

Pointwise convergence to initial data

Keith Rogers



August 28, 2016

Summary

- ▶ Part 1: Set-up and introduction to the PDEs.
- ▶ Part 2: Convergence for the heat equation.
- ▶ Part 3: Convergence for the Schrödinger equation.
- ▶ Part 4: Counterexample for the Schrödinger equation.
- ▶ Part 5: Decay of the Fourier transform of fractal measures.
- ▶ Part 6: Convergence for the wave equation.

Summary

- ▶ Part 1: Set-up and introduction to the PDEs.
- ▶ Part 2: Convergence for the heat equation.
- ▶ Part 3: Convergence for the Schrödinger equation.
- ▶ Part 4: Counterexample for the Schrödinger equation.
- ▶ Part 5: Decay of the Fourier transform of fractal measures.
- ▶ Part 6: Convergence for the wave equation.

Summary

- ▶ Part 1: Set-up and introduction to the PDEs.
- ▶ Part 2: Convergence for the heat equation.
- ▶ Part 3: Convergence for the Schrödinger equation.
- ▶ Part 4: Counterexample for the Schrödinger equation.
- ▶ Part 5: Decay of the Fourier transform of fractal measures.
- ▶ Part 6: Convergence for the wave equation.

Summary

- ▶ Part 1: Set-up and introduction to the PDEs.
- ▶ Part 2: Convergence for the heat equation.
- ▶ Part 3: Convergence for the Schrödinger equation.
- ▶ Part 4: Counterexample for the Schrödinger equation.
- ▶ Part 5: Decay of the Fourier transform of fractal measures.
- ▶ Part 6: Convergence for the wave equation.

Summary

- ▶ Part 1: Set-up and introduction to the PDEs.
- ▶ Part 2: Convergence for the heat equation.
- ▶ Part 3: Convergence for the Schrödinger equation.
- ▶ Part 4: Counterexample for the Schrödinger equation.
- ▶ Part 5: Decay of the Fourier transform of fractal measures.
- ▶ Part 6: Convergence for the wave equation.

Summary

- ▶ Part 1: Set-up and introduction to the PDEs.
- ▶ Part 2: Convergence for the heat equation.
- ▶ Part 3: Convergence for the Schrödinger equation.
- ▶ Part 4: Counterexample for the Schrödinger equation.
- ▶ Part 5: Decay of the Fourier transform of fractal measures.
- ▶ Part 6: Convergence for the wave equation.

Summary

- ▶ Part 1: Set-up and introduction to the PDEs.
- ▶ Part 2: Convergence for the heat equation.
- ▶ Part 3: Convergence for the Schrödinger equation.
- ▶ Part 4: Counterexample for the Schrödinger equation.
- ▶ Part 5: Decay of the Fourier transform of fractal measures.
- ▶ Part 6: Convergence for the wave equation.

Part 1:

Set-up and introduction to the PDEs

The heat equation

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

Taking the Fourier transform of the equation we obtain

$$\begin{cases} \partial_t \hat{u}(\xi) = -|\xi|^2 \hat{u}(\xi) \\ \hat{u}(\xi) = \hat{u}_0(\xi). \end{cases}$$

Solving the ODE this yields

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

Inverting the Fourier transform, we write

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

The heat equation

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

Taking the Fourier transform of the equation we obtain

$$\begin{cases} \partial_t \hat{u}(\xi) = -|\xi|^2 \hat{u}(\xi) \\ \hat{u}(\xi) = \hat{u}_0(\xi). \end{cases}$$

Solving the ODE this yields

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

Inverting the Fourier transform, we write

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

The heat equation

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

Taking the Fourier transform of the equation we obtain

$$\begin{cases} \partial_t \hat{u}(\xi) = -|\xi|^2 \hat{u}(\xi) \\ \hat{u}(\xi) = \hat{u}_0(\xi). \end{cases}$$

Solving the ODE this yields

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

Inverting the Fourier transform, we write

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

The heat equation

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

Taking the Fourier transform of the equation we obtain

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

Solving the ODE this yields

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

Inverting the Fourier transform, we write

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

The heat equation

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

Taking the Fourier transform of the equation we obtain

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

Solving the ODE this yields

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

Inverting the Fourier transform, we write

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

The Schrödinger equation

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

Taking the Fourier transform of the equation we obtain

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

Solving the ODE this yields

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

Inverting the Fourier transform, we write

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

The Schrödinger equation

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

Taking the Fourier transform of the equation we obtain

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

Solving the ODE this yields

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

Inverting the Fourier transform, we write

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

The Schrödinger equation

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

Taking the Fourier transform of the equation we obtain

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

Solving the ODE this yields

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

Inverting the Fourier transform, we write

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

The Schrödinger equation

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

Taking the Fourier transform of the equation we obtain

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

Solving the ODE this yields

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

Inverting the Fourier transform, we write

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

The Schrödinger equation

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

Taking the Fourier transform of the equation we obtain

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

Solving the ODE this yields

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

Inverting the Fourier transform, we write

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

The wave equation

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

Taking the Fourier transform of the equation we obtain

$$\begin{cases} \partial_{tt}\hat{u}(\xi) = -|\xi|^2\hat{u}(\xi) \\ \hat{u}(\xi) = \hat{u}_0(\xi) \\ \partial_t\hat{u}(\xi) = \hat{u}_1(\xi). \end{cases}$$

Solving the ODE this yields

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

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 wave equation

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

Taking the Fourier transform of the equation we obtain

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

Solving the ODE this yields

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

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 wave equation

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

Taking the Fourier transform of the equation we obtain

$$\begin{cases} \partial_{tt}\hat{u}(\xi) = -|\xi|^2\hat{u}(\xi) \\ \hat{u}(\xi) = \hat{u}_0(\xi) \\ \partial_t\hat{u}(\xi) = \hat{u}_1(\xi). \end{cases}$$

Solving the ODE this yields

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

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 wave equation

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

Taking the Fourier transform of the equation we obtain

$$\begin{cases} \partial_{tt}\hat{u}(\xi) = -|\xi|^2\hat{u}(\xi) \\ \hat{u}(\xi) = \hat{u}_0(\xi) \\ \partial_t\hat{u}(\xi) = \hat{u}_1(\xi). \end{cases}$$

Solving the ODE this yields

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

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 wave equation

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

Taking the Fourier transform of the equation we obtain

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

Solving the ODE this yields

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

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

We take the initial data u_0 in the Bessel potential space

$$\begin{aligned} H^s(\mathbb{R}^n) &:= (1 - \Delta)^{-s/2} L^2(\mathbb{R}^n) \\ &:= \{ f : \widehat{f} = (1 + |\cdot|^2)^{-s/2} \widehat{g}, \quad \widehat{g} \in L^2(\mathbb{R}^n) \} \end{aligned}$$

with norm

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

or in the Riesz potential space

$$\begin{aligned} \dot{H}^s(\mathbb{R}^n) &:= (-\Delta)^{-s/2} L^2(\mathbb{R}^n) \\ &:= \{ f : \widehat{f} = |\cdot|^{-s} \widehat{g}, \quad \widehat{g} \in L^2(\mathbb{R}^n) \}, \end{aligned}$$

with norm

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

The initial data

We take the initial data u_0 in the Bessel potential space

$$\begin{aligned} H^s(\mathbb{R}^n) &:= (1 - \Delta)^{-s/2} L^2(\mathbb{R}^n) \\ &:= \{ f : \widehat{f} = (1 + |\cdot|^2)^{-s/2} \widehat{g}, \quad \widehat{g} \in L^2(\mathbb{R}^n) \} \end{aligned}$$

with norm

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

or in the Riesz potential space

$$\begin{aligned} \dot{H}^s(\mathbb{R}^n) &:= (-\Delta)^{-s/2} L^2(\mathbb{R}^n) \\ &:= \{ f : \widehat{f} = |\cdot|^{-s} \widehat{g}, \quad \widehat{g} \in L^2(\mathbb{R}^n) \}, \end{aligned}$$

with norm

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

The initial data

We take the initial data u_0 in the Bessel potential space

$$\begin{aligned} H^s(\mathbb{R}^n) &:= (1 - \Delta)^{-s/2} L^2(\mathbb{R}^n) \\ &:= \{ f : \widehat{f} = (1 + |\cdot|^2)^{-s/2} \widehat{g}, \quad \widehat{g} \in L^2(\mathbb{R}^n) \} \end{aligned}$$

with norm

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

or in the Riesz potential space

$$\begin{aligned} \dot{H}^s(\mathbb{R}^n) &:= (-\Delta)^{-s/2} L^2(\mathbb{R}^n) \\ &:= \{ f : \widehat{f} = |\cdot|^{-s} \widehat{g}, \quad \widehat{g} \in L^2(\mathbb{R}^n) \}, \end{aligned}$$

with norm

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

The initial data

We take the initial data u_0 in the Bessel potential space

$$\begin{aligned} H^s(\mathbb{R}^n) &:= (1 - \Delta)^{-s/2} L^2(\mathbb{R}^n) \\ &:= \{ f : \widehat{f} = (1 + |\cdot|^2)^{-s/2} \widehat{g}, \quad \widehat{g} \in L^2(\mathbb{R}^n) \} \end{aligned}$$

with norm

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

or in the Riesz potential space

$$\begin{aligned} \dot{H}^s(\mathbb{R}^n) &:= (-\Delta)^{-s/2} L^2(\mathbb{R}^n) \\ &:= \{ f : \widehat{f} = |\cdot|^{-s} \widehat{g}, \quad \widehat{g} \in L^2(\mathbb{R}^n) \}, \end{aligned}$$

with norm

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

The initial data

We take the initial data u_0 in the Bessel potential space

$$\begin{aligned} H^s(\mathbb{R}^n) &:= (1 - \Delta)^{-s/2} L^2(\mathbb{R}^n) \\ &:= \{ f : \widehat{f} = (1 + |\cdot|^2)^{-s/2} \widehat{g}, \quad \widehat{g} \in L^2(\mathbb{R}^n) \} \end{aligned}$$

with norm

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

or in the Riesz potential space

$$\begin{aligned} \dot{H}^s(\mathbb{R}^n) &:= (-\Delta)^{-s/2} L^2(\mathbb{R}^n) \\ &:= \{ f : \widehat{f} = |\cdot|^{-s} \widehat{g}, \quad \widehat{g} \in L^2(\mathbb{R}^n) \}, \end{aligned}$$

with norm

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

The initial data

We take the initial data u_0 in the Bessel potential space

$$\begin{aligned} H^s(\mathbb{R}^n) &:= (1 - \Delta)^{-s/2} L^2(\mathbb{R}^n) \\ &:= \{ f : \widehat{f} = (1 + |\cdot|^2)^{-s/2} \widehat{g}, \quad \widehat{g} \in L^2(\mathbb{R}^n) \} \end{aligned}$$

with norm

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

or in the Riesz potential space

$$\begin{aligned} \dot{H}^s(\mathbb{R}^n) &:= (-\Delta)^{-s/2} L^2(\mathbb{R}^n) \\ &:= \{ f : \widehat{f} = |\cdot|^{-s} \widehat{g}, \quad \widehat{g} \in L^2(\mathbb{R}^n) \}, \end{aligned}$$

with norm

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

Lemma (Pointwise convergence for smooth data)

Let u solve the heat or Schrödinger equation with $u_0 \in \dot{H}^s(\mathbb{R}^n)$ with $n/2 < s < n/2 + 2$. Then

$$\lim_{t \rightarrow 0} u(x, t) = u_0(x) \quad \text{for all } x \in \mathbb{R}^n.$$

Proof: $\widehat{u}_0 = |\cdot|^{-s} \widehat{g}$ with $g \in L^2$

$$\begin{aligned} (2\pi)^{n/2} |e^{t\Delta} u_0(x) - u_0(x)| &= \left| \int \frac{\widehat{g}(\xi) e^{ix \cdot \xi} (e^{-t|\xi|^2} - 1)}{|\xi|^s} d\xi \right| \\ &\leq \|\widehat{g}\|_2 \left(\int \frac{|e^{-t|\xi|^2} - 1|^2}{|\xi|^{2s}} d\xi \right)^{1/2} \\ &= t^{s/2 - n/4} \|g\|_2 \left(\int \frac{|e^{-|y|^2} - 1|^2}{|y|^{2s}} dy \right)^{1/2} \\ &= t^{s/2 - n/4} \|f\|_{\dot{H}^s} \left(\int \frac{\min\{|y|^2, 1\}^2}{|y|^{2s}} dy \right)^{1/2} \\ &\leq C_s t^{s/2 - n/4} \|f\|_{\dot{H}^s}. \end{aligned}$$

The same calculation works for the Schrödinger equation

Lemma (Pointwise convergence for smooth data)

Let u solve the heat or Schrödinger equation with $u_0 \in \dot{H}^s(\mathbb{R}^n)$ with $n/2 < s < n/2 + 2$. Then

$$\lim_{t \rightarrow 0} u(x, t) = u_0(x) \quad \text{for all } x \in \mathbb{R}^n.$$

Proof: $\widehat{u}_0 = |\cdot|^{-s} \widehat{g}$ with $g \in L^2$

$$\begin{aligned} (2\pi)^{n/2} |e^{t\Delta} u_0(x) - u_0(x)| &= \left| \int \frac{\widehat{g}(\xi) e^{ix \cdot \xi} (e^{-t|\xi|^2} - 1)}{|\xi|^s} d\xi \right| \\ &\leq \|\widehat{g}\|_2 \left(\int \frac{|e^{-t|\xi|^2} - 1|^2}{|\xi|^{2s}} d\xi \right)^{1/2} \\ &= t^{s/2 - n/4} \|g\|_2 \left(\int \frac{|e^{-|y|^2} - 1|^2}{|y|^{2s}} dy \right)^{1/2} \\ &= t^{s/2 - n/4} \|f\|_{\dot{H}^s} \left(\int \frac{\min\{|y|^2, 1\}^2}{|y|^{2s}} dy \right)^{1/2} \\ &\leq C_s t^{s/2 - n/4} \|f\|_{\dot{H}^s}. \end{aligned}$$

The same calculation works for the Schrödinger equation

Lemma (Pointwise convergence for smooth data)

Let u solve the heat or Schrödinger equation with $u_0 \in \dot{H}^s(\mathbb{R}^n)$ with $n/2 < s < n/2 + 2$. Then

$$\lim_{t \rightarrow 0} u(x, t) = u_0(x) \quad \text{for all } x \in \mathbb{R}^n.$$

Proof: $\widehat{u}_0 = |\cdot|^{-s} \widehat{g}$ with $g \in L^2$

$$\begin{aligned} (2\pi)^{n/2} |e^{t\Delta} u_0(x) - u_0(x)| &= \left| \int \frac{\widehat{g}(\xi) e^{ix \cdot \xi} (e^{-t|\xi|^2} - 1)}{|\xi|^s} d\xi \right| \\ &\leq \|\widehat{g}\|_2 \left(\int \frac{|e^{-t|\xi|^2} - 1|^2}{|\xi|^{2s}} d\xi \right)^{1/2} \\ &= t^{s/2 - n/4} \|g\|_2 \left(\int \frac{|e^{-|y|^2} - 1|^2}{|y|^{2s}} dy \right)^{1/2} \\ &= t^{s/2 - n/4} \|f\|_{\dot{H}^s} \left(\int \frac{\min\{|y|^2, 1\}^2}{|y|^{2s}} dy \right)^{1/2} \\ &\leq C_s t^{s/2 - n/4} \|f\|_{\dot{H}^s}. \end{aligned}$$

The same calculation works for the Schrödinger equation

Lemma (Pointwise convergence for smooth data)

Let u solve the heat or Schrödinger equation with $u_0 \in \dot{H}^s(\mathbb{R}^n)$ with $n/2 < s < n/2 + 2$. Then

$$\lim_{t \rightarrow 0} u(x, t) = u_0(x) \quad \text{for all } x \in \mathbb{R}^n.$$

Proof: $\widehat{u}_0 = |\cdot|^{-s} \widehat{g}$ with $g \in L^2$

$$\begin{aligned} (2\pi)^{n/2} |e^{t\Delta} u_0(x) - u_0(x)| &= \left| \int \frac{\widehat{g}(\xi) e^{ix \cdot \xi} (e^{-t|\xi|^2} - 1)}{|\xi|^s} d\xi \right| \\ &\leq \|\widehat{g}\|_2 \left(\int \frac{|e^{-t|\xi|^2} - 1|^2}{|\xi|^{2s}} d\xi \right)^{1/2} \\ &= t^{s/2 - n/4} \|g\|_2 \left(\int \frac{|e^{-|y|^2} - 1|^2}{|y|^{2s}} dy \right)^{1/2} \\ &= t^{s/2 - n/4} \|f\|_{\dot{H}^s} \left(\int \frac{\min\{|y|^2, 1\}^2}{|y|^{2s}} dy \right)^{1/2} \\ &\leq C_s t^{s/2 - n/4} \|f\|_{\dot{H}^s}. \end{aligned}$$

The same calculation works for the Schrödinger equation

Lemma (Pointwise convergence for smooth data)

Let u solve the heat or Schrödinger equation with $u_0 \in \dot{H}^s(\mathbb{R}^n)$ with $n/2 < s < n/2 + 2$. Then

$$\lim_{t \rightarrow 0} u(x, t) = u_0(x) \quad \text{for all } x \in \mathbb{R}^n.$$

Proof: $\widehat{u}_0 = |\cdot|^{-s} \widehat{g}$ with $g \in L^2$

$$\begin{aligned} (2\pi)^{n/2} |e^{t\Delta} u_0(x) - u_0(x)| &= \left| \int \frac{\widehat{g}(\xi) e^{ix \cdot \xi} (e^{-t|\xi|^2} - 1)}{|\xi|^s} d\xi \right| \\ &\leq \|\widehat{g}\|_2 \left(\int \frac{|e^{-t|\xi|^2} - 1|^2}{|\xi|^{2s}} d\xi \right)^{1/2} \\ &= t^{s/2 - n/4} \|g\|_2 \left(\int \frac{|e^{-|y|^2} - 1|^2}{|y|^{2s}} dy \right)^{1/2} \\ &= t^{s/2 - n/4} \|f\|_{\dot{H}^s} \left(\int \frac{\min\{|y|^2, 1\}^2}{|y|^{2s}} dy \right)^{1/2} \\ &\leq C_s t^{s/2 - n/4} \|f\|_{\dot{H}^s}. \end{aligned}$$

The same calculation works for the Schrödinger equation

Lemma (Pointwise convergence for smooth data)

Let u solve the heat or Schrödinger equation with $u_0 \in \dot{H}^s(\mathbb{R}^n)$ with $n/2 < s < n/2 + 2$. Then

$$\lim_{t \rightarrow 0} u(x, t) = u_0(x) \quad \text{for all } x \in \mathbb{R}^n.$$

Proof: $\widehat{u}_0 = |\cdot|^{-s} \widehat{g}$ with $g \in L^2$

$$\begin{aligned} (2\pi)^{n/2} |e^{t\Delta} u_0(x) - u_0(x)| &= \left| \int \frac{\widehat{g}(\xi) e^{ix \cdot \xi} (e^{-t|\xi|^2} - 1)}{|\xi|^s} d\xi \right| \\ &\leq \|\widehat{g}\|_2 \left(\int \frac{|e^{-t|\xi|^2} - 1|^2}{|\xi|^{2s}} d\xi \right)^{1/2} \\ &= t^{s/2 - n/4} \|g\|_2 \left(\int \frac{|e^{-|y|^2} - 1|^2}{|y|^{2s}} dy \right)^{1/2} \\ &= t^{s/2 - n/4} \|f\|_{\dot{H}^s} \left(\int \frac{\min\{|y|^2, 1\}^2}{|y|^{2s}} dy \right)^{1/2} \\ &\leq C_s t^{s/2 - n/4} \|f\|_{\dot{H}^s}. \end{aligned}$$

Lemma (Pointwise convergence for smooth data)

Let u solve the heat or Schrödinger equation with $u_0 \in \dot{H}^s(\mathbb{R}^n)$ with $n/2 < s < n/2 + 2$. Then

$$\lim_{t \rightarrow 0} u(x, t) = u_0(x) \quad \text{for all } x \in \mathbb{R}^n.$$

Proof: $\widehat{u}_0 = |\cdot|^{-s} \widehat{g}$ with $g \in L^2$

$$\begin{aligned} (2\pi)^{n/2} |e^{t\Delta} u_0(x) - u_0(x)| &= \left| \int \frac{\widehat{g}(\xi) e^{ix \cdot \xi} (e^{-t|\xi|^2} - 1)}{|\xi|^s} d\xi \right| \\ &\leq \|\widehat{g}\|_2 \left(\int \frac{|e^{-t|\xi|^2} - 1|^2}{|\xi|^{2s}} d\xi \right)^{1/2} \\ &= t^{s/2 - n/4} \|g\|_2 \left(\int \frac{|e^{-|y|^2} - 1|^2}{|y|^{2s}} dy \right)^{1/2} \\ &= t^{s/2 - n/4} \|f\|_{\dot{H}^s} \left(\int \frac{\min\{|y|^2, 1\}^2}{|y|^{2s}} dy \right)^{1/2} \\ &\leq C_s t^{s/2 - n/4} \|f\|_{\dot{H}^s}. \end{aligned}$$

Lemma (Pointwise convergence for smooth data)

Let u solve the heat or Schrödinger equation with $u_0 \in \dot{H}^s(\mathbb{R}^n)$ with $n/2 < s < n/2 + 2$. Then

$$\lim_{t \rightarrow 0} u(x, t) = u_0(x) \quad \text{for all } x \in \mathbb{R}^n.$$

Proof: $\widehat{u}_0 = |\cdot|^{-s} \widehat{g}$ with $g \in L^2$

$$\begin{aligned} (2\pi)^{n/2} |e^{t\Delta} u_0(x) - u_0(x)| &= \left| \int \frac{\widehat{g}(\xi) e^{ix \cdot \xi} (e^{-t|\xi|^2} - 1)}{|\xi|^s} d\xi \right| \\ &\leq \|\widehat{g}\|_2 \left(\int \frac{|e^{-t|\xi|^2} - 1|^2}{|\xi|^{2s}} d\xi \right)^{1/2} \\ &= t^{s/2 - n/4} \|g\|_2 \left(\int \frac{|e^{-|y|^2} - 1|^2}{|y|^{2s}} dy \right)^{1/2} \\ &= t^{s/2 - n/4} \|f\|_{\dot{H}^s} \left(\int \frac{\min\{|y|^2, 1\}^2}{|y|^{2s}} dy \right)^{1/2} \\ &\leq C_s t^{s/2 - n/4} \|f\|_{\dot{H}^s}. \end{aligned}$$

Lebesgue a.e. convergence for data in L^2

Recall that the Hardy–Littlewood maximal operator M is defined by

$$Mf = \sup_{r>0} \frac{1}{|B(0,r)|} \mathbf{1}_{B(0,r)} * |f|,$$

and that it is bounded from $L^2(\mathbb{R}^n)$ to $L^2(\mathbb{R}^n)$.

This allows one to conclude that

$$\lim_{r \rightarrow 0} \frac{1}{|B(0,r)|} \mathbf{1}_{B(0,r)} * f(x) \rightarrow f(x), \quad \text{a.e. } x \in \mathbb{R}^n,$$

for all $f \in L^2(\mathbb{R}^n)$.

Later, I will remind you how to prove this using the L^2 -bound.

Lebesgue a.e. convergence for data in L^2

Recall that the Hardy–Littlewood maximal operator M is defined by

$$Mf = \sup_{r>0} \frac{1}{|B(0,r)|} \mathbf{1}_{B(0,r)} * |f|,$$

and that it is bounded from $L^2(\mathbb{R}^n)$ to $L^2(\mathbb{R}^n)$.

This allows one to conclude that

$$\lim_{r \rightarrow 0} \frac{1}{|B(0,r)|} \mathbf{1}_{B(0,r)} * f(x) \rightarrow f(x), \quad \text{a.e. } x \in \mathbb{R}^n,$$

for all $f \in L^2(\mathbb{R}^n)$.

Later, I will remind you how to prove this using the L^2 -bound.

Lebesgue a.e. convergence for data in L^2

Recall that the Hardy–Littlewood maximal operator M is defined by

$$Mf = \sup_{r>0} \frac{1}{|B(0,r)|} \mathbf{1}_{B(0,r)} * |f|,$$

and that it is bounded from $L^2(\mathbb{R}^n)$ to $L^2(\mathbb{R}^n)$.

This allows one to conclude that

$$\lim_{r \rightarrow 0} \frac{1}{|B(0,r)|} \mathbf{1}_{B(0,r)} * f(x) \rightarrow f(x), \quad \text{a.e. } x \in \mathbb{R}^n,$$

for all $f \in L^2(\mathbb{R}^n)$.

Later, I will remind you how to prove this using the L^2 -bound.

Lebesgue a.e. convergence for data in L^2

Recall that the Hardy–Littlewood maximal operator M is defined by

$$Mf = \sup_{r>0} \frac{1}{|B(0,r)|} \mathbf{1}_{B(0,r)} * |f|,$$

and that it is bounded from $L^2(\mathbb{R}^n)$ to $L^2(\mathbb{R}^n)$.

This allows one to conclude that

$$\lim_{r \rightarrow 0} \frac{1}{|B(0,r)|} \mathbf{1}_{B(0,r)} * f(x) \rightarrow f(x), \quad \text{a.e. } x \in \mathbb{R}^n,$$

for all $f \in L^2(\mathbb{R}^n)$.

Later, I will remind you how to prove this using the L^2 -bound.

Now

$$e^{-|\cdot|^2} \leq \sum_{j \geq 0} 2^{-j(n+1)} \mathbf{1}_{B(0,2^j)}$$

so that

$$e^{-|\cdot|^2/t} \leq \sum_{j \geq 0} 2^{-j(n+1)} \mathbf{1}_{B(0,t^{1/2}2^j)}$$

so that

$$\frac{1}{t^{n/2}} e^{-|\cdot|^2/t} \leq \sum_{j \geq 0} 2^{-j} \frac{1}{|B(0,t^{1/2}2^j)|} \mathbf{1}_{B(0,t^{1/2}2^j)}.$$

Thus

$$\sup_{t>0} |e^{t\Delta} f| = \sup_{t>0} \left| \frac{1}{t^{n/2}} e^{|\cdot|^2/t} * f \right| \leq \sum_{j \geq 0} 2^{-j} Mf \leq 2Mf.$$

So the L^2 -bound for M gives an L^2 maximal estimate for the heat equation which allows us to conclude that

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

using the same argument, which I will remind you of soon.

Now

$$e^{-|\cdot|^2} \leq \sum_{j \geq 0} 2^{-j(n+1)} \mathbf{1}_{B(0,2^j)}$$

so that

$$e^{-|\cdot|^2/t} \leq \sum_{j \geq 0} 2^{-j(n+1)} \mathbf{1}_{B(0,t^{1/2}2^j)}$$

so that

$$\frac{1}{t^{n/2}} e^{-|\cdot|^2/t} \leq \sum_{j \geq 0} 2^{-j} \frac{1}{|B(0,t^{1/2}2^j)|} \mathbf{1}_{B(0,t^{1/2}2^j)}.$$

Thus

$$\sup_{t>0} |e^{t\Delta} f| = \sup_{t>0} \left| \frac{1}{t^{n/2}} e^{|\cdot|^2/t} * f \right| \leq \sum_{j \geq 0} 2^{-j} Mf \leq 2Mf.$$

So the L^2 -bound for M gives an L^2 maximal estimate for the heat equation which allows us to conclude that

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

using the same argument, which I will remind you of soon.

Now

$$e^{-|\cdot|^2} \leq \sum_{j \geq 0} 2^{-j(n+1)} \mathbf{1}_{B(0,2^j)}$$

so that

$$e^{-|\cdot|^2/t} \leq \sum_{j \geq 0} 2^{-j(n+1)} \mathbf{1}_{B(0,t^{1/2}2^j)}$$

so that

$$\frac{1}{t^{n/2}} e^{-|\cdot|^2/t} \leq \sum_{j \geq 0} 2^{-j} \frac{1}{|B(0,t^{1/2}2^j)|} \mathbf{1}_{B(0,t^{1/2}2^j)}.$$

Thus

$$\sup_{t>0} |e^{t\Delta} f| = \sup_{t>0} \left| \frac{1}{t^{n/2}} e^{-|\cdot|^2/t} * f \right| \leq \sum_{j \geq 0} 2^{-j} Mf \leq 2Mf.$$

So the L^2 -bound for M gives an L^2 maximal estimate for the heat equation which allows us to conclude that

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

using the same argument, which I will remind you of, soon.

Now
$$e^{-|\cdot|^2} \leq \sum_{j \geq 0} 2^{-j(n+1)} \mathbf{1}_{B(0,2^j)}$$

so that

$$e^{-|\cdot|^2/t} \leq \sum_{j \geq 0} 2^{-j(n+1)} \mathbf{1}_{B(0,t^{1/2}2^j)}$$

so that

$$\frac{1}{t^{n/2}} e^{-|\cdot|^2/t} \leq \sum_{j \geq 0} 2^{-j} \frac{1}{|B(0,t^{1/2}2^j)|} \mathbf{1}_{B(0,t^{1/2}2^j)}.$$

Thus

$$\sup_{t>0} |e^{t\Delta} f| = \sup_{t>0} \left| \frac{1}{t^{n/2}} e^{|\cdot|^2/t} * f \right| \leq \sum_{j \geq 0} 2^{-j} Mf \leq 2Mf.$$

So the L^2 -bound for M gives an L^2 maximal estimate for the heat equation which allows us to conclude that

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

using the same argument, which I will remind you of, soon.

Now
$$e^{-|\cdot|^2} \leq \sum_{j \geq 0} 2^{-j(n+1)} \mathbf{1}_{B(0,2^j)}$$

so that

$$e^{-|\cdot|^2/t} \leq \sum_{j \geq 0} 2^{-j(n+1)} \mathbf{1}_{B(0,t^{1/2}2^j)}$$

so that

$$\frac{1}{t^{n/2}} e^{-|\cdot|^2/t} \leq \sum_{j \geq 0} 2^{-j} \frac{1}{|B(0, t^{1/2}2^j)|} \mathbf{1}_{B(0,t^{1/2}2^j)}.$$

Thus

$$\sup_{t>0} |e^{t\Delta} f| = \sup_{t>0} \left| \frac{1}{t^{n/2}} e^{|\cdot|^2/t} * f \right| \leq \sum_{j \geq 0} 2^{-j} Mf \leq 2Mf.$$

So the L^2 -bound for M gives an L^2 maximal estimate for the heat equation which allows us to conclude that

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

using the same argument, which I will remind you of soon.

Hausdorff measure

Let $A \subseteq \mathbb{R}^n$ be a borel set, $0 < \alpha < n$ and

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

Definition

The α -Hausdorff measure of A is

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

Hausdorff measure

Let $A \subseteq \mathbb{R}^n$ be a borel set, $0 < \alpha < n$ and

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

Definition

The α -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 α_0 such that

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

Definition

α_0 is the Hausdorff dimension of the set A :

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

Hausdorff dimension

Remark

There exists a unique α_0 such that

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

Definition

α_0 is the Hausdorff dimension of the set A :

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

Definition (Frostman measures)

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

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

$$\begin{aligned} E_{\alpha'}(\mu) &:= \iint \frac{d\mu(x)d\mu(y)}{|x-y|^{\alpha'}} = \int \sum_{j=0}^{\infty} \int_{A(y, 2^{-j})} \frac{d\mu(x)}{|x-y|^{\alpha'}} d\mu(y) \\ &\leq \int \sum_{j=0}^{\infty} c_\alpha(\mu) 2^{-j\alpha} 2^{j\alpha'} d\mu(y) \\ &\lesssim c_\alpha(\mu) \|\mu\| < \infty \quad \text{if } \alpha > \alpha'. \end{aligned}$$

Lemma (Frostman)

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

- ▶ $\mathcal{H}^\alpha(A) > 0$;
- ▶ there is an α -dimensional measure μ such that $\mu(A) > 0$.

Definition (Frostman measures)

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

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

$$\begin{aligned} E_{\alpha'}(\mu) &:= \iint \frac{d\mu(x)d\mu(y)}{|x-y|^{\alpha'}} = \int \sum_{j=0}^{\infty} \int_{A(y, 2^{-j})} \frac{d\mu(x)}{|x-y|^{\alpha'}} d\mu(y) \\ &\leq \int \sum_{j=0}^{\infty} c_\alpha(\mu) 2^{-j\alpha} 2^{j\alpha'} d\mu(y) \\ &\lesssim c_\alpha(\mu) \|\mu\| < \infty \quad \text{if } \alpha > \alpha'. \end{aligned}$$

Lemma (Frostman)

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

- ▶ $\mathcal{H}^\alpha(A) > 0$;
- ▶ there is an α -dimensional measure μ such that $\mu(A) > 0$.

Definition (Frostman measures)

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

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

$$\begin{aligned} E_{\alpha'}(\mu) &:= \iint \frac{d\mu(x)d\mu(y)}{|x-y|^{\alpha'}} = \int \sum_{j=0}^{\infty} \int_{A(y, 2^{-j})} \frac{d\mu(x)}{|x-y|^{\alpha'}} d\mu(y) \\ &\leq \int \sum_{j=0}^{\infty} c_\alpha(\mu) 2^{-j\alpha} 2^{j\alpha'} d\mu(y) \\ &\lesssim c_\alpha(\mu) \|\mu\| < \infty \quad \text{if } \alpha > \alpha'. \end{aligned}$$

Lemma (Frostman)

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

- ▶ $\mathcal{H}^\alpha(A) > 0$;
- ▶ there is an α -dimensional measure μ such that $\mu(A) > 0$.

Definition (Frostman measures)

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

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

$$\begin{aligned} E_{\alpha'}(\mu) &:= \iint \frac{d\mu(x)d\mu(y)}{|x-y|^{\alpha'}} = \int \sum_{j=0}^{\infty} \int_{A(y, 2^{-j})} \frac{d\mu(x)}{|x-y|^{\alpha'}} d\mu(y) \\ &\leq \int \sum_{j=0}^{\infty} c_\alpha(\mu) 2^{-j\alpha} 2^{j\alpha'} d\mu(y) \\ &\lesssim c_\alpha(\mu) \|\mu\| < \infty \quad \text{if } \alpha > \alpha'. \end{aligned}$$

Lemma (Frostman)

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

- ▶ $\mathcal{H}^\alpha(A) > 0$;
- ▶ there is an α -dimensional measure μ such that $\mu(A) > 0$.

Definition (Frostman measures)

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

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

$$\begin{aligned} E_{\alpha'}(\mu) &:= \iint \frac{d\mu(x)d\mu(y)}{|x-y|^{\alpha'}} = \int \sum_{j=0}^{\infty} \int_{A(y, 2^{-j})} \frac{d\mu(x)}{|x-y|^{\alpha'}} d\mu(y) \\ &\leq \int \sum_{j=0}^{\infty} c_\alpha(\mu) 2^{-j\alpha} 2^{j\alpha'} d\mu(y) \\ &\lesssim c_\alpha(\mu) \|\mu\| < \infty \quad \text{if } \alpha > \alpha'. \end{aligned}$$

Lemma (Frostman)

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

- ▶ $\mathcal{H}^\alpha(A) > 0$;
- ▶ there is an α -dimensional measure μ such that $\mu(A) > 0$.

Control of singularities

Lemma

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

$$\|f\|_{L^1(d\mu)} \leq C_\mu \|f\|_{\dot{H}^s}.$$

Proof: $f = I_s * g$ with $g \in L^2$ and $\widehat{I_s} = |\cdot|^{-s}$. Suffices to prove

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

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

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

Control of singularities

Lemma

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

$$\|f\|_{L^1(d\mu)} \leq C_\mu \|f\|_{\dot{H}^s}.$$

Proof: $f = I_s * g$ with $g \in L^2$ and $\widehat{I_s} = |\cdot|^{-s}$. Suffices to prove

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

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

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

Control of singularities

Lemma

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

$$\|f\|_{L^1(d\mu)} \leq C_\mu \|f\|_{\dot{H}^s}.$$

Proof: $f = I_s * g$ with $g \in L^2$ and $\widehat{I_s} = |\cdot|^{-s}$. Suffices to prove

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

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

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

Control of singularities

Lemma

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

$$\|f\|_{L^1(d\mu)} \leq C_\mu \|f\|_{\dot{H}^s}.$$

Proof: $f = I_s * g$ with $g \in L^2$ and $\widehat{I_s} = |\cdot|^{-s}$. Suffices to prove

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

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

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

Thus it remains to prove that

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

By Plancherel's theorem,

$$\|I_s * \mu\|_{L^2(\mathbb{R}^n)}^2 = \|\widehat{I_s \mu}\|_{L^2(\mathbb{R}^n)}^2 = \int \widehat{\mu}(\xi) \overline{\widehat{\mu}(\xi)} \widehat{I_{2s}}(\xi) d\xi.$$

Recalling that $I_{2s}(x) = C_{n,s}|x|^{-(n-2s)}$,

$$\begin{aligned} \|I_s * \mu\|_{L^2(\mathbb{R}^n)}^2 &= \int \mu * I_{2s}(y) d\mu(y) \\ &= C_{n,s} \iint \frac{d\mu(x)d\mu(y)}{|x-y|^{n-2s}} = C_{n,s} E_{n-2s}(\mu), \end{aligned}$$

and we are done. □

Thus it remains to prove that

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

By Plancherel's theorem,

$$\|I_s * \mu\|_{L^2(\mathbb{R}^n)}^2 = \|\widehat{I_s \mu}\|_{L^2(\mathbb{R}^n)}^2 = \int \widehat{\mu}(\xi) \overline{\widehat{\mu}(\xi)} \widehat{I_{2s}}(\xi) d\xi.$$

Recalling that $I_{2s}(x) = C_{n,s}|x|^{-(n-2s)}$,

$$\begin{aligned} \|I_s * \mu\|_{L^2(\mathbb{R}^n)}^2 &= \int \mu * I_{2s}(y) d\mu(y) \\ &= C_{n,s} \iint \frac{d\mu(x)d\mu(y)}{|x-y|^{n-2s}} = C_{n,s} E_{n-2s}(\mu), \end{aligned}$$

and we are done. □

Thus it remains to prove that

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

By Plancherel's theorem,

$$\|I_s * \mu\|_{L^2(\mathbb{R}^n)}^2 = \|\widehat{I_s \mu}\|_{L^2(\mathbb{R}^n)}^2 = \int \widehat{\mu}(\xi) \overline{\widehat{\mu}(\xi)} \widehat{I_{2s}}(\xi) d\xi.$$

Recalling that $I_{2s}(x) = C_{n,s}|x|^{-(n-2s)}$,

$$\begin{aligned} \|I_s * \mu\|_{L^2(\mathbb{R}^n)}^2 &= \int \mu * I_{2s}(y) d\mu(y) \\ &= C_{n,s} \iint \frac{d\mu(x)d\mu(y)}{|x-y|^{n-2s}} = C_{n,s}E_{n-2s}(\mu), \end{aligned}$$

and we are done. □

Thus it remains to prove that

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

By Plancherel's theorem,

$$\|I_s * \mu\|_{L^2(\mathbb{R}^n)}^2 = \|\widehat{I_s \mu}\|_{L^2(\mathbb{R}^n)}^2 = \int \widehat{\mu}(\xi) \overline{\widehat{\mu}(\xi)} \widehat{I_{2s}}(\xi) d\xi.$$

Recalling that $I_{2s}(x) = C_{n,s}|x|^{-(n-2s)}$,

$$\begin{aligned} \|I_s * \mu\|_{L^2(\mathbb{R}^n)}^2 &= \int \mu * I_{2s}(y) d\mu(y) \\ &= C_{n,s} \iint \frac{d\mu(x)d\mu(y)}{|x-y|^{n-2s}} = C_{n,s} E_{n-2s}(\mu), \end{aligned}$$

and we are done. □

Optimality of the control of singularities lemma

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

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

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

Optimality of the control of singularities lemma

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

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

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

Optimality of the control of singularities lemma

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

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

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

Optimality of the control of singularities lemma

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

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

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

Optimality of the control of singularities lemma

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

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

So there is initial data $u_0 \in \dot{H}^s(\mathbb{R}^n)$ 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 α -dimensional μ ,

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

Then, for all $u_0 \in \dot{H}^s$,

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

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

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

whenever μ is α -dimensional.

Proposition (Maximal estimates imply convergence)

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

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

Then, for all $u_0 \in \dot{H}^s$,

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

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

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

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

$$\mu \left\{ x \in \mathbb{R}^n \mid \lim_{t \rightarrow 0} u(t, x) \neq u_0(x) \right\} = 0$$

whenever μ is α -dimensional.

Proposition (Maximal estimates imply convergence)

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

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

Then, for all $u_0 \in \dot{H}^s$,

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

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

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

whenever μ is α -dimensional.

Proposition (Maximal estimates imply convergence)

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

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

Then, for all $u_0 \in \dot{H}^s$,

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

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

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

whenever μ is α -dimensional.

Proposition (Maximal estimates imply convergence)

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

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

Then, for all $u_0 \in \dot{H}^s$,

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

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

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

whenever μ is α -dimensional.

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

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

where u_h denotes the solution with initial data h .

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

$$|u(\cdot, t) - u_0| \leq |u_{u_0-h}(\cdot, t)| + |u_h(\cdot, t) - h| + |h - u_0|.$$

Then,

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

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

$$|u(\cdot, t) - u_0| \leq |u_{u_0-h}(\cdot, t)| + |u_h(\cdot, t) - h| + |h - u_0|.$$

Then,

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

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

$$|u(\cdot, t) - u_0| \leq |u_{u_0-h}(\cdot, t)| + |u_h(\cdot, t) - h| + |h - u_0|.$$

Then,

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

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

$$\mu\{x : \lim_{t \rightarrow 0} |u(x, t) - u_0(x)| > \lambda\} \leq \lambda^{-1} C_\mu \|u_0 - h\|_{\dot{H}^s(\mathbb{R}^n)} \leq \lambda^{-1} C_\mu \varepsilon.$$

Letting ε tend to zero, then λ tend to zero, we are done. \square

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

$$|u(\cdot, t) - u_0| \leq |u_{u_0-h}(\cdot, t)| + |u_h(\cdot, t) - h| + |h - u_0|.$$

Then,

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

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

$$\mu\{x : \lim_{t \rightarrow 0} |u(x, t) - u_0(x)| > \lambda\} \leq \lambda^{-1} C_\mu \|u_0 - h\|_{\dot{H}^s(\mathbb{R}^n)} \leq \lambda^{-1} C_\mu \varepsilon.$$

Letting ε tend to zero, then λ tend to zero, we are done. \square

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

$$|u(\cdot, t) - u_0| \leq |u_{u_0-h}(\cdot, t)| + |u_h(\cdot, t) - h| + |h - u_0|.$$

Then,

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

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

$$\mu\{x : \lim_{t \rightarrow 0} |u(x, t) - u_0(x)| > \lambda\} \leq \lambda^{-1} C_\mu \|u_0 - h\|_{\dot{H}^s(\mathbb{R}^n)} \leq \lambda^{-1} C_\mu \varepsilon.$$

Letting ε tend to zero, then λ tend to zero, we are done. \square

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

$$|u(\cdot, t) - u_0| \leq |u_{u_0-h}(\cdot, t)| + |u_h(\cdot, t) - h| + |h - u_0|.$$

Then,

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

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

$$\mu\{x : \lim_{t \rightarrow 0} |u(x, t) - u_0(x)| > \lambda\} \leq \lambda^{-1} C_\mu \|u_0 - h\|_{\dot{H}^s(\mathbb{R}^n)} \leq \lambda^{-1} C_\mu \varepsilon.$$

Letting ε tend to zero, then λ tend to zero, we are done. \square

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

$$|u(\cdot, t) - u_0| \leq |u_{u_0-h}(\cdot, t)| + |u_h(\cdot, t) - h| + |h - u_0|.$$

Then,

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

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

$$\mu\{x : \lim_{t \rightarrow 0} |u(x, t) - u_0(x)| > \lambda\} \leq \lambda^{-1} C_\mu \|u_0 - h\|_{\dot{H}^s(\mathbb{R}^n)} \leq \lambda^{-1} C_\mu \varepsilon.$$

Letting ε tend to zero, then λ tend to zero, we are done. \square

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

$$|u(\cdot, t) - u_0| \leq |u_{u_0-h}(\cdot, t)| + |u_h(\cdot, t) - h| + |h - u_0|.$$

Then,

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

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

$$\mu\{x : \lim_{t \rightarrow 0} |u(x, t) - u_0(x)| > \lambda\} \leq \lambda^{-1} C_\mu \|u_0 - h\|_{\dot{H}^s(\mathbb{R}^n)} \leq \lambda^{-1} C_\mu \varepsilon.$$

Letting ε tend to zero, then λ tend to zero, we are done. \square

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 α -dimensional μ ,

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

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

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

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

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

Squaring out the integral, it will suffice to show that

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

Theorem (Maximal estimate for the heat equation)

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

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

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

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

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

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

Squaring out the integral, it will suffice to show that

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

Theorem (Maximal estimate for the heat equation)

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

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

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

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

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

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

Squaring out the integral, it will suffice to show that

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

Theorem (Maximal estimate for the heat equation)

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

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

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

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

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

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

Squaring out the integral, it will suffice to show that

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

Theorem (Maximal estimate for the heat equation)

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

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

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

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

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

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

Squaring out the integral, it will suffice to show that

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

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

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

uniformly for all choices of $t(x), t(y) > 0$. Recalling that $\widehat{|\cdot|^{-2s}} = C_{n,s} |\cdot|^{2s-n}$, 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 λ . By changing variables, this would follow from

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

which can be checked by direct calculation. □

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

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

uniformly for all choices of $t(x), t(y) > 0$. Recalling that $\widehat{|\cdot|^{-2s}} = C_{n,s} |\cdot|^{2s-n}$, 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 λ . By changing variables, this would follow from

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

which can be checked by direct calculation. □

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

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

uniformly for all choices of $t(x), t(y) > 0$. Recalling that $\widehat{|\cdot|^{-2s}} = C_{n,s} |\cdot|^{2s-n}$, 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 λ . By changing variables, this would follow from

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

which can be checked by direct calculation. □

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

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

uniformly for all choices of $t(x), t(y) > 0$. Recalling that $\widehat{|\cdot|^{-2s}} = C_{n,s} |\cdot|^{2s-n}$, 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 λ . By changing variables, this would follow from

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

which can be checked by direct calculation. □

Corollary

Let $0 < s < n/2$ and let u be a solution to the heat equation with initial data $u_0 \in \dot{H}^s$. Then

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

As we saw before, $u_0 \in \dot{H}^s$ can be singular on a set of dimension less than $n - 2s$ and so this is optimal.

Corollary

Let $0 < s < n/2$ and let u be a solution to the heat equation with initial data $u_0 \in \dot{H}^s$. Then

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

As we saw before, $u_0 \in \dot{H}^s$ can be singular on a set of dimension less than $n - 2s$ and so this is optimal.

Part 3:

Convergence for the Schrödinger equation

Lebesgue a.e. convergence for Schrödinger

Studied by many authors:

Carleson (1979), Dahlberg–Kenig (1982), Cowling (1983), Carbery (1985), Sjölin (1987), Vega (1988), Bourgain (1992/95), Moyua–Vargas–Vega (1996/99), Tao–Vargàs (2000), Tao (2003), Lee (2006), Bourgain (2013).

Best known sufficient condition for Lebesgue a.e. convergence:

- ▶ $s \geq 1/4$ in dimension $n = 1$ (Carleson);
- ▶ $s > \frac{1}{2} - \frac{1}{4n}$ in dimension $n \geq 2$ (Lee, Bourgain).

Best known necessary condition for Lebesgue a.e. convergence:

- ▶ $s \geq 1/4$ in dimension $n = 1$ (Dahlberg–Kenig);
- ▶ $s \geq \frac{1}{2} - \frac{1}{n+2}$ in dimension $n \geq 2$ (Lucà–R.).

Lebesgue a.e. convergence for Schrödinger

Studied by many authors:

Carleson (1979), Dahlberg–Kenig (1982), Cowling (1983), Carbery (1985), Sjölin (1987), Vega (1988), Bourgain (1992/95), Moyua–Vargas–Vega (1996/99), Tao–Vargas (2000), Tao (2003), Lee (2006), Bourgain (2013).

Best known sufficient condition for Lebesgue a.e. convergence:

- ▶ $s \geq 1/4$ in dimension $n = 1$ (Carleson);
- ▶ $s > \frac{1}{2} - \frac{1}{4n}$ in dimension $n \geq 2$ (Lee, Bourgain).

Best known necessary condition for Lebesgue a.e. convergence:

- ▶ $s \geq 1/4$ in dimension $n = 1$ (Dahlberg–Kenig);
- ▶ $s \geq \frac{1}{2} - \frac{1}{n+2}$ in dimension $n \geq 2$ (Lucà–R.).

Lebesgue a.e. convergence for Schrödinger

Studied by many authors:

Carleson (1979), Dahlberg–Kenig (1982), Cowling (1983), Carbery (1985), Sjölin (1987), Vega (1988), Bourgain (1992/95), Moyua–Vargas–Vega (1996/99), Tao–Vargas (2000), Tao (2003), Lee (2006), Bourgain (2013).

Best known sufficient condition for Lebesgue a.e. convergence:

- ▶ $s \geq 1/4$ in dimension $n = 1$ (Carleson);
- ▶ $s > \frac{1}{2} - \frac{1}{4n}$ in dimension $n \geq 2$ (Lee, Bourgain).

Best known necessary condition for Lebesgue a.e. convergence:

- ▶ $s \geq 1/4$ in dimension $n = 1$ (Dahlberg–Kenig);
- ▶ $s \geq \frac{1}{2} - \frac{1}{n+2}$ in dimension $n \geq 2$ (Lucà–R.).

Maximal estimate for the Schrödinger equation

Theorem (Barceló–Bennett–Carbery–R.)

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

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

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 $n/4 \leq s < n/2$. □

Maximal estimate for the Schrödinger equation

Theorem (Barceló–Bennett–Carbery–R.)

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

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

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 $n/4 \leq s < n/2$. □

Maximal estimate for the Schrödinger equation

Theorem (Barceló–Bennett–Carbery–R.)

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

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

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 $n/4 \leq s < n/2$. □

Maximal estimate for the Schrödinger equation

Theorem (Barceló–Bennett–Carbery–R.)

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

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

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 $n/4 \leq s < n/2$ and let u be a solution to the Schrödinger equation with initial data $u_0 \in \dot{H}^s$. Then

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

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 we cannot go below this regularity in higher dimensions either via a fractal version of the **Lucà–R.**-necessary condition.

Corollary

Let $n/4 \leq s < n/2$ and let u be a solution to the Schrödinger equation with initial data $u_0 \in \dot{H}^s$. Then

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

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 we cannot go below this regularity in higher dimensions either via a fractal version of the Lucà–R.-necessary condition.

Corollary

Let $n/4 \leq s < n/2$ and let u be a solution to the Schrödinger equation with initial data $u_0 \in \dot{H}^s$. Then

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

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 we cannot go below this regularity in higher dimensions either via a fractal version of the **Lucà–R.**-necessary condition.

Corollary

Let $n/4 \leq s < n/2$ and let u be a solution to the Schrödinger equation with initial data $u_0 \in \dot{H}^s$. Then

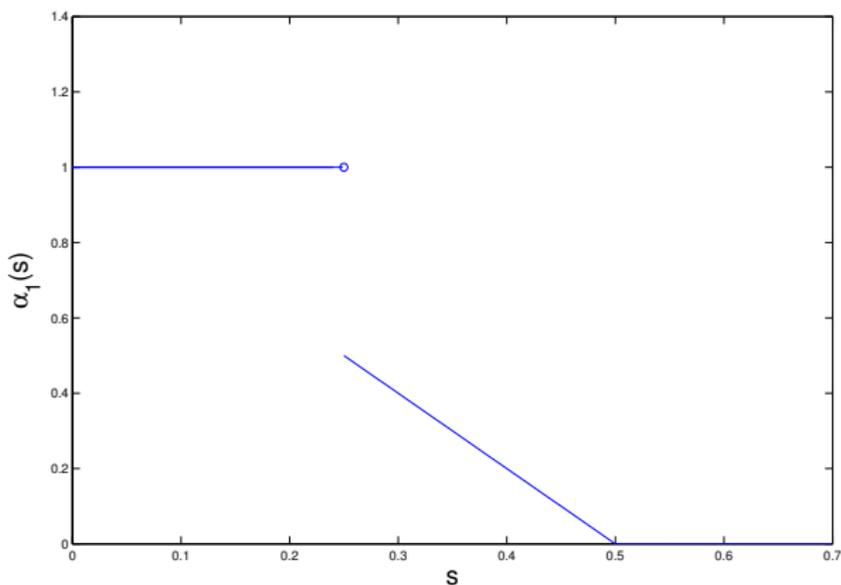
$$\dim \left\{ x \in \mathbb{R}^n \quad \lim_{t \rightarrow 0} u(t, x) \neq u_0(x) \right\} \leq n - 2s.$$

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 we cannot go below this regularity in higher dimensions either via a fractal version of the **Lucà–R.**-necessary condition.

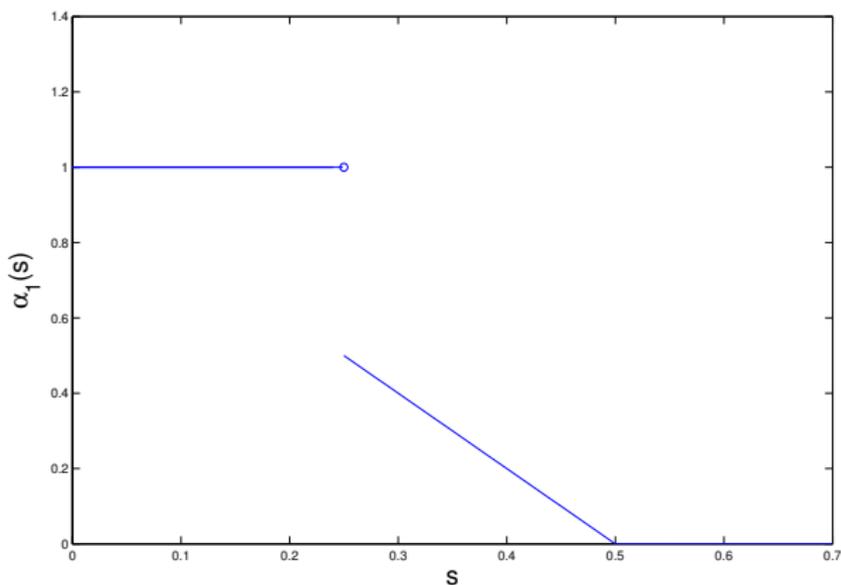
$$\alpha_n(s) := \sup_{u_0 \in H^s(\mathbb{R}^n)} \dim \left\{ x \in \mathbb{R}^n : \lim_{t \rightarrow 0} u(x, t) \neq u_0(x) \right\}$$



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

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

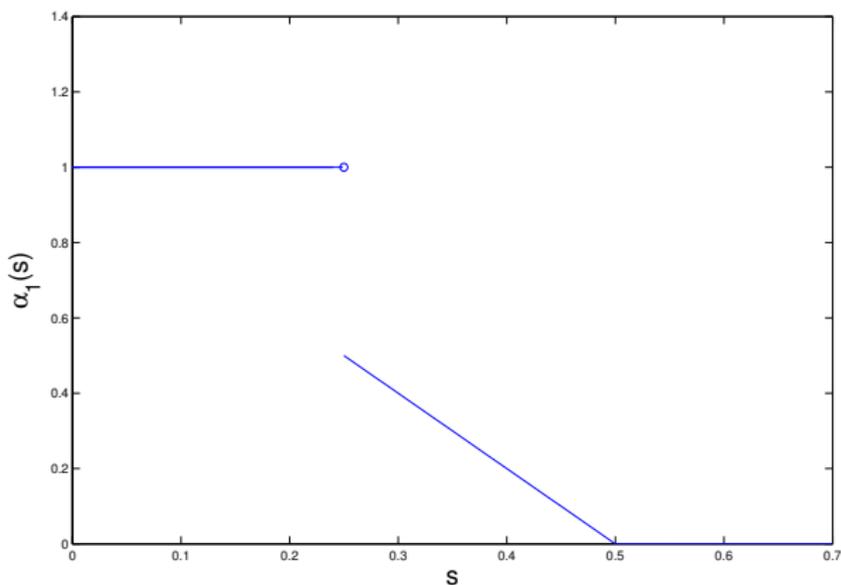
$$\alpha_n(s) := \sup_{u_0 \in H^s(\mathbb{R}^n)} \dim \left\{ x \in \mathbb{R}^n : \lim_{t \rightarrow 0} u(x, t) \neq u_0(x) \right\}$$



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

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

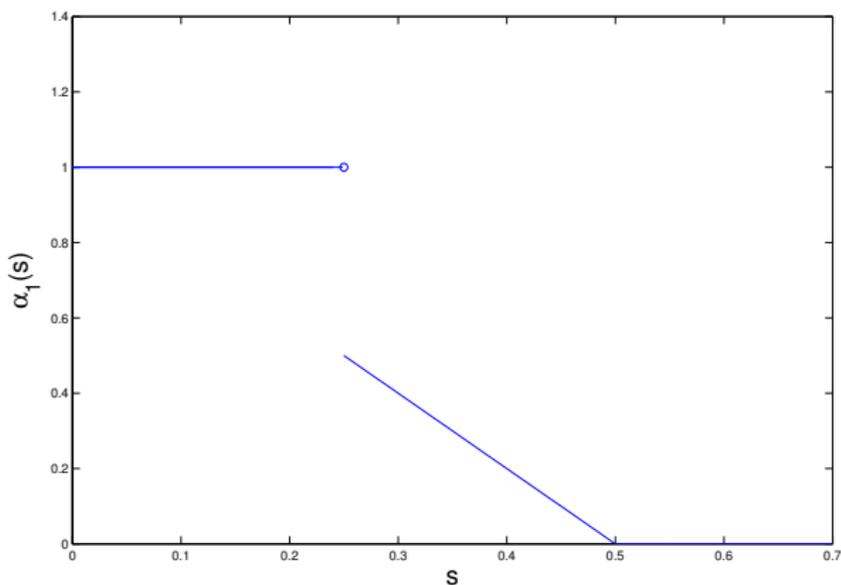
$$\alpha_n(s) := \sup_{u_0 \in H^s(\mathbb{R}^n)} \dim \left\{ x \in \mathbb{R}^n : \lim_{t \rightarrow 0} u(x, t) \neq u_0(x) \right\}$$



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

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

$$\alpha_n(s) := \sup_{u_0 \in H^s(\mathbb{R}^n)} \dim \left\{ x \in \mathbb{R}^n : \lim_{t \rightarrow 0} u(x, t) \neq u_0(x) \right\}$$



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

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

Part 4:

Counterexample for the
Schrödinger equation:

lower bounds for α_n

$$\alpha_n(s) := \sup_{u_0 \in H^s(\mathbb{R}^n)} \dim \left\{ x \in \mathbb{R}^n : \lim_{t \rightarrow 0} e^{it\Delta} u_0(x) \neq u_0(x) \right\}$$

Theorem (Lucà-R.)

$$\alpha_n(s) \geq \begin{cases} n & , \text{ when } s \leq \frac{n}{2(n+2)} \\ n + 1 - \frac{2(n+2)s}{n} & , \text{ when } \frac{n}{2(n+2)} \leq s \leq \frac{n}{4} \\ n - 2s & , \text{ when } \frac{n}{4} \leq s \leq \frac{n}{2} \\ 0 & , \text{ when } \frac{n}{2} \leq s \end{cases}$$

$$\alpha_n(s) := \sup_{u_0 \in H^s(\mathbb{R}^n)} \dim \left\{ x \in \mathbb{R}^n : \lim_{t \rightarrow 0} e^{it\Delta} u_0(x) \neq u_0(x) \right\}$$

Theorem (Lucà–R.)

$$\alpha_n(s) \geq \begin{cases} n & , \text{ when } s \leq \frac{n}{2(n+2)} \\ n + 1 - \frac{2(n+2)s}{n} & , \text{ when } \frac{n}{2(n+2)} \leq s \leq \frac{n}{4} \\ n - 2s & , \text{ when } \frac{n}{4} \leq s \leq \frac{n}{2} \\ 0 & , \text{ when } \frac{n}{2} \leq s \end{cases}$$

$$\alpha_n(s) \geq n \text{ when } s < \frac{n}{2(n+2)}$$

This bound follows from:

Theorem (Lucà–R.)

Suppose that

$$\lim_{t \rightarrow 0} e^{it\Delta} u_0(x) = u_0(x), \quad \text{a.e. } x \in \mathbb{R}^n$$

for all $u_0 \in H^s(\mathbb{R}^n)$. *Then*

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

which improves **Dahlberg–Kenig** for $n \geq 3$ (coinciding when $n = 2$).

$$\alpha_n(s) \geq n \text{ when } s < \frac{n}{2(n+2)}$$

This bound follows from:

Theorem (Lucà–R.)

Suppose that

$$\lim_{t \rightarrow 0} e^{it\Delta} u_0(x) = u_0(x), \quad \text{a.e. } x \in \mathbb{R}^n$$

for all $u_0 \in H^s(\mathbb{R}^n)$. Then

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

which improves **Dahlberg–Kenig** for $n \geq 3$ (coinciding when $n = 2$).

$$\alpha_n(s) \geq n \text{ when } s < \frac{n}{2(n+2)}$$

This bound follows from:

Theorem (Lucà–R.)

Suppose that

$$\lim_{t \rightarrow 0} e^{it\Delta} u_0(x) = u_0(x), \quad \text{a.e. } x \in \mathbb{R}^n$$

for all $u_0 \in H^s(\mathbb{R}^n)$. Then

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

which improves **Dahlberg–Kenig** for $n \geq 3$ (coinciding when $n = 2$).

Proof

Lemma (Nikišin–Stein maximal principle)

Modulo endpoints:

$$\lim_{t \rightarrow 0} e^{it\Delta} u_0(x) = u_0(x), \quad \text{a.e. } x \in \mathbb{R}^n,$$

for all $u_0 \in H^s(\mathbb{R}^n)$ if and only if there is a constant C such that

$$\left\| \sup_{0 < t < 1} |e^{it\Delta} u_0| \right\|_{L^2(B(0,1))} \leq C \|u_0\|_{H^s(\mathbb{R}^n)}.$$

So it suffices to show that $s \geq \frac{n}{2(n+2)}$ is necessary for

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

whenever $\text{supp } \widehat{f} \subset \{\xi : |\xi| \leq R\}$.

Proof

Lemma (Nikišin–Stein maximal principle)

Modulo endpoints:

$$\lim_{t \rightarrow 0} e^{it\Delta} u_0(x) = u_0(x), \quad \text{a.e. } x \in \mathbb{R}^n,$$

for all $u_0 \in H^s(\mathbb{R}^n)$ if and only if there is a constant C such that

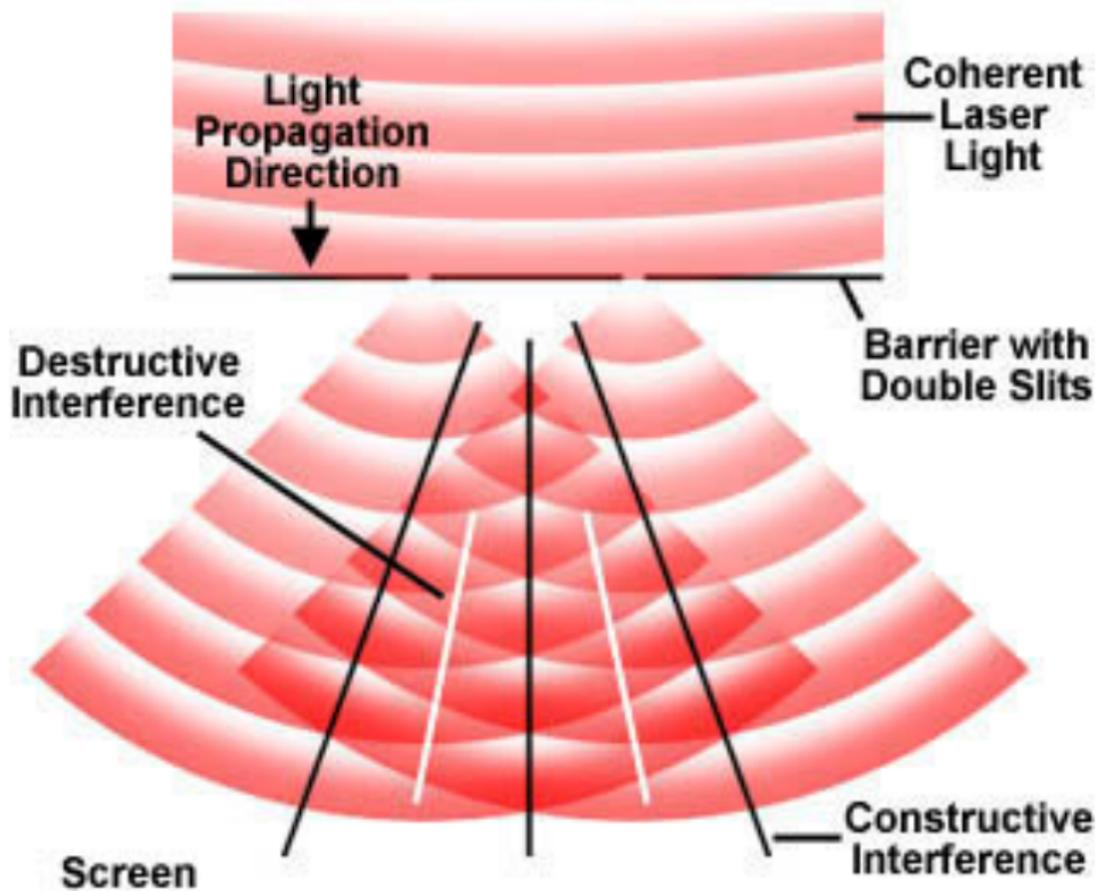
$$\left\| \sup_{0 < t < 1} |e^{it\Delta} u_0| \right\|_{L^2(B(0,1))} \leq C \|u_0\|_{H^s(\mathbb{R}^n)}.$$

So it suffices to show that $s \geq \frac{n}{2(n+2)}$ is necessary for

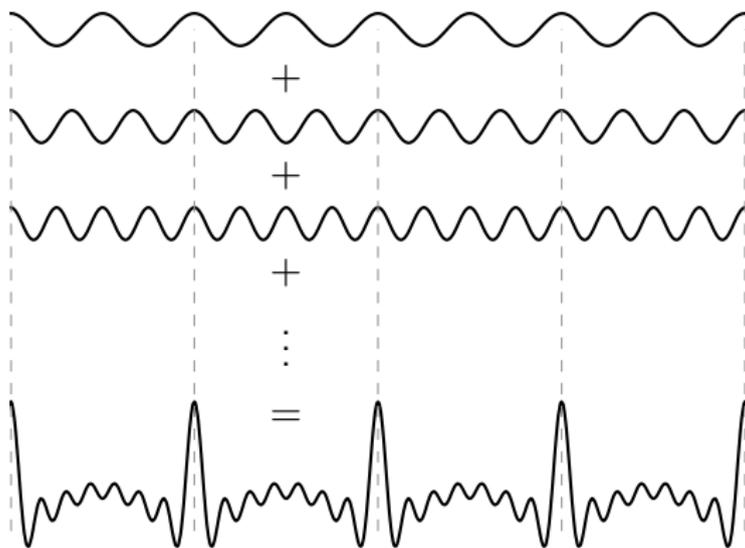
$$\left\| \sup_{0 < t < 1} |e^{it\Delta} f| \right\|_{L^2(B(0,1))} \lesssim R^s \|f\|_2,$$

whenever $\text{supp } \widehat{f} \subset \{\xi : |\xi| \leq R\}$.

Young's Double Slit Experiment



Constructive interference



The initial data

We consider the frequencies

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

for $0 < \kappa < \frac{1}{n+2}$,

and initial data defined by

$$\hat{f} = \frac{1}{\sqrt{|\Omega|}} \chi_{\Omega}, \quad \text{so that} \quad \|f\|_2 = 1.$$

This data was introduced in the context of Mattila's question by Barceló–Bennett–Carbery–Ruiz–Vilela (2007).

Note that

$$|\Omega| \simeq \text{number of frequencies} \simeq R^{n\kappa}.$$

The initial data

We consider the frequencies

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

for $0 < \kappa < \frac{1}{n+2}$,

and initial data defined by

$$\hat{f} = \frac{1}{\sqrt{|\Omega|}} \chi_{\Omega}, \quad \text{so that} \quad \|f\|_2 = 1.$$

This data was introduced in the context of Mattila's question by Barceló–Bennett–Carbery–Ruiz–Vilela (2007).

Note that

$$|\Omega| \simeq \text{number of frequencies} \simeq R^{n\kappa}.$$

The initial data

We consider the frequencies

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

for $0 < \kappa < \frac{1}{n+2}$,

and initial data defined by

$$\hat{f} = \frac{1}{\sqrt{|\Omega|}} \chi_{\Omega}, \quad \text{so that} \quad \|f\|_2 = 1.$$

This data was introduced in the context of Mattila's question by Barceló–Bennett–Carbery–Ruiz–Vilela (2007).

Note that

$$|\Omega| \simeq \text{number of frequencies} \simeq R^{n\kappa}.$$

The initial data

We consider the frequencies

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

for $0 < \kappa < \frac{1}{n+2}$,

and initial data defined by

$$\hat{f} = \frac{1}{\sqrt{|\Omega|}} \chi_{\Omega}, \quad \text{so that} \quad \|f\|_2 = 1.$$

This data was introduced in the context of Mattila's question by [Barceló–Bennett–Carbery–Ruiz–Vilela \(2007\)](#).

Note that

$$|\Omega| \simeq \text{number of frequencies} \simeq R^{n\kappa}.$$

The initial data

We consider the frequencies

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

for $0 < \kappa < \frac{1}{n+2}$,

and initial data defined by

$$\hat{f} = \frac{1}{\sqrt{|\Omega|}} \chi_{\Omega}, \quad \text{so that} \quad \|f\|_2 = 1.$$

This data was introduced in the context of Mattila's question by [Barceló–Bennett–Carbery–Ruiz–Vilela](#) (2007).

Note that

$$|\Omega| \simeq \text{number of frequencies} \simeq R^{n\kappa}.$$

Periodic constructive interference

The constructive interference reappears periodically in time:

$$|e^{it\Delta}f(x)| \gtrsim \sqrt{|\Omega|} \quad \text{for all } (x, t) \in X \times T,$$

where

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

and

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

Periodic constructive interference

The constructive interference reappears periodically in time:

$$|e^{it\Delta}f(x)| \gtrsim \sqrt{|\Omega|} \quad \text{for all } (x, t) \in X \times T,$$

where

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

and

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

Periodically coherent solutions

X is the dual-set of Ω :

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

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

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

So that there is no cancellation in the integral:

$$e^{it\Delta}f(x) \simeq \frac{1}{\sqrt{|\Omega|}} \int_{\Omega} e^{ix \cdot \xi - it|\xi|^2} d\xi \simeq \frac{|\Omega|}{\sqrt{|\Omega|}}.$$

Periodically coherent solutions

X is the dual-set of Ω :

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

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

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

So that there is no cancellation in the integral:

$$e^{it\Delta}f(x) \simeq \frac{1}{\sqrt{|\Omega|}} \int_{\Omega} e^{ix \cdot \xi - it|\xi|^2} d\xi \simeq \frac{|\Omega|}{\sqrt{|\Omega|}}.$$

Periodically coherent solutions

X is the dual-set of Ω :

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

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

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

So that there is no cancellation in the integral:

$$e^{it\Delta} f(x) \simeq \frac{1}{\sqrt{|\Omega|}} \int_{\Omega} e^{ix \cdot \xi - it|\xi|^2} d\xi \simeq \frac{|\Omega|}{\sqrt{|\Omega|}}.$$

Periodically coherent solutions

X is the dual-set of Ω :

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

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

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

So that there is no cancellation in the integral:

$$e^{it\Delta} f(x) \simeq \frac{1}{\sqrt{|\Omega|}} \int_{\Omega} e^{ix \cdot \xi - it|\xi|^2} d\xi \simeq \frac{|\Omega|}{\sqrt{|\Omega|}}.$$

Periodically coherent solutions

X is the dual-set of Ω :

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

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

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

So that there is no cancellation in the integral:

$$e^{it\Delta} f(x) \simeq \frac{1}{\sqrt{|\Omega|}} \int_{\Omega} e^{ix \cdot \xi - it|\xi|^2} d\xi \simeq \frac{|\Omega|}{\sqrt{|\Omega|}}.$$

Periodically coherent solutions

X is the dual-set of Ω :

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

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

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

So that there is no cancellation in the integral:

$$e^{it\Delta} f(x) \simeq \frac{1}{\sqrt{|\Omega|}} \int_{\Omega} e^{ix \cdot \xi - it|\xi|^2} d\xi \simeq \frac{|\Omega|}{\sqrt{|\Omega|}}.$$

Periodically coherent solutions

Thus

$$|e^{it\Delta}f(x)| \gtrsim \sqrt{|\Omega|} \quad \text{for all } (x, t) \in X \times T,$$

But the interference always reappears in the same places so

$$\sup_{0 < t < 1} |e^{it\Delta}f(x)| \gtrsim \sqrt{|\Omega|} \quad \text{only for } x \in X.$$

Periodically coherent solutions

Thus

$$|e^{it\Delta}f(x)| \gtrsim \sqrt{|\Omega|} \quad \text{for all } (x, t) \in X \times T,$$

But the interference always reappears in the same places so

$$\sup_{0 < t < 1} |e^{it\Delta}f(x)| \gtrsim \sqrt{|\Omega|} \quad \text{only for } x \in X.$$

Travelling periodically coherent solutions

Instead we take

$$f_{\theta}(x) = e^{i\frac{1}{2}\theta \cdot x} f(x), \quad \text{where } \theta \in \mathbb{S}^{n-1}$$

so that

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

which yields

$$\sup_{0 < t < 1} |e^{it\Delta} f_{\theta}(x)| \gtrsim \sqrt{|\Omega|} \quad \text{for all } x \in \bigcup_{t \in T} X + t\theta.$$

Travelling periodically coherent solutions

Instead we take

$$f_{\theta}(x) = e^{i\frac{1}{2}\theta \cdot x} f(x), \quad \text{where } \theta \in \mathbb{S}^{n-1}$$

so that

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

which yields

$$\sup_{0 < t < 1} |e^{it\Delta} f_{\theta}(x)| \gtrsim \sqrt{|\Omega|} \quad \text{for all } x \in \bigcup_{t \in T} X + t\theta.$$

Lemma

Let $0 < \kappa < \frac{1}{n+2}$. Then there exists $\theta \in \mathbb{S}^{n-1}$ 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 $\kappa \geq \frac{1}{n+2}$.

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

Lemma (Lucà-R.)

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

$$|y - t\theta| \leq R^{-(1-\delta)/n} \log R.$$

Lemma

Let $0 < \kappa < \frac{1}{n+2}$. Then there exists $\theta \in \mathbb{S}^{n-1}$ 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 $\kappa \geq \frac{1}{n+2}$.

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

Lemma (Lucà-R.)

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

$$|y - t\theta| \leq R^{-(1-\delta)/n} \log R.$$

Lemma

Let $0 < \kappa < \frac{1}{n+2}$. Then there exists $\theta \in \mathbb{S}^{n-1}$ 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 $\kappa \geq \frac{1}{n+2}$.

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

Lemma (Lucà–R.)

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

$$|y - t\theta| \leq R^{-(1-\delta)/n} \log R.$$

Conclusion of the proof

Plugging into the maximal estimate,

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

recalling that

$$\sup_{0 < t < 1} |e^{it\Delta} f_\theta| \geq \sqrt{|\Omega|} \quad \text{on } B(0, 1/2),$$

we obtain

$$\sqrt{|\Omega|} \lesssim R^s \|f_\theta\|_2.$$

Then as $|\Omega| \gtrsim R^{n\kappa}$ and $\|f_\theta\|_2 = 1$, this yields

$$\Rightarrow s \geq \frac{n\kappa}{2} \quad \text{and then we take } \kappa \rightarrow \frac{1}{n+2}.$$

Conclusion of the proof

Plugging into the maximal estimate,

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

recalling that

$$\sup_{0 < t < 1} |e^{it\Delta} f_\theta| \geq \sqrt{|\Omega|} \quad \text{on } B(0, 1/2),$$

we obtain

$$\sqrt{|\Omega|} \lesssim R^s \|f_\theta\|_2.$$

Then as $|\Omega| \gtrsim R^{n\kappa}$ and $\|f_\theta\|_2 = 1$, this yields

$$\Rightarrow s \geq \frac{n\kappa}{2} \quad \text{and then we take } \kappa \rightarrow \frac{1}{n+2}.$$

Conclusion of the proof

Plugging into the maximal estimate,

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

recalling that

$$\sup_{0 < t < 1} |e^{it\Delta} f_\theta| \geq \sqrt{|\Omega|} \quad \text{on } B(0, 1/2),$$

we obtain

$$\sqrt{|\Omega|} \lesssim R^s \|f_\theta\|_2.$$

Then as $|\Omega| \gtrsim R^{n\kappa}$ and $\|f_\theta\|_2 = 1$, this yields

$$\Rightarrow s \geq \frac{n\kappa}{2} \quad \text{and then we take } \kappa \rightarrow \frac{1}{n+2}.$$

Conclusion of the proof

Plugging into the maximal estimate,

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

recalling that

$$\sup_{0 < t < 1} |e^{it\Delta} f_\theta| \geq \sqrt{|\Omega|} \quad \text{on } B(0, 1/2),$$

we obtain

$$\sqrt{|\Omega|} \lesssim R^s \|f_\theta\|_2.$$

Then as $|\Omega| \gtrsim R^{n\kappa}$ and $\|f_\theta\|_2 = 1$, this yields

$$\Rightarrow s \geq \frac{n\kappa}{2} \quad \text{and then we take } \kappa \rightarrow \frac{1}{n+2}.$$

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

This follows from:

Theorem (Lucà–Rogers)

Let $n/2 \leq \alpha \leq n$ and suppose that, for all $u_0 \in H^s(\mathbb{R}^n)$,

$$\lim_{t \rightarrow 0} e^{it\Delta} u_0(x) = u_0(x)$$

for all x off an α -dimensional set. Then

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

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

This follows from:

Theorem (Lucà–Rogers)

Let $n/2 \leq \alpha \leq n$ and suppose that, for all $u_0 \in H^s(\mathbb{R}^n)$,

$$\lim_{t \rightarrow 0} e^{it\Delta} u_0(x) = u_0(x)$$

for all x off an α -dimensional set. Then

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

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

This follows from:

Theorem (Lucà–Rogers)

Let $n/2 \leq \alpha \leq n$ and suppose that, for all $u_0 \in H^s(\mathbb{R}^n)$,

$$\lim_{t \rightarrow 0} e^{it\Delta} u_0(x) = u_0(x)$$

for all x off an α -dimensional set. 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

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

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 $\frac{1}{2} < \kappa < \frac{n-\alpha+1}{2}$.

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

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

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 $\frac{1}{2} < \kappa < \frac{n-\alpha+1}{2}$.

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

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

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 $\frac{1}{r+2} < \kappa < \frac{n-\alpha+1}{n+2}$.

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

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

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 $\frac{1}{r+2} < \kappa < \frac{n-\alpha+1}{n+2}$.

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

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

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 α -dimensional.

For this we use that the limit is ' α -Hausdorff dense'.

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 α -dimensional.

For this we use that the limit is ' α -Hausdorff dense'.

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 α -dimensional.

For this we use that the limit is ' α -Hausdorff dense'.

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 α -dimensional.

For this we use that the limit is ' α -Hausdorff dense'.

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 α -dimensional.

For this we use that the limit is ' α -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

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

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

Part 5:

Decay for the Fourier transform of fractal measures

$\widehat{\delta_{x_n=0}}(R(\bar{\xi}, \xi_n)) = \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^{n-1}} e^{iR\bar{x}\cdot\bar{\xi}} d\bar{x}$ is independent of ξ_n .

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 β for which

$$\|\widehat{\mu}(R\cdot)\|_{L^2(S^{n-1})}^2 \lesssim c_\alpha(\mu) \|\mu\| R^{-\beta}$$

whenever $R > 1$ and μ is α -dimensional and supported in $B(0, 1)$.

Question (Mattila (1987))

Who is $\beta_n(\alpha)$?

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

$$\|(gd\sigma)^\vee(R\cdot)\|_{L^1(d\mu)} \lesssim \sqrt{c_\alpha(\mu) \|\mu\|} R^{-\beta/2} \|g\|_{L^2(S^{n-1})}.$$

$\widehat{\delta_{x_n=0}}(R(\bar{\xi}, \xi_n)) = \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^{n-1}} e^{iR\bar{x} \cdot \bar{\xi}} d\bar{x}$ is independent of ξ_n .

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 β for which

$$\|\widehat{\mu}(R \cdot)\|_{L^2(\mathbb{S}^{n-1})}^2 \lesssim c_\alpha(\mu) \|\mu\| R^{-\beta}$$

whenever $R > 1$ and μ is α -dimensional and supported in $B(0, 1)$.

Question (Mattila (1987))

Who is $\beta_n(\alpha)$?

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

$$\|(gd\sigma)^\vee(R \cdot)\|_{L^1(d\mu)} \lesssim \sqrt{c_\alpha(\mu) \|\mu\|} R^{-\beta/2} \|g\|_{L^2(\mathbb{S}^{n-1})}.$$

$\widehat{\delta_{x_n=0}}(R(\bar{\xi}, \xi_n)) = \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^{n-1}} e^{iR\bar{x} \cdot \bar{\xi}} d\bar{x}$ is independent of ξ_n .

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 β for which

$$\|\widehat{\mu}(R \cdot)\|_{L^2(\mathbb{S}^{n-1})}^2 \lesssim c_\alpha(\mu) \|\mu\| R^{-\beta}$$

whenever $R > 1$ and μ is α -dimensional and supported in $B(0, 1)$.

Question (Mattila (1987))

Who is $\beta_n(\alpha)$?

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

$$\|(gd\sigma)^\vee(R \cdot)\|_{L^1(d\mu)} \lesssim \sqrt{c_\alpha(\mu) \|\mu\|} R^{-\beta/2} \|g\|_{L^2(\mathbb{S}^{n-1})}.$$

$\widehat{\delta_{x_n=0}}(R(\bar{\xi}, \xi_n)) = \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^{n-1}} e^{iR\bar{x} \cdot \bar{\xi}} d\bar{x}$ is independent of ξ_n .

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 β for which

$$\|\widehat{\mu}(R \cdot)\|_{L^2(\mathbb{S}^{n-1})}^2 \lesssim c_\alpha(\mu) \|\mu\| R^{-\beta}$$

whenever $R > 1$ and μ is α -dimensional and supported in $B(0, 1)$.

Question (Mattila (1987))

Who is $\beta_n(\alpha)$?

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

$$\|(gd\sigma)^\vee(R \cdot)\|_{L^1(d\mu)} \lesssim \sqrt{c_\alpha(\mu) \|\mu\|} R^{-\beta/2} \|g\|_{L^2(\mathbb{S}^{n-1})}.$$

$\widehat{\delta_{x_n=0}}(R(\bar{\xi}, \xi_n)) = \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^{n-1}} e^{iR\bar{x} \cdot \bar{\xi}} d\bar{x}$ is independent of ξ_n .

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 β for which

$$\|\widehat{\mu}(R \cdot)\|_{L^2(\mathbb{S}^{n-1})}^2 \lesssim c_\alpha(\mu) \|\mu\| R^{-\beta}$$

whenever $R > 1$ and μ is α -dimensional and supported in $B(0, 1)$.

Question (Mattila (1987))

Who is $\beta_n(\alpha)$?

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

$$\|(gd\sigma)^\vee(R \cdot)\|_{L^1(d\mu)} \lesssim \sqrt{c_\alpha(\mu) \|\mu\|} R^{-\beta/2} \|g\|_{L^2(\mathbb{S}^{n-1})}.$$

$\widehat{\delta_{x_n=0}}(R(\bar{\xi}, \xi_n)) = \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^{n-1}} e^{iR\bar{x}\cdot\bar{\xi}} d\bar{x}$ is independent of ξ_n .

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 β for which

$$\|\widehat{\mu}(R\cdot)\|_{L^2(\mathbb{S}^{n-1})}^2 \lesssim c_\alpha(\mu)\|\mu\| R^{-\beta}$$

whenever $R > 1$ and μ is α -dimensional and supported in $B(0, 1)$.

Question (Mattila (1987))

Who is $\beta_n(\alpha)$?

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

$$\|(gd\sigma)^\vee(R\cdot)\|_{L^1(d\mu)} \lesssim \sqrt{c_\alpha(\mu)\|\mu\|} R^{-\beta/2} \|g\|_{L^2(\mathbb{S}^{n-1})}.$$

Previous results

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

$$\beta_n(\alpha) \geq \begin{cases} \alpha, & \alpha \in (0, \frac{n-1}{2}], \\ \frac{n-1}{2}, & \alpha \in [\frac{n-1}{2}, \frac{n}{2}], \\ \alpha - 1 + \frac{n+2-2\alpha}{4}, & \alpha \in [\frac{n}{2}, \frac{n+2}{2}], \\ \alpha - 1, & \alpha \in [\frac{n+2}{2}, n], \end{cases} \quad \begin{array}{l} \text{Mattila (1987)} \\ \\ \text{Erdogan (2005)} \\ \text{Sjolin (1993).} \end{array}$$

Previous results

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

$$\beta_n(\alpha) \geq \begin{cases} \alpha, & \alpha \in (0, \frac{n-1}{2}], \\ \frac{n-1}{2}, & \alpha \in [\frac{n-1}{2}, \frac{n}{2}], \\ \alpha - 1 + \frac{n+2-2\alpha}{4}, & \alpha \in [\frac{n}{2}, \frac{n+2}{2}], \\ \alpha - 1, & \alpha \in [\frac{n+2}{2}, n], \end{cases} \quad \begin{array}{l} \text{Mattila (1987)} \\ \\ \text{Erdoğ an (2005)} \\ \\ \text{Sjölin (1993).} \end{array}$$

Lemma (Bridging lemma)

Let u be a solution to $\partial_t u = i(-\Delta)^{m/2} u$ with initial data $u \in \dot{H}^s(\mathbb{R}^n)$ with $0 < s < n/2$. Then if $\beta_n(\alpha) > n - 2s$, then

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

Proof: It will suffice to prove, for all α -dimensional μ ,

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

Writing $\hat{f} = |\cdot|^{-s} \hat{g}$ and using polar coordinates,

$$\begin{aligned} & (2\pi)^{n/2} |e^{it(-\Delta)^{m/2}} f(x)| \\ &= \left| \int_{\mathbb{R}^n} e^{-it|\xi|^m} |\xi|^{-s} \hat{g}(\xi) e^{ix \cdot \xi} d\xi \right| \\ &= \left| \int_0^\infty e^{-itR^m} R^{n-1-s} \int_{\mathbb{S}^{n-1}} \hat{g}(R\omega) e^{iR x \cdot \omega} d\sigma(\omega) dR \right|. \end{aligned}$$

Lemma (Bridging lemma)

Let u be a solution to $\partial_t u = i(-\Delta)^{m/2} u$ with initial data $u \in \dot{H}^s(\mathbb{R}^n)$ with $0 < s < n/2$. Then if $\beta_n(\alpha) > n - 2s$, then

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

Proof: It will suffice to prove, for all α -dimensional μ ,

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

Writing $\hat{f} = |\cdot|^{-s} \hat{g}$ and using polar coordinates,

$$\begin{aligned} & (2\pi)^{n/2} |e^{it(-\Delta)^{m/2}} f(x)| \\ &= \left| \int_{\mathbb{R}^n} e^{-it|\xi|^m} |\xi|^{-s} \hat{g}(\xi) e^{ix \cdot \xi} d\xi \right| \\ &= \left| \int_0^\infty e^{-itR^m} R^{n-1-s} \int_{\mathbb{S}^{n-1}} \hat{g}(R\omega) e^{iR x \cdot \omega} d\sigma(\omega) dR \right|. \end{aligned}$$

Lemma (Bridging lemma)

Let u be a solution to $\partial_t u = i(-\Delta)^{m/2} u$ with initial data $u \in \dot{H}^s(\mathbb{R}^n)$ with $0 < s < n/2$. Then if $\beta_n(\alpha) > n - 2s$, then

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

Proof: It will suffice to prove, for all α -dimensional μ ,

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

Writing $\hat{f} = |\cdot|^{-s} \hat{g}$ and using polar coordinates,

$$\begin{aligned} & (2\pi)^{n/2} |e^{it(-\Delta)^{m/2}} f(x)| \\ &= \left| \int_{\mathbb{R}^n} e^{-it|\xi|^m} |\xi|^{-s} \hat{g}(\xi) e^{ix \cdot \xi} d\xi \right| \\ &= \left| \int_0^\infty e^{-itR^m} R^{n-1-s} \int_{\mathbb{S}^{n-1}} \hat{g}(R\omega) e^{iR x \cdot \omega} d\sigma(\omega) dR \right|. \end{aligned}$$

Lemma (Bridging lemma)

Let u be a solution to $\partial_t u = i(-\Delta)^{m/2} u$ with initial data $u \in \dot{H}^s(\mathbb{R}^n)$ with $0 < s < n/2$. Then if $\beta_n(\alpha) > n - 2s$, then

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

Proof: It will suffice to prove, for all α -dimensional μ ,

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

Writing $\hat{f} = |\cdot|^{-s} \hat{g}$ and using polar coordinates,

$$\begin{aligned} & (2\pi)^{n/2} |e^{it(-\Delta)^{m/2}} f(x)| \\ &= \left| \int_{\mathbb{R}^n} e^{-it|\xi|^m} |\xi|^{-s} \hat{g}(\xi) e^{ix \cdot \xi} d\xi \right| \\ &= \left| \int_0^\infty e^{-itR^m} R^{n-1-s} \int_{\mathbb{S}^{n-1}} \hat{g}(R\omega) e^{iR x \cdot \omega} d\sigma(\omega) dR \right|. \end{aligned}$$

Lemma (Bridging lemma)

Let u be a solution to $\partial_t u = i(-\Delta)^{m/2} u$ with initial data $u \in \dot{H}^s(\mathbb{R}^n)$ with $0 < s < n/2$. Then if $\beta_n(\alpha) > n - 2s$, then

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

Proof: It will suffice to prove, for all α -dimensional μ ,

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

Writing $\hat{f} = |\cdot|^{-s} \hat{g}$ and using polar coordinates,

$$\begin{aligned} & (2\pi)^{n/2} |e^{it(-\Delta)^{m/2}} f(x)| \\ &= \left| \int_{\mathbb{R}^n} e^{-it|\xi|^m} |\xi|^{-s} \hat{g}(\xi) e^{ix \cdot \xi} d\xi \right| \\ &= \left| \int_0^\infty e^{-itR^m} R^{n-1-s} \int_{\mathbb{S}^{n-1}} \hat{g}(R\omega) e^{iR x \cdot \omega} d\sigma(\omega) dR \right|. \end{aligned}$$

$$|e^{it(-\Delta)^{m/2}} f(x)| \lesssim \int_0^\infty R^{n-1-s} \left| \int_{\mathbb{S}^{n-1}} \widehat{g}(R\omega) e^{iRx \cdot \omega} d\sigma(\omega) \right| dR,$$

so that, by Fubini,

$$\left\| \sup_{t \in \mathbb{R}} |e^{it(-\Delta)^{m/2}} f| \right\|_{L^1(d\mu)} \lesssim \int_0^\infty R^{n-1-s} \|(\widehat{g}(R \cdot) d\sigma)^\vee(R \cdot)\|_{L^1(d\mu)} dR.$$

By the dual version of the Mattila inequality,

$$\|(\widehat{g}(R \cdot) d\sigma)^\vee(R \cdot)\|_{L^1(d\mu)} \leq C_\mu (1+R)^{-\beta/2} \|\widehat{g}(R \cdot)\|_{L^2(\mathbb{S}^{n-1})}.$$

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

$$\left\| \sup_{t \in \mathbb{R}} |e^{it(-\Delta)^{m/2}} f| \right\|_{L^1(d\mu)} \leq C_\mu \int_0^\infty \frac{R^{n-1-s}}{(1+R)^{\beta/2}} \|\widehat{g}(R \cdot)\|_{L^2(\mathbb{S}^{n-1})} dR.$$

Finally, by Cauchy–Schwarz,

$$\begin{aligned} &\leq C_\mu \left(\int_0^\infty \frac{R^{n-1-2s}}{(1+R)^\beta} dR \right)^{1/2} \left(\int_0^\infty \|\widehat{g}(R \cdot)\|_{L^2(\mathbb{S}^{n-1})}^2 R^{n-1} dR \right)^{1/2} \\ &\leq C_\mu \|g\|_{L^2(\mathbb{R}^n)}, \end{aligned}$$

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

$$|e^{it(-\Delta)^{m/2}} f(x)| \lesssim \int_0^\infty R^{n-1-s} \left| \int_{\mathbb{S}^{n-1}} \widehat{g}(R\omega) e^{iR x \cdot \omega} d\sigma(\omega) \right| dR,$$

so that, by Fubini,

$$\left\| \sup_{t \in \mathbb{R}} |e^{it(-\Delta)^{m/2}} f| \right\|_{L^1(d\mu)} \lesssim \int_0^\infty R^{n-1-s} \left\| (\widehat{g}(R \cdot) d\sigma)^\vee(R \cdot) \right\|_{L^1(d\mu)} dR.$$

By the dual version of the Mattila inequality,

$$\left\| (\widehat{g}(R \cdot) d\sigma)^\vee(R \cdot) \right\|_{L^1(d\mu)} \leq C_\mu (1+R)^{-\beta/2} \|\widehat{g}(R \cdot)\|_{L^2(\mathbb{S}^{n-1})}.$$

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

$$\left\| \sup_{t \in \mathbb{R}} |e^{it(-\Delta)^{m/2}} f| \right\|_{L^1(d\mu)} \leq C_\mu \int_0^\infty \frac{R^{n-1-s}}{(1+R)^{\beta/2}} \|\widehat{g}(R \cdot)\|_{L^2(\mathbb{S}^{n-1})} dR.$$

Finally, by Cauchy-Schwarz,

$$\begin{aligned} &\leq C_\mu \left(\int_0^\infty \frac{R^{n-1-2s}}{(1+R)^\beta} dR \right)^{1/2} \left(\int_0^\infty \|\widehat{g}(R \cdot)\|_{L^2(\mathbb{S}^{n-1})}^2 R^{n-1} dR \right)^{1/2} \\ &\leq C_\mu \|g\|_{L^2(\mathbb{R}^n)}, \end{aligned}$$

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

$$|e^{it(-\Delta)^{m/2}} f(x)| \lesssim \int_0^\infty R^{n-1-s} \left| \int_{\mathbb{S}^{n-1}} \widehat{g}(R\omega) e^{iR x \cdot \omega} d\sigma(\omega) \right| dR,$$

so that, by Fubini,

$$\left\| \sup_{t \in \mathbb{R}} |e^{it(-\Delta)^{m/2}} f| \right\|_{L^1(d\mu)} \lesssim \int_0^\infty R^{n-1-s} \left\| (\widehat{g}(R \cdot) d\sigma)^\vee(R \cdot) \right\|_{L^1(d\mu)} dR.$$

By the dual version of the Mattila inequality,

$$\left\| (\widehat{g}(R \cdot) d\sigma)^\vee(R \cdot) \right\|_{L^1(d\mu)} \leq C_\mu (1+R)^{-\beta/2} \|\widehat{g}(R \cdot)\|_{L^2(\mathbb{S}^{n-1})}.$$

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

$$\left\| \sup_{t \in \mathbb{R}} |e^{it(-\Delta)^{m/2}} f| \right\|_{L^1(d\mu)} \leq C_\mu \int_0^\infty \frac{R^{n-1-s}}{(1+R)^{\beta/2}} \|\widehat{g}(R \cdot)\|_{L^2(\mathbb{S}^{n-1})} dR.$$

Finally, by Cauchy-Schwarz,

$$\begin{aligned} &\leq C_\mu \left(\int_0^\infty \frac{R^{n-1-2s}}{(1+R)^\beta} dR \right)^{1/2} \left(\int_0^\infty \|\widehat{g}(R \cdot)\|_{L^2(\mathbb{S}^{n-1})}^2 R^{n-1} dR \right)^{1/2} \\ &\leq C_\mu \|g\|_{L^2(\mathbb{R}^n)}, \end{aligned}$$

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

$$|e^{it(-\Delta)^{m/2}} f(x)| \lesssim \int_0^\infty R^{n-1-s} \left| \int_{\mathbb{S}^{n-1}} \widehat{g}(R\omega) e^{iRx \cdot \omega} d\sigma(\omega) \right| dR,$$

so that, by Fubini,

$$\left\| \sup_{t \in \mathbb{R}} |e^{it(-\Delta)^{m/2}} f| \right\|_{L^1(d\mu)} \lesssim \int_0^\infty R^{n-1-s} \left\| (\widehat{g}(R \cdot) d\sigma)^\vee(R \cdot) \right\|_{L^1(d\mu)} dR.$$

By the dual version of the Mattila inequality,

$$\left\| (\widehat{g}(R \cdot) d\sigma)^\vee(R \cdot) \right\|_{L^1(d\mu)} \leq C_\mu (1+R)^{-\beta/2} \|\widehat{g}(R \cdot)\|_{L^2(\mathbb{S}^{n-1})}.$$

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

$$\left\| \sup_{t \in \mathbb{R}} |e^{it(-\Delta)^{m/2}} f| \right\|_{L^1(d\mu)} \leq C_\mu \int_0^\infty \frac{R^{n-1-s}}{(1+R)^{\beta/2}} \|\widehat{g}(R \cdot)\|_{L^2(\mathbb{S}^{n-1})} dR.$$

Finally, by Cauchy-Schwarz,

$$\begin{aligned} &\leq C_\mu \left(\int_0^\infty \frac{R^{n-1-2s}}{(1+R)^\beta} dR \right)^{1/2} \left(\int_0^\infty \|\widehat{g}(R \cdot)\|_{L^2(\mathbb{S}^{n-1})}^2 R^{n-1} dR \right)^{1/2} \\ &\leq C_\mu \|g\|_{L^2(\mathbb{R}^n)}, \end{aligned}$$

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

$$|e^{it(-\Delta)^{m/2}} f(x)| \lesssim \int_0^\infty R^{n-1-s} \left| \int_{\mathbb{S}^{n-1}} \widehat{g}(R\omega) e^{iRx \cdot \omega} d\sigma(\omega) \right| dR,$$

so that, by Fubini,

$$\left\| \sup_{t \in \mathbb{R}} |e^{it(-\Delta)^{m/2}} f| \right\|_{L^1(d\mu)} \lesssim \int_0^\infty R^{n-1-s} \left\| (\widehat{g}(R \cdot) d\sigma)^\vee(R \cdot) \right\|_{L^1(d\mu)} dR.$$

By the dual version of the Mattila inequality,

$$\left\| (\widehat{g}(R \cdot) d\sigma)^\vee(R \cdot) \right\|_{L^1(d\mu)} \leq C_\mu (1+R)^{-\beta/2} \|\widehat{g}(R \cdot)\|_{L^2(\mathbb{S}^{n-1})}.$$

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

$$\left\| \sup_{t \in \mathbb{R}} |e^{it(-\Delta)^{m/2}} f| \right\|_{L^1(d\mu)} \leq C_\mu \int_0^\infty \frac{R^{n-1-s}}{(1+R)^{\beta/2}} \|\widehat{g}(R \cdot)\|_{L^2(\mathbb{S}^{n-1})} dR.$$

Finally, by Cauchy-Schwarz,

$$\begin{aligned} &\leq C_\mu \left(\int_0^\infty \frac{R^{n-1-2s}}{(1+R)^\beta} dR \right)^{1/2} \left(\int_0^\infty \|\widehat{g}(R \cdot)\|_{L^2(\mathbb{S}^{n-1})}^2 R^{n-1} dR \right)^{1/2} \\ &\leq C_\mu \|g\|_{L^2(\mathbb{R}^n)}, \end{aligned}$$

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

$$|e^{it(-\Delta)^{m/2}} f(x)| \lesssim \int_0^\infty R^{n-1-s} \left| \int_{\mathbb{S}^{n-1}} \widehat{g}(R\omega) e^{iR x \cdot \omega} d\sigma(\omega) \right| dR,$$

so that, by Fubini,

$$\left\| \sup_{t \in \mathbb{R}} |e^{it(-\Delta)^{m/2}} f| \right\|_{L^1(d\mu)} \lesssim \int_0^\infty R^{n-1-s} \left\| (\widehat{g}(R \cdot) d\sigma)^\vee(R \cdot) \right\|_{L^1(d\mu)} dR.$$

By the dual version of the Mattila inequality,

$$\left\| (\widehat{g}(R \cdot) d\sigma)^\vee(R \cdot) \right\|_{L^1(d\mu)} \leq C_\mu (1+R)^{-\beta/2} \|\widehat{g}(R \cdot)\|_{L^2(\mathbb{S}^{n-1})}.$$

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

$$\left\| \sup_{t \in \mathbb{R}} |e^{it(-\Delta)^{m/2}} f| \right\|_{L^1(d\mu)} \leq C_\mu \int_0^\infty \frac{R^{n-1-s}}{(1+R)^{\beta/2}} \|\widehat{g}(R \cdot)\|_{L^2(\mathbb{S}^{n-1})} dR.$$

Finally, by Cauchy–Schwarz,

$$\begin{aligned} &\leq C_\mu \left(\int_0^\infty \frac{R^{n-1-2s}}{(1+R)^\beta} dR \right)^{1/2} \left(\int_0^\infty \|\widehat{g}(R \cdot)\|_{L^2(\mathbb{S}^{n-1})}^2 R^{n-1} dR \right)^{1/2} \\ &\leq C_\mu \|g\|_{L^2(\mathbb{R}^n)}, \end{aligned}$$

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



Part 6:

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

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

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 $(u_0, u_1) \in \dot{H}^s \times \dot{H}^{s-1}$.

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

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

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 $(u_0, u_1) \in \dot{H}^s \times \dot{H}^{s-1}$.

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

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

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 $(u_0, u_1) \in \dot{H}^s \times \dot{H}^{s-1}$.

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

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

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 $(u_0, u_1) \in \dot{H}^s \times \dot{H}^{s-1}$.

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

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

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 $(u_0, u_1) \in \dot{H}^s \times \dot{H}^{s-1}$.

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

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

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 $(u_0, u_1) \in \dot{H}^s \times \dot{H}^{s-1}$

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

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

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 $(u_0, u_1) \in \dot{H}^s \times \dot{H}^{s-1}$.

Corollary (of bridging lemma and Sjölin's estimate)

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

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

Proof: By the result of Sjölin, $\beta(\alpha) \geq \alpha - 1$ so that $\beta(\alpha) > n - 2s$ as long as $\alpha > n - 2s + 1$. Thus, by the bridging lemma,

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

□

Corollary (of the corollary)

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

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

Corollary (of bridging lemma and Sjölin's estimate)

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

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

Proof: By the result of Sjölin, $\beta(\alpha) \geq \alpha - 1$ so that $\beta(\alpha) > n - 2s$ as long as $\alpha > n - 2s + 1$. Thus, by the bridging lemma,

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

□

Corollary (of the corollary)

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

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

Corollary (of bridging lemma and Sjölin's estimate)

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

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

Proof: By the result of Sjölin, $\beta(\alpha) \geq \alpha - 1$ so that $\beta(\alpha) > n - 2s$ as long as $\alpha > n - 2s + 1$. Thus, by the bridging lemma,

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

□

Corollary (of the corollary)

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

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

Theorem (Lucà–R.)

Let $n \geq 3$. Then

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

This is an improvement in the range $n/2 + 1 \leq \alpha < n$.

The proof takes advantage of:

- ▶ 'multilinear restriction' estimates due to Bennett–Carbery–Tao
- ▶ 'decomposition' of Bourgain–Guth.
- ▶ 'interpolation' with the argument of Sjölin.

Corollary

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

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

Thus the solution cannot diverge on spheres.

Theorem (Lucà–R.)

Let $n \geq 3$. Then

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

This is an improvement in the range $n/2 + 1 \leq \alpha < n$.

The proof takes advantage of:

- ▶ 'multilinear restriction' estimates due to Bennett–Carbery–Tao
- ▶ 'decomposition' of Bourgain–Guth.
- ▶ 'interpolation' with the argument of Sjölin.

Corollary

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

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

Thus the solution cannot diverge on spheres.

Theorem (Lucà–R.)

Let $n \geq 3$. Then

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

This is an improvement in the range $n/2 + 1 \leq \alpha < n$.

The proof takes advantage of:

- ▶ 'multilinear restriction' estimates due to Bennett–Carbery–Tao
- ▶ 'decomposition' of Bourgain–Guth.
- ▶ 'interpolation' with the argument of Sjölin.

Corollary

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

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

Thus the solution cannot diverge on spheres.

Theorem (Lucà–R.)

Let $n \geq 3$. Then

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

This is an improvement in the range $n/2 + 1 \leq \alpha < n$.

The proof takes advantage of:

- ▶ 'multilinear restriction' estimates due to **Bennett–Carbery–Tao**
- ▶ 'decomposition' of **Bourgain–Guth**.
- ▶ 'interpolation' with the argument of **Sjölin**.

Corollary

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

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

Thus the solution cannot diverge on spheres.

Theorem (Lucà–R.)

Let $n \geq 3$. Then

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

This is an improvement in the range $n/2 + 1 \leq \alpha < n$.

The proof takes advantage of:

- ▶ 'multilinear restriction' estimates due to **Bennett–Carbery–Tao**
- ▶ 'decomposition' of **Bourgain–Guth**.
- ▶ 'interpolation' with the argument of **Sjölin**.

Corollary

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

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

Thus the solution cannot diverge on spheres.

Theorem (Lucà–R.)

Let $n \geq 3$. Then

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

This is an improvement in the range $n/2 + 1 \leq \alpha < n$.

The proof takes advantage of:

- ▶ ‘multilinear restriction’ estimates due to **Bennett–Carbery–Tao**
- ▶ ‘decomposition’ of **Bourgain–Guth**.
- ▶ ‘interpolation’ with the argument of **Sjölin**.

Corollary

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

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

Thus the solution cannot diverge on spheres.

Theorem (Lucà–R.)

Let $n \geq 3$. Then

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

This is an improvement in the range $n/2 + 1 \leq \alpha < n$.

The proof takes advantage of:

- ▶ ‘multilinear restriction’ estimates due to **Bennett–Carbery–Tao**
- ▶ ‘decomposition’ of **Bourgain–Guth**.
- ▶ ‘interpolation’ with the argument of **Sjölin**.

Corollary

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

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

Thus the solution cannot diverge on spheres.

Arigatou gozaimasu!