Closability of Quadratic Forms Associated to Invariant Probability Measures of SPDEs *

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Abstract

By using the integration by parts formula of a Markov operator, the closability of quadratic forms associated to the corresponding invariant probability measure is proved. The general result is applied to the study of semilinear SPDEs, infinite-dimensional stochastic Hamiltonian systems, and semilinear SPDEs with delay.

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1 Introduction

Let \mathbb{B} be a separable Banach space and μ a reference probability measure on \mathbb{B} . For any $k \in \mathbb{B}$, let ∂_k denote the directional derivative along k. According to [8], the form

$$\mathscr{E}_k(f,g) := \mu((\partial_k f)(\partial_k g)) := \int_{\mathbb{R}} (\partial_k f)(\partial_k g) d\mu, \quad f,g \in C_b^2(\mathbb{B}),$$

is closable on $L^2(\mu)$ if $\rho_s:=\frac{\mathrm{d}\mu(sk+\cdot)}{\mathrm{d}\mu}$ exists for any s such that $s\mapsto\rho_s$ is lower semi-continuous μ -a.e.; i.e. for some fixed μ -versions of $\rho_s,s\in\mathbb{R}$,

$$\liminf_{s \to t} \rho_s(x) \ge \rho_t(x), \quad \mu-\text{a.e. } x, \ t \in \mathbb{R}.$$

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In this paper, we aim to investigate the closability of \mathcal{E}_k for μ being the invariant probability measure of a (degenerate/delay) semilinear SPDE. Since in this case the above lower semi-continuity condition is hard to check, in this paper we make use of the integration by parts formula for the associated Markov semigroup in the line of [10] using coupling arguments.

The main motivation to study the closability of $E\mathscr{E}_k$ (respectively of ∂_k) on $L^2(\mu)$ is that it leads to a concept of weak differentiablity on \mathbb{B} with respect to μ and one can define the corresponding Sobolev space on \mathbb{B} in $L^p(\mu)$, $p \in [1, \infty)$. In particular, one can analyze the generator of a Markov process (e.g. arising from a solution of an SPDE) on these Sobolev spaces when μ is its (infinitesimally) invariant measure, see e.g. [7] for details.

Before considering specific models of SPDEs, we first introduce a general result on the closability of \mathcal{E}_k using the integration by parts formula. To this end, we consider a family of \mathbb{B} -valued random variables $\{X^x\}_{x\in\mathbb{B}}$ measurable in x, and let P(x, dy) be the distribution of X^x for $x \in \mathcal{B}$. Then we have the following Markov operator on $\mathcal{B}_b(\mathbb{B})$:

$$Pf(x) := \int_{\mathbb{B}} f(y)P(x, dy) = \mathbb{E}f(X^x), \quad x \in \mathbb{B}, f \in \mathscr{B}_b(\mathbb{B}).$$

A probability measure μ on \mathbb{B} is called an invariant measure of P if $\mu(Pf) = \mu(f)$ for all $f \in \mathscr{B}_b(\mathbb{B})$.

P1.1 Proposition 1.1. Assume that the Markov operator P has an invariant probability measure μ . Let $k \in \mathbb{B}$. If there exists a family of real random variables $\{M_x\}_{x \in \mathbb{B}}$ measurable in x such that $M \in L^2(\mathbb{P} \times \mu)$, i.e.

$$\boxed{1.0} \quad (1.1) \qquad \qquad (\mathbb{P} \times \mu)(|M|^2) := \int_{\mathbb{R}} \mathbb{E}|M_x|^2 \mu(\mathrm{d}x) < \infty;$$

and the integration by parts formula

holds, then $(\mathscr{E}_k, C_b^2(\mathbb{B}))$ is closable in $L^2(\mu)$.

Proof. Since μ is P-invariant, by (1.1) and (1.2) we have

$$\mu(\partial_k f) = \int_{\mathbb{R}} P(\partial_k f)(x)\mu(\mathrm{d}x) = (\mathbb{P} \times \mu) (f(X^{\cdot})M_{\cdot}), \quad f \in C_b^2(\mathbb{B}).$$

So,

$$\mathcal{E}_k(f,g) := \mu((\partial_k f)(\partial_k g)) = \mu(\partial_k \{f \partial_k g\}) - \mu(f \partial_k^2 g)$$

= $(\mathbb{P} \times \mu)(\{f \partial_k g\}(X)M) - \mu(f \partial_k^2 g), \quad f, g \in C_b^2(\mathbb{B}).$

It is standard that this implies the closability of the form $(\mathcal{E}_k, C_b^2(\mathbb{B}))$ in $L^2(\mu)$. Indeed, for $\{f_n\}_{n\geq 1}\subset C_b^2(\mathbb{B})$ with $f_n\to 0$ and $\partial_k f_n\to Z$ in $L^2(\mu)$, it suffices to prove that Z=0.

Since $\mu(f_n^2) \to 0$ and $(\mathbb{P} \times \mu)(|f_n \partial_k g|^2(X)|) = \mu(|f_n \partial_k g|^2)$ as μ is P-invariant, the above formula yields

$$\begin{aligned} &|\mu(Zg)| = \lim_{n \to \infty} |\mu(g\partial_k f_n)| \\ &= \lim_{n \to \infty} \left| (\mathbb{P} \times \mu) \left(\{f_n \partial_k g\}(X^{\cdot}) M_{\cdot} \right) - \mu(f_n \partial_k^2 g) \right| \\ &\leq \liminf_{n \to \infty} \left\{ \sqrt{(\mathbb{P} \times \mu) \left(|f_n \partial_k g|^2 (X^{\cdot}) \right) \cdot (\mathbb{P} \times \mu) (|M_{\cdot}|^2)} + \sqrt{\mu(f_n^2) \mu(|\partial_k^2 g|^2)} \right\} \\ &\leq \liminf_{n \to \infty} \left\{ \|\partial_k g\|_{\infty} \sqrt{\mu(f_n^2) \cdot (\mathbb{P} \times \mu) (|M_{\cdot}|^2)} + \|\partial_k^2 g\|_{\infty} \sqrt{\mu(f_n^2)} \right\} = 0, \quad g \in C_b^2(\mathbb{B}). \end{aligned}$$

Therefore, Z = 0.

Remark 1.1. The integration by parts formula (1.2) implies the estimate

$$|\mu(\partial_k f)|^2 \le (\mathbb{P} \times \mu)(|M|^2)\mu(f^2).$$

As the main result in [3] (Theorem 10), this type of estimate, called Fomin derivative estimate of the invariant measure, was derived as the main result for the following semi-linear SPDE on $\mathbb{H} := L^2(\mathcal{O})$ for any bounded open domain $\mathcal{O} \subset \mathbb{R}^n$ for $1 \leq n \leq 3$:

$$dX(t) = [\Delta X(t) + p(X(t))]dt + (-\Delta)^{-\gamma/2}dW(t),$$

where Δ is the Dirichlet Laplacian on \mathcal{O} , p is a decreasing polynomial with odd degree, $\gamma \in (\frac{n}{2} - 1, 1)$, and $W_t(t)$ is the cylindrical Brownian motion on \mathbb{H} . The main point of the study is to apply the Bismut-Elworthy-Li derivative formula and the following formula for the semigroup P_t^{α} for the Yoshida approximation of this SPDE (see [3, Proposition 7]):

$$P_t^{\alpha} \partial_k f = \partial_k P_t^{\alpha} - \int_0^t P_{t-s}(\partial_{Ak+\partial_k p} P_s^{\alpha} f) ds.$$

In this paper we will establish the integration by parts formula of type (1.2) for the associated semigroup which implies the estimate (1.3). Our results apply to a general framework where the operator $(-\Delta)^{-\gamma/2}$ is replaced by a suitable linear operator σ (see Section 2) which can be degenerate (see Section 3), and the drift p(x) is replaced by a general map b which may include a time delay (see Section 4). However, the price we have to pay for the generalization is that the drift b should be regular enough.

2 Semilinear SPDEs

Let $(\mathbb{H}, \langle \cdot, \cdot \rangle, |\cdot|)$ be a real separable Hilbert space, and $(W(t))_{t\geq 0}$ a cylindrical Wiener process on \mathbb{H} with respect to a complete probability space $(\Omega, \mathscr{F}, \mathbb{P})$ with the natural filtration $\{\mathscr{F}_t\}_{t\geq 0}$. Let $\mathscr{L}(\mathbb{H})$ and $\mathscr{L}_{HS}(\mathbb{H})$ be the spaces of all linear bounded operators and Hilbert-Schmidt operators on H respectively. Let $\|\cdot\|$ and $\|\cdot\|_{HS}$ denote the operator norm and the Hilbert-Schmidt norm respectively.

Consider the following semilinear SPDE

$$dX(t) = \{AX(t) + b(X(t))\}dt + \sigma dW(t),$$

where

- (A1) $(A, \mathcal{D}(A))$ is a negatively definite self-adjoint linear operator on \mathbb{H} with compact resolvent.
- (A2) Let \mathbb{H}^{-2} be the completion of \mathbb{H} under the inner product

$$\langle x, y \rangle_{\mathbb{H}^{-2}} := \langle A^{-1}x, A^{-1}y \rangle.$$

Let $b: \mathbb{H} \to \mathbb{H}^{-2}$ be such that

$$\int_0^1 |e^{tA}b(0)| dt < \infty, |e^{tA}(b(x) - b(y))| \le \gamma(t)|x - y|, x, y \in \mathbb{H}, t > 0$$

holds for some positive $\gamma \in C((0,\infty))$ with $\int_0^1 \gamma(t) dt < \infty$.

(A3)
$$\sigma \in \mathcal{L}(\mathbb{H})$$
 with $\operatorname{Ker}(\sigma \sigma^*) = \{0\}$ and $\int_0^1 \|e^{tA}\sigma\|_{HS}^2 dt < \infty$.

According to (A1), the spectrum of A is discrete with negative eigenvalues. Let $0 < \lambda_0 \le \cdots \le \lambda_n \cdots$ be all eigenvalues of -A counting the multiplicities, and let $\{e_i\}_{i\ge 1}$ be the corresponding unit eigen-basis. Denote $\mathbb{H}_{A,n} = \text{span}\{e_i : 1 \le i \le n\}, n \ge 1$. Then $\mathbb{H}_A := \bigcup_{n=1}^{\infty} \mathbb{H}_{A,n}$ is a dense subspace of \mathbb{H} . In assumption (A2) we have used the fact that for any t > 0, the operator e^{tA} extends uniquely to a bounded linear operator from \mathbb{H}^{-2} to \mathbb{H} , which is again denoted by e^{tA} .

Due to assumptions (A1), (A2) and (A3), by a standard iteration argument we conclude that for any $x \in \mathbb{H}$ the equation (2.1) has a unique mild solution $X^x(t)$ such that $X^x(0) = x$ (see [4]). Let

$$P_t f(x) = \mathbb{E}f(X^x(t)), \quad f \in \mathscr{B}_b(\mathbb{H}), x \in \mathbb{H}$$

be the associated Markov semigroup.

Let

$$||x||_{\sigma} = \inf\{|y|: y \in \mathbb{H}, \sqrt{\sigma\sigma^*}y = x\}, x \in \mathbb{H},$$

where $\inf \emptyset := \infty$ by convention. Then $||x||_{\sigma} < \infty$ if and only if $x \in \operatorname{Im}(\sigma)$.

- T2.1 Theorem 2.1. Assume that P_t has an invariant probability measure μ and $\mathbb{H}_A \subset \operatorname{Im}(\sqrt{\sigma\sigma^*})$.
 - (1) For any $k \in \mathbb{H}_A$ such that

**0 (2.2)
$$\sup_{x \in \mathbb{H}} \|\partial_k b(x)\|_{\sigma} := \sup_{x \in \mathbb{H}} \limsup_{\varepsilon \downarrow 0} \frac{\|b(x + \varepsilon k) - b(x)\|_{\sigma}}{\varepsilon} < \infty,$$

the form $(\mathcal{E}_k, C_b^2(\mathbb{H}))$ is closable in $L^2(\mu)$.

(2) If $\sigma\sigma^*$ is invertible and $b: \mathbb{H} \to \mathbb{H}$ is Lipschitz continuous, then $(\mathcal{E}_k, C_b^2(\mathbb{H}))$ is closable in $L^2(\mu)$ for any $k \in \mathcal{D}(A)$.

Proof. Since $d\tilde{W}_t := (\sigma \sigma^*)^{-1/2} \sigma dW_t$ is also a cylindrical Brownian motion and $\sigma dW_t = \sqrt{\sigma \sigma^*} d\tilde{W}_t$, we may and do assume that σ is non-negatively definite.

(1) Without loss of generality, we may and do assume that k is an eigenvector of A, i.e. $Ak = \lambda k$ for some $\lambda \in \mathbb{R}$. We first prove the case where b is Fréchet differentiable along the direction k. By $Ak = \lambda k$ we have

$$k(t) := \int_0^t e^{sA} k ds = \frac{e^{\lambda t} - 1}{\lambda} k, \quad t \ge 0,$$

where for $\lambda = 0$ we set $\frac{e^{\lambda t}-1}{\lambda} = t$. Due to $||k||_{\sigma} < \infty$ and (2.2), the proof of [10, Theorem 5.1(1)] leads to the integration by parts formula

$$P_T(\partial_k f)(x) = \mathbb{E}\left\{f(X^x(T))M_{x,T}\right\}, \quad f \in C_b^1(\mathbb{H}), x \in \mathbb{H}, T > 0,$$

where

$$M_{x,T} := \frac{\lambda}{e^{\lambda T} - 1} \int_0^T \left\langle \sigma^{-1} \left(k - \frac{e^{\lambda t} - 1}{\lambda} (\partial_k b)(X^x(t)) \right), dW(t) \right\rangle.$$

Since (2.2) implies

$$\boxed{\text{FH}} \quad (2.4) \qquad \int_{\mathbb{B}} \mathbb{E}|M_{x,T}|^2 \, \mu(\mathrm{d}x) \leq \frac{\lambda^2}{(\mathrm{e}^{\lambda T} - 1)^2} \int_0^T \left\| \sigma^{-1} \left(k - \frac{\mathrm{e}^{\lambda t} - 1}{\lambda} \partial_k b \right) \right\|_{\infty}^2 \mathrm{d}t < \infty,$$

 $(\mathscr{E}_k, C_b^2(\mathbb{H}))$ is closable in $L^2(\mu)$ according to Proposition 1.1. In general, for any $\varepsilon > 0$ let

$$b_{\varepsilon}(x) = \frac{1}{\sqrt{2\pi\varepsilon}} \int_{\mathbb{R}} b(x+rk) \exp\left[-\frac{r^2}{2\varepsilon}\right] dr, \quad x \in \mathbb{H}.$$

Then for any $\varepsilon > 0$, b_{ε} is Fréchet differentiable along k and (2.2) holds uniformly in ε with b_{ε} replacing b. Let P_t^{ε} be the semigroup for the solution $X_{\varepsilon}(t)$ associated to equation (2.1) with b_{ε} replacing b. By simple calculations we have:

- (i) $\lim_{\varepsilon \downarrow 0} \mathbb{E}|X_{\varepsilon}^{x}(t) X^{x}(t)|^{2} = 0, \ t \geq 0, x \in \mathbb{H}.$
- (ii) For any T > 0, the family

$$M_{\cdot,T}^{\varepsilon} := \frac{\lambda}{\mathrm{e}^{\lambda T} - 1} \int_{0}^{T} \left\langle \sigma^{-1} \left(k - \frac{\mathrm{e}^{\lambda t} - 1}{\lambda} (\partial_{k} b_{\varepsilon}) (X_{\varepsilon}^{\cdot}(t)) \right), \, \mathrm{d}W(t) \right\rangle, \quad \varepsilon > 0$$

is bounded in $L^2(\mathbb{P} \times \mu)$; i.e. $\sup_{\varepsilon>0} \int_{\mathbb{B}} \mathbb{E} |M_{x,T}|^2 \mu(\mathrm{d}x) < \infty$.

(iii)
$$P_T^{\varepsilon}(\partial_k f)(x) = \mathbb{E}(f(X_{\varepsilon}^x(T)M_{x,T}^{\varepsilon}), f \in C_b^1(\mathbb{H}), \varepsilon > 0.$$

So, there exist $M_{\cdot,T} \in L^2(\mathbb{P} \times \mu)$ and a sequence $\varepsilon_n \downarrow 0$ such that $M_{\cdot,T}^{\varepsilon_n} \to M_{\cdot,T}$ weakly in $L^2(\mathbb{P} \times \mu)$. Thus, by taking $n \to \infty$ in (iii) and using (i), we prove (2.3) for μ -a.e. $x \in \mathbb{B}$. Then the proof of the first assertion is completed as in the first case.

(2) Since σ is invertible, (A3) implies $\alpha := \sum_{i=1}^{\infty} \frac{1}{\lambda_i} < \infty$. Next, since the Lipschitz constant $\|\partial b\|_{\infty}$ of b is finite, the integration by parts formula (2.3) also implies explicit Fomin derivative estimates on the invariant probability measure, which were investigated recently in [3]. Indeed, it follows from (2.3) and (2.4) that

$$|\mu(\partial_k f)| = \inf_{T>0} |\mu(P_T(\partial_k f))| \le \inf_{T>0} \sqrt{\mu(P_T f^2)} \left(\int_{\mathbb{B}} \mathbb{E} |M_{x,T}|^2 \, \mu(\mathrm{d}x) \right)^{\frac{1}{2}}$$

$$\le |k| \cdot ||f||_{L^2(\mu)} \inf_{T>0} \frac{\lambda}{\mathrm{e}^{\lambda T} - 1} \left(\int_0^T \left\| \sigma^{-1} \left(I - \frac{\mathrm{e}^{\lambda t} - 1}{\lambda} \partial b \right) \right\|_{\infty}^2 \mathrm{d}t \right)^{\frac{1}{2}}, \quad Ak = \lambda k.$$

By taking $k = e_i, T = \lambda_i^{-1}$ and $\lambda = -\lambda_i$ in the above estimate, for any $k \in \mathcal{D}(A)$ we have

$$|\mu(\partial_{k}f)| \leq \sum_{i=1}^{\infty} |\langle k, e_{i} \rangle \mu(\partial_{e_{i}}f)| \leq \left(\sum_{i=1}^{\infty} \lambda_{i}^{2} \langle k, e_{i} \rangle^{2}\right)^{\frac{1}{2}} \left(\sum_{i=1}^{\infty} \frac{1}{\lambda_{i}^{2}} \mu(\partial_{e_{i}}f)^{2}\right)^{\frac{1}{2}}$$

$$\leq |Ak| \left(\sum_{i=1}^{\infty} \frac{\|\sigma^{-1}\|^{2}}{\lambda_{i}(e-1)^{2}} \left(1 + \frac{e-1}{\lambda_{i}} \|\partial b\|_{\infty}\right)^{2}\right)^{\frac{1}{2}} \|f\|_{L^{2}(\mu)}$$

$$\leq C|Ak| \cdot \|f\|_{L^{2}(\mu)},$$

where $C := \frac{\|\sigma^{-1}\|\sqrt{\alpha}}{e-1} \left(1 + \frac{e-1}{\lambda_1} \|\partial b\|_{\infty}\right)$. This implies the closablity of $(\mathcal{E}_k, C_b^2(\mathbb{H}))$ as explained in the proof of Proposition 1.1. Indeed, if $\{f_n\}_{n\geq 1} \subset C_b^2(\mathbb{B})$ satisfies $f_n \to 0$ and $\partial_k f_n \to Z$ in $L^2(\mu)$, then (2.5) implies

$$|\mu(gZ)| = \lim_{n \to \infty} |\mu(g\partial_k f_n)| = \lim_{n \to \infty} |\mu(\partial_k (f_n g) - \mu(f_n \partial_k g)|$$

$$\leq C|Ak| \lim_{n \to \infty} \sqrt{\mu((f_n g)^2)} = 0, \quad g \in C_b^2(\mathbb{B}),$$

so that Z=0.

To conclude this section, let us recall a result concerning existence and stability of the invariant probability measure. Let $W_a(t) = \int_0^t \mathrm{e}^{A(t-s)} \sigma \mathrm{d}W(s), t \geq 0$. Assume that b is Lipschitz continuous and $\int_0^\infty \|\mathrm{e}^{tA}\sigma\|_{HS}^2 \mathrm{d}t < \infty$. We have

$$\sup_{t>0} \mathbb{E}(\|W_A(t)\|^2 + |b(W_A(t))|^2) < \infty.$$

Therefore, by [5, Theorem 2.3], if there exist $c_1 > 0, c_2 \in \mathbb{R}$ with $c_1 + c_2 > 0$ such that

$$\langle A(x-y), x-y \rangle \le -c_1|x-y|^2, \ \langle b(x)-b(y), x-y \rangle \le -c_2|x-y|^2, \ x,y \in \mathbb{H},$$

then P_t has a unique invariant probability measure such that $\lim_{t\to\infty} P_t f = \mu(f)$ holds for $f \in C_b(\mathbb{H})$.

3 Stochastic Hamiltonian systems on Hilbert spaces

Let $\tilde{\mathbb{H}}$ and \mathbb{H} be two separable Hilbert spaces. Consider the following stochastic differential equation for Z(t) := (X(t), Y(t)) on $\tilde{\mathbb{H}} \times \mathbb{H}$:

where $B \in \mathcal{L}(\mathbb{H} \to \tilde{\mathbb{H}})$, $(A, \mathcal{D}(A))$ satisfies **(A1)**, σ satisfies **(A3)**, W(t) is the cylindrical Brownian motion on \mathbb{H} , and $b:[0,\infty)\times \tilde{\mathbb{H}}\times \mathbb{H} \to \mathbb{H}^{-2}$ satisfies: for any T>0 there exists $\gamma \in C((0,T])$ with $\int_0^T \gamma(t) dt < \infty$ such that

$$\sup_{s \in [0,T]} \int_0^T |\mathrm{e}^{tA} b(s,0)| \mathrm{d}t < 1,$$

$$\sup_{s \in [0,T]} |\mathrm{e}^{tA} (b(s,z) - b(s,z'))| \le \gamma(t) |z-z'|, \quad t \in [0,T], z,z' \in \tilde{\mathbb{H}} \times \mathbb{H}.$$

Obviously, for any initial data $z := (x, y) \in \mathbb{H}$, the equation has a unique mild solution $Z^z(t)$. Let P_t be the associated Markov semigroup.

When $\tilde{\mathbb{H}}$ and \mathbb{H} are finite-dimensional, the integration by parts formula of P_t has been established in [10, Theorem 3.1]. Here, we extend this result to the present infinite-dimensional setting.

P3.1 Proposition 3.1. Assume that $BB^* \in \mathcal{L}(\tilde{\mathbb{H}})$ with $Ker(BB^*) = \{0\}$. Let T > 0 and $k := (k_1, k_2) \in Im(BB^*) \times \mathbb{H}$ be such that

EG (3.3)
$$Ak_2 = \theta_2 k_2, \quad AB^* (BB^*)^{-1} k_1 = \theta_1 B^* (BB^*)^{-1} k_1$$

for some constants $\theta_1, \theta_2 \in \mathbb{R}$. For any $\phi, \psi \in C^1([0,T])$ such that

$$\boxed{\text{PH}} \quad (3.4) \qquad \phi(0) = \phi(T) = \psi(0) = \psi(T) - 1 = \int_0^T \mathrm{e}^{\theta_2 t} \psi(t) \mathrm{d}t = 0, \quad \int_0^T \phi(t) \mathrm{e}^{\theta_1 t} \mathrm{d}t = \mathrm{e}^{\theta_1 T},$$

$$let$$

$$h(t) = B^*(BB^*)^{-1}k_1 \int_0^t \phi'(s)e^{\theta_1(s-T)}ds + k_2 \int_0^t \psi'(s)e^{\theta_2(s-T)}ds,$$

$$\tilde{h}(t) = \phi(t)e^{\theta_1(t-T)}B^*(BB^*)^{-1}k_1 + \psi(t)e^{\theta_2(t-T)}k_2,$$

$$\Theta(t) = \left(\int_0^t B\tilde{h}(s)ds, \ \tilde{h}(t)\right), \quad t \in [0,T].$$

If for any $t \in [0,T]$, $b(s,\cdot)$ is Fréchet differentiable along $\Theta(t)$ such that

$$\int_0^T \sup_{z \in \tilde{\mathbb{H}} \times \mathbb{H}} \left\| h'(t) - (\partial_{\Theta(t)} b(t, \cdot))(z) \right\|_{\sigma}^2 dt < \infty,$$

then for any $f \in C_b^1(\tilde{\mathbb{H}} \times \mathbb{H})$,

$$P_T(\partial_k f) = \mathbb{E}\bigg\{f(Z(T))\int_0^T \Big\langle (\sigma\sigma^*)^{-1/2} \big\{h'(t) - (\partial_{\Theta(t)}b(t,\cdot))(Z(t))\big\}, \ \mathrm{d}W(t)\Big\rangle\bigg\}.$$

Proof. As explained in the proof of Theorem 2.1, we simply assume that $\sigma = \sqrt{\sigma \sigma^*}$. Let $(X^0(t), Y^0(t)) = (X(t), Y(t))$ solve (3.1) with initial data (x, y), and for $\varepsilon \in (0, 1]$ let $(X^{\varepsilon}(t), Y^{\varepsilon}(t))$ solve the equation

Then it is easy to see from (3.3) and (3.4) that

$$Y^{\varepsilon}(t) - Y(t) = \varepsilon \int_0^t e^{(t-s)A} h'(s) ds$$

$$= \varepsilon B^* (BB^*)^{-1} k_1 \int_0^t \phi'(s) e^{\theta_1(s-T)} e^{\theta_1(t-s)} ds + \varepsilon k_2 \int_0^t \psi'(s) e^{\theta_2(s-T)} e^{\theta_2(t-s)} ds$$

$$= \varepsilon \left(\phi(t) e^{\theta_1(t-T)} B^* (BB^*)^{-1} k_1 + \psi(t) e^{\theta_2(t-T)} k_2\right) = \varepsilon \tilde{h}(t),$$

and hence,

$$X^{\varepsilon}(t) - X(t) = \varepsilon \int_0^t B\tilde{h}(s) ds$$
$$= \varepsilon \left(k_1 \int_0^t \phi(r) e^{\theta_1(r-T)} dr + (Bk_2) \int_0^t \psi(r) e^{\theta_2(r-T)} dr \right).$$

So,

X3 (3.7)
$$X^{\varepsilon}(t) - X(t) = \varepsilon \Theta(t), \quad t \in [0, T],$$

and in particular

$$(X^{\varepsilon}(T), Y^{\varepsilon}(T)) = (X(T), Y(T)) + \varepsilon k$$

due to (3.4). Next,

$$\xi_{\varepsilon}(s) = \varepsilon h'(s) + b(s, X(s), Y(s)) - b(s, X^{\varepsilon}(s), Y^{\varepsilon}(s))$$

and

$$R_{\varepsilon} = \exp\left[-\int_0^T \left\langle \sigma^{-1} \xi_{\varepsilon}(s), dW(s) \right\rangle - \frac{1}{2} \int_0^T |\sigma^{-1} \xi_{\varepsilon}(s)|^2 ds\right].$$

We reformulate (3.6) as

$$\begin{cases}
dX^{\varepsilon}(t) = BY^{\varepsilon}(t)dt, & X^{\varepsilon}(0) = x, \\
dY^{\varepsilon}(t) = \sigma dW^{\varepsilon}(t) + \{b(t, X^{\varepsilon}(t), Y^{\varepsilon}(t)) + AY^{\varepsilon}(t)\}dt, & Y^{\varepsilon}(0) = y,
\end{cases}$$

where by (3.5) and (3.7),

$$W^{\varepsilon}(t) := W(t) + \int_0^t \sigma^{-1} \xi_{\varepsilon}(s) \mathrm{d}s, \quad t \in [0, T]$$

is a cylindrical Brownian motion under the weighted probability measure $\mathbb{Q}_{\varepsilon} := R_{\varepsilon}\mathbb{P}$. Since $|\xi_{\varepsilon}|$ is uniformly bounded on [0,T], by the dominated convergence theorem and (3.7), for any $f \in C_b^1(\tilde{\mathbb{H}} \times \mathbb{H})$ we obtain

$$P_{T}(\partial_{k}f) = \lim_{\varepsilon \to 0} \mathbb{E} \frac{f((X(T), Y(T)) + \varepsilon k) - f((X(t), Y(t)))}{\varepsilon}$$

$$= \lim_{\varepsilon \to 0} \mathbb{E} \frac{f((X^{\varepsilon}(T), Y^{\varepsilon}(T))) - R_{\varepsilon}f((X^{\varepsilon}(T), Y^{\varepsilon}(T)))}{\varepsilon}$$

$$= \mathbb{E} \left(f(Z(T)) \lim_{\varepsilon \to 0} \frac{1 - R_{\varepsilon}}{\varepsilon} \right)$$

$$= \mathbb{E} \left(f(Z(T)) \int_{0}^{T} \left\langle \sigma^{-1} \left\{ h'(t) - (\partial_{\Theta(t)}b)(Z(t)) \right\}, dW(t) \right\rangle \right).$$

To apply this result, we present here a specific choice of (ϕ, ψ) such that (3.4) holds:

$$\phi(t) = \frac{e^{\theta_1 T} t(T - t)}{\int_0^T s(T - s) e^{\theta_1 s} ds}, \quad \psi(t) = \frac{e^{\theta_2 (T - t)}}{T} \left(\frac{3t^2}{T} - 2t\right), \quad t \in [0, T].$$

Theorem 3.2. Let $\tilde{\mathbb{H}} = \mathbb{H} = \mathbb{H}$ and $\operatorname{Ker}(B) = \{0\}$. Let $b(t, \cdot) = b$ do not dependent on t such that P_t has an invariant probability measure μ . If

$$\underbrace{ \text{*A2}}_{(x,y) \in \mathbb{H} \times \mathbb{H}} \lim_{r \downarrow 0} \frac{\|b(x+rB^{-1}\tilde{k},y+rk) - b(x,y)\|_{\sigma}}{r} < \infty, \quad (\tilde{k},k) \in (B\mathbb{H}_A) \times \mathbb{H}_A,$$

Then for any $(k_1, k_2) \in (B\mathbb{H}_A) \times \mathbb{H}_A$, the form $(\mathscr{E}_k, C_b^2(\mathbb{H} \times \mathbb{H}))$ is closable in $L^2(\mu)$.

Proof. It suffices to prove for $k=(k_1,k_2)$ such that $B^{-1}k_1$ and k_2 are eigenvectors of A, i.e. $AB^{-1}k_1=\theta_1B^{-1}k_1$ and $Ak_2=\theta_2k_2$ hold for some $\theta_1,\theta_2\in\mathbb{R}$. As explained above there exists T>0 such that (3.4) holds for some $\phi,\psi\in C^{\infty}([0,T])$. Moreover, as explained in the proof of Theorem 2.1, by taking

$$b_{\varepsilon}(s, x, y) = \frac{1}{\sqrt{2\pi\varepsilon}} \int_{\mathbb{R}} b((x, y) + r\Theta(s)) \exp\left[-\frac{r^2}{2\varepsilon}\right] dr, \quad s \in [0, T], (x, y) \in \mathbb{H} \times \mathbb{H}$$

for $\varepsilon > 0$, such that (3.11) holds uniformly in $\varepsilon > 0$ and $s \in [0, T]$ with $b_{\varepsilon}(s, \cdot)$ replacing b, we may and do assume that $b(s, \cdot)$ is Fréchet differentiable along $\Theta(s)$. Then the integration by parts formula in Proposition 3.1 holds, and due to (3.11) we have

$$M_{\cdot,T} := \int_0^T \left\langle (\sigma \sigma^*)^{-1/2} \left\{ h'(t) - (\partial_{\Theta(t)} b(t, \cdot))(Z(t)) \right\}, \ dW(t) \right\rangle \in L^2(\mathbb{P} \times \mu).$$

Therefore, by Proposition 1.1, the form $(\mathcal{E}_k, C_b^2(\mathbb{H} \times \mathbb{H}))$ is closable on $L^2(\mu)$.

Below are typical examples of the stochastic Hamiltonian system with invariant probability measure such that Theorem 3.2 applies.

Example 3.1. Let $\tilde{\mathbb{H}} = \mathbb{H} = \mathbb{H}$.

(1) Let $\mathbb{H} = \mathbb{R}^d$ for some $d \geq 1$. When $\sigma = B = I$, $A \leq -\lambda I$ for some $\lambda > 0$ is a negatively definite $d \times d$ -matrix, and $b(x,y) = A^{-1}\nabla V(x)$ for some $V \in C^2(\mathbb{R}^d)$ such that $\int_{\mathbb{R}^d} e^{-V(x)} dx < \infty$. Then the unique invariant probability measure of P_t is

$$\mu(\mathrm{d}x,\mathrm{d}y) = C\mathrm{e}^{-V(x) + \frac{\lambda}{2}\langle Ay, y \rangle} \mathrm{d}x\mathrm{d}y,$$

where C > 0 is the normalization. See [2, 6, 9] for the study of hypercoercivity of the associated semigroup P_t with respect to μ , as well as [12] for the stronger property of hypercontractivity.

(2) In the infinite-dimensional setting, let $\sigma = B = I$ and A be negatively definite such that A^{-1} is of trace class. Take $b(x,y) = A^{-1}Qx$ for some positively definite self-adjoint operator $\mathbb Q$ on $\mathbb H$ such that Q^{-1} is of trace class and

$$\int_0^1 \|\mathbf{e}^{tA} A^{-1} Q\| \mathrm{d}t < 1.$$

Then it is easy to see that

$$\mu(dx, dy) = N_{Q^{-1}}(dx)N_{-A^{-1}}(dy)$$

is an invariant probability measure.

(3) More generally, let $\sigma = B = I$ and

$$b(x,y) = \tilde{b}(x) := A^{-1}\nabla V(x), \quad (x,y) \in \mathbb{H} \times \mathbb{H}_A$$

for some Fréchet differentiable $V: \mathbb{H}_A \to \mathbb{R}$ such that (3.11) holds. For any $n \geq 1$, let

$$V_n(r) = V \circ \varphi_n(r), \ \varphi_n(r) = \sum_{i=1}^n r_i e_i, \ r = (r_1, \dots, r_n) \in \mathbb{R}^n.$$

If $\int_{\mathbb{R}^n} e^{-V_n(r)} dr < \infty$ and when $n \to \infty$ the probability measure

$$\nu_n(D) := \frac{1}{\int_{\mathbb{R}^n} e^{-V_n(r)} dr} \int_{\varphi_n^{-1}(D)} e^{-V_n(r)} dr, \quad D \in \mathscr{B}(\mathbb{H})$$

converges weakly to some probability measure ν , then $\mu := \nu \times N_{-A^{-1}}$ is an invariant probability measure of P_t . This can be confirmed by (1) and a finite-dimensional approximation argument. Indeed, let $\pi_n : \mathbb{H} \to \mathbb{H}_{A,n}$ be the orthogonal projection, and let $A_n = \pi_n A, W_n = \pi_n W$ and $b_n(x,y) = \pi_n \nabla V(x)$. Let $X_n(t)$ solve the finite-dimensional equation

$$\begin{cases} dX_n(t) = Y_n(t)dt, \\ dY_n(t) = \{A_nY_n(t) + b_n(X_n(t))\}dt + dW_n(t) \end{cases}$$

with $(X_n(0), Y_n(0)) = (\pi_n X(0), \pi_n Y(0))$. Then the proof of [11, Theorem 2.1] yields that for every $t \ge 0$,

$$\lim_{n \to \infty} \mathbb{E}(|X_n(t) - X(t)|^2 + |Y_n(t) - Y(t)|^2) = 0$$

uniformly in the initial data $(X(0), Y(0)) \in \mathbb{H} \times \mathbb{H}$. Thus, letting $P_t^{(n)}$ be the semigroup for $(X_n(t), Y_n(t))$, we have

$$\lim_{n\to\infty} \sup_{(x,y)\in\mathbb{H}\times\mathbb{H}} |P_t^{(n)}f(\pi_n x, \pi_n y) - P_t f(x,y)| = 0, \quad f \in C_b^1(\mathbb{H}\times\mathbb{H}).$$

Combining this with the assertion in (1) and noting that $\nu_n \times (N_{-A^{-1}} \circ \pi_n^{-1}) \to \mu$ weakly as $n \to \infty$, we conclude that μ is an invariant probability measure of P_t .

4 Semilinear SPDEs with delay

For fixed $\tau > 0$, let $\mathscr{C}_{\tau} = C([-\tau, 0]; \mathbb{H})$ be equipped with the uniform norm $\|\eta\|_{\infty} := \sup_{\theta \in [-\tau, 0]} |\eta(\theta)|$. For any $\xi \in C([-\tau, \infty); \mathbb{H})$, we define $\xi \in C([0, \infty); \mathscr{C}_{\tau})$ by letting

$$\xi_t(\theta) = \xi(t+\theta), \quad \theta \in [-\tau, 0], t \ge 0.$$

Consider the following stochastic differential equation with delay:

$$dX(t) = \{AX(t) + b(X_t)\}dt + \sigma dW(t), \quad X_0 \in \mathscr{C}_{\tau},$$

where $(A, \mathcal{D}(A))$ satisfies **(A1)**, σ satisfies **(A3)**, and $b : \mathscr{C}_{\tau} \to \mathbb{H}$ satisfies: for any T > 0 there exists $\gamma \in C((0,T])$ with $\int_0^T \gamma(t) dt < \infty$ such that

$$\boxed{*4} \quad (4.2) \quad \int_0^T \sup_{s \in [0,T]} |\mathrm{e}^{tA} b(s,0)|^2 \mathrm{d}t < \infty, \quad |\mathrm{e}^{tA} (b(s,\xi) - b(s,\eta))|^2 \le \gamma(t) \|\xi - \eta\|_\infty^2, \quad t,s \in [0,T].$$

Then for any initial datum $\xi \in \mathscr{C}_{\tau}$, the equation has a unique mild solution $X^{\xi}(t)$ with $X_0 = \xi$. Let P_t be the Markov semigroup for the segment solution X_t .

$$\mathscr{C}_{\tau}^{1} = \left\{ \eta \in \mathscr{C}_{\tau} : \eta(\theta) \in \mathscr{D}(A) \text{ for } \theta \in [-\tau, 0], \int_{-\tau}^{0} \left(|A\eta(\theta)|^{2} + |\eta'(\theta)|^{2} \right) d\theta < \infty \right\}.$$

The following result is an extension of [10, Theorem 4.1(1)] to the infinite-dimensional setting.

P4.1 Proposition 4.1. For any $\eta \in \mathscr{C}_{\tau}^{1}$ and $T > \tau$, let

$$\Gamma(t) := \begin{cases} \frac{1}{T - \tau} e^{(s + \tau - T)A} \eta(-\tau), & \text{if } s \in [0, T - \tau], \\ \eta'(s - T) - A \eta(s - T), & \text{if } s \in (T - \tau, T], \end{cases}$$

and

$$\Theta(t) := \int_0^{t \vee 0} \Gamma(s) \mathrm{d}s, \quad t \in [-\tau, T].$$

If $b(t,\cdot)$ is Fréchet differentiable along Θ_t for $t \in [0,T]$ such that

$$\sup_{\xi \in \mathscr{C}_{\tau}} \int_{0}^{T} \left\| \Gamma(t) - (\nabla_{\Theta_{t}} b(T, \cdot))(\xi) \right\|_{\sigma}^{2} dt < \infty,$$

then

PP (4.4

$$P_T(\partial_{\eta} f) = \mathbb{E}\left(f(X_T) \int_0^T \left\langle (\sigma \sigma^*)^{-1/2} \left(\Gamma(t) - (\nabla_{\Theta_t} b(t, \cdot))(X_t)\right), dW(t) \right\rangle \right), \quad f \in C_b^1(\mathscr{C}_\tau).$$

Proof. Simply let $\sigma = \sqrt{\sigma \sigma^*}$ as in the proof of Theorem 2.1. For any $\varepsilon \in (0,1)$, let $X^{\varepsilon}(t)$ solve the equation

$$\mathbf{E}^{\mathsf{r}} \quad (4.5) \qquad \mathrm{d}X^{\varepsilon}(t) = \{AX^{\varepsilon}(t) + b(t, X_t) + \varepsilon\Gamma(t)\}\mathrm{d}t + \sigma\mathrm{d}W(t), \quad X_0^{\varepsilon} = X_0.$$

We have

$$X^{\varepsilon}(t) - X(t) = \varepsilon \int_{0}^{t^{+}} e^{(t-s)A} \Gamma(s) ds$$

$$= \frac{\varepsilon t^{+}}{T - \tau} e^{(\tau - T)A} \eta(-\tau) 1_{[-\tau, T - \tau)}(t) + \varepsilon \eta(t - T) 1_{[T - \tau, T]}(t), \quad t \in [-\tau, T].$$

In particular, we have $X_T^{\varepsilon} - X_T = \varepsilon \eta$. To formulate P_T using X_T^{ε} , rewrite (4.5) by

$$dX^{\varepsilon}(t) = \{AX^{\varepsilon}(t) + b(t, X_t^{\varepsilon})\}dt + \sigma dW_{\varepsilon}(t), \quad X_0^{\varepsilon} = X_0,$$

where

$$W_{\varepsilon}(t) := W(t) + \int_0^t \xi_{\varepsilon}(s) ds, \quad \xi_{\varepsilon}(s) := b(s, X_s) - b(s, X_s^{\varepsilon}) + \varepsilon \Gamma(s).$$

By (4.3) and the Girsanov theorem, we see that $\{W_{\varepsilon}(t)\}_{t\in[0,T]}$ is a cylindrical Brownian motion on \mathbb{H} under the probability measure $d\mathbb{Q}_{\varepsilon} := R_{\varepsilon}d\mathbb{P}$, where

$$R_{\varepsilon} := \exp\left[\int_{0}^{T} \left\langle \sigma^{-1}\left(b(t, X_{t}^{\varepsilon}) - b(t, X_{t}) - \varepsilon\Gamma(t)\right), \, dW(t)\right\rangle\right].$$

Then

$$\mathbb{E}(f(X_T)) = P_T f = \mathbb{E}(R_{\varepsilon} f(X_T^{\varepsilon})).$$

Combining this with $X_T^{\varepsilon} = X_T + \varepsilon \eta$ and using (4.6), we arrive at

$$P_{T}(\partial_{\eta}f) = \lim_{\varepsilon \downarrow 0} \frac{1}{\varepsilon} \mathbb{E} \{ f(X_{T} + \varepsilon \eta) - f(X_{T}) \} = \lim_{\varepsilon \downarrow 0} \frac{1}{\varepsilon} \mathbb{E} \{ f(X_{T}^{\varepsilon}) - R_{\varepsilon}f(X_{T}^{\varepsilon}) \}$$
$$= \mathbb{E} \Big(f(X_{T}) \lim_{\varepsilon \downarrow 0} \frac{1 - R_{\varepsilon}}{\varepsilon} \Big) = \mathbb{E} \Big\{ f(X_{T}) \int_{0}^{T} \Big\langle \sigma^{-1} \big(\Gamma(t) - (\nabla_{\Theta_{t}}b(t, \cdot))(X_{t}) \big), dW(t) \Big\rangle \Big\}.$$

Theorem 4.2. Let $b(t,\cdot) = b$ be independent of t such that P_t has an invariant probability measure μ . If $\text{Im}(\sigma) \supset \mathbb{H}_A$ and

then for any $\eta \in \mathscr{C}^1_{\tau} \cap (\cup_{n\geq 1} C([-\tau,0];\mathbb{H}_{A,n}))$, which is dense in \mathscr{C}_{τ} , the form

$$\mathscr{E}_{\eta}(f,g) := \int_{\mathscr{C}_{\tau}} (\partial_{\eta} f)(\partial_{\eta} g) d\mu, \quad f,g \in C_b^2(\mathscr{C}_{\tau})$$

is closable in $L^2(\mu)$.

Proof. For any $\varepsilon \in (0,1)$ let

$$b_{\varepsilon}(t,\xi) = \frac{1}{\sqrt{2\pi\varepsilon}} \int_{\mathbb{R}} b(\xi + r\Theta_t) \exp\left[-\frac{r^2}{2\varepsilon}\right] dr, \quad \xi \in \mathscr{C}_{\tau}.$$

Then $b_{\varepsilon}(t,\cdot)$ is Féchet differentiable along Θ_t and (4.7) holds uniformly in ε with $b_{\varepsilon}(t,\cdot)$ replacing b. Moreover, $\eta \in \mathscr{C}^1_{\tau} \cap \left(\cup_{n \geq 1} C([-\tau,0];\mathbb{H}_n) \right)$ implies that $\Theta_t \in \mathscr{C}^1_{\tau} \cap \left(\cup_{n \geq 1} C([-\tau,0];\mathbb{H}_n) \right)$ and (4.7) holds uniformly in $t \in [0,T]$ and $\varepsilon \in (0,1)$ with Θ_t and $b_{\varepsilon}(t,\cdot)$ replacing η and b respectively. Combining this with $\text{Im}(\sigma) \supset \mathbb{H}_A$, we conclude that (4.3) holds uniformly in ε with b_{ε} replacing b. Therefore, as explained in the proof of Theorem 2.1, we may assume that b is Fréchet differentiable along $\Theta_t, t \in [0,T]$, and by Proposition 4.1 the integration by parts formula (4.4) holds. Moreover, (4.7) implies

$$M_{\cdot,T} := \int_0^T \left\langle (\sigma \sigma^*)^{-1/2} \left(\Gamma(t) - (\nabla_{\Theta_t} b(t, \cdot))(X_t) \right), dW(t) \right\rangle \in L^2(\mathbb{P} \times \mu).$$

Then the proof is finished by Proposition 1.1.

Finally, we introduce the following example to illustrate Theorem 4.2.

Example 4.1. Let $b(\xi) = F(\xi(-\tau)), \xi \in \mathscr{C}_{\tau}$, for some $F \in C_b^1(\mathbb{H})$. If σ is Hilbert-Schmidt and

$$\langle x, Ax + F(y) - F(y') \rangle \le -\lambda_1 |x|^2 + \lambda_2 |y - y'|^2, \quad x, y \in \mathbb{H},$$

for some constants $\lambda_1 > \lambda_2 \geq 0$, then according to [1, Theorem 4.9] P_t has a unique invariant probability measure μ . If moreover $\text{Im}(\sigma) \supset \mathbb{H}_A$ and for any $y \in \mathbb{H}_A$ there exists a constant

$$\limsup_{\varepsilon \downarrow 0} \sup_{x \in \mathbb{H}} \frac{\|F(x + \varepsilon y) - F(x)\|_{\sigma}}{\varepsilon} < \infty,$$

then by Theorem 4.2, for any $\eta \in \mathscr{C}_{\tau}^1 \cap (\bigcup_{n\geq 1} C([-\tau,0];\mathbb{H}_{A,n}))$ the form $(\mathscr{E}_{\eta}, C_b^2(\mathscr{C}_{\tau}))$ is closable on $L^2(\mu)$.

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