gronwall's lemma we obtain that

$$(E(\sup_{t \in [0, T]} \|\theta(t)\|_{L^p}^{pq}))^{1/q} \leq [2\|\theta_0\|_{L^p}^p + \varepsilon CT + c(p)(q\varepsilon)^{p/2} T] \exp [CT\varepsilon + c(p)T(q\varepsilon)^{p/2}].$$

Letting $n \rightarrow \infty$ we get

$$(E(\sup_{t \in [0, T]} \|\theta^\varepsilon(t)\|_{L^p}^{pq}))^{1/q} \leq [2\|\theta_0\|_{L^p}^p + \varepsilon CT + c(p)(q\varepsilon)^{p/2} T] \exp [CT\varepsilon + c(p)T(q\varepsilon)^{p/2}].$$

Since

$$P(\sup_{0 \leq t \leq T} \|\theta^\varepsilon(t)\|_{L^p}^p > M) \leq M^{-q} E(\sup_{t \in [0, T]} \|\theta^\varepsilon(t)\|_{L^p}^{pq}),$$

letting $q = 2/\varepsilon$ we get

$$\begin{aligned}
\varepsilon \log P(\sup_{0 \leq t \leq T} \|\theta^\varepsilon(t)\|_{L^p}^p > M) & \leq -2 \log M + 2 \log(E(\sup_{t \in [0, T]} \|\theta^\varepsilon(t)\|_{L^p}^{pq}))^{1/q} \\
& \leq -2 \log M + 2 \log(2\|\theta_0\|_{L^p}^p + \varepsilon CT + CT) + 2CT\varepsilon + 2CT.
\end{aligned}$$

Hence the proof is complete. \square

Since $H^\delta \cap L^p$ is dense in L^p , there exists a sequence $\{\theta_0^n\} \subset H^\delta \cap L^p$ such that

$$\lim_{n \rightarrow \infty} \|\theta_0^n - \theta_0\|_{L^p} = 0.$$

Let θ_n^ε be the solution of (4.1) with initial value θ_0^n . From the proof of Lemma 4.3, it follows that

$$\lim_{M \rightarrow \infty} \sup_n \sup_{0 < \varepsilon \leq 1} \varepsilon \log P(\sup_{0 \leq t \leq T} \|\theta_n^\varepsilon(t)\|_{L^p}^p > M) = -\infty. \quad (4.12)$$

Let v_n^ε be the solution of (4.3) with initial value θ_0^n . By the same argument as in (4.8) and Lemma 4.3 we have the following result.

Lemma 4.4 For every $n \in \mathbb{Z}^+$,

$$\lim_{M \rightarrow \infty} \sup_{0 < \varepsilon \leq 1} \varepsilon \log P(\sup_{0 \leq t \leq T} (\|v_n^\varepsilon(t)\|_{H^\delta}^2 + \|v_n^\varepsilon(t)\|_{L^p}^p) > M) = -\infty.$$

Lemma 4.5 For every $\eta > 0$,

$$\lim_{n \rightarrow \infty} \sup_{0 < \varepsilon \leq 1} \varepsilon \log P(\sup_{0 \leq t \leq T} \|\theta_n^\varepsilon(t) - \theta^\varepsilon(t)\|_{H^{-1/2}}^2 > \eta) = -\infty.$$

Proof For $M > 0$, we define the following stopping time for $N_0 = \frac{\alpha}{\alpha - \frac{1}{2} - \frac{1}{p}}$:

$$\bar{\tau}_{\varepsilon, M} = \inf\{t \geq 0 : \int_0^t \|\theta^\varepsilon(s)\|_{L^p}^{N_0} ds > M\}.$$

Clearly, we have

$$\begin{aligned} & P(\sup_{0 \leq t \leq T} \|\theta_n^\varepsilon(t) - \theta^\varepsilon(t)\|_{H^{-1/2}}^2 > \eta, \int_0^T \|\theta^\varepsilon(s)\|_{L^p}^{N_0} ds \leq M) \\ & \leq P(\sup_{0 \leq t \leq T \wedge \bar{\tau}_{\varepsilon, M}} \|\theta_n^\varepsilon(t) - \theta^\varepsilon(t)\|_{H^{-1/2}}^2 > \eta). \end{aligned} \quad (4.13)$$

Let k be a positive constant and $N_0 = \frac{\alpha}{\alpha - \frac{1}{2} - \frac{1}{p}}$. Then applying Ito's formula to

$$e^{-k\varepsilon \int_0^{t \wedge \bar{\tau}_{\varepsilon, M}} \|\theta^\varepsilon(s)\|_{L^p}^{N_0} ds} |\Lambda^{-1/2}(\theta^\varepsilon(t \wedge \bar{\tau}_{\varepsilon, M}) - \theta_n^\varepsilon(t \wedge \bar{\tau}_{\varepsilon, M}))|^2$$

we get

$$\begin{aligned} & e^{-k\varepsilon \int_0^{t \wedge \bar{\tau}_{\varepsilon, M}} \|\theta^\varepsilon(s)\|_{L^p}^{N_0} ds} |\Lambda^{-1/2}(\theta^\varepsilon(t \wedge \bar{\tau}_{\varepsilon, M}) - \theta_n^\varepsilon(t \wedge \bar{\tau}_{\varepsilon, M}))|^2 \\ & + 2\varepsilon k \int_0^{t \wedge \bar{\tau}_{\varepsilon, M}} e^{-k\varepsilon \int_0^s \|\theta^\varepsilon(r)\|_{L^p}^{N_0} dr} |\Lambda^{\alpha - \frac{1}{2}}(\theta^\varepsilon(s) - \theta_n^\varepsilon(s))|^2 ds \\ = & |\Lambda^{-\frac{1}{2}}(\theta_0 - \theta_0^n)|^2 - k\varepsilon \int_0^{t \wedge \bar{\tau}_{\varepsilon, M}} e^{-k\varepsilon \int_0^s \|\theta^\varepsilon(r)\|_{L^p}^{N_0} dr} \|\theta^\varepsilon(s)\|_{L^p}^{N_0} |\Lambda^{-\frac{1}{2}}(\theta^\varepsilon(s) - \theta_n^\varepsilon(s))|^2 ds \\ & - 2\varepsilon \int_0^{t \wedge \bar{\tau}_{\varepsilon, M}} e^{-k\varepsilon \int_0^s \|\theta^\varepsilon(r)\|_{L^p}^{N_0} dr} \langle u^\varepsilon(s) \cdot \nabla \theta^\varepsilon(s) - u_n^\varepsilon(s) \cdot \nabla \theta_n^\varepsilon(s), \Lambda^{-1}(\theta^\varepsilon(s) - \theta_n^\varepsilon(s)) \rangle ds \\ & + 2\sqrt{\varepsilon} \int_0^{t \wedge \bar{\tau}_{\varepsilon, M}} e^{-k\varepsilon \int_0^s \|\theta^\varepsilon(r)\|_{L^p}^{N_0} dr} \langle \Lambda^{-1/2}(\theta^\varepsilon(s) - \theta_n^\varepsilon(s)), \Lambda^{-1/2}(G(\theta^\varepsilon(s)) - G(\theta_n^\varepsilon(s))) \rangle dW(s) \\ & + \varepsilon \int_0^{t \wedge \bar{\tau}_{\varepsilon, M}} e^{-k\varepsilon \int_0^s \|\theta^\varepsilon(r)\|_{L^p}^{N_0} dr} \|\Lambda^{-1/2}(G(\theta^\varepsilon(s)) - G(\theta_n^\varepsilon(s)))\|_{L_2(U, H)}^2 ds, \end{aligned}$$

where u_n^ε satisfies (1.3) with θ replaced by θ_n^ε .

Note that

$$\begin{aligned} & \langle u^\varepsilon \cdot \nabla \theta^\varepsilon - u_n^\varepsilon \cdot \nabla \theta_n^\varepsilon, \Lambda^{-1}(\theta^\varepsilon - \theta_n^\varepsilon) \rangle \\ &= \langle (u_n^\varepsilon - u^\varepsilon) \cdot \nabla \theta_n^\varepsilon, \Lambda^{-1}(\theta_n^\varepsilon - \theta^\varepsilon) \rangle + \langle u^\varepsilon \cdot \nabla(\theta_n^\varepsilon - \theta^\varepsilon), \Lambda^{-1}(\theta_n^\varepsilon - \theta^\varepsilon) \rangle. \end{aligned}$$

Moreover, we also have (cf.e.g. [27])

$$\langle (u_n^\varepsilon - u^\varepsilon) \cdot \nabla \theta_n^\varepsilon, \Lambda^{-1}(\theta_n^\varepsilon - \theta^\varepsilon) \rangle = 0 \quad (4.14)$$

and

$$\begin{aligned} & |\langle u^\varepsilon \cdot \nabla(\theta_n^\varepsilon - \theta^\varepsilon), \Lambda^{-1}(\theta_n^\varepsilon - \theta^\varepsilon) \rangle| = |\langle u^\varepsilon \cdot \nabla \Lambda^{-1}(\theta_n^\varepsilon - \theta^\varepsilon), \theta_n^\varepsilon - \theta^\varepsilon \rangle| \\ & \leq \|u^\varepsilon\|_{L^p} \|\theta_n^\varepsilon - \theta^\varepsilon\|_{L^{p'}} \|\nabla \Lambda^{-1}(\theta_n^\varepsilon - \theta^\varepsilon)\|_{L^{p'}} \\ & \leq C \|u^\varepsilon\|_{L^p} \|\theta_n^\varepsilon - \theta^\varepsilon\|_{H^{1/p}} \|\nabla \Lambda^{-1}(\theta_n^\varepsilon - \theta^\varepsilon)\|_{H^{1/p}} \\ & \leq C \|u^\varepsilon\|_{L^p} \|\Lambda^{-1}(\theta_n^\varepsilon - \theta^\varepsilon)\|_{H^{1+\frac{1}{p}}}^2 \\ & \leq C \|\theta^\varepsilon\|_{L^p} \|\Lambda^{-1}(\theta_n^\varepsilon - \theta^\varepsilon)\|_{H^{1/2}}^{2/N} \|\Lambda^{-1}(\theta_n^\varepsilon - \theta^\varepsilon)\|_{H^{\frac{1}{2}+\alpha}}^{2(1-\frac{1}{N})} \\ & \leq \kappa |\Lambda^{\alpha-\frac{1}{2}}(\theta_n^\varepsilon - \theta^\varepsilon)|^2 + C_0 \|\theta^\varepsilon\|_{L^p}^{N_0} |\Lambda^{-1/2}(\theta_n^\varepsilon - \theta^\varepsilon)|^2, \end{aligned} \quad (4.15)$$

where $\frac{1}{p} + \frac{2}{p'} = 1$ and we used that $\operatorname{div} u^\varepsilon = 0$ in the first equality and $H^{1/p} \hookrightarrow L^{p'}$ in the second inequality, the interpolation inequality in the forth inequality and Young's inequality in the last inequality.

Therefore, by (4.14), (4.15) and S.3)

$$\begin{aligned} & e^{-k\varepsilon} \int_0^{t \wedge \bar{\tau}_{\varepsilon, M}} \|\theta^\varepsilon(s)\|_{L^p}^{N_0} ds |\Lambda^{-1/2}(\theta^\varepsilon(t \wedge \bar{\tau}_{\varepsilon, M}) - \theta_n^\varepsilon(t \wedge \bar{\tau}_{\varepsilon, M}))|^2 \\ & + 2\varepsilon \kappa \int_0^{t \wedge \bar{\tau}_{\varepsilon, M}} e^{-k\varepsilon} \int_0^s \|\theta^\varepsilon(r)\|_{L^p}^{N_0} dr |\Lambda^{\alpha-\frac{1}{2}}(\theta^\varepsilon(s) - \theta_n^\varepsilon(s))|^2 ds \\ & \leq |\Lambda^{-\frac{1}{2}}(\theta_0 - \theta_0^n)|^2 - k\varepsilon \int_0^{t \wedge \bar{\tau}_{\varepsilon, M}} e^{-k\varepsilon} \int_0^s \|\theta^\varepsilon\|_{L^p}^{N_0} dr \|\theta^\varepsilon(s)\|_{L^p}^{N_0} |\Lambda^{-\frac{1}{2}}(\theta^\varepsilon(s) - \theta_n^\varepsilon(s))|^2 ds \\ & + 2\varepsilon \int_0^{t \wedge \bar{\tau}_{\varepsilon, M}} e^{-k\varepsilon} \int_0^s \|\theta^\varepsilon\|_{L^p}^{N_0} dr (\kappa |\Lambda^{\alpha-\frac{1}{2}}(\theta_n^\varepsilon(s) - \theta^\varepsilon(s))|^2 + C_0 \|\theta^\varepsilon(s)\|_{L^p}^{N_0} |\Lambda^{-1/2}(\theta_n^\varepsilon(s) - \theta^\varepsilon(s))|^2) ds \\ & + 2\sqrt{\varepsilon} \int_0^{t \wedge \bar{\tau}_{\varepsilon, M}} e^{-k\varepsilon} \int_0^s \|\theta^\varepsilon\|_{L^p}^{N_0} dr \langle \Lambda^{-1/2}(\theta^\varepsilon(s) - \theta_n^\varepsilon(s)), \Lambda^{-1/2}(G(\theta^\varepsilon(s)) - G(\theta_n^\varepsilon(s))) \rangle dW(s) \\ & + C\varepsilon \int_0^{t \wedge \bar{\tau}_{\varepsilon, M}} e^{-k\varepsilon} \int_0^s \|\theta^\varepsilon\|_{L^p}^{N_0} dr |\Lambda^{-1/2}(\theta^\varepsilon(s) - \theta_n^\varepsilon(s))|^2 ds. \end{aligned}$$

Choosing $k > 2C_0$ and using (4.6) we have

$$\begin{aligned} & (E[\sup_{0 \leq s \leq t \wedge \bar{\tau}_{\varepsilon, M}} e^{-k\varepsilon} \int_0^s \|\theta^\varepsilon(r)\|_{L^p}^{N_0} dr |\Lambda^{-1/2}(\theta^\varepsilon(s) - \theta_n^\varepsilon(s))|^2]^q)^{2/q} \\ & \leq 2 |\Lambda^{-\frac{1}{2}}(\theta_0 - \theta_0^n)|^4 \\ & + C(q\varepsilon + t\varepsilon^2) \int_0^t (E[\sup_{0 \leq r \leq s \wedge \bar{\tau}_{\varepsilon, M}} e^{-k\varepsilon} \int_0^s \|\theta^\varepsilon(r)\|_{L^p}^{N_0} dr |\Lambda^{-1/2}(\theta^\varepsilon(s) - \theta_n^\varepsilon(s))|^2]^q)^{2/q} ds. \end{aligned}$$

Applying Gronwall's lemma we obtain

$$(E[\sup_{0 \leq s \leq T \wedge \bar{\tau}_{\varepsilon, M}} e^{-k\varepsilon \int_0^s \|\theta^\varepsilon(r)\|_{L^p}^{N_0} dr} |\Lambda^{-1/2}(\theta^\varepsilon(s) - \theta_n^\varepsilon(s))|^2]^q)^{2/q} \leq 2|\Lambda^{-\frac{1}{2}}(\theta_0 - \theta_0^n)|^4 e^{CT(q\varepsilon + \varepsilon^2 T)}.$$

Hence we have

$$(E[\sup_{0 \leq s \leq T \wedge \bar{\tau}_{\varepsilon, M}} |\Lambda^{-1/2}(\theta^\varepsilon(s) - \theta_n^\varepsilon(s))|^2]^q)^{2/q} \leq 2e^{2kM} |\Lambda^{-\frac{1}{2}}(\theta_0 - \theta_0^n)|^4 e^{CT(q\varepsilon + \varepsilon^2 T)}.$$

Fixing M and taking $q = 2/\varepsilon$ we get

$$\begin{aligned} & \sup_{0 < \varepsilon \leq 1} \varepsilon \log P(\sup_{0 \leq t \leq T \wedge \bar{\tau}_{\varepsilon, M}} \|\theta_n^\varepsilon(t) - \theta^\varepsilon(t)\|_{H^{-1/2}}^2 > \eta) \\ & \leq \sup_{0 < \varepsilon \leq 1} \varepsilon \log \frac{E[\sup_{0 \leq s \leq T \wedge \bar{\tau}_{\varepsilon, M}} |\Lambda^{-1/2}(\theta^\varepsilon(s) - \theta_n^\varepsilon(s))|^{2q}]}{\eta^q} \\ & \leq 2kM + \log 2|\Lambda^{-\frac{1}{2}}(\theta_0 - \theta_0^n)|^4 - 2 \log \eta + C \\ & \rightarrow -\infty, \text{ as } n \rightarrow \infty. \end{aligned} \tag{4.16}$$

By Lemma 4.3, for any $R > 0$ there exists a constant M such that for every $\varepsilon \in (0, 1]$ the following inequality holds:

$$P(\int_0^T \|\theta^\varepsilon(t)\|_{L^p}^{N_0} dt > M) \leq P(\sup_{0 \leq t \leq T} \|\theta^\varepsilon(t)\|_{L^p}^p > (\frac{M}{T})^{p/N_0}) \leq e^{-R/\varepsilon}. \tag{4.17}$$

For such M , according to (4.13) and (4.16), there exists a constant N_2 such that for every $n \geq N_2$,

$$\sup_{0 < \varepsilon \leq 1} \varepsilon \log P(\sup_{0 \leq t \leq T} \|\theta_n^\varepsilon(t) - \theta^\varepsilon(t)\|_{H^{-1/2}} > \eta, \int_0^T \|\theta^\varepsilon(t)\|_{L^p}^{N_0} dt \leq M) \leq -R. \tag{4.18}$$

Combining (4.17) and (4.18) we conclude that there exists a positive integer N_2 such that for every $n \geq N_2$ and $\varepsilon \in (0, 1]$

$$P(\sup_{0 \leq t \leq T} \|\theta_n^\varepsilon(t) - \theta^\varepsilon(t)\|_{H^{-1/2}}^2 > \eta) \leq 2e^{-R/\varepsilon}.$$

Since R is arbitrary, the assertion of the lemma follows. \square

The next lemma can be proved similarly as Lemma 4.5.

Lemma 4.6 For every $\eta > 0$,

$$\lim_{n \rightarrow \infty} \sup_{0 < \varepsilon \leq 1} \varepsilon \log P(\sup_{0 \leq t \leq T} \|v_n^\varepsilon(t) - v^\varepsilon(t)\|_{H^{-1/2}}^2 > \eta) = -\infty.$$

Lemma 4.7 For every $\eta > 0$ and every positive integer n ,

$$\lim_{\varepsilon \rightarrow 0} \varepsilon \log P(\sup_{0 \leq t \leq T} \|\theta_n^\varepsilon(t) - v_n^\varepsilon(t)\|_{H^{-1/2}}^2 > \eta) = -\infty.$$

Proof For $M > 0$, we define the following stopping times:

$$\tau_{\varepsilon, M}^n = \inf\{t : \|v_n^\varepsilon(t)\|_{H^\delta}^2 + \|v_n^\varepsilon(t)\|_{L^p}^p > M\}.$$

Then we have

$$\begin{aligned} & P\left(\sup_{0 \leq t \leq T} \|\theta_n^\varepsilon(t) - v_n^\varepsilon(t)\|_{H^{-1/2}}^2 > \eta, \sup_{0 \leq t \leq T} (\|v_n^\varepsilon(t)\|_{H^\delta}^2 + \|v_n^\varepsilon(t)\|_{L^p}^p) \leq M\right) \\ & \leq P\left(\sup_{0 \leq t \leq T \wedge \tau_{\varepsilon, M}^n} \|\theta_n^\varepsilon(t) - v_n^\varepsilon(t)\|_{H^{-1/2}}^2 > \eta\right). \end{aligned} \quad (4.19)$$

Applying Itô's formula to $|\Lambda^{-1/2}(v_n^\varepsilon(t \wedge \tau_{\varepsilon, M}^n) - \theta_n^\varepsilon(t \wedge \tau_{\varepsilon, M}^n))|^2$ we get

$$\begin{aligned} & |\Lambda^{-1/2}(v_n^\varepsilon(t \wedge \tau_{\varepsilon, M}^n) - \theta_n^\varepsilon(t \wedge \tau_{\varepsilon, M}^n))|^2 + 2\varepsilon\kappa \int_0^{t \wedge \tau_{\varepsilon, M}^n} |\Lambda^{\alpha-1/2}(v_n^\varepsilon(s) - \theta_n^\varepsilon(s))|^2 ds \\ & = 2\varepsilon \int_0^{t \wedge \tau_{\varepsilon, M}^n} \langle A_\alpha v_n^\varepsilon(s), \Lambda^{-1}(v_n^\varepsilon(s) - \theta_n^\varepsilon(s)) \rangle ds + 2\varepsilon \int_0^{t \wedge \tau_{\varepsilon, M}^n} \langle u_n^\varepsilon \cdot \nabla \theta_n^\varepsilon, \Lambda^{-1}(v_n^\varepsilon - \theta_n^\varepsilon) \rangle ds \\ & \quad + 2\sqrt{\varepsilon} \int_0^{t \wedge \tau_{\varepsilon, M}^n} \langle \Lambda^{-1/2}(v_n^\varepsilon - \theta_n^\varepsilon), \Lambda^{-1/2}(G(v_n^\varepsilon) - G(\theta_n^\varepsilon)) dW \rangle \\ & \quad + \varepsilon \int_0^{t \wedge \tau_{\varepsilon, M}^n} \|\Lambda^{-1/2}(G(v_n^\varepsilon) - G(\theta_n^\varepsilon))\|_{L_2(U, H)}^2 ds. \end{aligned}$$

Note that by similar arguments as in (4.14) and (4.15), we have

$$\begin{aligned} & |\langle u_n^\varepsilon \cdot \nabla \theta_n^\varepsilon, \Lambda^{-1}(v_n^\varepsilon - \theta_n^\varepsilon) \rangle| \\ & = |(\langle u_n^\varepsilon - u_{v_n}^\varepsilon \cdot \nabla \theta_n^\varepsilon, \Lambda^{-1}(\theta_n^\varepsilon - v_n^\varepsilon) \rangle \\ & \quad + \langle u_{v_n}^\varepsilon \cdot \nabla(\theta_n^\varepsilon - v_n^\varepsilon), \Lambda^{-1}(\theta_n^\varepsilon - v_n^\varepsilon) \rangle \\ & \quad + \langle u_{v_n}^\varepsilon \cdot \nabla v_n^\varepsilon, \Lambda^{-1}(\theta_n^\varepsilon - v_n^\varepsilon) \rangle)| \\ & \leq \frac{\kappa}{2} |\Lambda^{\alpha-1/2}(\theta_n^\varepsilon - v_n^\varepsilon)|^2 + C \|v_n^\varepsilon\|_{L^p}^{N_0} |\Lambda^{-1/2}(\theta_n^\varepsilon - v_n^\varepsilon)|^2 + C \|v_n^\varepsilon\|_{H^\delta}^4, \end{aligned}$$

where $u_{v_n}^\varepsilon$ satisfies (1.3) with θ replaced by v_n^ε . Here in the last step for the last term we use the following estimate:

$$\begin{aligned} |\langle u_{v_n}^\varepsilon \cdot \nabla v_n^\varepsilon, \Lambda^{-1}(\theta_n^\varepsilon - v_n^\varepsilon) \rangle| & = |\langle u_{v_n}^\varepsilon \cdot \nabla \Lambda^{-1}(\theta_n^\varepsilon - v_n^\varepsilon), v_n^\varepsilon \rangle| \leq \|\theta_n^\varepsilon - v_n^\varepsilon\|_{L^{p_1}} \|v_n^\varepsilon\|_{L^{p_2}}^2 \\ & \leq |\Lambda^{\alpha-1/2}(\theta_n^\varepsilon - v_n^\varepsilon)| |\Lambda^\delta v_n^\varepsilon|^2, \end{aligned}$$

where $\frac{1}{p_1} + \frac{2}{p_2} = 1$, $\frac{1}{p_1} + \frac{\alpha-1/2}{2} = \frac{1}{2}$ and we use $H^{\alpha-1/2} \subset L^{p_1}$ and $H^\delta \subset L^{p_2}$ since $\delta \geq (\frac{3}{4} - \frac{\alpha}{2})$.

Therefore, by S.3)

$$\begin{aligned} & |\Lambda^{-1/2}(v_n^\varepsilon(t \wedge \tau_{\varepsilon, M}^n) - \theta_n^\varepsilon(t \wedge \tau_{\varepsilon, M}^n))|^2 + 2\varepsilon\kappa \int_0^{t \wedge \tau_{\varepsilon, M}^n} |\Lambda^{\alpha-1/2}(v_n^\varepsilon - \theta_n^\varepsilon)|^2 ds \\ & \leq 2\varepsilon \int_0^{t \wedge \tau_{\varepsilon, M}^n} \frac{\kappa}{2} |\Lambda^{\alpha-1/2}(v_n^\varepsilon - \theta_n^\varepsilon)|^2 + C |\Lambda^{\alpha-1/2} v_n^\varepsilon|^2 ds \\ & \quad + 2\varepsilon \int_0^{t \wedge \tau_{\varepsilon, M}^n} \frac{\kappa}{2} |\Lambda^{\alpha-1/2}(\theta_n^\varepsilon - v_n^\varepsilon)|^2 + C \|v_n^\varepsilon\|_{L^p}^{N_0} |\Lambda^{-1/2}(\theta_n^\varepsilon - v_n^\varepsilon)|^2 + C \|v_n^\varepsilon\|_{H^\delta}^4 ds \\ & \quad + 2\sqrt{\varepsilon} \int_0^{t \wedge \tau_{\varepsilon, M}^n} \langle \Lambda^{-1/2}(v_n^\varepsilon - \theta_n^\varepsilon), \Lambda^{-1/2}(G(v_n^\varepsilon) - G(\theta_n^\varepsilon)) dW \rangle \\ & \quad + \varepsilon C \int_0^{t \wedge \tau_{\varepsilon, M}^n} |\Lambda^{-1/2}(v_n^\varepsilon - \theta_n^\varepsilon)|^2 ds. \end{aligned}$$

Then Gronwall's lemma yields that

$$\begin{aligned} & |\Lambda^{-1/2}(v_n^\varepsilon(t \wedge \tau_{\varepsilon, M}^n) - \theta_n^\varepsilon(t \wedge \tau_{\varepsilon, M}^n))|^2 \leq \left[2C\varepsilon \int_0^{t \wedge \tau_{\varepsilon, M}^n} (|\Lambda^{\alpha-1/2} v_n^\varepsilon|^2 + \|v_n^\varepsilon\|_{H^\delta}^4) ds \right. \\ & \left. + 2\sqrt{\varepsilon} \left| \int_0^{t \wedge \tau_{\varepsilon, M}^n} \langle \Lambda^{-1/2}(v_n^\varepsilon - \theta_n^\varepsilon), \Lambda^{-1/2}(G(v_n^\varepsilon) - G(\theta_n^\varepsilon)) dW \rangle \right| \right] e^{\varepsilon C \int_0^{t \wedge \tau_{\varepsilon, M}^n} \|v_n^\varepsilon\|_{L^p}^{N_0} ds + Ct\varepsilon}. \end{aligned}$$

Using (4.6) we have

$$\begin{aligned} & (E[\sup_{0 \leq s \leq t \wedge \tau_{\varepsilon, M}^n} |\Lambda^{-1/2}(v_n^\varepsilon(s) - \theta_n^\varepsilon(s))|^2]^q)^{2/q} \\ & \leq C e^{\varepsilon C t M^{N_0/p} + Ct\varepsilon} \left[(\varepsilon M t + \varepsilon M^2 t)^2 + q\varepsilon \int_0^t (E[\sup_{0 \leq r \leq s \wedge \tau_{\varepsilon, M}^n} |\Lambda^{-1/2}(v_n^\varepsilon(r) - \theta_n^\varepsilon(r))|^2]^q)^{2/q} ds \right]. \end{aligned}$$

By Gronwall's lemma we obtain that

$$\begin{aligned} & (E[\sup_{0 \leq s \leq T \wedge \tau_{\varepsilon, M}^n} |\Lambda^{-1/2}(v_n^\varepsilon(s) - \theta_n^\varepsilon(s))|^2]^q)^{2/q} \\ & \leq C e^{\varepsilon C T M^{N_0/p} + CT\varepsilon} (\varepsilon M T + \varepsilon M^2 T)^2 \exp \left[C q T \varepsilon e^{\varepsilon C T M^{N_0/p} + CT\varepsilon} \right]. \end{aligned}$$

Fixing M and taking $q = 2/\varepsilon$ we have

$$\begin{aligned} & \varepsilon \log P(\sup_{0 \leq t \leq T \wedge \tau_{\varepsilon, M}^n} \|\theta_n^\varepsilon(t) - v_n^\varepsilon(t)\|_{H^{-1/2}}^2 > \eta) \\ & \leq \varepsilon \log \frac{E[\sup_{0 \leq s \leq T \wedge \tau_{\varepsilon, M}^n} |\Lambda^{-1/2}(v_n^\varepsilon(s) - \theta_n^\varepsilon(s))|^{2q}]}{\eta^q} \\ & \leq \log C (\varepsilon M T + \varepsilon M^2 T)^2 - 2 \log \eta + CT e^{\varepsilon C T M^{N_0/p} + CT\varepsilon} + \varepsilon C M^{N_0/p} + CT\varepsilon \\ & \rightarrow -\infty, \text{ as } \varepsilon \rightarrow 0. \end{aligned} \tag{4.20}$$

Therefore, there exists ε_0 such that for every ε satisfying $0 < \varepsilon \leq \varepsilon_0$,

$$P(\sup_{0 \leq t \leq T} \|\theta_n^\varepsilon(t) - v_n^\varepsilon(t)\|_{H^{-1/2}}^2 > \eta, \sup_{0 \leq t \leq T} (\|v_n^\varepsilon(t)\|_{H^\delta}^2 + \|v_n^\varepsilon(t)\|_{L^p}^p) \leq M) \leq e^{-R/\varepsilon}. \tag{4.21}$$

By Lemma 4.4 and (4.21), we know that there exists ε_0 such that for every ε satisfying $0 < \varepsilon \leq \varepsilon_0$ we have

$$P(\sup_{0 \leq t \leq T} \|\theta_n^\varepsilon(t) - v_n^\varepsilon(t)\|_{H^{-1/2}}^2 > \eta) \leq 2e^{-R/\varepsilon}.$$

Since R is arbitrary, the desired result follows. \square

Now we can finish the proof of Theorem 4.2.

Proof of Theorem 4.2 By Lemmas 4.5 and 4.6, we have for every $R > 0$ there exists N_2 such that

$$P(\sup_{0 \leq t \leq T} \|\theta_{N_2}^\varepsilon(t) - \theta^\varepsilon(t)\|_{H^{-1/2}}^2 > \frac{\eta}{3}) \leq e^{-R/\varepsilon} \text{ for any } \varepsilon \in (0, 1];$$

and

$$P(\sup_{0 \leq t \leq T} \|v_{N_2}^\varepsilon(t) - v^\varepsilon(t)\|_{H^{-1/2}}^2 > \frac{\eta}{3}) \leq e^{-R/\varepsilon} \text{ for any } \varepsilon \in (0, 1].$$

For such N_2 , according to Lemma 4.7, there exists ε_0 such that for every ε satisfying $0 < \varepsilon \leq \varepsilon_0$,

$$P\left(\sup_{0 \leq t \leq T} \|\theta_{N_2}^\varepsilon(t) - v_{N_2}^\varepsilon(t)\|_{H^{-1/2}}^2 > \frac{\eta}{3}\right) \leq e^{-R/\varepsilon}.$$

Therefore, for every ε satisfying $0 < \varepsilon \leq \varepsilon_0$ we have

$$P\left(\sup_{0 \leq t \leq T} \|\theta^\varepsilon(t) - v^\varepsilon(t)\|_{H^{-1/2}}^2 > \eta\right) \leq 3e^{-R/\varepsilon}.$$

Since R is arbitrary, we have

$$\lim_{\varepsilon \rightarrow 0} \varepsilon \log P\left(\sup_{0 \leq t \leq T} |\Lambda^{-1/2}(\theta^\varepsilon(t) - v^\varepsilon(t))|^2 > \eta\right) = -\infty,$$

i.e. (4.10) holds. Hence the proof of Theorem 4.2 is complete. \square

Appendix

Theorem A.1 Suppose that A.1)-A.3) hold. Then for any $\theta_0 \in H^\delta \cap L^p$ with p in Hypothesis 3.5 iii), (3.8) has a unique solution

$$\theta_v \in L^\infty([0, T], H^\delta \cap L^p) \cap L^2([0, T], H^{\delta+\alpha}) \cap C([0, T], H^{-\beta})$$

and it has the following estimate:

$$\sup_{t \in [0, T]} (|\Lambda^\delta \theta_v(t)|^2 + \|\theta_v(t)\|_{L^p}^p) + \int_0^T |\Lambda^{\delta+\alpha} \theta_v(s)|^2 ds \leq C, \quad (\text{A.4})$$

where C is some constant only depending on $|\Lambda^\delta \theta_0|$, $\|\theta_0\|_{L^p}$, T and $\int_0^T |v|_U^2 ds$.

Proof In the following we will assume that $\delta < 1$. The case for $\delta \geq 1$ is similar.

[Step 1] We first establish the existence of solutions of the following equation

$$\frac{d\theta(t)}{dt} + A_\alpha \theta(t) + w(t) \cdot \nabla \theta(t) = k_\sigma * G(\theta(t))v(t), \quad (\text{A.5})$$

$$\theta(0) = \theta_0 \in H^3$$

with a given smooth function $w(t)$ which satisfies $\operatorname{div} w(t) = 0$ and $\sup_{t \in [0, T]} \|w(t)\|_{C^3} \leq C$. Here $k_\sigma * G(\theta)$ means for $y \in U$, $k_\sigma * G(\theta)(y) = k_\sigma * (G(\theta)(y))$, where k_σ is the periodic Poisson kernel in \mathbb{T}^2 given by $\widehat{k}_\sigma(\zeta) = e^{-\sigma|\zeta|}$, $\zeta \in \mathbb{Z}^2$.

Then we have the following apriori estimate

$$\frac{d}{dt} |\Lambda^3 \theta|^2 + 2\kappa |\Lambda^{3+\alpha} \theta|^2 \leq 2|\langle w \cdot \nabla \theta, \Lambda^6 \theta \rangle| + 2|\langle \Lambda^3 \theta, \Lambda^3 k_\delta * G(\theta)v \rangle|.$$

By Lemmas 2.1 and 2.2 we have that

$$|\langle \Lambda^{3-\alpha}(w \cdot \nabla \theta), \Lambda^{3+\alpha} \theta \rangle| \leq C \|w\|_{C^3(\mathbb{T}^2)} |\Lambda^{4-\alpha} \theta| |\Lambda^{3+\alpha} \theta| \leq C |\Lambda^3 \theta|^2 + \kappa |\Lambda^{3+\alpha} \theta|^2,$$

where in the last inequality we use the interpolation inequality and Young's inequality.

Note that we also have

$$|\Lambda^3 k_\sigma * G(\theta)v| \leq C(\sigma) \|G(\theta)\|_{L^2(U,H)} |v|_U \leq C|v|_U (|\Lambda^\alpha \theta| + 1).$$

Thus,

$$\frac{d}{dt} |\Lambda^3 \theta|^2 + \kappa |\Lambda^{3+\alpha} \theta|^2 \leq C|v|_U (|\Lambda^3 \theta|^2 + 1) + C|\Lambda^3 \theta|^2.$$

Then by the standard Galerkin approximation we obtain that there exists a solution $\theta \in L^\infty([0, T], H^3) \cap L^2([0, T], H^{3+\alpha}) \cap C([0, T], H^1)$ of (A.5).

[Step 2] Now we construct an approximation of (3.8).

We pick a smooth $\phi \geq 0$, with $\text{supp } \phi \subset [1, 2]$ and $\int_0^\infty \phi = 1$, and for $\sigma > 0$ let

$$U_\sigma[\theta](t) := \int_0^\infty \phi(\tau) (k_\sigma * R^\perp \theta)(t - \sigma\tau) d\tau,$$

where k_σ is the periodic Poisson Kernel in \mathbb{T}^2 given by $\widehat{k_\sigma}(\zeta) = e^{-\sigma|\zeta|}$, $\zeta \in \mathbb{Z}^2$, and we set $\theta(t) = 0$ for $t < 0$.

We take a sequence $\delta_n \downarrow 0$ and consider the equation

$$\frac{d\theta_n(t)}{dt} + A_\alpha \theta_n(t) + u_n(t) \cdot \nabla \theta_n(t) = k_{\delta_n} * G(\theta_n)v(t) \quad (\text{A.6})$$

with initial data $\theta_n(0) = k_{\delta_n} * \theta_0$ and $u_n = U_{\delta_n}[\theta_n]$.

For a fixed n , this is a linear equation in θ_n on each subinterval $[t_k^n, t_{k+1}^n]$ with $t_k^n = k\delta_n$, since u_n is determined by the values of θ_n on the two previous subintervals.

By [Step 1], we obtain the existence of a solution to (A.6) for fixed n . Moreover by (A.1) the solution satisfies the following L^p norm estimate:

$$\begin{aligned} \frac{d}{dt} \|\theta_n\|_{L^p}^p &= p \int |\theta_n|^{p-2} \theta_n (k_{\delta_n} * G(\theta_n)v - u_n \cdot \nabla \theta_n - \Lambda^{2\alpha} \theta_n) dx \\ &\leq p \int |\theta_n|^{p-2} \theta_n k_{\delta_n} * G(\theta_n)v dx \leq C|v|_U (\|\theta_n\|_{L^p}^p + 1). \end{aligned}$$

Here in the first inequality we use $\text{div} u_n = 0$ and $\int |\theta_n|^{p-2} \theta_n \Lambda^{2\alpha} \theta_n \geq 0$ (c.f. [27, Lemma 3.2]). Then Gronwall's lemma implies that

$$\sup_{t \in [0, T]} \|\theta_n\|_{L^p}^p \leq (\|\theta_0\|_{L^p}^p + \int_0^T |v|_U dt) \exp\left(\int_0^T |v|_U dt\right).$$

By (2.1) we have

$$\sup_{t \in [0, T]} \|u_n\|_{L^p}^p \leq C(\|\theta_0\|_{L^p}^p + \int_0^T |v|_U dt) \exp\left(\int_0^T |v|_U dt\right).$$

Now we prove the uniform H^δ estimate:

$$\frac{d}{dt} |\Lambda^\delta \theta_n|^2 + 2\kappa |\Lambda^{\delta+\alpha} \theta_n|^2 \leq 2|\langle \Lambda^\delta (u_n \cdot \nabla \theta_n), \Lambda^\delta \theta_n \rangle| + 2|\langle \Lambda^\delta \theta_n, \Lambda^\delta k_{\delta_n} * G(\theta_n)v \rangle|.$$

By [27, Proposition 3.6] we have

$$|\langle \Lambda^\delta(u_n \cdot \nabla \theta_n), \Lambda^\delta \theta_n \rangle| \leq \frac{\kappa}{4} |\Lambda^{\delta+\alpha} \theta_n|^2 + \frac{\kappa}{8} |\Lambda^{\delta+\alpha} u_n|^2 + C \|u_n\|_{L^p}^{N_0} |\Lambda^\delta \theta_n|^2 + C \|\theta_n\|_{L^p}^{N_0} |\Lambda^\delta u_n|^2,$$

where $N_0 = \frac{\alpha}{\alpha - \frac{1}{2} - \frac{1}{p}}$.

By A.2) we also obtain

$$|\langle \Lambda^\delta \theta_n, \Lambda^\delta k_{\delta_n} * G(\theta_n) v \rangle| \leq C |v|_U |\Lambda^\delta \theta_n| (|\Lambda^{\delta+\alpha} \theta_n| + 1) \leq \varepsilon |\Lambda^{\delta+\alpha} \theta_n|^2 + C(|v|_U^2 |\Lambda^\delta \theta_n|^2 + 1).$$

Thus,

$$|\Lambda^\delta \theta_n(t)|^2 + \kappa \int_0^t |\Lambda^{\delta+\alpha} \theta_n|^2 ds \leq \int_0^t [2C \|u_n\|_{L^p}^{N_0} |\Lambda^\delta \theta_n|^2 + C \|\theta_n\|_{L^p}^{N_0} |\Lambda^\delta u_n|^2 + C(|v|_U^2 |\Lambda^\delta \theta_n|^2 + 1)] ds.$$

Note that we have (here we cannot control $|\Lambda^\delta u_n|$ by $|\Lambda^\delta \theta_n|$ pointwisely in time)

$$\int_0^t |\Lambda^\delta u_n|^2 ds \leq C \int_0^t |\Lambda^\delta \theta_n|^2 ds.$$

Using Gronwall's inequality and L^p norm estimate above we obtain the uniform H^δ estimate for θ_n .

Then by standard argument we know that θ_n converges to the solution θ_v of (3.8), which implies (A.4). The proof of uniqueness is the same as in [27, Theorem 3.7] by A.3). \square

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