

# Remarks on the construction of invariant measures for KdV

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We denote by  $d\mu_s$  a probability (Gaussian) measure on the spaces  $H^{s-1/2-\epsilon}$  for  $\epsilon > 0$

In fact the measure  $d\mu_s = e^{-\|u\|_{H^s}^2} du$  can be constructed in a rigorous way by a finite dimensional approximation via a general Kolmogorov result, and it is in principle a measure on  $C^\infty$ .

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A more concrete representation of the probability space is given by the randomized vector

$$\varphi_s(x, \omega) = \sum_{n \in \mathbf{Z} \setminus \{0\}} \frac{\varphi_n(\omega)}{|n|^s} e^{inx}$$

where  $(\varphi_n(\omega))$  is a sequence of centered complex gaussian variables defined on a probability space  $(\Omega, \mathcal{A}, p)$

Then for every function  $F(u)$  defined on the support of the measure  $d\mu_s = e^{-\|u\|_s^2} du$  we have

$$\int F(u) d\mu_s = \int_{\Omega} F(\varphi_s(\omega)) dp$$

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As a consequence one can deduce that  $d\mu_s(H^{s-1/2}) = 0$  and the space  $H^{s-1/2-\epsilon}$  is support for  $d\mu_s$

We also have  $d\mu_1(H^{1/2}) = 0$  and  $d\mu_1(\mathcal{F}L^{r,1/2}) = 1$  for  $r > 2$  where

$$\|u\|_{\mathcal{F}L^{r,1/2}} = \|\{\hat{u}(n)n^{1/2}\}\|_{l^r}$$

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Let us consider a general Hamiltonian PDE

$$u_t + Au = F(u)$$

$$u(0, x) = u_0(x)$$

$$u(0, t) = u(2\pi, t)$$

with an associated conservation law

$$E_s(u) = \|u\|_{H^s}^2 + R_s(u)$$

For example the KdV, BO, DNLS satisfy this assumption.

Typical probabilistic statements that can be proved for the Cauchy problems above are of the following type.

### Theorem

*The Cauchy problem is G.W.P. for a.e.  $u_0$  w.r.t.  $d\mu_s = e^{-\|u\|_{H^s}^2} du$*

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## Theorem

The measure  $e^{-R_s(u)} e^{-\|u\|_{H^s}^2} d\mathbf{u} = e^{-R_s(u)} d\mu_s$  is invariant along the flow of the PDE.

**Consequence: Poincaré Recurrence Theorem** Let  $T_t$  be a dynamical system on a topological space  $X$  endowed with a finite invariant measure  $d\mu$ , then for a.e.  $d\mu$  initial data in the phase space there is a sequence  $t_n \rightarrow \infty$  such that  $T_{t_n}(u_0) \rightarrow u_0$

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$\Phi_N^t$  is the flow associated with

$$\partial_t u_N + Au_N = \pi_N(Fu_N)$$

$$u_N(0) = \pi_N(u_0)$$

where  $\pi_N$  is the projections on the modes with frequency  $\leq N$  and  $u_N$  are trigonometric polynomials of degree at most  $N$ .

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- 1 There is a suitable Banach space  $X$  which is a support of  $d\mu_s = e^{-\|u\|_{H^s}^2} du$  where the Cauchy problem is L.W.P. (i.e. there is an unique local solution in the space  $\mathcal{C}(0, T(\|u_0\|_X)); X$ )
- 2 Finite dimensional approximation:

$$\|\Phi_N^t(u_0) - \Phi^t(u_0)\|_{L^\infty((0,T);X)} \rightarrow 0 \text{ as } N \rightarrow \infty$$

- 3 The energy  $E_s(\Phi_N^t u_0)$  is (almost) preserved along the evolution as long as  $N \rightarrow \infty$  for a.e.  $u_0 \in X$ .



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## Lemma

Assume that

$$\lim_{N \rightarrow \infty} \|G_N\|_{L^q(d\mu_s)} = 0$$

where

$$G_N(u_0) = \frac{d}{dt} (E_s(\Phi_N^t(u_0)))|_{t=0}$$

Then we have the following:

$$\lim_{N \rightarrow \infty} \sup_{\substack{t \in [0, t_0] \\ A \in \text{Borel} \subset \text{supp}(d\mu_s)}} \left| \frac{d}{dt} \int_{\Phi_N^t(A)} e^{-R_s(\pi_N u)} d\mu_s(u) \right| = 0$$

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## Reduction of the analysis at time $t = 0$

We have

$$\begin{aligned} & \frac{d}{dt} \left( \int_{\phi_N^t(A)} e^{-R_s(\pi_N u)} d\mu \right)_{t=\bar{t}} \\ &= \lim_{h \rightarrow 0} h^{-1} \left( \int_{\Phi_N^{\bar{t}+h}(A)} e^{-R_s(\pi_N u)} d\mu - \int_{\phi_N^{\bar{t}}(A)} e^{-R_s(\pi_N u)} d\mu \right) \\ &= \lim_{h \rightarrow 0} h^{-1} \left( \int_{\Phi_N^h \circ \Phi_N^{\bar{t}}(A)} e^{-R_s(\pi_N u)} d\mu - \int_{\phi_N^{\bar{t}}(A)} e^{-R_s(\pi_N u)} d\mu \right) \end{aligned}$$

and hence

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**Analysis for KdV** We compute concretely  $\frac{d}{dt} E_m(\Phi_N^t(u))|_{t=0}$  where

$$E_m(u) = \int (\partial_x^m u)^2 dx + \int u (\partial_x^{m-1} u)^2 dx + \dots$$

Roughly speaking, in general, it is sufficient to compute (by Leibnitz rule) the time derivative of the density of  $E$  and to replace  $u_t$  by  $\pi_{>N}(F(u^N(t)))$ . More precisely to compute  $\frac{d}{dt} E_m(u^N(t))|_{t=0}$  (for KdV) we get:

$$\begin{aligned} \frac{d}{dt} E_m(u^N(t)) = & \int 2(\partial_x^m u_t^N) \partial_x^m u^N + u_t^N (\partial_x^{m-1} u^N(t))^2 \\ & + 2 \int u^N (\partial_x^{m-1} u^N) \partial_x^{m-1} u_t^N dx + \dots |_{u_t^N = u_{xx}^N + u^N \partial_x u^N - \pi_{>N}(u^N \partial_x u^N)} = \\ & \int \pi_{>N}(u^N \partial_x u^N) (\partial_x^{m-1} u^N)^2 dx + 2 \int u^N (\partial_x^{m-1} u^N) \pi_{>N} \partial_x^{m-1} (u^N \partial_x u^N) dx \end{aligned}$$

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By the computation above the worst term to be estimated is

$$\int u^N (\partial_x^{m-1} u^N) \partial_x^{m-1} \pi_{>N}(u^N \partial_x u^N) dx \Big|_{u^N = \sum_{n \in \mathbf{Z} \setminus \{0\}, |n| < N} \frac{\varphi_n(\omega)}{|n|^m} e^{inx}}$$

then we get

$$\int \sum_{\substack{|j_1|, |j_2|, |j_3|, |j_4| < N \\ |j_3 + j_4| > N}} \frac{\varphi_{j_1}(\omega)}{|j_1|^m} \frac{\varphi_{j_2}(\omega)}{|j_2|} \frac{\varphi_{j_3}(\omega)}{|j_3|^m} \varphi_{j_4}(\omega) e^{i(j_1 + j_2 + j_3 + j_4)x} dx$$

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and hence

by integration  $dx$  we get

$$\dots = \sum_{\substack{|j_1|, |j_2|, |j_3|, |j_4| < N \\ |j_3 + j_4| > N, j_1 + j_2 + j_3 + j_4 = 0}} \frac{\varphi_{j_1}(\omega)}{|j_1|^m} \frac{\varphi_{j_2}(\omega)}{|j_2|} \frac{\varphi_{j_3}(\omega)}{|j_3|^m} \varphi_{j_4}(\omega)$$

By Minkowski

$$\begin{aligned} \|G_N(u)\|_{L^q(d\mu_m)} &= \|G_N(\varphi(\omega))\|_{L^q(dp_\omega)} \\ &\leq \sum_{\substack{|j_1|, |j_2|, |j_3|, |j_4| < N \\ |j_1 + j_2| > N}} \frac{1}{|j_1|^m} \frac{1}{|j_2|} \frac{1}{|j_3|^m} \rightarrow 0 \text{ as } N \rightarrow \infty \end{aligned}$$

Thank you!