

# Pointwise convergence to initial data

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**Warning:** Many arbitrarily small  $\varepsilon$  have been identified with 0.

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I will review recent progress for Carleson's question for the Schrödinger equation  $i\partial_t u + \Delta u = 0$  with initial data  $u_0$  in the Sobolev space  $H^s(\mathbb{R}^n)$ . That is, for which  $s$  can we be sure that  $u(x, t) \rightarrow u_0(x)$  as  $t \rightarrow 0$  for almost every  $x \in \mathbb{R}^n$ . First I will present examples, due to Bourgain, Lucà and myself, which show that  $s \geq \frac{1}{2} - \frac{1}{2(n+1)}$  is necessary. We will see that spread out interference-type behaviour becomes a problem when  $n \geq 2$ . I will then present maximal estimates, due to Du, Guth, Li and Zhang, which show that  $s > \frac{1}{2} - \frac{1}{2(n+1)}$  is sufficient. Loosely speaking, these are a consequence of Strichartz-type estimates that improve for spread out solutions. We will also consider the fractal dimension version of the problem and the analogous questions for other PDE.

# Summary

- ▶ Part 1: Set-up and introduction to the PDEs.
- ▶ Part 2: Convergence for the heat equation.
- ▶ Part 3: Decay of the Fourier transform of fractal measures.
- ▶ Part 4: Convergence for the wave equation.
- ▶ Part 5(a): Counterexample for the Schrödinger equation.
- ▶ Part 5(b): Convergence for the Schrödinger equation.

# Part 0:

## Basic properties of the Fourier transform

For Schwartz functions  $f : \mathbb{R}^n \rightarrow \mathbb{C}$  and  $g : \mathbb{R}^n \rightarrow \mathbb{C}$ , we write

$$\widehat{f}(\xi) := \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^n} e^{-ix \cdot \xi} f(x) dx, \quad g^\vee(x) := \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^n} e^{ix \cdot \xi} g(\xi) d\xi,$$

so that

$$f = (\widehat{f})^\vee, \quad (\text{Inversion formula})$$

$$\widehat{\Delta f}(\xi) = -|\xi|^2 \widehat{f}(\xi), \quad \text{where } \Delta := \frac{\partial^2}{\partial x_1^2} + \dots + \frac{\partial^2}{\partial x_n^2},$$

$$(\widehat{f} \widehat{g})^\vee(x) = f * g(x) := \int_{\mathbb{R}^n} f(x-y) g(y) dy,$$

$$\int_{\mathbb{R}^n} f(x) \overline{g(x)} dx = \int_{\mathbb{R}^n} \widehat{f}(\xi) \overline{\widehat{g}(\xi)} d\xi, \quad (\text{Plancherel})$$

and

$$\|f\|_{L^2} := \left( \int_{\mathbb{R}^n} |f(x)|^2 dx \right)^{1/2} = \left( \int_{\mathbb{R}^n} |\widehat{f}(\xi)|^2 d\xi \right)^{1/2} = \|\widehat{f}\|_{L^2}.$$

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so that

$$f = (\widehat{f})^\vee, \quad (\text{Inversion formula})$$

$$\widehat{\Delta f}(\xi) = -|\xi|^2 \widehat{f}(\xi), \quad \text{we write } \varphi(\Delta)f := (\varphi(-|\cdot|^2)\widehat{f})^\vee,$$

$$(\widehat{f} \widehat{g})^\vee(x) = f * g(x) := \int_{\mathbb{R}^n} f(x-y) g(y) dy,$$

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# Part 1:

## PDEs to ODEs using the Fourier transform

# The heat equation

$$\begin{cases} \partial_t u = \Delta u & \text{in } \mathbb{R}^n \times (0, \infty) \\ u(\cdot, 0) = u_0 & \text{in } \mathbb{R}^n. \end{cases}$$

Taking the Fourier transform of the equation we obtain

$$\begin{cases} \partial_t \widehat{u}(\xi, t) = -|\xi|^2 \widehat{u}(\xi, t) \\ \widehat{u}(\xi, 0) = \widehat{u}_0(\xi). \end{cases}$$

Solving the ODE (with fixed  $\xi$ ) this yields

$$\widehat{u}(\xi, t) = e^{-t|\xi|^2} \widehat{u}_0(\xi).$$

Inverting the Fourier transform, we write

$$u(x, t) = e^{t\Delta} u_0(x) := \left( e^{-t|\cdot|^2} \widehat{u}_0 \right)^\vee.$$



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$$u(x, t) = e^{t\Delta} u_0(x) := \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^n} e^{ix \cdot \xi} e^{-t|\xi|^2} \widehat{u}_0(\xi) d\xi.$$

# The Schrödinger equation

$$\begin{cases} \partial_t u = i\Delta u & \text{in } \mathbb{R}^n \times \mathbb{R} \\ u(\cdot, 0) = u_0 & \text{in } \mathbb{R}^n. \end{cases}$$

Taking the Fourier transform of the equation we obtain

$$\begin{cases} \partial_t \widehat{u}(\xi) = -i|\xi|^2 \widehat{u}(\xi, t) \\ \widehat{u}(\xi, 0) = \widehat{u}_0(\xi). \end{cases}$$

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## The wave equation

$$\begin{cases} \partial_{tt}u &= \Delta u & \text{in } \mathbb{R}^n \times \mathbb{R} \\ u(\cdot, 0) &= u_0 & \text{in } \mathbb{R}^n \\ \partial_t u(\cdot, 0) &= u_1 & \text{in } \mathbb{R}^n. \end{cases}$$

Taking the Fourier transform of the equation we obtain

$$\begin{cases} \partial_{tt}\widehat{u}(\xi, t) &= -|\xi|^2\widehat{u}(\xi, t) \\ \widehat{u}(\xi, 0) &= \widehat{u}_0(\xi) \\ \partial_t\widehat{u}(\xi, 0) &= \widehat{u}_1(\xi). \end{cases}$$

Solving the ODE this yields

$$\widehat{u}(\xi, t) = \cos(t|\xi|)\widehat{u}_0(\xi) + \frac{\sin(t|\xi|)}{|\xi|}\widehat{u}_1(\xi).$$

Inverting the Fourier transform, we write

$$u(\cdot, t) = \cos(t\sqrt{-\Delta})u_0 + \frac{\sin(t\sqrt{-\Delta})}{\sqrt{-\Delta}}u_1.$$

## The initial data

The Bessel potential  $G_s$  is defined via its Fourier transform by

$$\widehat{G}_s(\xi) := (1 + |\cdot|^2)^{-s/2}$$

and satisfies

$$G_s(x) \leq c_{n,s} |x|^{-(n-s)}.$$

We take the initial data  $u_0$  in the Bessel potential space

$$H^s(\mathbb{R}^n) := \{ G_s * g : g \in L^2(\mathbb{R}^n) \}$$

with norm

$$\|f\|_{H^s} = \left( \int_{\mathbb{R}^n} (1 + |\xi|^2)^s |\widehat{f}(\xi)|^2 d\xi \right)^{1/2} = \|g\|_{L^2}.$$

## Lemma (Pointwise convergence for smooth data)

Let  $f \in H^s(\mathbb{R}^n)$  with  $s > n/2$ . Then

$$\lim_{t \rightarrow 0} e^{t\Delta} f(x) = f(x) \quad \text{for all } x \in \mathbb{R}^n.$$

**Proof:** Taking  $s < n/2 + 2$ ,

$$\begin{aligned} (2\pi)^{n/2} |e^{t\Delta} f(x) - f(x)| &= \left| \int \widehat{f}(\xi) |\xi|^s \frac{e^{ix \cdot \xi} (e^{-t|\xi|^2} - 1)}{|\xi|^s} d\xi \right| \\ &\leq \left( \int |\widehat{f}(\xi)|^2 |\xi|^{2s} d\xi \right)^{1/2} \left( \int \frac{|e^{-t|\xi|^2} - 1|^2}{|\xi|^{2s}} d\xi \right)^{1/2} \\ &\leq t^{\frac{s-n/2}{2}} \|f\|_{H^s} \left( \int \frac{|e^{-|y|^2} - 1|^2}{|y|^{2s}} dy \right)^{1/2} \\ &= t^{\frac{s-n/2}{2}} \|f\|_{H^s} \left( \int \frac{\min\{|y|^2, 1\}^2}{|y|^{2s}} dy \right)^{1/2} \\ &\leq C_s t^{\frac{s-n/2}{2}} \|f\|_{H^s}. \end{aligned}$$

A similar calculation works for the Schrödinger equation.



# Hausdorff measure

For  $A \subseteq \mathbb{R}^n$  a Borel set,

$$\mathcal{H}_\delta^\alpha(A) := \inf \left\{ \sum_i \delta_i^\alpha : A \subset \bigcup_i B(x_i, \delta_i), \quad \delta_i < \delta \right\}.$$

## Definition

The  $\alpha$ -Hausdorff measure of  $A$  is

$$\mathcal{H}^\alpha(A) := \lim_{\delta \rightarrow 0} \mathcal{H}_\delta^\alpha(A).$$

# Hausdorff dimension

## Remark

*There exists a unique  $\alpha_0 \in [0, n]$  such that*

$$\mathcal{H}^\alpha(A) = \begin{cases} \infty & \text{if } \alpha < \alpha_0 \\ 0 & \text{if } \alpha > \alpha_0. \end{cases}$$

## Definition

$\alpha_0$  is the Hausdorff dimension of the set  $A$ :

$$\dim(A) := \alpha_0.$$

## Definition (Frostman measures)

We say a positive Borel measure  $\mu$  with  $\text{supp}(\mu) \subset B(0, 1)$  is  $\alpha$ -dimensional if

$$c_\alpha(\mu) := \sup_{\substack{x \in \mathbb{R}^n \\ r > 0}} \frac{\mu(B(x, r))}{r^\alpha} < \infty.$$

$$\begin{aligned} E_{\alpha'}(\mu) &:= \iint \frac{d\mu(x)d\mu(y)}{|x-y|^{\alpha'}} = \int \sum_{j=-1}^{\infty} \int_{|x-y| \sim 2^{-j}} \frac{d\mu(x)}{|x-y|^{\alpha'}} d\mu(y) \\ &\leq \int \sum_{j=-1}^{\infty} c_\alpha(\mu) 2^{-j\alpha} 2^{j\alpha'} d\mu(y) \\ &\lesssim c_\alpha^2(\mu) < \infty \quad \text{if } \alpha > \alpha'. \end{aligned}$$

## Lemma (Frostman)

Let  $A \subset \mathbb{R}^n$  be a Borel set. The following are equivalent:

- ▶  $\mathcal{H}^\alpha(A) = 0$ ;
- ▶  $\mu(A) = 0$  for all  $\alpha$ -dimensional  $\mu$ .



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**Lemma (Frostman)**  $\mu(A) \leq \sum_i \mu(B(x_i, \delta_i)) \leq c_\alpha(\mu) \sum_i \delta_i^\alpha$

Let  $A \subset \mathbb{R}^n$  be a Borel set. The following are equivalent:

- ▶  $\mathcal{H}^\alpha(A) = 0$ ;
- ▶  $\mu(A) = 0$  for all  $\alpha$ -dimensional  $\mu$ .

# Control of singularities

## Lemma

Let  $0 < s < n/2$  and  $\alpha > n - 2s$ . Then, for all  $\alpha$ -dimensional  $\mu$ ,

$$\|f\|_{L^1(d\mu)} \lesssim c_\alpha(\mu) \|f\|_{H^s}.$$

Thus if  $f \in H^s(\mathbb{R}^n)$ , then

$$\mu\{x \in \mathbb{R}^n : f(x) = \infty\} = 0 \quad \forall \alpha\text{-dimensional } \mu$$

whenever  $\alpha > n - 2s$ , so that by Frostman's lemma,

$$\mathcal{H}^\alpha\{x \in \mathbb{R}^n : f(x) = \infty\} = 0,$$

whenever  $\alpha > n - 2s$ , so that

$$\dim\{x \in \mathbb{R}^n : f(x) = \infty\} \leq n - 2s.$$

# Control of singularities

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$$\|f\|_{L^1(d\mu)} \lesssim c_\alpha(\mu) \|f\|_{H^s}.$$

**Proof:** Writing  $f = G_s * g = ((1 + |\cdot|^2)^{-s/2} \widehat{g})^\vee$ , it suffices to prove

$$\|G_s * g\|_{L^1(d\mu)} \lesssim \sqrt{E_{n-2s}(\mu)} \|g\|_{L^2(\mathbb{R}^n)}.$$

By Fubini's theorem and the Cauchy-Schwarz inequality,

$$\begin{aligned} \|G_s * g\|_{L^1(d\mu)} &\leq \int \int G_s(x-y) d\mu(x) |g(y)| dy \\ &\leq \|G_s * \mu\|_{L^2(\mathbb{R}^n)} \|g\|_{L^2(\mathbb{R}^n)}. \end{aligned}$$

Thus it remains to prove that

$$\|G_s * \mu\|_{L^2(\mathbb{R}^n)}^2 \lesssim E_{n-2s}(\mu).$$

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By two applications of Plancherel's theorem,

$$\begin{aligned} \|G_s * \mu\|_{L^2(\mathbb{R}^n)}^2 &= \|(1 + |\cdot|^2)^{-s/2} \widehat{\mu}\|_{L^2(\mathbb{R}^n)}^2 \\ &= \int \widehat{\mu}(\xi) (1 + |\xi|^2)^{-s} \overline{\widehat{\mu}(\xi)} d\xi \\ &\leq \int \mu * G_{2s}(y) d\mu(y) \\ &\leq c_{n,s} \iint \frac{d\mu(x) d\mu(y)}{|x - y|^{n-2s}} = c_{n,s} E_{n-2s}(\mu), \end{aligned}$$

where we used that  $((1 + |\cdot|^2)^{-s})^\vee = G_{2s} \leq c_{n,s} |\cdot|^{-(n-2s)}$ . □

## Optimality of the control of singularities lemma

If  $\dim(A) = \alpha$  with  $\alpha < n - 2s$ , then we can take a  $\gamma$  such that

$$\alpha < \gamma < n - 2s.$$

Then

$$\mathbf{1}_{B(0,1)} \text{dist}(\cdot, A)^{-\gamma} \in L^2(\mathbb{R}^n)$$

but on the other hand

$$u_0 := G_s * \left[ \mathbf{1}_{B(0,1)} \text{dist}(\cdot, A)^{-\gamma} \right] = \infty \quad \text{on } A.$$

So there is initial data  $u_0 \in H^s(\mathbb{R}^n)$  which is singular on a set of dimension  $\alpha < n - 2s$ .

## Proposition (Maximal estimates imply convergence)

Let  $\alpha > \alpha_0 \geq n - 2s$ . Suppose that, for all  $\alpha$ -dimensional  $\mu$ ,

$$\left\| \sup_{0 < t < 1} |u(\cdot, t)| \right\|_{L^1(d\mu)} \leq C_\mu \|u_0\|_{H^s}.$$

Then, for all  $u_0 \in H^s$ ,

$$\dim \left\{ x \in \mathbb{R}^n \quad \lim_{t \rightarrow 0} u(t, x) \neq u_0(x) \right\} \leq \alpha_0.$$

**Proof:** We are required to prove that for all  $\alpha > \alpha_0$ ,

$$\mathcal{H}^\alpha \left\{ x \in \mathbb{R}^n \quad \lim_{t \rightarrow 0} u(t, x) \neq u_0(x) \right\} = 0$$

whenever  $u_0 \in H^s$ . By Frostman's lemma, this follows by showing

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$$\mu \left\{ x \in \mathbb{R}^n \quad \lim_{t \rightarrow 0} |u(t, x) - u_0(x)| > \lambda \right\} = 0, \quad \forall \lambda > 0,$$

whenever  $\mu$  is  $\alpha$ -dimensional.

Take  $h \in H^{n/2+1}$  such that  $\|u_0 - h\|_{H^s} < \varepsilon$ , and note that

$$|u(x, t) - u_0(x)| \leq |u(x, t) - u_h(x, t)| + |u_h(x, t) - h(x)| + |h(x) - u_0(x)|,$$

where  $u_h$  denotes the solution with initial data  $h$ . Then

$$\begin{aligned} & \mu\{x : \limsup_{t \rightarrow 0} |u(x, t) - u_0(x)| > \lambda\} \\ & \leq \mu\{x : \limsup_{t \rightarrow 0} |u(x, t) - u_h(x, t)| > \lambda/3\} \\ & + \mu\{x : \limsup_{t \rightarrow 0} |u_h(x, t) - h(x)| > \lambda/3\} \\ & + \mu\{x : \limsup_{t \rightarrow 0} |h(x) - u_0(x)| > \lambda/3\}. \end{aligned}$$



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$$\begin{aligned} & \mu\{x : \limsup_{t \rightarrow 0} |u(x, t) - u_0(x)| > \lambda\} \lambda \\ & \leq \mu\{x : \sup_{0 < t < 1} |u_{u_0-h}(x, t)| > \lambda/3\} \lambda \\ & + 0 \\ & + \mu\{x : |h(x) - u_0(x)| > \lambda/3\} \lambda. \end{aligned}$$

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$$\begin{aligned} & \mu\{x : \limsup_{t \rightarrow 0} |u(x, t) - u_0(x)| > \lambda\} \lambda \\ & \leq \left\| \sup_{0 < t < 1} |u_{u_0-h}| \right\|_{L^1(d\mu)} \\ & + 0 \\ & + \mu\{x : |h(x) - u_0(x)| > \lambda/3\} \lambda. \end{aligned}$$

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We use [the maximal estimate](#) for the first term and the third term can be bounded by the [control of singularities lemma](#) so that

$$\mu\{x : \lim_{t \rightarrow 0} |u(x, t) - u_0(x)| > \lambda\} \lambda \leq C_\mu \|u_0 - h\|_{H^s(\mathbb{R}^n)} \leq C_\mu \varepsilon.$$

Letting  $\varepsilon$  tend to zero, then  $\lambda$  tend to zero, we are done. □



# Part 2:

## Convergence for the heat equation

## Theorem (Maximal estimate for the heat equation)

Let  $0 < s < n/2$  and  $\alpha > n - 2s$ . Then, for all  $\alpha$ -dimensional  $\mu$ ,

$$\left\| \sup_{0 < t < 1} |e^{t\Delta} f| \right\|_{L^1(d\mu)} \leq C_\mu \|f\|_{H^s}.$$

**Proof:** By linearising the operator, it will suffice to prove

$$\left| \iint e^{ix \cdot \xi} e^{-t(x)|\xi|^2} \widehat{f}(\xi) d\xi w(x) d\mu(x) \right|^2 \lesssim E_{n-2s}(\mu) \|f\|_{H^s}^2,$$

whenever  $t : \mathbb{R}^n \rightarrow (0, \infty)$  and  $w : \mathbb{R}^n \rightarrow \mathbb{S}^1$  are measurable.

Now, by Fubini and Cauchy-Schwarz, the LHS is bounded by

$$\int |\widehat{f}(\xi)|^2 (1 + |\xi|^2)^s d\xi \int \left| \int e^{ix \cdot \xi} e^{-t(x)|\xi|^2} w(x) d\mu(x) \right|^2 \frac{d\xi}{(1 + |\xi|^2)^s}.$$

Squaring out the integral, it will suffice to show that

$$\iiint e^{i(x-y) \cdot \xi} e^{-(t(x)+t(y))|\xi|^2} \frac{d\xi}{(1 + |\xi|^2)^s} w(x) w(y) d\mu(x) d\mu(y) \lesssim E_{n-2s}(\mu).$$

Thus, it remains to prove that, for  $0 < s < n/2$ ,

$$\left| \int e^{i(x-y)\cdot\xi} e^{-(t(x)+t(y))|\xi|^2} \frac{d\xi}{(1+|\xi|^2)^s} \right| \lesssim \frac{1}{|x-y|^{n-2s}}$$

uniformly for all choices of  $t(x), t(y) > 0$ . Recalling that

$$(e^{-\lambda|\cdot|^2})^\vee = \frac{1}{\lambda^{n/2}} e^{-|\cdot|^2/\lambda} \quad \text{and} \quad ((1+|\cdot|^2)^{-s})^\vee =: G_{2s} \leq c_{n,s} |\cdot|^{-(n-2s)},$$

this would follow from

$$\frac{1}{\lambda^{n/2}} e^{-|\cdot|^2/\lambda} * \frac{1}{|\cdot|^{n-2s}} \lesssim \frac{1}{|\cdot|^{n-2s}}.$$

uniformly in  $\lambda$ . By changing variables, this is equivalent to

$$e^{-|\cdot|^2} * \frac{1}{|\cdot|^{n-2s}} \lesssim \frac{1}{|\cdot|^{n-2s}},$$

which can be checked by direct calculation. □

## Corollary

Let  $f \in H^s$  with  $0 < s < n/2$ . Then

$$\dim \left\{ x \in \mathbb{R}^n : \lim_{t \rightarrow 0} e^{t\Delta} f(x) \neq f(x) < \infty \right\} \leq n - 2s.$$

As we saw before,  $f \in H^s$  can be singular on a set of dimension less than  $n - 2s$  and so this is optimal.

# Part 3:

## Decay for the Fourier transform of fractal measures

$\widehat{\delta_{x_1=0}}(\xi_1, \bar{\xi}) = \int_{\mathbb{R}^{n-1}} e^{-i\bar{x}\cdot\bar{\xi}} d\bar{x}$  is independent of  $\xi_1$ .

Thus, the Fourier transform of certain  $(n - 1)$ -dimensional measures do not decay in every direction.

But perhaps they decay on average.....

Let  $\beta_n(\alpha)$  denote the supremum of the numbers  $\beta$  for which

$$\frac{1}{|\mathbb{S}_r^{n-1}|} \int_{\mathbb{S}_r^{n-1}} |\widehat{\mu}(\omega)|^2 d\sigma_r(\omega) \lesssim C_\mu (1+r)^{-\beta}$$

whenever  $r > 0$  and  $\mu$  is  $\alpha$ -dimensional and supported in  $B(0, 1)$ .

### Question (Mattila (1987))

What is  $\beta_n(\alpha)$  ?

Equivalently  $\beta_n(\alpha)$  is the supremum of the numbers  $\beta$  for which

$$\|(gd\sigma_r)^\vee\|_{L^1(d\mu)} \lesssim C_\mu r^{\frac{n-1}{2}} (1+r)^{-\beta/2} \|g\|_{L^2(\mathbb{S}_r^{n-1})}.$$

## Best known results

$$\beta_2(\alpha) = \begin{cases} \alpha, & \alpha \in (0, 1/2], \\ 1/2, & \alpha \in [1/2, 1], \\ \alpha/2, & \alpha \in [1, 2], \end{cases} \quad \begin{array}{l} \text{Mattila (1987)} \\ \\ \text{Wolff (1999).} \end{array}$$

$$\beta_3(\alpha) \geq \begin{cases} \alpha, & \alpha \in (0, 1], \\ 1, & \alpha \in [1, \frac{3}{2}], \\ \alpha - \frac{\alpha}{3}, & \alpha \in [\frac{3}{2}, 3], \end{cases} \quad \begin{array}{l} \text{Mattila (1987)} \\ \\ \text{Du-Guth-Ou-Wang-Wilson-Zhang.} \end{array}$$

$$\beta_{n \geq 4}(\alpha) \geq \begin{cases} \alpha, & \alpha \in (0, \frac{n-1}{2}], \\ \frac{n-1}{2}, & \alpha \in [\frac{n-1}{2}, \frac{n}{2}], \\ \alpha - \frac{\alpha}{n}, & \alpha \in [\frac{n}{2}, n], \end{cases} \quad \begin{array}{l} \text{Mattila (1987)} \\ \\ \text{Du-Zhang.} \end{array}$$

## The initial data

The Riesz potential  $I_s$  is defined via its Fourier transform by

$$\widehat{I_s}(\xi) := |\cdot|^{-s}$$

and satisfies

$$I_s(x) \leq c_{n,s}|x|^{-(n-s)}.$$

We now take the initial data in the Riesz potential space

$$\dot{H}^s(\mathbb{R}^n) := \{ I_s * g : g \in L^2(\mathbb{R}^n) \}$$

with norm

$$\|f\|_{\dot{H}^s} = \left( \int_{\mathbb{R}^n} |\xi|^{2s} |\widehat{f}(\xi)|^2 d\xi \right)^{1/2} = \|g\|_{L^2}.$$



## Lemma (Bridging lemma)

Let  $f \in \dot{H}^s(\mathbb{R}^n)$  with  $0 < s < n/2$  and  $\beta_n(\alpha) > n - 2s$ . Then

$$\dim \left\{ x \in \mathbb{R}^n : \lim_{t \rightarrow 0} e^{it(-\Delta)^{m/2}} f(x) \neq f(x) \right\} \leq \alpha.$$

**Proof:** It will suffice to prove, for all  $\alpha$ -dimensional  $\mu$ ,

$$\left\| \sup_{0 < t < 1} |e^{it(-\Delta)^{m/2}} f| \right\|_{L^1(d\mu)} \lesssim C_\mu \|f\|_{\dot{H}^s(\mathbb{R}^n)}.$$

Writing in polar coordinates,

$$\begin{aligned} |e^{it(-\Delta)^{m/2}} f(x)| &= \left| \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^n} e^{-it|\xi|^m} e^{ix \cdot \xi} \widehat{f}(\xi) d\xi \right| \\ &= \left| \frac{1}{(2\pi)^{n/2}} \int_0^\infty e^{-itr^m} \int_{\mathbb{S}_r^{n-1}} e^{ix \cdot \omega} \widehat{f}(\omega) d\sigma_r(\omega) dr \right| \\ &\leq \frac{1}{(2\pi)^{n/2}} \int_0^\infty \left| \int_{\mathbb{S}_r^{n-1}} e^{ix \cdot \omega} \widehat{f}(\omega) d\sigma_r(\omega) \right| dr. \end{aligned}$$

## Lemma (Bridging lemma)

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**Proof:** It will suffice to prove, for all  $\alpha$ -dimensional  $\mu$ ,

$$\left\| \sup_{0 < t < 1} |e^{it(-\Delta)^{m/2}} f| \right\|_{L^1(d\mu)} \lesssim C_\mu \|f\|_{\dot{H}^s(\mathbb{R}^n)}.$$

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$$|e^{it(-\Delta)^{m/2}}f(x)| \lesssim \int_0^\infty |(\widehat{f}d\sigma_r)^\vee(x)| dr,$$

so that, by Fubini,

$$\left\| \sup_{0 < t < 1} |e^{it(-\Delta)^{m/2}}f| \right\|_{L^1(d\mu)} \lesssim \int_0^\infty \|(\widehat{f}d\sigma_r)^\vee\|_{L^1(d\mu)} dr.$$

By the dual version of the Mattila inequality,

$$\|(\widehat{f}d\sigma_r)^\vee\|_{L^1(d\mu)} \leq C_\mu r^{\frac{n-1}{2}} (1+r)^{-\beta/2} \|\widehat{f}\|_{L^2(\mathbb{S}_r^{n-1})}.$$

for all  $\beta < \beta_n(\alpha)$ , so that

$$\left\| \sup_{0 < t < 1} |e^{it(-\Delta)^{m/2}}f| \right\|_{L^1(d\mu)} \leq C_\mu \int_0^\infty r^{\frac{n-1}{2}} (1+r)^{-\beta/2} \|\widehat{f}\|_{L^2(\mathbb{S}_r^{n-1})} dr.$$

Finally, by Cauchy-Schwarz,

$$\begin{aligned} &\leq C_\mu \left( \int_0^\infty (1+r)^{-\beta} r^{n-1-2s} dr \right)^{1/2} \left( \int_0^\infty \|\widehat{f}\|_{L^2(\mathbb{S}_r^{n-1})}^2 r^{2s} dr \right)^{1/2} \\ &\leq C_\mu \|f\|_{\dot{H}^s(\mathbb{R}^n)}, \end{aligned}$$

where for the final inequality we must take  $\beta > n - 2s$ .



# Part 4:

## Convergence for the wave equation

Recall that, with initial data  $u(\cdot, 0) = u_0$  and  $\partial_t u(\cdot, 0) = u_1$ , the solution to the wave equation satisfies

$$\begin{aligned}
 \widehat{u}(\xi, t) &= \cos(t|\xi|)\widehat{u}_0(\xi) + \frac{\sin(t|\xi|)}{|\xi|}\widehat{u}_1(\xi) \\
 &= \frac{1}{2}(e^{it|\xi|} + e^{-it|\xi|})\widehat{u}_0(\xi) + \frac{1}{2}\frac{(e^{it|\xi|} - e^{-it|\xi|})}{i|\xi|}\widehat{u}_1(\xi) \\
 &= e^{it|\xi|}\frac{1}{2}\left(\widehat{u}_0(\xi) + \frac{\widehat{u}_1(\xi)}{i|\xi|}\right) + e^{-it|\xi|}\frac{1}{2}\left(\widehat{u}_0(\xi) - \frac{\widehat{u}_1(\xi)}{i|\xi|}\right) \\
 &=: e^{it|\xi|}\widehat{f}_+(\xi) + e^{-it|\xi|}\widehat{f}_-(\xi).
 \end{aligned}$$

With this notation, we can write

$$u(\cdot, t) = e^{it(-\Delta)^{1/2}} f_+ + e^{-it(-\Delta)^{1/2}} f_-.$$

If the initial data is in  $\dot{H}^s \times \dot{H}^{s-1}$ , both  $f_+$  and  $f_-$  belong to  $\dot{H}^s$ .

Thus convergence of  $e^{it(-\Delta)^{1/2}} f$  to  $f$  for all  $f \in \dot{H}^s$  implies convergence of  $u(\cdot, t)$  to  $u_0$  for all  $(u_0, u_1) \in \dot{H}^s \times \dot{H}^{s-1}$ .

Now  $\beta(\alpha) \geq \frac{n-1}{n}\alpha$ , so if  $\alpha > \frac{n}{n-1}(n-2s)$  then  $\beta(\alpha) > n-2s$ .  
Thus, by the bridging lemma,

### Corollary

*Let  $u$  be a solution to the Schrödinger equation with initial data in  $H^s$  or to the wave equation with initial data in  $\dot{H}^s \times \dot{H}^{s-1}$ . Then*

$$\dim \left\{ x \in \mathbb{R}^n : \lim_{t \rightarrow 0} u(x, t) \neq u_0(x) \right\} \leq \frac{n}{n-1}(n-2s).$$

In particular,

### Corollary

*Let  $u$  be a solution to the Schrödinger equation with initial data in  $H^1$  or to the wave equation with initial data in  $\dot{H}^1 \times L^2$ . Then*

$$\dim \left\{ x \in \mathbb{R}^n : \lim_{t \rightarrow 0} u(x, t) \neq u_0(x) \right\} < n-1.$$

# Part 5: The Schrödinger equation

# Lebesgue a.e. convergence for Schrödinger

In 1979, **Carleson** asked for which  $s$  is it true that

$$\lim_{t \rightarrow 0} e^{it\Delta} f(x) = f(x), \quad \text{a.e. } x \in \mathbb{R}^n, \quad \forall f \in H^s(\mathbb{R}^n)?$$

Improvements made by:

**Carleson** (1979), **Dahlberg-Kenig** (1982),  
**Carbery/Cowling** (1985/83), **Sjölin/Vega** (1987/88),  
**Bourgain** (1991/92), **Moyua-Vargas-Vega** (1996/99),  
**Tao-Vargas-Vega** (1998), **Tao-Vargas** (2000), **Tao** (2003),  
**Lee** (2006), **Bourgain** (2013), **Lucà-R.** (2015), **Bourgain** (2016),  
**Du-Guth-Li** (2017), **Du-Guth-Li-Zhang** (2018), **Du-Zhang** (2018).



# Lebesgue a.e. convergence for Schrödinger

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Best known sufficient condition for convergence:

- ▶  $s \geq 1/4$  with  $n = 1$  (**Carleson**);
- ▶  $s > 1/3$  with  $n = 2$  (**Du-Guth-Li**);
- ▶  $s > \frac{1}{2} - \frac{1}{2(n+1)}$  with  $n \geq 3$  (**Du-Zhang**).

Best known necessary condition for convergence:

- ▶  $s \geq 1/4$  with  $n = 1$  (**Dahlberg-Kenig**);
- ▶  $s \geq \frac{1}{2} - \frac{1}{2(n+1)}$  with  $n \geq 2$  (**Bourgain**).

Part 5(a):

$s \geq \frac{1}{2} - \frac{1}{2(n+1)}$  is necessary for  
Lebesgue a.e. convergence

# Proof

## Lemma (Nikišin-Stein maximal principle)

$$\lim_{t \rightarrow 0} e^{it\Delta} f(x) = f(x), \quad \text{a.e. } x \in \mathbb{R}^n,$$

for all  $f \in H^s(\mathbb{R}^n)$  if and only if there is a constant  $C$  such that

$$\left\| \sup_{0 < t < 1} |e^{it\Delta} f| \right\|_{L^2(B(0,1))} \leq C \|f\|_{H^s(\mathbb{R}^n)}$$

for all  $f \in H^s(\mathbb{R}^n)$ .

So it suffices to show that, if

$$\left\| \sup_{0 < t < 1} |e^{it\Delta} f| \right\|_{L^2(B(0,1))} \lesssim R^s \|f\|_2,$$

whenever  $\text{supp } \widehat{f} \subset \{\xi : |\xi| \leq R\}$ , then  $s \geq \frac{1}{2} - \frac{1}{2(n+1)} = \frac{n}{2(n+1)}$ .

## The concentrated example

Consider initial data  $f$  defined by

$$\widehat{f}(\xi) = \mathbf{1}_{|\xi| \leq \frac{1}{10}R^{1/2}} \quad \text{so that} \quad \|f\|_2 \leq R^{n/4}.$$

Then, if  $(x, t) \in X \times T$ , where

$$X := B(0, R^{-1/2}) \quad \text{and} \quad T := (0, R^{-1}],$$

there is no cancellation in the integral:

$$|e^{it\Delta} f(x)| = \left| \frac{1}{(2\pi)^{n/2}} \int_{|\xi| \leq \frac{1}{10}R^{1/2}} e^{ix \cdot \xi} e^{-it|\xi|^2} d\xi \right| \geq cR^{n/2}.$$

# The travelling concentrated example

Instead Dahlberg-Kenig took

$$f_{dk}(x) = e^{i\frac{1}{2}x\cdot\theta} f(x),$$

where  $\theta \in \mathbb{R}^n$ , so that

$$|e^{it\Delta} f_{dk}(x)| = |e^{it\Delta} f(x - t\theta)| \geq cR^{n/4}$$

whenever

$$x \in X + t\theta \quad \text{and} \quad t \in T = (0, R^{-1}).$$

This yields

$$\sup_{0 < t < 1} |e^{it\Delta} f_{dk}(x)| \geq cR^{n/2}$$

whenever

$$x \in \bigcup_{t \in T} X + t\theta.$$

When  $n = 1$ , we can take  $\theta = R$ , so that

$$(0, 1) \subset \bigcup_{t \in T} X + t\theta.$$

## Conclusion that $s \geq 1/4$ is necessary when $n = 1$

Plugging into the maximal estimate,

$$\left\| \sup_{0 < t < 1} |e^{it\Delta} f_{dk}| \right\|_{L^2(0,1)} \leq CR^s \|f_{dk}\|_2,$$

and recalling that when  $x \in (0, 1)$ ,

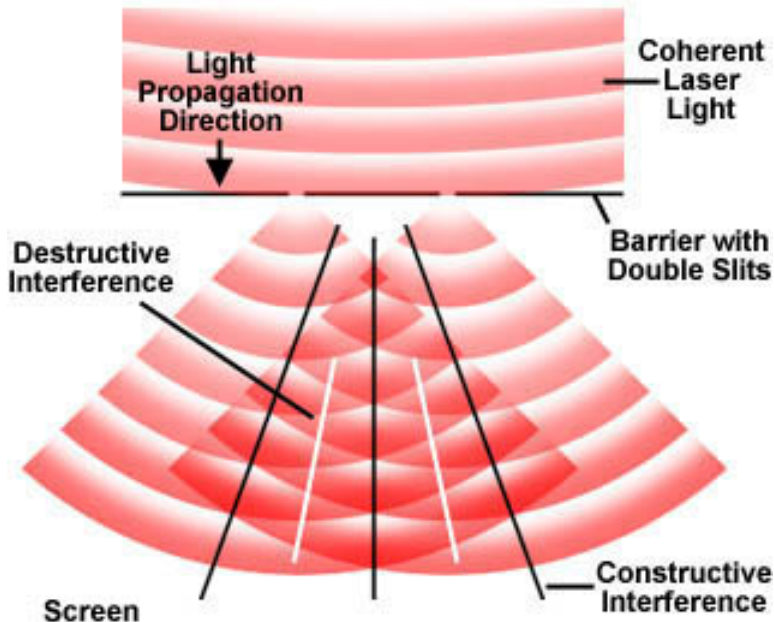
$$\sup_{0 < t < 1} |e^{it\Delta} f_{dk}(x)| \geq cR^{1/2} \quad \text{and} \quad \|f_{dk}\|_2 \leq R^{1/4},$$

we obtain

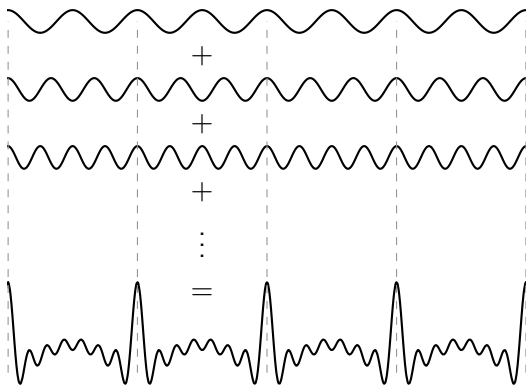
$$cR^{1/4} \leq CR^s.$$

Letting  $R \rightarrow \infty$ , we see that  $s \geq 1/4$ . □

# Young's Double Slit Experiment



# Constructive interference with different frequencies





# The Barceló-Bennett-Carbery-Ruiz-Vilela example

Consider the frequencies

$$\Omega := \left\{ \xi \in 2\pi R^{1-\kappa} \mathbb{Z}^n : |\xi| \leq R \right\} + B\left(0, \frac{1}{10}\right),$$

for  $0 < \kappa < 1$ ,

and initial data defined by

$$\widehat{f_{bbcrv}} = \mathbf{1}_\Omega,$$

so that

$$\|f_{bbcrv}\|_2 = \sqrt{|\Omega|} \leq R^{\frac{n\kappa}{2}}.$$

This was originally considered in the context of **Mattila's** question regarding decay of the Fourier transform of measures.

## Periodic constructive interference

The interference pattern reappears periodically for a short time:

$$|e^{it\Delta} f_{bbcrv}(x)| \geq c|\Omega|,$$

whenever  $(x, t) \in X \times T$ ,

where

$$X := \{x \in R^{\kappa-1}\mathbb{Z}^n : |x| \leq 1\} + B(0, R^{-1}),$$

and

$$T := \left\{ t \in \frac{1}{2\pi} R^{2(\kappa-1)}\mathbb{Z} : 0 < t < R^{-1} \right\}.$$

## Periodic constructive interference

In order to avoid cancellation in the integral

$$|e^{it\Delta} f_{bbcrv}(x)| = \left| \frac{1}{(2\pi)^{n/2}} \int_{\Omega} e^{ix \cdot \xi} e^{-it|\xi|^2} d\xi \right| \geq c|\Omega|,$$

this time  $X$  is in some sense the dual-set of  $\Omega$ :

$$x \cdot \xi \in (R^{\kappa-1}\mathbb{Z}^n) \cdot (2\pi R^{1-\kappa}\mathbb{Z}^n) = 2\pi\mathbb{Z}.$$

and  $T$  is some sense the dual-set of  $\Omega \cdot \Omega$ :

$$t\xi \cdot \xi \in \left( \frac{1}{2\pi} R^{2(\kappa-1)}\mathbb{Z} \right) (2\pi R^{1-\kappa}\mathbb{Z}^n) \cdot (2\pi R^{1-\kappa}\mathbb{Z}^n) = 2\pi\mathbb{Z}.$$

# Periodic constructive interference

Thus

$$|e^{it\Delta} f_{bbcrv}(x)| \geq c|\Omega|$$

whenever  $(x, t) \in X \times T$ .

But the interference always reappears in the same places so

$$\sup_{0 < t < 1} |e^{it\Delta} f_{bbcrv}(x)| \geq c|\Omega|$$

only for  $x \in X$ .

## The travelling interference example

Instead we take

$$f_\theta(x) = e^{i\frac{1}{2}x \cdot \theta} f(x),$$

where  $\theta \in \mathbb{R}^n$ , so that

$$|e^{it\Delta} f_\theta(x)| = |e^{it\Delta} f(x - t\theta)|,$$

which yields

$$\sup_{0 < t < 1} |e^{it\Delta} f_\theta(x)| \geq c|\Omega|$$

whenever

$$x \in \bigcup_{t \in T} X + t\theta.$$

If  $n = 1$  and  $\kappa < 1/3$ , we can take  $\theta = R^\kappa$  so that

$$(0, 1) \subset \bigcup_{t \in T} X + t\theta.$$

## Lemma (Lucà-R.)

If  $0 < \kappa < \frac{1}{n+2}$ , then there exists  $\theta \in \mathbb{R}^n$  such that

$$B(0, 1/2) \subset \bigcup_{t \in T} X + t\theta.$$

This is optimal in the sense that it is not true for  $\kappa > \frac{1}{n+2}$ .

After scaling and quotienting out  $\mathbb{Z}^n$ , this follows from **quantitative ergodic theory** on the torus  $\mathbb{T}^n$ .

## Lemma (Lucà-R.)

Let  $0 < \delta < 1$ . Then, there exists  $\theta \in \mathbb{S}^{n-1}$  such that for all  $y \in \mathbb{T}^n$  there is a  $t \in R^\delta \mathbb{Z} \cap (0, R)$  such that

$$\|y - t\theta\| \leq R^{-\frac{1-\delta}{n}}.$$

## Conclusion that $s \geq \frac{n}{2(n+2)}$ is necessary

Plugging into the maximal estimate,

$$\left\| \sup_{0 < t < 1} |e^{it\Delta} f_\theta| \right\|_{L^2(B(0,1))} \leq CR^s \|f_\theta\|_2,$$

and recalling that when  $x \in B(0, 1/2)$ ,

$$\sup_{0 < t < 1} |e^{it\Delta} f_\theta(x)| \geq c|\Omega| \quad \text{and} \quad \|f_\theta\|_2 = \sqrt{|\Omega|},$$

we obtain

$$c\sqrt{|\Omega|} \leq CR^s.$$

Then as  $|\Omega| \geq R^{n\kappa}$ , this yields

$$cR^{\frac{n\kappa}{2}} \leq CR^s.$$

Letting  $\kappa \rightarrow \frac{1}{n+2}$  and  $R \rightarrow \infty$ , we see that  $s \geq \frac{n}{2(n+2)}$ .



## Combining the examples

Writing  $x = (x_1, \bar{x}) \in \mathbb{R}^n$ , we consider

$$f(x) = f_{dk}(x_1)f_\theta(\bar{x})$$

with  $\kappa < \frac{1}{2(n+1)}$  and  $\theta \in \mathbb{R}^{n-1}$ .

Note that

$$e^{it\Delta}f(x) = e^{it\Delta}f_{dk}(x_1)e^{it\Delta}f_\theta(\bar{x}).$$

In order to make the first factor large, we must take  $t$  near to  $x_1/R$ .

Thus we do not have as many good times as before.

However, we have taken fewer waves than before (smaller  $\kappa$ ).

By the ergodic lemma we can still find a  $\theta \in \mathbb{R}^{n-1}$  and enough good  $t$ 's (near to  $x_1/R$ ), such that the integral of  $e^{it\Delta}f_\theta(\bar{x})$  has no cancellation for all  $\bar{x} \in B(0, 1/2)$ .



## Conclusion that $s \geq \frac{n}{2(n+1)}$ is necessary

Plugging into the maximal estimate,

$$\left\| \sup_{0 < t < 1} |e^{it\Delta} f_{dk} e^{it\Delta} f_{\theta}| \right\|_{L^2((0,1) \times B(0,1))} \leq CR^s \|f_{dk}\|_2 \|f_{\theta}\|_2,$$

and recalling that when  $(x_1, \bar{x}) \in (0, 1) \times B(0, 1/2)$ ,

$$\sup_{0 < t < 1} |e^{it\Delta} f_{dk} e^{it\Delta} f_{\theta}| \geq cR^{1/2} |\Omega| \quad \text{and} \quad \|f_{dk}\|_2 \|f_{\theta}\|_2 \leq R^{1/4} \sqrt{|\Omega|}$$

we obtain

$$cR^{1/4} \sqrt{|\Omega|} \leq CR^s.$$

Then as  $|\Omega| \geq cR^{(n-1)\kappa}$ , we see that

$$s \geq 1/4 + \frac{(n-1)\kappa}{2}.$$

Finally we let  $\kappa \rightarrow \frac{1}{2(n+1)}$ , so that  $s \geq \frac{n+1}{4(n+1)} + \frac{n-1}{4(n+1)} = \frac{n}{2(n+1)}.$   $\square$

Part 5(b):

$s > \frac{n}{2(n+1)}$  is sufficient for  
Lebesgue a.e. convergence.

## Proof

By summing a geometric series, it suffices to show

$$\int_{B(0,1)} \sup_{0 < t < 1} |e^{it\Delta} f(x)|^2 dx \lesssim R^{\frac{n}{n+1}} \|f\|_2^2.$$

whenever  $\text{supp } \widehat{f} \subset \{\xi \in \mathbb{R}^n : R \leq |\xi| \leq 2R\}$ .

By scaling, this can be rewritten

$$\int_{B(0,R)} \sup_{0 < t < R^2} |e^{it\Delta} f(x)|^2 dx \lesssim R^{\frac{n}{n+1}} \|f\|_2^2.$$

whenever  $\text{supp } \widehat{f} \subset \{\xi \in \mathbb{R}^n : 1 \leq |\xi| \leq 2\}$ .

By [Lee's](#) temporal localisation lemma, this would follow from

$$\int_{B(0,R)} \sup_{0 < t < R} |e^{it\Delta} f(x)|^2 dx \lesssim R^{\frac{n}{n+1}} \|f\|_2^2.$$

whenever  $\text{supp } \widehat{f} \subset \{\xi \in \mathbb{R}^n : 1 \leq |\xi| \leq 2\}$ .

Covering  $B(0, R) \times [0, R]$  by disjoint cubes  $Q \times I$  of sidelength 1,

$$\begin{aligned} \int_{B(0,R)} \sup_{0 < t < R} |e^{it\Delta} f(x)|^2 dx &\lesssim \sum_Q \int_Q \int_0^R |e^{it\Delta} f(x)|^2 dt dx \\ &\lesssim \sum_{Q,I} \left( \int_{Q \times I} |e^{it\Delta} f(x)|^{p_n} dx dt \right)^{\frac{2}{p_n}}, \end{aligned}$$

where  $p_n = \frac{2(n+1)}{n-1}$ .

Summing over all the cubes ( $\lesssim R^{n+1}$ ) for which

$$\int_{Q \times I} |e^{it\Delta} f(x)|^{p_n} dx dt \lesssim R^{-(n+1)} \|f\|_2^{p_n},$$

we get a good enough bound.

On the other hand, as  $|e^{it\Delta} f(x)| \leq \|\widehat{f}\|_1 \leq \|f\|_2$ , we have

$$\int_{Q \times I} |e^{it\Delta} f(x)|^{p_n} dx dt \lesssim \|f\|_2^{p_n}.$$

## The pigeonhole principle

We can divide the remaining cubes  $Q \times I$  into  $(n+1) \log R$  classes  $Q_j$  for which

$$2^{-j-1} \|f\|_2^{p_n} < \int_{Q \times I} |e^{it\Delta} f(x)|^{p_n} dx dt \leq 2^{-j} \|f\|_2^{p_n}.$$

Now, leaving only a single  $Q \times I$  for each  $Q$ , we have

$$\int_{B(0,R)} \sup_{0 < t < R} |e^{it\Delta} f(x)|^2 dx \lesssim \sum_j \sum_{Q \times I \in Q_j} \int_{Q \times I} |e^{it\Delta} f(x)|^2 dx dt$$

so we can find a single  $j$  for which

$$\int_{B(0,R)} \sup_{0 < t < R} |e^{it\Delta} f(x)|^2 dx \lesssim \log R \sum_{Q \times I \in Q_j} \int_{Q \times I} |e^{it\Delta} f(x)|^2 dx dt.$$

## Theorem (Spread-improving Strichartz estimate)

Let  $p_n = \frac{2(n+1)}{n-1}$ . Then

$$\left( \sum_{Q \times I \in \mathcal{Q}_j} \int_{Q \times I} |e^{it\Delta} f(x)|^{p_n} dx dt \right)^{1/p_n} \lesssim (\#\mathcal{Q}_j)^{-\frac{1}{n+1}} R^{\frac{n}{2(n+1)}} \|f\|_2$$

Using this, the proof is completed by

$$\begin{aligned} & \left( \sum_{Q \times I \in \mathcal{Q}_j} \int_{Q \times I} |e^{it\Delta} f(x)|^2 dx dt \right)^{1/2} \\ & \leq \left| \bigcup_{Q \times I \in \mathcal{Q}_j} Q \times I \right|^{\frac{1}{n+1}} \left( \sum_{Q \times I \in \mathcal{Q}_j} \int_{Q \times I} |e^{it\Delta} f(x)|^{p_n} dx dt \right)^{1/p_n} \\ & \lesssim (\#\mathcal{Q}_j)^{\frac{1}{n+1}} (\#\mathcal{Q}_j)^{-\frac{1}{n+1}} R^{\frac{n}{2(n+1)}} \|f\|_2 \\ & \lesssim R^{\frac{n}{2(n+1)}} \|f\|_2. \end{aligned}$$



# An ingredient for spread-improving Strichartz: Decoupling

## Theorem (Bourgain-Demeter)

Let  $q_d = \frac{2(d+2)}{d}$  and write  $f = \sum_{\tau} f_{\tau}$ , where  $\widehat{f_{\tau}}$  are supported on pieces of diameter  $R^{-1/2}$ . Then

$$\left( \int_{B(0,R)} |e^{it\Delta} f(x)|^{q_d} dx \right)^{\frac{1}{q_d}} \lesssim \left( \sum_{\tau} \left( \int_{B(0,R)} |e^{it\Delta} f_{\tau}(x)|^{q_d} dx \right)^{\frac{2}{q_d}} \right)^{\frac{1}{2}}.$$

This is used in  $d = n - 1$  dimensions after a dimension reduction.

## Part 5(c):

# Refined convergence for the Schrödinger equation



# Maximal estimate for the Schrödinger equation

## Theorem

Let  $n/4 \leq s < n/2$  and  $\alpha > n - 2s$ . Then, for all  $\alpha$ -dimensional  $\mu$ ,

$$\left\| \sup_{0 < t < 1} |e^{it\Delta} f| \right\|_{L^1(d\mu)} \leq C_\mu \|f\|_{\dot{H}^s}.$$

**Proof:** By the same proof as for the heat equation, one finally arrives to the inequality

$$\left| e^{-i|\cdot|^2} * \frac{1}{|\cdot|^{n-2s}} \right| \leq C_{n-2s} \frac{1}{|\cdot|^{n-2s}},$$

This can also be shown to be true by more difficult direct calculation as long as  $n/4 \leq s < n/2$ . □

## Corollary

Let  $f \in H^s$  with  $n/4 \leq s < n/2$ . Then

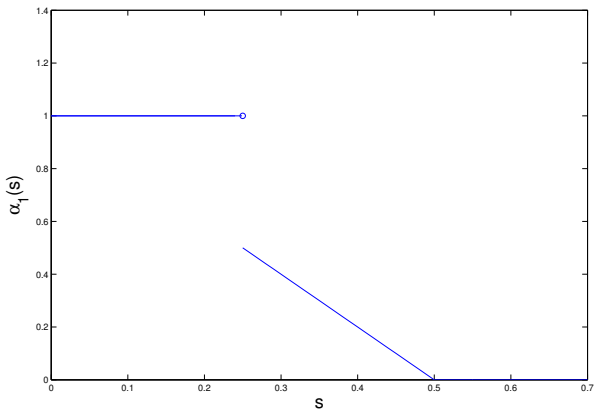
$$\dim \left\{ x \in \mathbb{R}^n \quad \lim_{t \rightarrow 0} e^{it\Delta} f(x) \neq f(x) \right\} \leq n - 2s.$$

Again this is sharp in the range  $s \geq n/4$ .

We cannot go below this regularity in one dimension due to the necessary condition of **Dahlberg-Kenig**.

In the next section we will see that neither can we go below this regularity in higher dimensions using a fractal version of the **Lucà-R.**-necessary condition.

$$\alpha_n(s) := \sup_{f \in H^s(\mathbb{R}^n)} \dim \left\{ x \in \mathbb{R}^n \mid \lim_{t \rightarrow 0} e^{it\Delta} f(x) \neq f(x) \right\}$$



$$\alpha_n(s) = n - 2s, \quad n/4 \leq s \leq n/2.$$

What about lower regularity ( $s < n/4$ ) in higher dimensions?

## Best known bounds in higher dimensions

$$\alpha_n(s) \leq \begin{cases} n & , \quad s \in [0, \frac{n}{2(n+1)}) \\ n + 1 - \frac{2(n+1)s}{n} & , \quad s \in [\frac{n}{2(n+1)}, \frac{n}{4}) \\ n - 2s & , \quad s \in [\frac{n}{4}, \frac{n}{2}] \end{cases} \quad \begin{array}{l} \text{(Du-Guth-Li, Du-Zhang)} \\ \text{(Barceló-Bennett-Carbery-R.)} \end{array}$$

$$\alpha_n(s) \geq \begin{cases} n & , \quad s \in [0, \frac{n}{2(n+1)}) \quad \text{(Dahlberg-Kenig, Bourgain)} \\ n + \frac{n}{n-1} - \frac{2(n+1)s}{n-1} & , \quad s \in [\frac{n}{2(n+1)}, \frac{n+1}{8}) \quad \text{(Lucà-R.)} \\ n + 1 - \frac{2(n+2)s}{n} & , \quad s \in [\frac{n+1}{8}, \frac{n}{4}) \quad \text{(Lucà-R.)} \\ n - 2s & , \quad s \in [\frac{n}{4}, \frac{n}{2}] \quad \text{(Žubrinčić)} \end{cases}$$

$$\alpha_n(s) \geq n + 1 - \frac{2(n+2)s}{n} \quad \text{when} \quad \frac{n}{2(n+2)} \leq s \leq \frac{n}{4}$$

This follows from:

### Theorem (Lucà-R.)

Let  $n/2 \leq \alpha \leq n$  and suppose that

$$\dim \left\{ x \in \mathbb{R}^n \quad \lim_{t \rightarrow 0} e^{it\Delta} f(x) \neq f(x) \right\} < \alpha$$

whenever  $f \in H^s(\mathbb{R}^n)$ . Then

$$s \geq \frac{n}{2(n+2)} (n - \alpha + 1).$$

## Proof

The **Nikišin-Stein** maximal principle does not hold in this context, and so we first give a direct proof of the Lebesgue measure result.

We consider a sum of the previous initial data

$$f := \sum_{j>1} f_{\theta_j}, \quad \theta_j \in \mathbb{S}^{n-1},$$

where we take  $R = 2^j$  and normalise in a different way, so that

$$f_{\theta_j}(x) := e^{i\frac{1}{2}\theta_j \cdot x} f_j(x), \quad \widehat{f}_j = 2^{-j(n\kappa - \varepsilon)} \chi_{\Omega_j},$$

$$\Omega_j := \left\{ \xi \in 2\pi 2^{j(1-\kappa)} \mathbb{Z}^n : |\xi| \leq 2^j \right\} + B(0, \frac{1}{10}).$$

Note that  $|\Omega_j| \simeq 2^{jn\kappa}$ , so that  $\|f_j\|_{H^s} \simeq 2^{-j\frac{n\kappa}{2} + j\varepsilon + js}$ .

Then if  $s < \frac{n\kappa}{2} - \varepsilon$  we can sum so that  $f \in H^s$ .

To generalise to the fractal case we will take  $\frac{1}{n+2} \leq \kappa < \frac{n-\alpha+1}{n+2}$ .

By the previous calculations, for all  $x \in E_j := \cup_{t \in T_j} X_j + t\theta_j$ , where

$$X_j := \{x \in 2^{j(\kappa-1)}\mathbb{Z}^n : |x| \leq 2\} + B(0, 2^{-j}),$$

$$T_j := \left\{ t \in \frac{1}{2\pi} 2^{2j(\kappa-1)}\mathbb{Z} : 0 < t < 2^{-j} \right\},$$

there is a  $t_j(x) \in T_j$  such that  $|e^{it_j(x)\Delta} f_{\theta_j}(x)| \gtrsim 2^{j\varepsilon}$ .

One can also show (essentially) that  $|e^{it_j(x)\Delta} \sum_{k \neq j} f_{\theta_k}(x)| \leq C$ .

If  $\kappa < \frac{1}{n+2}$ , then  $B(0, 1/2) \subset \bigcap_{j>1} E_j$ , and we are done.

If  $\kappa \geq \frac{1}{n+2}$ , we consider the limit set

$$\limsup_{j \rightarrow \infty} E_j := \bigcap_{j>1} \bigcup_{k>j} E_k$$

and prove that this is  $\alpha$ -dimensional.

For this we use that the limit is ' $\alpha$ -Hausdorff dense'.

## Falconer's density theorem

Consider the Hausdorff content  $\mathcal{H}_\infty^\alpha$  defined by

$$\mathcal{H}_\infty^\alpha(E) := \inf \left\{ \sum_i \delta_i^\alpha : E \subset \bigcup_i B(x_i, \delta_i) \right\}.$$

Theorem (Falconer (1985))

Suppose that, for all balls  $B_r \subset B(0, 1)$  of radius  $r$ ,

$$\liminf_{j \rightarrow \infty} \mathcal{H}_\infty^\alpha(E_j \cap B(x, r)) \geq cr^\alpha. \quad (\dagger)$$

Then  $\dim(\limsup_{j \rightarrow \infty} E_j) \geq \alpha$ .

The proof is completed by checking the density condition  $(\dagger)$  with  $E_j = \bigcup_{t \in T_j} X_j + t\theta_j$  using a variant of the ergodic lemma.  $\square$