Time-splitting for the random nonlinear Schrödinger equation

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Contents



The nonlinear Schrödinger equation with random dispersion

Time-splitting scheme for NLS with random dispersion

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NLS with random dispersion

$$\begin{cases}
\frac{\partial u}{\partial t}(t,x) = i \overset{\circ}{W}(t) \frac{\partial^2 u}{\partial x^2}(t,x) + g(u(t,x)) & \forall (t,x) \in (0,1] \times \mathbb{R} \\
u(t=0,x) = u_0(x)
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where

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Introduction

The white noise W

NLS with random dispersion

$$\begin{cases} \frac{\partial u}{\partial t}(t,x) = i \stackrel{\circ}{W}(t) \frac{\partial^2 u}{\partial x^2}(t,x) + g(u(t,x)) \quad \forall (t,x) \in (0,1] \times \mathbb{R} \\ u(t=0,x) = u_0(x) \end{cases}$$

Roughly speaking, \ddot{W} can be seen as a stationary Gaussian process satisfying for every $t, s \in [0, 1]$,

$$\mathbb{E}(\overset{\circ}{W}(t)) = 0 \quad \& \quad \mathbb{E}(\overset{\circ}{W}(s) \overset{\circ}{W}(t)) = \delta(t-s).$$

It corresponds to the normalized independent infinitesimal increments of a **Brownian motion** *W*, which is a Gaussian process satisfying for every $t, s \in [0, 1]$,

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Sample path of the white noise W

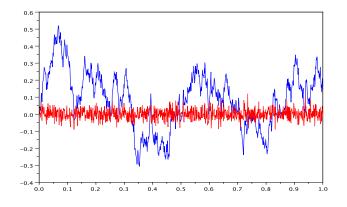


FIGURE: Blue : W; Red : \mathring{W} .

Image: Image:

NLS with random dispersion

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- z is the position on the fiber, τ is the time,
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- G. P. Agrawal Nonlinear fiber optics.
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NLS with random dispersion

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- Existence and uniqueness of solution.
- Time-splitting scheme for simulations.

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Contents



2 The nonlinear Schrödinger equation with random dispersion

Time-splitting scheme for NLS with random dispersion

The linear equation

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$$v(t = t_0, x) = v_0(x)$$

is solved by following the procedure :

• $v_0(x) \xrightarrow{FT} \widehat{v_0}(\xi)$

we solve the Fourier linear equation ∂v
∂t
(t,ξ) = -i w
(t)ξ²v
(t,ξ), which gives v
(t,ξ) = v
0(ξ) exp(-iξ²(W(t) - W(t₀))) (Itô-Stratonovich calculus)
v
(t,ξ) → v(t,x)

$$X(t_0,t)v_0(x)=v(t,x).$$

The linear equation

$$\frac{\partial v}{\partial t}(t,x) = i \overset{\circ}{W}(t) \frac{\partial^2 v}{\partial x^2}(t,x) \quad \forall (t,x) \in (0,1] \times \mathbb{R}$$
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 ∂t
 (t,ξ) = -i w
 (t)ξ²v
 (t,ξ), which gives v
 (t,ξ) = v
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 v
 (t,ξ) → v(t,x)

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NLS with random dispersion

$$\begin{cases} \frac{\partial u}{\partial t}(t,x) = i \overset{\circ}{W}(t) \frac{\partial^2 u}{\partial x^2}(t,x) + g(u(t,x)) \quad \forall (t,x) \in (0,1] \times \mathbb{R} \\ u(t=0,x) = u_0(x) \end{cases}$$

is written in the integral form :

NLS with random dispersion in integral form

$$u(t,x) = X(0,t)u_0(x) + \int_0^t X(\theta,t)g(u(\theta,x))d\theta \qquad \forall (t,x) \in (0,1] \times \mathbb{R}$$

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Let $u_0 \in L_x^2$. We assume that g is Lipschitz. Then, there exists a unique solution $u \in C([0, 1], L_x^2)$ a. s. Moreover, if $u_0 \in H_x^2$, g two times differentiable and its derivatives up the order 2 are bounded, then $u \in C([0, 1], H_x^2)$ a. e.

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Contents



The nonlinear Schrödinger equation with random dispersion

Time-splitting scheme for NLS with random dispersion

We introduce three operators :

• *S* such that $v(t, x) = S(t_0, t)v_0(x)$ solves

$$\frac{\partial v}{\partial t} = i \stackrel{\circ}{W} \frac{\partial^2 v}{\partial x^2} + g(v)$$
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(Remember that
$$u(nh, \cdot) = S((n-1)h, nh)u((n-1)h, \cdot)...)$$

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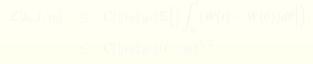
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$$||u_n^h-u(nh,\cdot)||_{L^2(\Omega,L^2(\mathbb{R},\mathbb{C}))}| \leq -C\sum_{j=1}^n \mathcal{L}[(j-1)h,jh,u((j-1)h,\cdot)]$$

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We aim to illustrate the previous result by simulation. We let $\varepsilon(h) = \|u_N^h - u(1, \cdot)\|_{L^2(\Omega, L^2(\mathbb{R}, \mathbb{C}))}$ for every $h = 2^{-m}$ where *m* is integer. The order of convergence will be given by

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Indeed : assume that $\varepsilon(h) \approx h^{\alpha^*}$ for $h \to 0$. Then

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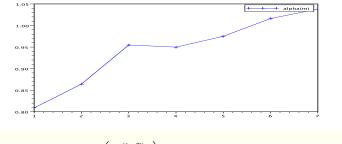
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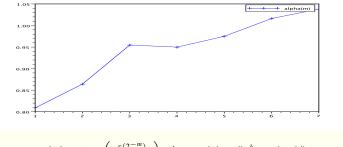
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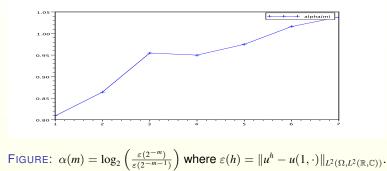
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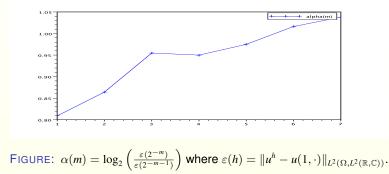
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In order to study more precisely the scheme we establish :

Theorem (M. (preprint 2011))

We assume that $v \in H^4$ and g sufficiently regular. Then for every $t_0 \le t \in [0, 1]$,

$$\|\mathcal{L}[t_0,t,v] - \mathcal{G}(v)\mathcal{I}(t_0,t)\|_{L^2(\Omega,L^2(\mathbb{R},\mathbb{C}))} \le C(t-t_0)^2$$

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We assume that $v \in H^4$ and g sufficiently regular. Then for every $t_0 \le t \in [0, 1]$,

$$\|\mathcal{L}[t_0,t,v] - \mathcal{G}(v)\mathcal{I}(t_0,t)\|_{L^2(\Omega,L^2(\mathbb{R},\mathbb{C}))} \le C(t-t_0)^2$$

with
$$\mathcal{G}(v) = i \frac{\partial^2 g(v)}{\partial x^2} - \left\langle (\nabla g)(v), i \frac{\partial^2 v}{\partial x^2} \right\rangle$$
 and $\mathcal{I}(t_0, t) = \int_{t_0}^t (W_t - W_\theta) d\theta$.

$$\left(\left\langle (\nabla g)(v), i\frac{\partial^2 v}{\partial x^2} \right\rangle = (\partial_1 g)(v) \Re \left(i\frac{\partial^2 v}{\partial x^2} \right) + (\partial_2 g)(v) \Im \left(i\frac{\partial^2 v}{\partial x^2} \right) \right)$$

Previously we established an estimate on the local error :

$$\mathcal{L}[t_0, t, v] = ||S(t_0, t)v - Y(t - t_0)X(t_0, t)v||_{L^2(\Omega, L^2(\mathbb{R}, \mathbb{C}))} \leq C(t - t_0)^{3/2}.$$

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Applying the norm $\|\cdot\|_{L^2}$, we get

$$||u_n^h - u(nh, \cdot)||_{L^2}^2 = \mathcal{E}_1(n, h) + 2\mathcal{R}\mathcal{E}_2(n, h)$$

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Finally, we get

$$egin{array}{rcl} \mathbb{E}(\mathcal{E}_2(n,h)) &=& \mathbb{E}\left(\sum_{j,k=1,j< k}^n \langle E_j^{(n,h)}, E_k^{(n,h)}
angle_{L^2}
ight) \ &\leq& C\sum_{j,k=1,j< k}^n h^4 \leq Ch^2, \end{array}$$

then

$$\left\| u_{n}^{h} - u(nh, \cdot) \right\|_{L^{2}(\Omega, L^{2}(\mathbb{R}, \mathbb{C}))}^{2} = \mathbb{E}[\left\| u_{n}^{h} - u(nh, \cdot) \right\|_{L^{2}}^{2}] \le Ch^{2}$$

This concludes the "proof".

Thank you !

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This concludes the "proof".

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