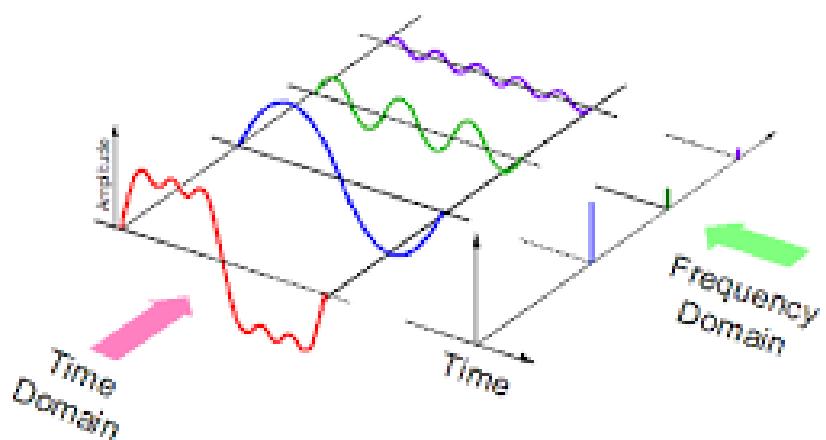

Harmonic Analysis

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Introduction

A brief overview over the material of this lecture:

1. Fourier Transform: Let $f : [0, 1] \rightarrow \mathbb{C}$ and periodic, then $f(x) = \sum_{n \in \mathbb{Z}} \hat{f}(n)e^{2\pi nix}$, where $\hat{f} = \int_0^1 f(x)e^{-2\pi nix} dx$. This decomposition decomposes f from sth uncountable to sth countable. Fourier transform in general for $f : \mathbb{R}^d \rightarrow \mathbb{C}$:

$$\hat{f}(k) = \int_{\mathbb{R}^d} f(x)e^{-2\pi kix} dx$$

Inverse Fourier transform:

$$\hat{f}(x) = \int_{\mathbb{R}^d} \hat{f}(k)e^{2\pi kix} dk$$

2. Convolution form: Fourier transform has the form $\hat{f}(k) = \int f(x)g(kx)dx$

- $f, g : \mathbb{R}^d \rightarrow \mathbb{C}$: $(f \star g)(x) = \int_{\mathbb{R}^d} f(x-y)g(y)dy$
 $= \int_{\mathbb{R}^d} f(y)g(x-y)dy = (g \star f)(x)$
- $\widehat{f \star g}(k) = (\hat{f} \cdot \hat{g})(k)$
- If f smooth: $\widehat{\partial x_j f}(k) = (2\pi i k_j) \hat{f}(k)$
- Define operator: $G : \begin{cases} L^2(\mathbb{R}^d) \rightarrow L^2(\mathbb{R}^d) \\ \widehat{Gf}(k) = \hat{g}(k)\hat{f}(k) \end{cases}$, then

$$Gf(x) = (g \star f)(x) = \int_{\mathbb{R}^d} g(x-y)f(y)dy$$

g is called kernel of G

- Example: $-\Delta u = f$ in $\mathbb{R}^d \longleftrightarrow u = (-\Delta)^{-1}f = g \star f$, where $\hat{g}(k) = \frac{1}{|2\pi k|^2}$. If \mathbb{R}^3 , then $g(x) = \frac{1}{4\pi|x|^2}$ is the Coulomb potential.

3. Hardy-Littlewood maximal function: $f \in L^1(\mathbb{R}^d)$,

$$M_f = \sup_{R>0} \fint_{B(x,R)} |f| = \sup_{R>0} \frac{1}{|B(x,R)|} \int_{B(x,R)} |f|$$

Then $M_f(x) \geq |f(x)|$ and

$$\|M_f\|_{L^p(\mathbb{R}^d)} \leq C \|f\|_{L^p(\mathbb{R}^d)} \text{ for all } p > 1 \text{ ("weak form" if } p = 1)$$

Theorem 0.1 (Lebesgue Differentiation Theorem). If $f \in L^1_{loc}$, then for a.e. x

$$\lim_{r \rightarrow 0} \fint_{B(x,r)} f = f(x)$$

To prove this one needs the so called "Covering Lemmas"

4. Theory of interpolation and decomposition:

$$T : \begin{cases} X_0 \rightarrow Y_0 \\ X_1 \rightarrow Y_1 \end{cases}, T : X_\Theta \rightarrow Y_\Theta, \Theta \in [0, 1]$$

5. Applications:

- Quantum mechanics: *Heisenberg uncertainty principle*, i.e.

$$(p, x) \mapsto \begin{cases} \Psi \in L^2(\mathbb{R}^d) \\ |\Psi(x)|^2 = \text{probability density in } x \\ |\hat{\Psi}(x)|^2 = \text{probability density in } p \end{cases}$$

Hardy uncertainty principle: $\widehat{e^{-\pi x^2}} = e^{-\pi k^2}$, if $f(x) \underset{\substack{\sim \\ |x| \rightarrow \infty}}{\lesssim} e^{-\pi \alpha x^2}$

$\Rightarrow \hat{f}(k) \gtrsim e^{-\pi k^2}$ as $|x| \rightarrow \infty$

Sobolev uncertainty principle: $f, g : \mathbb{R}^d \rightarrow \mathbb{C}$, operator $f(x)g(x)$ compact, if $f, g \rightarrow 0$ at ∞

$\Rightarrow H^1(\Omega_{bd}) \subset\subset L^2(\Omega)$

- PDE: $\min \mathcal{E}(u)$, $\mathcal{E}(u) = \int_{\mathbb{R}^d} |\nabla u|^2 + V|u|^2$

\rightsquigarrow Schrödinger eq. $-\Delta u + Vu = Eu$

Harmonic function $\Delta u = 0$

$\Delta u \geq u \Rightarrow u(x) \leq \fint_{B(x,r)} u$ (max principle) and maximum attained on boundary

- Potential theory \rightsquigarrow Newton thm: $\int_{\mathbb{R}^3} \int_{\mathbb{R}^3} \frac{f(x)g(y)}{|x-y|} = \frac{\int f \cdot \int g}{d}$

$\Delta \frac{1}{|x|} = 0$ for all $x \in \mathbb{R}^3 \setminus \{0\}$

- Number theory: $f : \mathbb{R}^d \rightarrow \mathbb{C}$

Poisson formula: if f smooth, decay fast at ∞ :

$$\sum_{x \in \mathbb{Z}^d} f(x) = \sum_{k \in \mathbb{Z}^d} \hat{f}(k)$$

Lets have a closer look at an example of the applications in number theory:

Gauss circle problem (*still unsolved*)

$$\begin{aligned} S_R &= \#\text{integer points inside circle } B(0, R) \\ &= \#\{(x_1, x_2) : x_1, x_2 \in \mathbb{Z}, x_1^2 + x_2^2 \leq R\} \end{aligned}$$

Conjecture: $|S_R - \pi R^2| \leq \mathcal{O}(R^{\frac{1}{2}+\varepsilon})$

Gauss: $\mathcal{O}(R)$

Sierpinski: $\mathcal{O}(R^{\frac{2}{3}})$

Huxley (2005): $\mathcal{O}(R^{0.63})$

Theorem 0.2 (Sierpinski). $|S_R - \pi R^2| \leq \mathcal{O}(R^{\frac{2}{3}})$

Proof. The following proof is due to Hugh Montgomery.

Step 1: $S_R = \sum_{x \in \mathbb{Z}^2} \mathbb{1}_{B_R}$. We need to replace the indicator function by some smooth

function. Use Convolution. Take $\varphi \in C_c^\infty$, $\varphi \geq 0$ radial $\varphi = 0$ outside B_1 and $\int \varphi = 1$

Define $\varphi_\varepsilon(x) = \varepsilon^{-2}(\frac{x}{\varepsilon})$, $\int \varphi_\varepsilon = 1$ $f_R(x) = (\varphi_\varepsilon \star \mathbb{1}_{B_R})(x) = \int_{B(y, R)} \varphi(x-y) dy$

Then $f_R \in C_c^\infty(\mathbb{R}^2)$ by Poisson summation formula

$$\sum_{x \in \mathbb{Z}^2} f_R(x) = \sum_{k \in \mathbb{Z}^2} \hat{f}_R(k)$$

We have
$$\begin{cases} \text{supp } \varphi_\varepsilon \subset B_\varepsilon \\ \text{supp } \mathbb{1}_{B_R} \subset B_R \\ \text{supp } f_R \subset B_{\varepsilon+R} \end{cases}$$

Easy to check: $f_R(x) = 1$, if $|x| < R - \varepsilon$, $f_R(x) = 0$, if $|x| > R + \varepsilon$ and

$$\tilde{S}_{R-\varepsilon} \leq S_R \leq \tilde{S}_{R+\varepsilon}$$

Step 2: Consider $\tilde{S}_R = \sum_{k \in \mathbb{Z}^2} \hat{f}_R(k)$

$$k = 0 : \hat{f}(0) = \int_{\mathbb{R}^2} f_R = \int \varphi_\varepsilon \star \mathbb{1}_{B_R} = (\int \varphi_\varepsilon)(\int \mathbb{1}_{B_R}) = \pi R^2$$

$$k \neq 0 : \hat{f}_R(k) = \widehat{\varphi_\varepsilon \star \mathbb{1}_{B_R}}(k) = \hat{\varphi}_\varepsilon(k) \hat{\mathbb{1}}_{B_R}(k)$$

We have

$$\varphi_\varepsilon = \int_{\mathbb{R}^2} \varepsilon^{-2} \varphi\left(\frac{x}{\varepsilon}\right) e^{-2\pi i k x} dx \quad (1)$$

$$\stackrel{\frac{x}{\varepsilon}=y}{=} \int_{\mathbb{R}^2} \varphi(y) e^{-2\pi i k \varepsilon y} dy \quad (2)$$

$$= \hat{\varphi}(\varepsilon k) \quad (3)$$

$$\Rightarrow |\hat{\varphi}_\varepsilon(k)| = |\hat{\varphi}(\varepsilon k)| \leq \frac{c_l}{|\varepsilon k|^l} \quad \forall l \geq 0$$

(since φ in C_c^∞ $\Rightarrow \hat{\varphi}(k)$ decays faster than any polynomial)

$$R^{-2}\hat{\mathbb{1}}_{B_R}(k) = \hat{\mathbb{1}}_{B_1}(Rk)$$

Remark. $|\hat{\mathbb{1}}_{B_1}(\xi)| \leq C|\xi|^{-\frac{d+1}{2}}$, $\xi \in \mathbb{R}^d$ is a Bessel function

Exercise. Define $\mathcal{J}_1(t) = \frac{1}{i\pi} \int_0^\pi e^{it\cos(\Theta)} \cos(\Theta) d\Theta$, then $|\mathcal{J}_1(t)| \leq Ct^{-\frac{1}{2}}$, $\forall t > 0$

$$\Rightarrow \hat{\mathbb{1}}_{B_R}(k) = R^2 |\hat{\mathbb{1}}_{B_1}(Rk)| \leq \frac{CR^2}{(R|k|)^{\frac{3}{2}}}$$

$$\text{In summary: } |\hat{f}_R(k)| \leq \frac{c_l R^{\frac{1}{2}}}{|k^{\frac{3}{2}}|} \cdot \frac{1}{|\varepsilon k|^l}, \quad \forall l \geq 0$$

$$\Rightarrow \sum_{k \neq 0} |\hat{f}_R(k)| = \sum_{0 < |k| < K} |\hat{f}_R(k)| + \sum_{|k| \geq K} |\hat{f}_R(k)| \quad (4)$$

$$\leq \sum_{0 < |k| < K} \frac{cR^{\frac{1}{2}}}{|k^{\frac{3}{2}}|} + \sum_{|k| \geq K} \frac{cR^{\frac{1}{2}}}{\varepsilon |k^{\frac{5}{2}}|} \quad (5)$$

$$\leq cR^{\frac{1}{2}}K^{\frac{1}{2}} + \frac{cR^{\frac{1}{2}}}{\varepsilon |K^{\frac{1}{2}}|}, \quad \forall K > 0 \quad (6)$$

Now, choose best K !

$$\text{Step 3: } S_R \geq \tilde{S}_{R-\varepsilon} \geq \pi(R - \varepsilon)^2 - C \frac{(R-\varepsilon)^{\frac{1}{2}}}{\varepsilon^{\frac{1}{2}}}$$

$$\text{Thus, } S_R \leq \tilde{S}_{R+\varepsilon} \leq \pi(R + \varepsilon)^2 + C \frac{(R+\varepsilon)^{\frac{1}{2}}}{\varepsilon^{\frac{1}{2}}}$$

$$\Rightarrow |S_R - \pi R^2| \leq C(R\varepsilon + \frac{R^{\frac{1}{2}}}{\varepsilon^{\frac{1}{2}}}) \underset{\text{opt over } \varepsilon > 0}{\rightarrow} CR^{\frac{2}{3}}$$

$$R\varepsilon \sim \frac{R^{\frac{1}{2}}}{\varepsilon^{\frac{1}{2}}} \sim \left(R\varepsilon \left(\frac{R^{\frac{1}{2}}}{\varepsilon^{\frac{1}{2}}} \right) \left(\frac{R^{\frac{1}{2}}}{\varepsilon^{\frac{1}{2}}} \right) \right)^{\frac{1}{3}} = R^{\frac{2}{3}}$$

□

Chapter 1

L^p -Spaces

1.1 Measures

(Σ, Ω) , $\Sigma \subset 2^\Omega = \{\text{subsets of } \Omega\}$

Σ is called a $\sigma-$ algebra, if $\left\{ \begin{array}{l} \Omega \in \Sigma \\ A \in \Sigma \Rightarrow A^c \in \Sigma \\ (A_i)_{i=0}^n \subset \Sigma \Rightarrow \bigcap_{i=0}^n A_i \in \Sigma \\ (A_i)_{i \in \mathbb{N}} \text{ countable} \Rightarrow \bigcup_{i \in \mathbb{N}} A_i \in \Sigma \end{array} \right.$

μ is called a (positive) measure, if $\mu : \Sigma \rightarrow [0, \infty]$, s.t. $\mu(\emptyset) = 0$ and if $(A_i)_{i \in \mathbb{N}}$ countable and $A_i \cap A_j = \emptyset$ for all $i \neq j$, then $\mu(\bigcup_{i \in \mathbb{N}} A_i) = \sum_{i \in \mathbb{N}} \mu(A_i)$

Example 1.1 (Dirac-Delta measure). μ_y a measure on \mathbb{R}^d , $y \in \mathbb{R}^d$ s.t.

$$\mu_y(A) = \begin{cases} 1, & \text{if } y \in A \\ 0, & \text{else} \end{cases}. \text{ Here } \Sigma = 2^{\mathbb{R}^d}$$

Example 1.2. (counting measure) $\mu(A) = \#A = |A|$ with $\Omega = \mathbb{R}^d$ and $\Sigma = 2^\Omega$

Remark. This measure is not sigma-finite

Definition 1.3. A measure μ is called $\sigma-$ finite, there is a countable family $(A_i)_{i \in \mathbb{N}}$, s.t. $\bigcup_{i \in \mathbb{N}} A_i = \Omega$ and $\mu A_i < \infty$ for all i

Example 1.4 (Borel measure). $(\mu, \Sigma, \Omega) \rightsquigarrow \Sigma = \text{smallest } \sigma\text{-Algebra s.t. it contains all open and closed sets. In } \mathbb{R}^d \text{ we have a uniquely defined measure s.t. } \mu([0, 1]^d) = 1 \text{ and } \forall x \in \mathbb{R}^d, A \in \Sigma : \mu(x + A) = \mu(A)$

Theorem 1.5. There is exactly one Borel measure μ

Example 1.6 (Lebesgue Measure). $(\mu, \Sigma, \Omega) \rightsquigarrow \Sigma = \text{Borel } \sigma\text{-Algebra} + \text{"sets of 0-measure"}$

$$\mu(A) = \mu(A_{\text{borel measurable}} + \text{"sets of 0-measure"}) := \mu(A_{\text{borel measurable}})$$

Definition 1.7. $A \subset \mathbb{R}^d$ is a set of 0-measure, if for every $\varepsilon > 0$ there is a Borel measurable set A_ε s.t. $A \subset A_\varepsilon$ and $\mu(A_\varepsilon) < \varepsilon$

Remark.

- The advantage of the Lebesgue σ -Algebra is its completeness, i.e. if $\mu(A) = 0$ for a Borel set A , then for all $B \subset A$: $\mu(B) = 0$.
- The disadvantage of the Lebesgue σ -Algebra is that we need to be careful with the "product property", i.e. if $A_1 \subset \mathbb{R}$ is not Lebesgue measurable and A_2 is a set of 0-measure, then $\mu_{\mathbb{R}^2}(A_1 \times A_2) = 0$, but $\mu_{\mathbb{R}^2}(A_1 \times A_2) = \underbrace{\mu_{\mathbb{R}}(A_1)}_{\text{not well defined}} \cdot \mu_{\mathbb{R}}(A_2) \not\in$
This is not nice when applying Fubini
- For two functions f, g we say that $f = g$ a.e. if there is a Lebesgue set of 0-measure A , s.t. $\forall x \in \mathbb{R}^d \setminus A : f(x) = g(x)$

Theorem 1.8 (Regularity of Lebesgue measure). If $A \subset \mathbb{R}^d$ is Lebesgue measurable, then

1. $|A| := \mu(A) = \inf\{|O| : O \text{ open and } A \subset O\}$ (outer regularity)
2. $|A| := \mu(A) = \sup\{|B| : B \text{ compact and } B \subset A\}$ (inner regularity)

1.2 Integration

(μ, Σ, Ω) measure space. Let $f : \Omega \rightarrow \mathbb{R}$

Definition 1.9. f is measurable, if for all $\lambda \in \mathbb{R}$ $x \in \Omega : \underbrace{f(x) > \lambda}_{\text{level set}}$ is measurable

In general, $f : \Omega \rightarrow \mathbb{C}$ measurable, iff $Re(f)$ and $Im(f)$ are measurable

Definition 1.10. Take $f : \Omega \rightarrow [0, \infty]$ measurable.

Define $\int_{\Omega} f d\mu := \int_0^{\infty} \mu(\{x \in \Omega : f(x) > \lambda\})$
Lebesgue integral Riemann integral

This works, since $\lambda \mapsto \mu(\{x \in \Omega : f(x) > \lambda\})$ is monotone

Remark. If $g : \mathbb{R} \rightarrow \mathbb{R}$ monotone, then g is continuous except on a countable set.

More precisely: \exists countable points $\{x_i\}_{i \in I} \subset \mathbb{R}$ s.t. g is continuous on $\mathbb{R} \setminus \{x_i\}_{i \in I}$ and at x_i :

$$g(x_i, -) = \lim_{y \nearrow x_i} g(y) \leq \lim_{y \searrow x_i} g(y) = g(x_i, +)$$

In general, if $f : \Omega \rightarrow \mathbb{R}$ and $f = f_+ - f_-$, s.t. $f_+ = \max(f, 0)$ and $f_- = \max(-f, 0)$ and assume $\int f_+$ and $\int f_-$ are finite, then we say that f is summable and define $\int f = \int f_+ - \int f_-$. Also if $f : \Omega \rightarrow \mathbb{C}$ and $Re f$ and $Im f$ are summable, then f is summable and $\int f = \int Re f + i \int Im f$

Theorem 1.11. (Layer cake representation) For all $1 \leq p < \infty$:

$$\int_{\Omega} |f(x)|^p d\mu(x) = \int_0^{\infty} p \lambda^{p-1} \mu(\{x \in \Omega : |f(x)| > \lambda\}) d\lambda$$

Remark. Think of Fubini:

$$\int_0^{\infty} p \lambda^{p-1} \mu(\{x \in \Omega : |f(x)| > \lambda\}) d\lambda = \int_0^{\infty} p \lambda^{p-1} \left(\int_{\Omega} \mathbb{1}_{\{|f(x)| > \lambda\}} d\mu(x) \right) d\lambda \quad (1.1)$$

$$= \int_{\Omega} \underbrace{\left(p \lambda^{p-1} \int_0^{\infty} \mathbb{1}_{\{|f(x)| > \lambda\}} d\lambda \right)}_{=|f(x)|^p} d\mu(x) \quad (1.2)$$

5 fundamental theorems:

1. Monotone c.v.
2. Dominated c.v.
3. Fatou's Lemma
4. Brezis-Lieb Lemma
5. Fubini

1.3 L^p spaces

Definition 1.12. $\|f\|_p = \begin{cases} (\int_{\Omega} |f|^p)^{\frac{1}{p}}, & \text{if } 1 \leq p < \infty \\ ess\sup |f|, & \text{if } p = \infty \end{cases}$

$ess\sup = "sup up to sets of 0-measure" = \inf\{\lambda : |f(x)| \leq \lambda \text{ a.e.}\} f \in L^p(dom(f)) \Leftrightarrow \|f\|_p < \infty$

Theorem 1.13. $L^p(\Omega)$ is a Banach space for all $p \in [1, \infty]$

Theorem 1.14 (Hölder inequality). If for all $f \in L^p$, $g \in L^q$, $1 \leq p, q \leq \infty$ s.t. $\frac{1}{p} + \frac{1}{q} = 1$, then

$$\left| \int_{\Omega} fg \right| \leq \|f\|_p \|g\|_q$$

Remark. Dual formulation: $\|f\|_p = \sup_{0 \neq g \in L^q} \frac{|\int_{\Omega} fg|}{\|g\|_q}$

Theorem 1.15 (Riesz-Representation). $(L^p(\Omega))^* = L^q(\Omega)$ for all $\frac{1}{p} + \frac{1}{q} = 1$, $1 \leq p, q < \infty$. Moreover, if (Ω, μ) is σ -finite, then $(L^1)^* = L^\infty$, but $(L^\infty)^*$ is in general much bigger than L^1 .

Definition 1.16 (Weak L^p). Take $f : \Omega \rightarrow \mathbb{C}$ measurable. We say $f \in L_w^p(\Omega)$, if

$$\sup_{\lambda > 0} \lambda |\{x \in \Omega : |f(x)| > \lambda\}|^{\frac{1}{p}} < \infty$$

and define

$$\|f\|_{p,w} = \sup_{\lambda > 0} \lambda^p |\{x \in \Omega : |f(x)| > \lambda\}|$$

Example 1.17. Take $f : x \mapsto \frac{1}{|x|}$ for $x \in \mathbb{R}^d$.

$$\begin{aligned} |\{f(x) > \lambda\}| &= |x : |x| < \frac{1}{\lambda}| = C \cdot \frac{1}{\lambda^d} \\ \Rightarrow \sup_{\lambda > 0} \lambda^d |\{f(x) > \lambda\}| &< \infty \\ \Rightarrow \frac{1}{|x|} &\in L_w^p(\mathbb{R}^d) \end{aligned}$$

Similarly for $\frac{1}{|x|^d} \in L_{1,w}$

Chapter 2

Hardy-Littlewood maximal functions

2.1 Hardy-Littlewood maximal inequalities

Motivation: We want to prove the Lebesgue differentiation theorem for L^1 functions.

Theorem 2.1 (Lebesgue Differentiation Theorem). If $f \in L^1(\mathbb{R}^d)$, then

$$\operatorname{f}_{B(x,r)} |f - f(x)| \xrightarrow{r \rightarrow 0} 0 \text{ a.e.}$$

The same holds, if $f \in L^1_{loc}(\mathbb{R}^d)$ or $f \in L^1_{loc}(\Omega)$ for some $\Omega \subset \mathbb{R}^d$

Definition 2.2 (Maximal function). If $f \in L^1_{loc}(\mathbb{R}^d)$, then define

$$Mf(x) := \sup_{r>0} \operatorname{f}_{B(x,r)} |f|$$

Remark. 1. By Lebesgue differentiation theorem, $Mf(x) \geq |f(x)|$ a.e.

2. If $f \not\equiv 0$, then $Mf \notin L^1(\mathbb{R}^d)$ even, if $f \in L^1(\mathbb{R}^d)$, since

$$Mf(x) \geq \operatorname{f}_{B(x,r)} |f| = \frac{1}{|B_1|r^d} \int_{B(x,r)} |f| \quad (2.1)$$

$$\geq \frac{1}{|B_1|(2|x|)^d} \int_{B(0,|x|)} |f(y)| dy \quad (2.2)$$

$$\geq \frac{\varepsilon}{|x|^d}, \text{ as } |x| \rightarrow \infty \quad (2.3)$$

for some $\varepsilon > 0$ (ε depends on f but not on x) Thus, $\int_{\mathbb{R}^d} Mf(x) dx \geq \int_{\mathbb{R}^d} \frac{\varepsilon}{|x|^d} dx = \infty$

We can not even expect $Mf \in L^1_{loc}(\mathbb{R}^d)$, if $f \in L^1(\mathbb{R}^d)$

Example 2.3. $f(x) = \frac{1}{|x||\ln x|^2} \in L^1(-1, 1)$. Note that $\frac{1}{|x||\ln x|} \notin L^1$

$$Mf(x) \geq \int_0^{2x} |f| = \frac{1}{2x} \int_0^{2x} \frac{1}{|x||\ln x|} \sim \frac{1}{|x||\ln x|} \notin L^1_{loc}(-1, 1)$$

We will prove that on the other hand Mf is "as nice as f "

Theorem 2.4 (Hardy-Littlewood inequality - L^1 weak form). If $f \in L^1(\mathbb{R}^d)$, then

$$|\{x : Mf(x) > \lambda\}| \leq \frac{C_d}{\lambda} \|f\|_{L^1}$$

Here we can take $C_d = 5^d$ (or 3^d).

Note that the bound is equivalent: $\|Mf\|_{L_w^1} \leq C_d \|f\|_{L^1}$

Proof (Lebesgue differentiation theorem by Hardy-Littlewood maximal inequality 2.1). Wlog assume $f \in L_c^1$, i.e. $f \in L^1$ and f has compact support.

Step 1 (Reformulate the statement): We want to prove

$$\limsup_{n \rightarrow 0} \operatorname{fint}_{B(x,r)} |f - f(x)| = 0 \text{ for a.e. } x$$

Define $A_\varepsilon := \{x : \limsup_{n \rightarrow 0} |f - f(x)| > \varepsilon\}$

We will prove that for all $\varepsilon > 0$: $|A_\varepsilon| = 0$

Then this implies $\bigcup_{n \in \mathbb{N}} A_{\frac{\varepsilon}{n}} = 0$ and $\bigcup_{n \in \mathbb{N}} A_{\frac{\varepsilon}{n}} = \{x : \limsup_{n \rightarrow 0} |f - f(x)| > 0\}$.

This implies our desired condition.

Step 2: Take $\varepsilon > 0$ and prove $|A_\varepsilon| = 0$

If f is continuous, then this is obvious. If $f \in L_c^1(\mathbb{R}^d)$, then $\exists \{f_n\}_{n \in \mathbb{N}} \subset C_c(\mathbb{R}^d)$ s.t. $f_n \rightarrow f$ in L^1 . Then:

$$A_\varepsilon \subset \{x : \limsup_{r \rightarrow 0} \operatorname{fint}_{B(x,r)} |f(y) - f(x)| dy > \varepsilon\}$$

By triangle inequality:

$$\begin{aligned} |f(y) - f(x)| &\leq |f(y) - f_n(y)| + |f_n(y) - f_n(x)| + |f_n(x) - f(x)| \\ \Rightarrow \operatorname{fint}_{B(x,r)} |f(y) - f(x)| dy &\leq \operatorname{fint}_{B(x,r)} |f(y) - f_n(y)| dy + \operatorname{fint}_{B(x,r)} |f_n(y) - f_n(x)| dy + |f_n(x) - f(x)| \end{aligned}$$

We have that for all n : $\fint_{B(x,r)} |f_n(y) - f_n(x)| dy \xrightarrow{r \rightarrow 0} 0$, since f_n is continuous. Moreover, by the definition of the maximal function:

$$\fint_{B(x,r)} |f(y) - f_n(y)| dy \leq \sup_{r \rightarrow 0} \fint_{B(x,r)} |f - f_n|(y) dy = M(f - f_n)(x)$$

Thus

$$A_\varepsilon \subset \{x : M(f - f_n)(x) + |f_n(x) - f(x)| > \varepsilon\}.$$

Note: $a + b > \varepsilon \Rightarrow$ either $a > \frac{\varepsilon}{2}$ or $b > \frac{\varepsilon}{2}$

$$\Rightarrow A_\varepsilon \subset \{x : M(f - f_n)(x) > \frac{\varepsilon}{2}\} \cup \{x : |f_n(x) - f(x)| > \frac{\varepsilon}{2}\}$$

We have: $|\{x : |f_n(x) - f(x)| > \frac{\varepsilon}{2}\}| \leq \int_{\mathbb{R}^d} \frac{|f_n(x) - f(x)|}{\frac{\varepsilon}{2}} \leq \frac{2}{\varepsilon} \|f_n - f\|_{L^1}$

$$|\{x : M(f - f_n)(x) > \frac{\varepsilon}{2}\}| \leq \frac{c_d}{\varepsilon} \|f_n - f\|_{L^1} \quad (\text{maximal inequality})$$

Conclusion: $|A_\varepsilon| \leq \frac{\tilde{c}}{\varepsilon} \|f_n - f\|_{L^1} \quad \forall n \in \mathbb{N}$

$$\Rightarrow |A_\varepsilon| \leq \lim_{n \rightarrow \infty} \frac{\tilde{c}}{\varepsilon} \|f_n - f\|_{L^1} = 0$$

□

Lemma 2.5 (Vitali covering lemma). Consider a collection of balls in \mathbb{R}^d , $\{B_i\}_{i \in I}$, s.t. $\sup_i \text{diam } B_i < \infty$. Then \exists a subcollection $I' \subset I$ s.t. $\{B_i\}_{i \in I'}$ contains only disjoint balls and

$$\bigcup_{i \in I} B_i \subset \bigcup_{i \in I'} 5B_i$$

where $5B(x, r) = B(x, 5r)$. Here 5 can be replaced by any $\nu > 3$

Proof. First, we only consider the case, where I is finite. Choose I' by induction:

1. B_1 is the largest ball in $\{B_i\}_{i \in I}$. Add B_1 to I'
2. Ignore all balls that intersect with B_1 . Note that $5B_1$ covers all balls that intersect with B_1
3. Repeat with remaining balls

The remaining set has the desired properties.

□

Proof (Hardy-Littlewood 2.4). We only consider the case where $f \in L_c^1$. Note that this is enough to prove 2.1.

Let $f \in L_c^1(\mathbb{R}^d)$. By the definition of the maximal function we get that for all $x \in \mathbb{R}^d$ $\exists r_x > 0$ s.t.

$$Mf(x) = \sup_{r>0} \fint_{B(x,r)} |f| \leq 2 \fint_{B(x,r_x)} |f|$$

$$\Rightarrow \int_{B(x, r_x)} |f| \geq \frac{1}{2} |B_r| Mf(x)$$

Idea: We want to decompose $\{x : Mf(x) > \lambda\}$ into disjoint balls $\{B(x, r_x)\}_{x \in I}$. In this step we need that $\sup_x r_x < \infty$. This is guaranteed by $f \in L_c^1$.

$$\Rightarrow \underbrace{|\{x : Mf(x) > \lambda\}|}_{|\bigcup_x B(x, r_x)| \leq \sum_x |B(r_x)| \lesssim \sum_x \frac{1}{\lambda} \int_{B(x, r_x)} |f|} \lesssim \frac{1}{\lambda} \sum_{x \in I} \int_{B(x, r_x)} |f| \leq \frac{1}{\lambda} \int_{\mathbb{R}^d} |f| = \frac{1}{\lambda} \|f\|_{L^1}$$

We have $A = \{x : Mf(x) > \lambda\} \subset \bigcup_{x \in A} B(x, r_x)$ $\xrightarrow{\text{Vitali covery lemma}} \exists A' \subset A$ s.t.

$$\bigcup_{x \in A} B(x, r_x) \subset \bigcup_{x \in A'} 5B(x, r_x)$$

and A' contains only disjoint balls.

$$\text{For any } x \in A' \subset A : \int_{B(x, r_x)} |f| \geq \frac{1}{2} |B(r_x)| \lambda$$

$$\sum_{x \in A'} \int_{B(x, r_x)} |f| \geq \sum_{x \in A'} \frac{1}{2} |B(r_x)| \lambda = \sum_{x \in A'} \frac{1}{2 \cdot 5^d} |B(5r_x)| \lambda \quad (2.4)$$

$$\geq \frac{1}{2 \cdot 5^d} \left| \bigcup_{x \in A'} B(5r_x) \right| \lambda \quad (2.5)$$

$$\geq |A| \quad (2.6)$$

$$\Rightarrow |A| \leq \frac{2 \cdot 5^d}{\lambda} \|f\|_{L^1}$$

□

Theorem 2.6 (Hardy-Littlewood maximal inequality - L^p strong version). If $f \in L^p(\mathbb{R}^d)$ for $1 < p \leq \infty$, then:

$$\|Mf\|_{L^p} \leq C_{d,p} \|f\|_{L^p}$$

Interpolation idea: $p = \infty$ is trivial. $Mf(x) = \sup_{r>0} \frac{1}{|B(x,r)|} \int_{B(x,r)} |f| \leq \|f\|_\infty$

The lower p is the harder it gets. In fact, we will see that the hardest case $p = 1$ in its weak form implies the theorem.

Remark. We can make $C_{d,p}$ independent of p if p is far away from 1 but $C_{d,p} \rightarrow \infty$ if $p = 1$ or $d \rightarrow \infty$.

Proof. Let $1 < p < \infty$. We use the layer-cake representation

$$\|Mf\|_{L^p}^p = \int_0^\infty p \lambda^{p-1} |\{Mf > \lambda\}| d\lambda$$

First try: Applying weak L^1 bound form 2.4

$$\Rightarrow \|Mf\|_{L^p}^p = c_{d,p} \int_0^\infty \lambda^{p-1} \frac{\|f\|_{L^1}}{\lambda} d\lambda = \infty, \text{ since } \int_0^\infty \lambda^s d\lambda = \infty \text{ for all } s > 0$$

Second try: First show $\{Mf > 0\} \subset \{Mf_{\frac{\lambda}{2}} > \frac{\lambda}{2}\}$ where $f_{\frac{\lambda}{2}} = f \cdot \mathbb{1}_{f > \frac{\lambda}{2}}$:
This holds because $f(x) = f_{\frac{\lambda}{2}}(x) + f(x)\mathbb{1}_{f \leq \frac{\lambda}{2}}$ for all x

$$\begin{aligned} \Rightarrow Mf(x) &= \sup_{r>0} \underset{B(x,r)}{\text{f}} f(y) dy \leq (\sup_{r>0} \underset{B(x,r)}{\text{f}} f_{\frac{\lambda}{2}}(y) dy) + \frac{\lambda}{2} = Mf_{\frac{\lambda}{2}}(x) + \frac{\lambda}{2} \\ \Rightarrow \text{If } Mf(x) > \frac{\lambda}{2} &\Rightarrow Mf_{\frac{\lambda}{2}} > \frac{\lambda}{2} \Rightarrow \{Mf > 0\} \subset \{Mf_{\frac{\lambda}{2}} > \frac{\lambda}{2}\} \end{aligned}$$

By the weak L^1 bound:

$$\begin{aligned} |\{x : Mf_{\frac{\lambda}{2}}(x) > \frac{\lambda}{2}\}| &\leq \frac{C_d}{\frac{\lambda}{2}} \left\| f_{\frac{\lambda}{2}} \right\|_{L^1} \\ &= \frac{2C_d}{\lambda} \int_{\mathbb{R}^d} f(x) \mathbb{1}_{f > \frac{\lambda}{2}} dx \end{aligned}$$

Then we conclude from the layer-cake formula:

$$\begin{aligned} \|Mf\|_{L^p}^p &\leq C_{d,p} \int_0^\infty \lambda^{p-1} \left(\frac{1}{\lambda} \int_{\mathbb{R}^d} f(x) \mathbb{1}_{f > \frac{\lambda}{2}} dx \right) d\lambda \\ &\stackrel{\text{Fubini}}{=} C_{d,p} \int_{\mathbb{R}^d} f(x) \underbrace{\left(\int_0^\infty \lambda^{p-2} \mathbb{1}_{f > \frac{\lambda}{2}} d\lambda \right)}_{\frac{2f(x)}{\lambda^{p-2}}} dx \\ &= \int_0^\infty \lambda^{p-2} d\lambda = c(f(x))^{p-1} \\ &= c \int_{\mathbb{R}^d} f^p \end{aligned}$$

□

Remark. (Interpolation) From the proof, we use 2 inequalities:

$$f(x) = \underbrace{f(x)\mathbb{1}_{f > \frac{\lambda}{2}}}_{\text{weak } L^1} + \underbrace{f(x)\mathbb{1}_{f \leq \frac{\lambda}{2}}}_{\text{strong } L^\infty}$$

The same idea can be used in a much more general setting \rightsquigarrow interpolation inequalities

2.2 Hardy-Littlewood-Sobolev inequality

Theorem 2.7 (Hardy-Littlewood-Sobolev inequality). Let $p, q > 1$ and $\frac{1}{p} + \frac{1}{q} + \frac{\lambda}{d} = 2$

$$\left| \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \frac{f(x)g(y)}{|x-y|^\lambda} dx dy \right| \leq c \|f\|_{L^p} \|g\|_{L^q}$$

where $c = c_{d,\lambda,p,q} < \infty$ independently of f, g

Remark. 1. This implies the standard Sobolev inequality. E.g. in $3D$: $\int_{\mathbb{R}^3} |\nabla u|^2 \geq c \left(\int_{\mathbb{R}^3} |u|^6 \right)^{\frac{1}{3}}$

The latter follows from HLS and $(-\delta)^{-1}f(x) = \int_{\mathbb{R}^3} \frac{1}{4\pi|x-y|} f(y) dy$

2. HLS inequality is also called weak young-ineq.

$$\|f * g\|_{L^r} \leq \|f\|_{L^p} \|g\|_{L^q}, \quad \forall r, p, q > 1 \text{ s.t. } \frac{1}{p} + \frac{1}{q} = 1 + \frac{1}{r}$$

$$\Leftrightarrow \left| \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} f(x)g(x-y)h(y) dx dy \right| \leq C \|f\|_{L^p} \|g\|_{L^r} \|h\|_{L^q}, \quad \forall r, p, q > 1 \text{ s.t. } \frac{1}{p} + \frac{1}{q} + \frac{1}{r} = 2$$

3. Scaling argument: Let $f_t(x) := f(tx)$, $t > 0$

LHS:

$$\begin{aligned} \left| \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \frac{f_t(x)g_t(y)}{|x-y|^\lambda} dx dy \right| &\stackrel{\hat{x}=tx, \hat{y}=ty}{=} \left| \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \frac{f(\hat{x})g(\hat{y})}{|\frac{\hat{x}}{t}-\frac{\hat{y}}{t}|^\lambda} d\hat{x} d\hat{y} \right| \\ &= t^{\lambda-2d} \left| \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \frac{f_t(x)g_t(y)}{|x-y|^\lambda} dx dy \right| \end{aligned}$$

LHS:

$$\|f_t\|_{L^p} \|g_t\|_{L^q} = t^{-d(\frac{1}{p} + \frac{1}{q})} \|f\|_{L^p} \|g\|_{L^q}$$

(just via substitution $\hat{x} = tx$)

Observation: If $A, B > 0$ and $At^\alpha \leq Bt^\beta \forall t > 0$, then $\alpha = \beta$

HLS ineq $\Rightarrow \lambda - 2d = -d(\frac{1}{p} + \frac{1}{q}) \Rightarrow \frac{\lambda}{d} - 2 = -(\frac{1}{p} + \frac{1}{q})$. So this is the only reasonable choice for λ, p, q .

Lemma 2.8. (Fefferman-de la Llave) Let $0 < \lambda < d$ and $x, y \in \mathbb{R}^d$ s.t. $x \neq y$. Then there exists a $c = c_{d,\lambda} > 0$ s.t.

$$\frac{1}{|x-y|^\lambda} = c_{d,\lambda} \int_0^\infty \int_{\mathbb{R}^d} \mathbb{1}_{B(t,r)}(x) \mathbb{1}_{B(z,r)}(y) dz \frac{dr}{r^{d+\lambda+1}}$$

Proof. The proof consists of showing that the function on the RHS $RHS(x, y)$ satisfies $RHS(tx, ty) = \frac{1}{t^\lambda} RHS(x, y)$. Then it follows that $RHS(x, y) = c \frac{1}{|x-y|^\lambda}$ for a constant $c > 0$ (exercise).

$$\begin{aligned}
 RHS(tx, ty) &= \int_0^\infty \int_{\mathbb{R}^d} \mathbb{1}_{B(x,r)}(tx) \mathbb{1}_{B(z,r)}(ty) dz \frac{dr}{r^{d+\lambda+1}} \\
 &= \int_0^\infty \int_{\mathbb{R}^d} \mathbb{1}_{B(tz,tr)}(tx) \mathbb{1}_{B(tz,tr)}(ty) \underbrace{d(tz)}_{=t^d dz} \frac{d(tr)}{(tr)^{d+\lambda+1}} \\
 &= \left(\int_0^\infty \int_{\mathbb{R}^d} \mathbb{1}_{B(t,r)}(x) \mathbb{1}_{B(z,r)}(y) dz \frac{dr}{r^{d+\lambda+1}} \right) \frac{1}{t^\lambda} \\
 &= \frac{1}{t^\lambda} RHS(x, y)
 \end{aligned}$$

□

proof of HLS by HL max inequality + FD lemma 2.8.

Assume $\|f\|_{L^p} = \|g\|_{L^q} = 1$ and $f, g \geq 0$. Then:

$$\begin{aligned}
 &\int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \frac{f(x)g(y)}{|x-y|^\lambda} dx dy \\
 &= c_{d,\lambda} \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} f(x)g(y) \left(\int_0^\infty \int_{\mathbb{R}^d} \mathbb{1}_{B(t,r)}(x) \mathbb{1}_{B(z,r)}(y) dz \frac{dr}{r^{d+\lambda+1}} \right) dx dy \\
 &= c_{d,\lambda} \int_{\mathbb{R}^d} \left[\int_0^\infty (f \star \mathbb{1}_{B(0,r)})(z) (g \star \mathbb{1}_{B(0,r)})(z) \right] dx dy
 \end{aligned}$$

We have the following bounds:

$$1. f \star \mathbb{1}_{B(0,r)}(z) \leq |B_r| Mf(z) \leq cr^d Mf(z).$$

$$\text{Identically: } g \star \mathbb{1}_{B(0,r)}(z) \leq cr^d Mg(z)$$

$$\Rightarrow F := (f \star \mathbb{1}_{B(0,r)})(z) (g \star \mathbb{1}_{B(0,r)})(z) \frac{1}{r^{d+\lambda+1}} \lesssim r^{d-\lambda+1} Mf(z) Mg(z) \quad \forall r > 0$$

(We will only use $r < R$)

2.

$$\|f \star \mathbb{1}_{B(0,r)}\|_{L^\infty} \stackrel{\text{Young ineq}}{\leq} \|f\|_{L^p} \|\mathbb{1}_{B(0,r)}\|_{L^{p'}} \lesssim r^{\frac{d}{p'}}$$

for $\frac{1}{p} + \frac{1}{p'} = 1$.

Identically: $\|f \star \mathbb{1}_{B(0,r)}\|_{L^\infty} \lesssim r^{\frac{d}{q'}}$ for $\frac{1}{q} + \frac{1}{q'} = 1$

$$\Rightarrow F \lesssim r^{\frac{d}{p'}} r^{\frac{d}{q'}} \frac{1}{r^{d+\lambda+1}} = \frac{1}{r^{1+d}}$$

We used that $\frac{d}{p} + \frac{d}{q} = 2d - \lambda$ and since we have Hölder conjugates, the same holds for p', q' instead of p, q .

$$\begin{aligned} \Rightarrow \int_0^\infty F dr &= \underbrace{\int_0^R F dr}_{\text{use first bound}} + \underbrace{\int_R^\infty F dr}_{\text{use second bound}} \\ &\lesssim \int_0^R r^{d-\lambda+1} Mf(z) Mg(z) dr + \int_R^\infty \frac{1}{r^{1+d}} dr \\ &\lesssim R^{d-\lambda} Mf(z) Mg(z) + \frac{1}{R^d}, \quad \forall R > 0 \end{aligned}$$

Optimize over R :

$$\begin{aligned} \Rightarrow \int_0^\infty F dr &\lesssim \left[(R^{d-\lambda} Mf(z) Mg(z))^d \left(\frac{1}{R^d}\right)^{d-\lambda} \right]^{\frac{1}{2d}-\lambda} \\ &= (Mf(z))^{\frac{d}{2d-\lambda}} (Mg(z))^{\frac{d}{2d-\lambda}} \end{aligned}$$

Finally:

$$\begin{aligned} \int_{\mathbb{R}^d} \int_0^\infty F dr dz &\lesssim \int_{\mathbb{R}^d} (Mf(z))^{\frac{d}{2d-\lambda}} (Mg(z))^{\frac{d}{2d-\lambda}} \\ &\leq \left(\int_{\mathbb{R}^d} (Mf(z))^{\frac{d}{2d-\lambda} \alpha} dz \right)^{\frac{1}{\alpha}} \left(\int_{\mathbb{R}^d} (Mg(z))^{\frac{d}{2d-\lambda} \beta} dz \right)^{\frac{1}{\beta}} \\ &= \|Mf\|_{L^p}^{\frac{p}{\alpha}} \|Mg\|_{L^q}^{\frac{q}{\alpha}} \stackrel{\text{HL max}}{\lesssim} \|f\|_{L^p}^{\frac{p}{\alpha}} \|g\|_{L^q}^{\frac{q}{\alpha}} \end{aligned}$$

for

$$1. \quad \frac{1}{\alpha} + \frac{1}{\beta} = 1$$

$$2. \quad \frac{d}{2d-\lambda} - \alpha = p, \quad \frac{d}{2d-\lambda} - \beta = q$$

$$\Rightarrow 1 = \frac{1}{\alpha} + \frac{1}{\beta} = \frac{d}{2d-\lambda} \frac{1}{p} + \frac{d}{2d-\lambda} \frac{1}{q} = \frac{1}{2d-\lambda} \left(\frac{d}{p} + \frac{d}{q} \right)$$

$$\Rightarrow d\left(\frac{1}{p} + \frac{1}{q} = d\left(2 - \frac{\lambda}{q}\right)\right)$$

□

2.3 Lieb-Oxford inequality

Theorem 2.9. Let $\Psi \in L^2(\mathbb{R}^{dN})$, $\|\Psi\|_{L^2} = 1$ and define the one-body density

$$\rho := \rho_\Psi(x) := \sum_{i=1}^N \int_{\mathbb{R}^{d(N-1)}} |\Psi(x_1, \dots, x_{i-1}, x, x_{i+1}, \dots, x_N)|^2 dx_1, \dots, dx_{i-1}, x_{i+1}, \dots, dx_N$$

Then $\forall 0 < \lambda < d$ we have

$$\int_{\mathbb{R}^{dN}} |\Psi(x_1, \dots, x_N)|^2 \sum_{1 \leq i < j \leq N} \frac{1}{|x_i - x_j|^\lambda} \geq \frac{1}{2} \int_{\mathbb{R}^{dN}} \frac{\rho_\Psi(x)\rho_\Psi(y)}{|x - y|^\lambda} dx dy - C_{d,\lambda} \int_{\mathbb{R}^d} \rho_\Psi^{1+\frac{\lambda}{d}} dx$$

Motivation: In QM $|\Psi|^2$ is the probability density of N quantum particles.

Density functional theory: Only consider $\rho \in L^1(\mathbb{R}^d)$, $\rho \geq 0$ and $\int_{\mathbb{R}^d} \rho dx = N$

If $|\Psi(x_1, \dots, x_N)|$ is symmetric, i.e. $|\Psi(x_1, \dots, x_N)| = |\Psi(x_{\sigma(1)}, \dots, x_{\sigma(N)})|$ for all $\sigma \in S_N$, then the one-body density is simply given by

$$\rho(x) = N \int_{\mathbb{R}^{d(N-1)}} |\Psi(x, x_2, \dots, x_N)|^2 dx_2, \dots, dx_N$$

Actually, there are two kind of particles:

1. Bosons: $|\Psi(x_1, \dots, x_N)| = |\Psi(x_{\sigma(1)}, \dots, x_{\sigma(N)})|$ for all $\sigma \in S_N$
2. Fermions: $|\Psi(x_1, \dots, x_N)| = (-1)^\sigma |\Psi(x_{\sigma(1)}, \dots, x_{\sigma(N)})|$ for all $\sigma \in S_N$

In density functional theory, people try to replace Ψ by ρ in various situations. E.g. if we compute the ground state energy of a system described by a Hamitionian H usually simplified

$$\inf_{\|\Psi\|_{L^2}=1} \langle \Psi, H\Psi \rangle = \inf_{\substack{0 \leq f \in L^2(\mathbb{R}^d) \\ \int f = N}} \underbrace{\inf_{\substack{\Psi: \rho_{Psi} = f \\ \text{In computational physics/chemistry,} \\ \text{one tries to find approximations} \\ \text{for this as a functional } \mathcal{E}(f)}} \langle \Psi, H\Psi \rangle}_{\text{direct energy on mean field energy}}$$

The L-O inequality suggests that for $H = \sum_{1 \leq i < j \leq N} \frac{1}{|x_i - x_j|^\lambda}$ ($\lambda = 1$ and $d = 3$ gives Coulomb interaction) we have the approximation

$$\mathcal{E}(f) = \underbrace{\frac{1}{2} \int_{\mathbb{R}^{dN}} \frac{f(x)f(y)}{|x - y|^\lambda} dx dy}_{\text{direct energy on mean field energy}} - \underbrace{C \int_{\mathbb{R}^d} f(x)^{1+\frac{\lambda}{d}} dx}_{\text{exchange energy}}$$

The effective formulas can be obtained by mean-field approximation:

- Bosons: Take $\Psi = u(x_1) \dots u(x_N)$

$$\begin{aligned} \int_{\mathbb{R}^{dN}} \frac{|\Psi(x_1, \dots, x_N)|^2}{|x_i - x_j|^\lambda} &= \frac{N(N-1)}{2} \int_{\mathbb{R}^{dN}} \frac{|u(x_1)|^2 \dots |u(x_N)|^2}{|x_1 - x_2|} dx_1 \dots dx_N \\ &= \frac{N(N-1)}{2} \int_{\mathbb{R}^{dN}} \frac{|u(x_1)|^2 |u(x_2)|^2}{|x_1 - x_2|} dx_1 dx_2 \end{aligned}$$

and $\rho = N|u(x)|^2 \Rightarrow \frac{(N-1)}{2N} \int_{\mathbb{R}^{dN}} \frac{\rho(x)\rho(y)}{|x-y|^\lambda} dxdy$ (leading term as $N \rightarrow \infty$)

- Fermions: Take $\Psi(x_1, \dots, x_N) = \frac{1}{\sqrt{N!}} \det(u_i(x_j))_{i,j}$ for $\{u_i\}_{i=1}^N$ orthonormal family in L^2

Exercise. For the Slater determinant $\rho(x) = \rho_\Psi(x) = \sum_{i=1}^N |u_i(x)|^2$ we have

$$\sum_{i < j} \int |\Psi|^2 \frac{1}{|x_i - x_j|^\lambda} = \frac{1}{2} \int_{\mathbb{R}^{Nd}} \frac{\rho(x)\rho(y)}{|x-y|^\lambda} - \frac{1}{2} \int \frac{|\sum_i u_i(x)u_i(y)|^2}{|x-y|^\lambda} dxdy$$

Proof of LO inequality using the HL maximal inequality. We use Fefferman-de la Llave approximation: $\forall 0 < \lambda < d$ we have

$$\frac{1}{|x-y|^\lambda} = C_{\lambda,d} \int_0^\infty \int_{\mathbb{R}^d} \mathbb{1}_{B(z,r)}(x) \mathbb{1}_{B(z,r)}(y) dz \frac{dr}{r^{\lambda+d+1}}$$

that can be applied to x_i and x_j s.t.

$$\int_{\mathbb{R}^{dN}} |\Psi|^2 \sum_{i < j} \frac{1}{|x_i - x_j|^\lambda} = C_{d,\lambda} \int_{\mathbb{R}^{dN}} \int_0^\infty \int_{\mathbb{R}^d} \sum_{i < j} \mathbb{1}_{B(z,r)}(x) \mathbb{1}_{B(z,r)}(y) |\Psi|^2 dz \frac{dr}{r^{\lambda+d+1}}$$

Note that

- $\sum_{i < j} X_i X_j = \frac{1}{2} (\sum_{i < j} X_i)^2 - \frac{1}{2} (\sum_{i < j} X_i^2)$ to be used with $X_i = \mathbb{1}_{B(z,r)}(x_i)$ so that $X_i^2 = X_i$
- $\sum_{i=1}^N \int_{\mathbb{R}^{dN}} |\Psi|^2 \mathbb{1}_{B(z,r)}(x_i) = \int_{\mathbb{R}^d} \rho \mathbb{1}_{B(0,r)}(x) dx = \rho \star \mathbb{1}_{B(0,r)}(z)$
- $\int_{\mathbb{R}^{dN}} |\Psi|^2 (\sum_{i=1}^N \mathbb{1}_{B(z,r)}(x_i))^2 \geq (\int_{\mathbb{R}^{dN}} |\Psi|^2 \sum_{i=1}^N \mathbb{1}_{B(z,r)}(x_i))^2$
(using that in general $(\int |\Psi| F^2)(\int |\Psi|^2) \geq (\int |\Psi|^2 F)^2$ and $\int |\Psi|^2 = 1$)

In summary, we get that

$$\int_{\mathbb{R}^{dN}} |\Psi|^2 \sum_{i < j} \frac{1}{|x_i - x_j|^\lambda} \geq \frac{C_{d,\lambda}}{2} \int_{\mathbb{R}^d} \int_0^\infty [(\rho \star \mathbb{1}_{B(0,r)})^2(z) - (\rho \star \mathbb{1}_{B(0,r)})(z)] dz \frac{dr}{r^{d+\lambda+1}}$$

Claim:

$$C_{\frac{d,\lambda}{2}} \int_{\mathbb{R}^d} \int_0^\infty (\rho * \mathbb{1}_{B(0,r)})^2(z) dz \frac{dr}{r^{d+\lambda+1}} = \frac{1}{2} \int_{\mathbb{R}^{2d}} \frac{\rho(x)\rho(y)}{|x-y|^\lambda} dx dy$$

(using Fefferman-de laLlave (exercise))

Consider the error term: by the HL maximal function:

$$(\rho * \mathbb{1}_{B(0,r)})(z) = \int_{B(z,r)} \rho(x) dx \leq |B(z,r)| M\rho(z) = Cr^d M\rho(z)$$

$$\int_0^\infty (\rho * \mathbb{1}_{B(0,r)})(z) \frac{dr}{r^{d+\lambda+1}} \leq C \int_0^\infty M\rho(z) \frac{dr}{r^{\lambda+1}}$$

BUT $\frac{1}{r^{\lambda+1}}$ is not integrable. We use the trivial fact that we can put a positive part

$$[(\rho * \mathbb{1}_{B(0,r)})^2(z) - (\rho * \mathbb{1}_{B(0,r)})(z)]_+ = (\rho * \mathbb{1}_{B(0,r)})^2(z) - \min\{(\rho * \mathbb{1}_{B(0,r)})(z), (\rho * \mathbb{1}_{B(0,r)})^2(z)\}$$

and by the maximal function again:

$$\begin{aligned} \int_0^\infty \min\{...\} \frac{dr}{r^{d+\lambda+1}} &= \int_0^L \dots + \int_L^\infty \dots \leq \int_0^L (M\rho)^2 \frac{r^{2d}}{r^{d+\lambda+1}} dr + \int_L^\infty M\rho \frac{r^d}{r^{d+\lambda+1}} dr \\ &\lesssim (M\rho)^2 L^{d-\lambda} + M\rho(z) L^{d-\lambda} + M\rho(z) L^{-\lambda} \lesssim (M\rho(z))^{1+\frac{\lambda}{d}} \end{aligned}$$

The last inequality follows from $M^2 L^{d-\lambda} \sim M L^{-\lambda}$. Optimizing over L yields:

$$\int_{\mathbb{R}}^d \int_0^\infty \min\{...\} \frac{dr}{r^{d+\lambda+1}} dz \lesssim \int_{\mathbb{R}^d} (M\rho(z))^{1+\frac{\lambda}{d}} dz \underset{\text{HL max ineq}}{\lesssim} \int_{\mathbb{R}^d} \rho(z)^{1+\frac{\lambda}{d}} dz$$

□

Remark. L-O: $\int |\Psi|^2 \sum_{i < j} \frac{1}{|x_i - x_j|} \geq \frac{1}{2} \int \frac{\rho(x)\rho(y)}{|x-y|^\lambda} - C_{L-O} \int_{\mathbb{R}^d} \rho(x)^{1+\frac{\lambda}{d}} dx$ The most interesting case is $d = 3$ and $\lambda = 1$

→ 1979 Lieb: using Newton's theorem for Coulomb $\rightsquigarrow C_{L-O} \simeq 8$

→ Lieb-Oxford 1980 $\rightsquigarrow C_{L-O} \simeq 1.68$

Numerically, the expected best constant is $C_{L-O} \simeq 1.45$

(2022 Lewin-Lieb-Seiringer $C_{L-O} \simeq 1.58$)

→ Our proof is due to Lieb-Solovej-Yngvason. This gives a very bad constant! ≥ 100

Open Problem: Find best constant for HL maximal inequality.

Only solved for $d = p = 1$: $|x \in \mathbb{R} : Mf(x) > \lambda| \leq \frac{C}{\lambda} \|f\|_{L^1}$. By Melas (2003) finding: $C = \frac{11+\sqrt{61}}{12} \simeq 1.567$

Dependence on d : $\|Mf\|_{L^p} \leq C_{d,p} \|f\|_{L^p} \quad \forall 1 < p \leq \infty$ In our proof (using Vitali's covering lemma) we get that $d \mapsto C_{d,p}$ grows exponentially.

By Stein's theorem one can see that $C_{d,p} = C_p < \infty$ and by Stein's lemma: $C_{s,d+1,p} \leq C_{s,d,p}$ for all $d \in \mathbb{N}$

Chapter 3

Interpolation Theory

Motivation:

Exercise.

$$\begin{aligned} 0 \leq A_0 \leq B_0 \\ 0 \leq A_1 \leq B_1 \end{aligned} \Rightarrow A_\theta \leq B_\theta \quad \forall \theta \in [0, 1]$$

with $A_\theta = A_0^{1-\theta} A_1^\theta, B_\theta = B_0^{1-\theta} B_1^\theta$

Exercise (Hölder).

$$|\int f g| \leq \|f\|_p \|g\|_{p'} \quad \frac{1}{p} + \frac{1}{p'} = 1, \quad 1 \leq p \leq \infty$$

Interpolation:

$$\|f\|_r \leq \|f\|_p^\theta \|g\|_q^{1-\theta} \quad \forall p < r < q \exists \theta \in (0, 1)$$

Thus: $L^p \cap L^q \subset L^r$

Reversely: $L^r \subset L^p + L^q$ as $f \in L^r \Rightarrow f = \underbrace{f \mathbb{1}_{|f|>\varepsilon}}_{\in L^p \cap L^r} + \underbrace{f \mathbb{1}_{|f|\leq\varepsilon}}_{\in L^q \cap L^r}$,

since $\int |f|^p \mathbb{1}_{|f|>\varepsilon} \leq \int \frac{|f|^r}{\varepsilon^{r-p}} < \infty$

Exercise (Young inequality).

$$\|f \star g\|_{L^r} \leq \|f\|_p \|g\|_q, \quad \frac{1}{p} + \frac{1}{q} = 1 + \frac{1}{r}$$

proof 1.

Duality: $\|f \star g\|_{L^r} = \sup_{\|h\|_{r'}=1} |\int (f \star g) h| = \sup_{\|h\|_{r'}=1} |\int f(x) g(y-x) h(y) dx dy| \leq \|f\|_p \|g\|_q \underbrace{\|h\|_r}_{=1}$ \square

proof 2.

Fix g s.t. $\|g\|_q = 1$ Define $T : f \mapsto f \star g$. We prove that $\|Tf\|_q \leq \|f\|_p \Leftrightarrow \|T\|_{L^p \rightarrow L^r} \leq 1$. We have two cases:

- $p = 1, r = q$

$$\|Tf\|_q \leq \|f\|_1 \Leftrightarrow \|f \star g\|_{L^q} \leq \|f\|_1 \|g\|_q$$

- $p = q', r = \infty$

$$\|Tf\|_{\infty} \leq \|f\|_{q'} \Leftrightarrow \|f * g\|_{\infty} \leq \|f\|_{q'} \|g\|_q$$

Desired claim: if $\|T\|_{L^1 \rightarrow L^q} \leq 1$, $\|T\|_{L^{q'} \rightarrow L^{\infty}} \leq 1 \stackrel{(\text{?})}{\Rightarrow} \|T\|_{L^p \rightarrow L^r} \leq 1$ \square

Exercise (Fourier Transform). $f \in L^1$, $\hat{f}(k) = \int_{\mathbb{R}^d} f(x) e^{-2\pi i kx} dx$
 $\Rightarrow \hat{f} \in L^{\infty}$, $\|\hat{f}\|_{\infty} \leq \|f\|_1$

Plancherel: $\|\mathcal{F}f\|_2 = \|\hat{f}\|_2 = \|f\|_2$

3.1 Complex Interpolation

Theorem 3.1 (Hausdorff-Young). The Fourier transform \mathcal{F} can be defined on $L^1 + L^2$ and it satisfies

$$\|\mathcal{F}f\|_{p'} \leq \|f\|_p, \quad \forall 1 \leq p \leq 2, \frac{1}{p} + \frac{1}{p'} = 1$$

"Formal proof": $\|\mathcal{F}\|_{L^1 \rightarrow L^{\infty}} \leq 1$, $\|\mathcal{F}\|_{L^2 \rightarrow L^2} = 1$
 $\Rightarrow \|\mathcal{F}\|_{L^p \rightarrow L^{p'}} \leq 1, \quad \forall p \in [1, 2]$ (by Riesz-Thorin)

Theorem 3.2 (Riesz-Thorin). Let (X, μ) , (Y, σ) be two measure spaces, both sigma finite. Let $T : L^{p_0}(X) + L^{p_1}(X) \rightarrow L^{q_0}(Y) + L^{q_1}(Y)$ be a linear operator, where $1 \leq p_0 < p_1 \leq \infty$, $1 \leq q_0, q_1 \leq \infty$ s.t.

$$\|T\|_{L^{p_0} \rightarrow L^{q_0}} \leq 1, \quad \|T\|_{L^{p_1} \rightarrow L^{q_1}} \leq 1$$

Then:

$$\|T\|_{L^{p_\theta} \rightarrow L^{q_\theta}} \leq 1$$

for any $\theta \in (0, 1)$,

$$\frac{1}{p_\theta} = \frac{1-\theta}{p_0} + \frac{\theta}{p_1}, \quad \frac{1}{q_\theta} = \frac{1-\theta}{q_0} + \frac{\theta}{q_1}$$

We use the maximum principle for analytic functions in complex planes!

Maximum principle: $f : \Omega \overset{\text{bounded, open}}{\subset} \mathbb{C} \rightarrow \mathbb{C}$, analytic in Ω , continuous on $\bar{\Omega}$. Then:

$$\sup_{z \in \bar{\Omega}} |f(z)| = \sup_{z \in \partial\Omega} |f(z)|$$

Lemma 3.3 (Hadamard's three lines theorem). Let f be analytic on $S = \{0 < \operatorname{Re} z < 1\}$ and f continuous on \overline{S} and $|f(z)| \leq Ce^{|z|}$ for all $z \in S$. Then:

$$\sup_{z \in S} |f(z)| = \sup_{z \in \overline{S}, \operatorname{Re} z \in \{0,1\}} |f(z)|$$

Proof. Step 1: Assume additionally $|f(z)| \rightarrow 0$ as $|z| \rightarrow \infty$. Take $l_n \rightarrow \infty$.

Then $\sup_{\operatorname{Im} z = \pm l_n} |f(z)| \rightarrow 0$ as $n \rightarrow \infty$.

By the maximum principle:

$$\begin{aligned} \sup_{z \in S_n} |f(z)| &\leq \sup_{z \in \partial S_n} |f(z)| \\ &\leq \max\left\{\sup_{\operatorname{Re} z \in \{0,1\}} |f(x)|, \sup_{\operatorname{Im} z \in \{\pm l_n\}} |f(x)|\right\} \end{aligned}$$

Take $n \rightarrow \infty$.

$$\sup_{z \in S} |f(z)| = \sup_{z \in \overline{S}, \operatorname{Re} z \in \{0,1\}} |f(z)|$$

Step 2: In general, if we only know $|f(z)| \leq Ce^{|z|}$, then we define

$$f_n(z) = f(z)e^{\frac{z^2-1}{n}} \quad \forall z \in S \quad \forall n \geq 1$$

Then:

- $|f_n(z)| = |f_n(x + iy)| \leq Ce^{|x|}e^{-\frac{y^2}{n}} \leq Ce^{-\frac{y^2}{n} + y + 1} \rightarrow 0$, as $|y| \rightarrow \infty$ or equivalently $|z| \rightarrow \infty$, where $z = x + iy$
- $|f_n(z)| \leq |f(z)|$ for all $z \in S$

By step 1, $\sup_{z \in S} |f_n(z)| \leq \sup_{\operatorname{Re} z \in \{0,1\}} |f_n(z)| \leq \sup_{\operatorname{Re} z \in \{0,1\}} |f(z)|$.

Take $n \rightarrow \infty$, use $f_n(z) \rightarrow f(z)$ pointwise \Rightarrow desired bound for f \square

Remark. The H-3-lines still holds if $\forall \varepsilon > 0 : e^{-\varepsilon|z|^2} f(z) \rightarrow 0$ as $|z| \rightarrow \infty$

proof for Hölder inequality using H3L. Assume $\|f\|_{L^p} = 1$, $\|g\|_{L^q} = 1$. We prove $\|f\|_{L^r} = 1$, $p < r < q$. First try:

$$F(z) = \int |f|^{p_z} \quad p_z = (1-z)p + zq$$

If $|f| > 0$ s.t. $|f|^{p_z}$ is well defined and $|f|^{p_z}$ is analytic, then $F(z)$ is also analytic and by H3L:

$$\begin{aligned} \int |f|^r &\leq \sup_{z \in S} |F(z)| \leq \sup_{\operatorname{Re} z \in \{0,1\}} |F(z)| \\ &\leq \max\left(\int |f|^p, \int |f|^q\right) = 1 \end{aligned}$$

There are two restrictions:

1. $|f|^{p_z}$ might not be well defined if $f = 0$
2. the center $|F(z)| \leq Ce^{|z|}$ is somewhere nontrivial

Second try: Assume f is a step function, namely $f(x) = \sum_{n=1}^M a_m e^{i\alpha_m} \mathbb{1}_{A_m}$, $a_m > 0$, $\alpha_m \in \mathbb{R}$, $A_m \subset \Omega$, (A_m) disjoint. Define

$$F(z) := \sum_{m=1}^M a_m^{P_z} |A_m|, \quad P_z = p(1-z) + qz \quad \forall z \in \mathbb{C}$$

$$\stackrel{H^3 L}{\Rightarrow} \int |f|^r \leq \max(\int |f|^p, \int |f|^q)$$

for all step functions f . To conclude, we use the density argument, $\{\text{step functions}\}$ is dense in $L^r(\Omega)$ if $\begin{cases} \Omega \text{ is sigma-finite} \\ r < \infty \end{cases}$ \square

Motivated by the proof above we want to use our new tools to prove Riesz-Thorin:

proof of Riesz-Thorin.

By duality: $\|Tf\|_{p_\Theta} \leq \|f\|_{p_\Theta} \quad \forall f \in L^{p_\Theta}$

$$\Leftrightarrow |\int_Y (Tf)(y)g(y)dy| \leq \|f\|_{L^{p_\Theta}} \|g\|_{L^{q_\Theta}}$$

We will prove the statement for step functions and then use a density argument.

Step 1: Take f, g step functions.

$$f(x) := \sum_{m=1}^M a_m e^{i\alpha_m} \mathbb{1}_{A_m}(x)$$

and

$$g(y) := \sum_{n=1}^N b_n e^{i\beta_n} \mathbb{1}_{B_n}(y)$$

for $a_m, b_m > 0$, $\alpha_m, \beta_m \in \mathbb{R}$, $\{A_m\}_m$ disjoint sets of finite measure, $\{B_n\}_n$ disjoint sets of finite measure.

For all $z \in \overline{S} = \{0 \leq \operatorname{Re} z \leq 1\}$ define

$$f_z(x) = \sum_{m=1}^M a_m^{P_z} e^{i\alpha_m} \mathbb{1}_{A_m}(x)$$

for $P_z = p_\Theta(\frac{1-z}{p_0} + \frac{z}{p_1})$ (chosen s.t. $P_\Theta = 1$)

$$g_z(y) = \sum_{n=1}^N b_n^{Q_z} e^{i\beta_n} \mathbb{1}_{B_n}(y)$$

for $Q_z = q'_\Theta \left(\frac{1-z}{q'_0} + \frac{z}{q'_1} \right)$ (chosen s.t. $Q_\Theta = 1$)

Define: $F(z) := \int_Y (Tf_z)(y)g_z(y)dy$. One can prove that F is analytic in S , $|F(z)| \leq e^{c|z|}$ and that F is continuous in \overline{S} .

By H3L Lemma:

$$|\int (Tf)g| = |F(\Theta)| \leq \sup_{Re z \in \{0,1\}} |F(z)|$$

So now we have to check the cases where $Re z = 0$ and $Re z = 1$.

$Re z = 0$:

$$\begin{aligned} f_z(x) &= \sum_{m=1}^M a_m^{P_z} e^{i\alpha_m} \mathbb{1}_{A_m}(x) \\ \Rightarrow \int |f_z(x)|^p dx &= \sum_{m=1}^M |a_m^{P_z}|^p |A_m| \\ &= \sum_{m=1}^M |a_m^{Re P_z}|^p |A_m| \\ &= \sum_{m=1}^M a_m^{p(\frac{p_\Theta}{p_0} + Re z(\frac{p_\Theta}{p_1} - \frac{p_\Theta}{p_0}))} |A_m| \\ &\stackrel{Re z=0}{=} \sum_{m=1}^M a_m^{p \frac{p_\Theta}{p_0}} |A_m| \end{aligned}$$

Analogously:

$$\begin{aligned} |F(z)| &= |\int_Y (Tf_z)g_z| \leq \|Tf_z\|_{L^{q_0}} \|g_z\|_{L^{q'_0}} \\ &\leq \|f_z\|_{L^{p_0}} \|g_z\|_{q'_0} \\ &= \underbrace{\left(\sum_{m=1}^M a_m^{p_\Theta} |A_m| \right)^{\frac{1}{p_0}}}_{=\|f\|_{L^{p_\Theta}}^{p_\Theta}=1} \underbrace{\left(\sum_{n=1}^N b_n^{q'_\Theta} |B_n| \right)^{\frac{1}{q'_0}}}_{=\|g\|_{L^{q'_\Theta}}^{q'_\Theta}=1} \\ &= 1 \end{aligned}$$

$Re z = 1$:

$$\begin{aligned} \int |f_z(x)|^p dx &= \sum_{m=1}^M a_m^{p \frac{p_\Theta}{p_1}} |A_m| \\ \int |g_z(y)|^p dy &= \sum_{n=1}^N b_n^{p \frac{q'_\Theta}{q'_1}} |B_n| \end{aligned}$$

$$\begin{aligned}
\Rightarrow |F(z)| = \left| \int_Y (Tf_z)g_z \right| &\leq \|Tf_z\|_{L^{q_1}} \|g_z\|_{L^{q'_1}} \\
&\leq \|f_z\|_{L^{p_1}} \|g_z\|_{L^{q'_1}} \\
&= \left(\sum_{m=1}^M a_m^{p_\Theta} |A_m| \right)^{\frac{1}{p_1}} \left(\sum_{n=1}^N b_n^{q'_\Theta} |B_n| \right)^{\frac{1}{q'_1}} \\
&= 1
\end{aligned}$$

Step 2 (Density Argument): We proved the statement for the case that f, g are step functions. By a standard density argument, this extends to all $g \in L^{q_\Theta}$, as

$$\sup_{g \text{ step function}} \frac{|\int hg|}{\|g\|_{L^{q'}}} = \|h\|_{L^q} \quad \forall 1 \leq q \leq \infty \text{ (exercise)}$$

Thus: $\|Tf\|_{L^{q_\Theta}} \leq \|f\|_{L^{p_\Theta}}$ for all f step functions.

We want to extend this to all $f \in L^{p_\Theta}$.

Easy case: Assume $f \in L^{p_\Theta} \cap L^{p_0}$ and $p_0 < p_\Theta < \infty$. Then find $\{f_n\}_{n \in \mathbb{N}}$ step functions s.t. $f_n \rightarrow f$ in L^{p_Θ} and L^{p_0} ($|f_n| \leq |f|$).

By assumption: $\|Tf_n - Tf\|_{L^{q_0}} \leq \|f_n - f\|_{L^{p_0}} \rightarrow 0$.

By step 1: $\|Tf_n - Tf\|_{L^{p_\Theta}} \leq \|f_n - f\|_{L^{p_\Theta}} \rightarrow 0$

Thus: $\left. \begin{array}{l} Tf_n \rightarrow Tf \text{ in } L^{L^{p_0}} \\ \{Tf_n\} \text{ Cauchy sequence in } L^{p_\Theta} \end{array} \right\} \Rightarrow Tf_n \rightarrow Tf \text{ in } L^{p_\Theta}$

Consequently, by step 1 again:

$$\|Tf\|_{L^{q_\Theta}} = \lim_{n \rightarrow \infty} \|Tf_n\|_{L^{q_\Theta}} \leq \lim_{n \rightarrow \infty} \|f_n\|_{L^{p_\Theta}} = \|f\|_{L^{p_\Theta}}$$

More general case: Assume $f \in L^{p_\Theta}$, $p_0 < p_\Theta < p_1$

Decompose: $f = \underbrace{f \mathbf{1}_{\{|f| > \varepsilon\}}}_{=: f_\varepsilon \in L^{p_0} \cap L^{p_\Theta}} + \underbrace{f \mathbf{1}_{\{|f| < \varepsilon\}}}_{\in L^{p_1}}$

Then: $Tf = Tf_\varepsilon + T\tilde{f}_\varepsilon$

$$\|Tf_\varepsilon\|_{L^{q_\Theta}} \leq \|f_\varepsilon\|_{L^{p_\Theta}} \leq \|f\|_{L^{p_\Theta}} \quad \forall \varepsilon > 0$$

$$\left\| T\tilde{f}_\varepsilon \right\|_{L^{q_1}} \leq \left\| \tilde{f}_\varepsilon \right\|_{L^{p_1}} \rightarrow 0 \quad \text{as } \varepsilon \rightarrow 0$$

□

We used:

Q: If $F = F_\varepsilon + G_\varepsilon$ and $\|F_\varepsilon\|_{L^q} \leq 1$ for all $\varepsilon > 0$ and $\|G_\varepsilon\|_{L^r} \rightarrow 0$ as $\varepsilon \rightarrow 0$. How can we conclude that $\|F\|_{L^q} \leq 1$?

Idea: If we only need an upper bound, then Fatou is helpful!

Apply: From $\|G_\varepsilon\|_{L^r} \rightarrow 0 \Rightarrow$ up to a subseq. $G_\varepsilon \rightarrow 0$ a.e.

$\Rightarrow F_\varepsilon \rightarrow F$ a.e.

$\Rightarrow \|F\|_{L^q} \leq \limsup \varepsilon \rightarrow 0 \|F_\varepsilon\|_{L^q} \leq 1$

In general: If $g_n \rightarrow g$ in L^r for $r < \infty$, then \exists subseq:

$$\begin{cases} g_n(x) \rightarrow g(x) \text{ a.e.} \\ |g_n(x)| \leq G(x) \in L^r \end{cases}$$

Note: If $p = p_0 = p_1$, then we cannot use $p_0 < p_\Theta < p_1$ but we can simply use Hölder:

$$\|Tf\|_{L^{q_\Theta}} \leq \max\{\|Tf\|_{L^{q_0}}, \|Tf\|_{L^{q_1}}\} \leq \|f\|_{L^p}$$

3.2 Real Interpolation

Theorem 3.4 (Marcinkiewicz Interpolation Theorem). Let X be a sigma finite measure space. Take $1 \leq p_0 < p_1 \leq \infty$. Assume T is a quasi linear map of measurable functions from X to measurable functions in X , i.e.

$$\begin{cases} |T(\lambda f)(y)| \leq |\lambda| |Tf(y)| \quad \forall \lambda \in \mathbb{R} \\ |T(f+g)(y)| \leq K(|Tf(y)| + |Tg(y)|) \end{cases}$$

for all $y \in X$. And

$$\begin{cases} \|T\|_{L^{p_0} \rightarrow L_w^{p_0}} \leq 1 \\ \|T\|_{L^{p_1} \rightarrow L_w^{p_1}} \leq 1 \end{cases}$$

Then:

$$\|T\|_{L^p \rightarrow L^p} \leq C_{p,K} \|Tf\|_{L^p} \leq C_p \|f\|_{L^p}$$

for all $p_0 < p < p_1$.

Motivation: HL max ineq: $\left. \begin{array}{l} \|Mf\|_{L_w^1} \leq C \|f\|_{L^1} \\ \|Mf\|_{L^\infty} \leq \|f\|_{L^\infty} \end{array} \right\} \Rightarrow \|Mf\|_{L^p} \leq \|f\|_{L^p}$

Recall: $f \in L_w^p \Leftrightarrow \|f\|_{L_w^p} = \sup_{\lambda > 0} \lambda |\{|f| > \lambda\}|^{\frac{1}{p}} < \infty$

Here $\|\cdot\|_{L_w^p}$ is a quasi norm, i.e.

$$\|f+g\|_{L_w^p} \leq K(\|f\|_{L_w^p} + \|g\|_{L_w^p})$$

Remark. We can take $C_{p,K} = 2K(\frac{p}{p-p_0} + \frac{p}{p_1-p})^{\frac{1}{p}} \rightarrow \infty$ as $p \rightarrow p_1$ or $p \rightarrow p_0$

In comparison to the Riesz-Thorin thm, the constant here is not too good but the thm applies to non-linear mappings!

proof of Marcinkiewicz Interpolation Theorem. Start with the layer-cake representation ($p_0 < p < p_1 < \infty$)

$$\|Tf\|_{L^p}^p = p \int_0^\infty \lambda^{p-1} |\{|Tf| > \lambda\}| d\lambda$$

$$\begin{aligned} \text{Decompose: } f &= \underbrace{f \mathbb{1}_{\{|f| > \frac{\lambda}{2}\}}}_{f_\lambda^>} + \underbrace{f \mathbb{1}_{\{|f| < \frac{\lambda}{2}\}}}_{f_\lambda^<} \\ \Rightarrow |Tf| &\leq K(|Tf_\lambda^>| + |Tf_\lambda^<|) \end{aligned}$$

$$\begin{aligned} |\{|Tf| > \lambda\}| &\leq |\{|Tf_\lambda^>| > \frac{\lambda}{2K}\}| + |\{|Tf_\lambda^<| > \frac{\lambda}{2K}\}| \\ &\lesssim \|Tf_\lambda^>\|_{L_w^{p_0}}^{p_0} + \frac{1}{\lambda^{p_1}} \|f_\lambda^<\|_{L^{p_1}}^{p_1} \\ &\lesssim \frac{1}{\lambda^{p_0}} \|f_\lambda^>\|_{L^{p_0}}^{p_0} + \frac{1}{\lambda^{p_1}} \|f_\lambda^<\|_{L^{p_1}}^{p_1} \\ &= \frac{1}{\lambda^{p_0}} \int |f|^{p_0} \mathbb{1}_{\{|f| > \frac{\lambda}{2}\}} + \frac{1}{\lambda^{p_1}} \int |f|^{p_1} \mathbb{1}_{\{|f| < \frac{\lambda}{2}\}} \end{aligned}$$

$$\begin{aligned} \Rightarrow \|Tf\|_{L^p}^p &\lesssim \int_0^\infty d\lambda \left[\frac{1}{\lambda^{p_0}} \int |f|^{p_0} \mathbb{1}_{\{|f| > \frac{\lambda}{2}\}} + \frac{1}{\lambda^{p_1}} \int |f|^{p_1} \mathbb{1}_{\{|f| < \frac{\lambda}{2}\}} \right] \\ &= \int |f|^{p_0} \underbrace{\int_0^\infty \lambda^{p-p_0-1} \mathbb{1}_{\{|f| > \frac{\lambda}{2}\}} d\lambda}_{|f|^{p-p_0}} + \int |f|^{p_1} \underbrace{\int_0^\infty \lambda^{p-p_1-1} \mathbb{1}_{\{|f| < \frac{\lambda}{2}\}} d\lambda}_{|f|^{p-p_0}} \\ &\lesssim \int |f|^p \end{aligned}$$

case $p_1 = \infty$ (exercise)

□

Chapter 4

Lorentz spaces

Motivation: X_0, X_1 2 Banach spaces compatible (i.e. $X_0 \cap X_1$ and $X_0 + X_1$ can be defined, $X_0, X_1 \subset Z$ vs)

$$X_0 \cup X_1 \subset X \subset X_0 + X_1$$

X is intermediate between $X_0 \cap X_1$ and $X_0 + X_1$.

With Lorentz spaces $L_{p,q}$ we want to study what's inbetween $L^{p,p} = L^p$ and $L^{p,\infty} = L_w^p$ if $p < q$. If $p > q$ then $L^{p,q}$ has a stronger norm than L^p .

Definition 4.1 (Lorentz space). Let $f : \mathbb{R}^d \rightarrow \mathbb{C}$ be measurable. $f \in L^{p,q} \Leftrightarrow \|f\|_{L^{p,q}} < \infty$

- $\|f\|_{L^{p,q}} := p^{\frac{1}{q}} \left\| |\lambda| \{ |f| > \lambda \}^{\frac{1}{p}} \right\|_{\mathbb{R}_+, \frac{d\lambda}{\lambda}} = \left(\int_0^\infty \lambda^{q-1} |\{ |f| > \lambda \}|^{\frac{q}{p}} d\lambda \right)^{\frac{1}{q}}$
- $\|f\|_{L^{p,\infty}} := \|f\|_{L_w^p}$

Definition 4.2 (Quasi-normed vector space). Let V be a vector space with field \mathbb{C} . We say that $(V, \|\cdot\|)$ is a quasi normed vector space, if $\|\cdot\| : V \rightarrow [0, \infty)$ satisfies:

- $\|x\| = 0 \Leftrightarrow x = 0$
- $\|\lambda x\| = |\lambda| \|x\|$ for all $\lambda \in \mathbb{C}, x \in V$
- $\|x + y\| \leq C(\|x\| + \|y\|)$ for all $x, y \in V, C \geq 1$ independent of x, y

Remark. It's easy to see that $\|\cdot\|_{L^{p,q}}$ is a quasi norm. (to see this use $|\{|f+g| > t\}| \leq |\{|f| > \frac{t}{2}\}| + |\{|g| > \frac{t}{2}\}|$)

Example 4.3. $L^{1,\infty} = L_w^1$ is a quasi normed space with $C = 2$. Recall that $\|f\|_{L^{1,\infty}} = \sup_{\lambda > 0} \lambda |\{f > \lambda\}|$.

Take $f, g \in L^{1,\infty}$. Consider $|\{|f + g| > \lambda\}| \leq |\{|f| > \frac{\lambda}{2}\}| + |\{|g| > \frac{\lambda}{2}\}|$

$$\begin{aligned} \lambda |\{|f + g| > \lambda\}| &\leq 2\left(\frac{\lambda}{2} |\{|f| > \frac{\lambda}{2}\}| + \frac{\lambda}{2} |\{|g| > \frac{\lambda}{2}\}|\right) \\ &\stackrel{\sup_{\lambda}}{\leq} 2(\|f\|_{L^{1,\infty}} + \|g\|_{L^{1,\infty}}) \end{aligned}$$

Remark. There exists no norm $|||\cdot|||$ on $L^{1,\infty}$ which is equivalent to $\|\cdot\|_{L^{1,\infty}}$, i.e. there are no $C_1, C_2 > 0$ s.t. for all $f \in L^{1,\infty}$:

$$C_1 \|f\|_{L^{1,\infty}} \leq |||f||| \leq C_2 \|f\|_{L^{1,\infty}}$$

Example 4.4. Define $f_{N,R}(x) = \frac{1}{\log N} \frac{1}{|x - \frac{R}{N}|}$ for $N \geq 1$, $R \in \{1, 2, \dots, N\}$. Easy to check: $f_{N,R} \in L^{1,\infty}(\mathbb{R})$ and:

$$\begin{aligned} |\{f_{N,R} > \lambda\}| &= |\{x \mid \frac{1}{\log N} \frac{1}{|x - \frac{R}{N}|} > \lambda\}| \\ &= |\{x \mid \frac{1}{\log N \lambda} > |x - \frac{R}{N}|\}| \\ &= \frac{2}{\log(N) \lambda} \\ \Rightarrow \|f_{N,R}\|_{L^{1,\infty}} &= \sup_{\lambda > 0} \lambda |f_{N,R} > \lambda| = \frac{2}{\log N} \rightarrow 0 \quad (\text{as } N \rightarrow 0) \end{aligned}$$

However, $F_N(0) = \frac{1}{N} \sum_{R=1}^N f_{N,R}(0) = \frac{1}{\log N} \sum_{R=1}^N \frac{1}{R} \geq 1 > 0 \forall N$

Exercise: similarly $\|F_N\|_{L^{1,\infty}} \not\rightarrow 0$ as $N \rightarrow \infty$. This shows: \nexists norm equivalent to $\|\cdot\|_{L^{1,\infty}}$. ($L^{1,\infty}$ is not locally convex, i.e. \nexists local basis of neighbourhood of 0 which consists of convex sets)

4.1 Aoki-Robewicz theorem

Theorem 4.5 (Aoki-Robewicz). Let $(V; \|\cdot\|)$ be a quasi normed vector space. Then it is metrizable, i.e. \exists a metric $d(x, y) = \Lambda(x - y)$ for a function $\Lambda : V \rightarrow [0, \infty]$ s.t.

- $\Lambda(x) = 0 \Leftrightarrow x = 0$
- $\Lambda(x) = \Lambda(-x)$
- $\Lambda(x + y) \leq \Lambda(x) + \Lambda(y)$

and

$$x_n \rightarrow x \text{ w.r.t. the metric} \Leftrightarrow \|x_n - x\| \rightarrow 0$$

Lemma 4.6. Let V be a quasi normed vector space and C be the constant of the quasi norm $\|\cdot\|$ and $\alpha > 0$ s.t. $(2C) = 2^\alpha$. Then $\forall n \in \mathbb{N}$, $\forall x_1, \dots, x_n \in V$:

$$\|x_1 + x_2 + \dots + x_n\|^{\frac{1}{\alpha}} \leq 4(\|x_1\|^{\frac{1}{\alpha}} + (\|x_2\|^{\frac{1}{\alpha}} + \dots + (\|x_n\|^{\frac{1}{\alpha}}))$$

Remark. If we use the assumption $\|x + y\| \leq C(\|x\| + \|y\|)$, then

$$\begin{aligned} \|x_1 + x_2 + \dots + x_n\| &\leq C\|x_1\| + C\|x_2 + \dots + x_n\| \\ &\leq C\|x_1\| + C^2\|x_2\| + C^2\|x_3 + \dots + x_n\| \\ &\leq C\|x_1\| + C^2\|x_2\| + \dots + C^{N-1}\|x_n\| \end{aligned}$$

proof of lemma. Define $H : V \rightarrow [0, \infty)$, $H(x) := \begin{cases} 0, & \text{if } x = 0 \\ 2^{j\alpha}, & \text{if } 2^{(j-1)\alpha} \leq \|x\| < 2^{j\alpha} \quad \forall j \in \mathbb{Z} \end{cases}$

(Dyadic decomposition)

Then: $\|x\| \leq \|H(x)\| \leq 2^\alpha \|x\| \quad \forall x \in V$

Claim: $\forall n \forall x_1, \dots, x_n : \|x_1, \dots, x_n\|^{\frac{1}{\alpha}} \leq 2(H(x_1)^{\frac{1}{\alpha}} + \dots + H(x_n)^{\frac{1}{\alpha}})$

We prove this by induction in n :

$$n = 1 : \|x\|_1 \leq H(x_1) \Rightarrow \|x\|_1^{\frac{1}{\alpha}} \leq H(x_1)^{\frac{1}{\alpha}}$$

$(n-1) \rightsquigarrow n$: Wlog, assume $\|x\|_1 \geq \|x\|_2 \geq \dots \geq \|x\|_n \Rightarrow H(x_1) \geq H(x_2) \geq \dots \geq H(x_n)$

Case 1: $\exists i_0 \in \{1, \dots, n-1\}$ s.t. $H(x_{i_0}) = H(x_{i_0+1}) = 2^{j_0\alpha}$

Then: $\|x_{i_0} + x_{i_0+1}\| \leq C(\|x_{i_0}\| + \|x_{i_0+1}\|) \leq \underbrace{H(x_{i_0}) + H(x_{i_0+1})}_{2C2^{j_0\alpha}} \underset{2C=2^\alpha}{=} 2^{(j_0+1)\alpha}$

$$\Rightarrow H(x_{i_0} + x_{i_0+1}) \leq 2^{(j_0+1)\alpha}$$

$\Rightarrow H(x_{i_0} + x_{i_0+1})^{\frac{1}{\alpha}} \leq 2^{j_0+1} = H(x_{i_0})^{\frac{1}{\alpha}} + H(x_{i_0+1})^{\frac{1}{\alpha}}$ By induction assumption:

$$\begin{aligned} \|x_1 + \dots + x_n\| &= \|x_1 + \dots + (x_{i_0} + x_{i_0+1}) + \dots + x_n\| \\ &\leq 2(H(x_1)^{\frac{1}{\alpha}} + \dots + \underbrace{H(x_{i_0} + x_{i_0+1})^{\frac{1}{\alpha}}}_{\leq H(x_{i_0})^{\frac{1}{\alpha}} + H(x_{i_0+1})^{\frac{1}{\alpha}}} + \dots + H(x_n)^{\frac{1}{\alpha}}) \end{aligned}$$

Case 2: $H(x_1) > H(x_2) > \dots > H(x_n)$
 $\Rightarrow H(x_i) \leq H(x_{i-1})2^\alpha \forall i \Rightarrow H(x_i) \leq 2^{-(i-1)\alpha}H(x_1) \forall i$

$$\begin{aligned} \Rightarrow \|x_1 + \dots + x_n\| &\leq C(\|x_1\| + \|x_2 + \dots + x_n\|) \\ &\leq \max(2C\|x_1\|, 2C\|x_2 + \dots + x_n\|) \\ &\leq \dots \leq \max(\underbrace{2C}_{=2^\alpha}, (2C)^2\|x_2\|, \dots, (2C)^{n-1}\|x_n\|) \\ &\leq \max(2^\alpha H(x_1), 2^{2\alpha} H(x_2), \dots, 2^{(n-1)\alpha} H(x_n)) \\ &= 2^\alpha H(x_1) \end{aligned}$$

$$\Rightarrow \|x_1 + x_2 + \dots + x_n\|^{\frac{1}{\alpha}} 2H(x_1)^{\frac{1}{\alpha}} \leq 2(H(x_1)^{\frac{1}{\alpha}} + \dots + H(x_n)^{\frac{1}{\alpha}})$$

Thus the claim is correct and this implies the conclusion.

$$\begin{aligned} \|x_1 + x_2 + \dots + x_n\|^{\frac{1}{\alpha}} &\leq 2(H(x_1)^{\frac{1}{\alpha}} + \dots + H(x_n)^{\frac{1}{\alpha}}) \\ &\leq 4(\|x_1\|^{\frac{1}{\alpha}} + \dots + \|x_n\|^{\frac{1}{\alpha}}) \end{aligned}$$

□

Remark. The difficulty is related to L^p spaces with $0 < p < 1$.

Proof of the Aoki-Robewicz theorem.

Define $\Lambda : V \rightarrow [0, \infty]$ by $\Lambda(x) = \inf\left\{\sum_{i=1}^n \|x_i\|^{\frac{1}{\alpha}} : \forall n \ \forall\{x_i\} : x = \sum_{i=1}^n \|x_i\|\right\}$,

where $2C = 2^\alpha$ and C is the constant from the quasi-norm.

Then:

$$\|x\|^{\frac{1}{\alpha}} \geq \Lambda(x) \geq \frac{1}{4} \|x\|^{\frac{1}{\alpha}}$$

(the second bound comes from lemma 4.6)

This function satisfies all desired properties:

- $\|f\|_{L^{p,q}} := p^{\frac{1}{q}} \left\| \lambda |\{|f| > \lambda\}|^{\frac{1}{p}} \right\|_{\mathbb{R}_+, \frac{d\lambda}{\lambda}} = \left(\int_0^\infty \lambda^{q-1} |\{|f| > \lambda\}|^{\frac{q}{p}} d\lambda \right)^{\frac{1}{q}}$
 - $\|f\|_{L^{p,\infty}} := \|f\|_{L_w^p}$. This is an easy consequence from the definition. In fact $\forall \varepsilon > 0 : \exists\{x_i\}, \{y_j\}$ s.t. $x = \sum_i \|x_i\|, y = \sum_j \|y_j\|$
- $$\begin{aligned} \Lambda(x) &\geq \sum_i \|x_i\|^{\frac{1}{\alpha}} - \varepsilon \\ \Lambda(y) &\geq \sum_j \|y_j\|^{\frac{1}{\alpha}} - \varepsilon \end{aligned}$$

$$\begin{aligned} \Lambda(x) + \Lambda(y) &\geq \sum_i \|x_i\|^{\frac{1}{\alpha}} - \varepsilon + \Lambda(y) \geq \sum_j \|y_j\|^{\frac{1}{\alpha}} - \varepsilon \\ &\geq \Lambda\left(\sum_i x_i + \sum_j y_j\right) - 2\varepsilon = \Lambda(x+y) - 2\varepsilon \end{aligned}$$

then $\varepsilon \rightarrow 0$

This allows us to define the distance function $d(x, y) = \Lambda(x - y)$, which satisfies all requirements of a metric $\Rightarrow (V, \|\cdot\|)$ is a metric space.

Further, we have that

$$x_n \rightarrow x \text{ in } (V, d) \Leftrightarrow d(x, y) \rightarrow 0 \Leftrightarrow \|x_n - x\| \rightarrow 0$$

Thus quasi-norm and metric are compatible. □

Remark. This topology, i.e. (V, d) is the unique topology which is compatible with the quasi-norm.

To be precise, if we are given the quasi-norm $\|\cdot\|$, then the only topology should come from the local basis of the neighbourhood of 0 given by $B_r(0) = \{x : \|x\| < r\}$ ($B_r(0)$ might not be open in (V, d))

Here the open sets are defined by

$$U \subset V \text{ is open iff } \forall x \in U \exists r_x > 0 \text{ s.t. } B_{r_x}(x) = \{y : |x - y| < r_x\} \subset U$$

Exercise. The $L^{1,\infty}$ quasi norm is not equivalent to any other norm.

4.2 Normability of Lorentz spaces

Q: For which $L^{p,q}$ do we have a norm which is equivalent to the quasi norm?

Theorem 4.7. $\forall 1 < p \leq \infty$ and $1 \leq q \leq \infty$. Then \exists norm in $L^{p,q}$ that is equivalent to the quasi-norm. We will construct $\|\cdot\|$ s.t.

$$\|f\|_{L^{p,q}} \leq \|\cdot\| \leq p' \|f\|_{L^{p,q}}, \quad \frac{1}{p} + \frac{1}{p'} = 1$$

To prove this, we need a rearranged version of f ! First, let us consider the basic rearrangement inequality:

Theorem 4.8. If we have real numbers $a_1 \geq a_2 \geq \dots \geq a_n, b_1 \geq b_2 \geq \dots \geq b_n$, then:

$$\sum_{i=1}^n a_i b_i \geq \sum_{i=1}^n a_i b_{\sigma(i)} \quad \forall \sigma \in S_n$$

[Basic rearrangement inequality/Chebychev sum inequality] E.g. $a_1 b_1 + a_2 b_2 \geq a_1 b_2 + a_2 b_1 \Leftrightarrow (a_1 - a_2)(b_1 - b_2) \geq 0$

Definition 4.9. Let $f : \mathbb{R}^d \rightarrow \mathbb{C}$. We define $f^* : \mathbb{R}^d \rightarrow [0, \infty]$ s.t.

1. f^* is radially symmetric decreasing, i.e. $f^*(x)$ depends only on $|x|$ and $|x| \mapsto f^*(x)$ is decreasing
2. $|\{f^* > t\}| = |\{|f| > t\}|$, for all $t > 0$.

Equivalently, we can use the layer cake representation to define f^* .

$$\begin{aligned} |f(x)| &= \int_0^\infty \mathbb{1}_{\{|f|>t\}}(x) dt \\ &\rightsquigarrow f^* := \int_0^\infty \mathbb{1}_{\{|f|>t\}^*} dt \end{aligned}$$

Where for a set $\Omega \subset \mathbb{R}^d$, $\Omega^* = \text{ball in } \mathbb{R}^d \text{ centered at } 0$, with $|\Omega| = |\Omega^*|$

Theorem 4.10 (Hardy-Littlewood rearrangement inequality). If $f, g : \mathbb{R}^d \rightarrow \mathbb{C}$, then

$$\left| \int_{\mathbb{R}^d} f g \right| \leq \int_{\mathbb{R}^d} f^* g^*$$

Proof. Consider $f, g \geq 0$. Then by layer cake

$$\begin{aligned} \int_{\mathbb{R}^d} f(x)g(x) &= \int_{\mathbb{R}^d} \left(\int_0^\infty \mathbb{1}_{\{f>t\}}(x) dt \right) \left(\int_0^\infty \mathbb{1}_{\{g>t\}}(x) dt \right) dx \\ &= \int_{\mathbb{R}^d} \left(\int_0^\infty \mathbb{1}_{\{f>t\}}(x) dt \right) \left(\int_0^\infty \mathbb{1}_{\{g>s\}}(x) ds \right) dx \\ &\leq \int_{\mathbb{R}^d} \left(\int_0^\infty \mathbb{1}_{\{f>t\}^*}(x) dt \right) \left(\int_0^\infty \mathbb{1}_{\{g>s\}^*}(x) ds \right) dx \\ &= \int_{\mathbb{R}^d} f^*(x)g^*(x) dx \end{aligned}$$

We used: $\int_{\mathbb{R}^d} \mathbb{1}_A \mathbb{1}_B = |A \cap B| = |A^* \cap B^*| = \int_{\mathbb{R}^d} \mathbb{1}_{A^*} \mathbb{1}_{B^*}$

And $|A \cap B| \leq \min(|A|, |B|) = \min(|A^*|, |B^*|) = |A^* \cap B^*|$

□

Definition 4.11 (Decreasing rearrangement). Let $f : \mathbb{R}^d \rightarrow \mathbb{C}$. Define $f_* : [0, \infty] \rightarrow [0, \infty]$ s.t.

- f_* is decreasing
- $|\{f_* > t\}|_{\mathbb{R}_+} = |\{|f| > t\}|_{\mathbb{R}^d}$

Example 4.12. If $f(x) = \frac{1}{|x|^p}$ for $x \in \mathbb{R}^d$, then $f^*(x) = f(x) = \frac{1}{|x|^p}$ and $f_*(t) = \frac{1}{(\frac{t}{|B_1|})^{\frac{p}{d}}}$ where $t \in \mathbb{R}_+$ and $|B_1|$ = volume of unit ball in \mathbb{R}^d as:

$$|\{f > \lambda\}| = |\left\{\frac{1}{|x|^p} > \lambda\right\}| = |\{|x| < \lambda^{-\frac{1}{p}}\}|_{\mathbb{R}^d} = \lambda^{-\frac{d}{p}} |B_1|$$

$$|\{f_* > \lambda\}|_{\mathbb{R}_+} = \left|\left\{\left(\frac{|B_1|}{t}\right)^{\frac{p}{d}} > \lambda\right\}\right|_{\mathbb{R}_+} = |B_1| \lambda^{\frac{-d}{p}}$$

Theorem 4.13 (Alternative definition of $\|\cdot\|_{L^{p,q}}$).

$p, q < \infty$:

$$\begin{aligned} \|f\|_{L^{p,q}} &\stackrel{\text{def}}{=} \left(p \int_0^\infty t^{q-1} |\{|f| > t\}|^{\frac{q}{p}} dt \right)^{\frac{1}{q}} \\ &\stackrel{\text{new}}{=} \left(\int_0^\infty \lambda^{\frac{q}{p}-1} f_*(\lambda)^q d\lambda \right)^{\frac{1}{q}} \\ &= \left\| \lambda^{\frac{1}{p}} f_*(\lambda) \right\|_{L^q(\mathbb{R}_+, \frac{d\lambda}{\lambda})} \end{aligned}$$

If $q = \infty$:

$$\begin{aligned} \|f\|_{L^{p,\infty}} &\stackrel{\text{def}}{=} \sup_{t>0} t |\{|f| > t\}|^{\frac{1}{p}} \\ &\stackrel{\text{new}}{=} \left\| \lambda^{\frac{1}{p}} f_*(\lambda) \right\|_{L^\infty(\mathbb{R}_+)} \end{aligned}$$

Proof. Case $q < \infty$:

From the def:

$$\begin{aligned} \|f\|_{L^{p,q}}^q &= p \int_0^\infty t^{q-1} |\{|f| > t\}|^{\frac{q}{p}} dt \\ &= p \int_0^\infty t^{q-1} \left(\frac{q}{q} \int_0^\infty \lambda^{\frac{q}{p}-1} \mathbb{1}_{\{f_*(\lambda) > t\}} d\lambda \right) dt \end{aligned}$$

$f_*(\lambda) > t \Leftrightarrow |\{|f| > t\}| > \lambda$ Duality formula (exercise)

$$\begin{aligned} &\stackrel{\text{Fubini}}{=} \int_0^\infty \left(q \int_0^\infty t^{q-1} \mathbb{1}_{\{f_*(\lambda) > t\}} dt \right) \lambda^{\frac{q}{p}-1} d\lambda \\ &= \int_0^\infty f_*(\lambda^{\frac{q}{p}-1} d\lambda) \end{aligned}$$

case $q = \infty$ (exercise) □

Lemma 4.14.

$$\int_0^t f_*(s)ds = \sup\left\{\int_E |f| : |E| \leq t\right\}$$

Proof. Consider step functions $f(x) = \sum_{i=1}^n a_i \mathbb{1}_{A_i}(x)$ with $a_1 > a_2 > \dots > a_n$ and $\{A_i\}$ disjoint.

- case $t < \sum_{i=1}^n |A_i|$ and we stop at some $k < n$ s.t.

$$\sup_{|E| \leq t} \int_E |f| = \sum_{i=1}^k +a_{k+1}(t - \sum_{i=1}^k |A_i|)$$

- else: $k = n$ and $a_{k+1} = 0$

(bathtub principle)

$$\begin{aligned} f_*(x) &= \sum_{i=1}^n a_i \mathbb{1}_{[B_{i-1}, B_i]}(s) \text{ where } B_i = \sum_{j=1}^i |A_j| \\ \Rightarrow \int_0^t f(s)ds &= \sum_{i=1}^k a_i(B_i - B_{i-1}) + a_{k+1}(t - B_k) \end{aligned}$$

Still we need to pass the limit for the general case $f \in L^{p,q}$. In the general case, we can choose a sequence of step functions $\{f_n\}$ s.t. $f_n \uparrow f$ point wise a.e.

$$\Rightarrow (f_n)_*(t) \uparrow f_*(t) \quad \forall t \quad (\text{exercise})$$

$$\Rightarrow \int_0^t (f_n)_*(s)ds \rightarrow \int_0^t f_*(s)ds \text{ by monotone convergence}$$

Moreover: $\sup_{|E| \leq t} \int_E |f| \uparrow \sup_{|E| \leq t} \int_E |f|$ by monotone convergence □

Lemma 4.15 (Hardy inequality).

If $g \in C^1(\mathbb{R}_+)$, $g(0) = 0$, then :

$$p' \|g\|_{L^p(\mathbb{R}_+)} \geq \left\| \frac{g}{|x|} \right\|_{L^p(\mathbb{R}_+)}$$

More generally:

$$p' \left(\int_0^\infty x^{\frac{q}{p}-1} |g'(x)|^q \right)^{\frac{1}{q}} \geq \left(\int_0^\infty x^{\frac{q}{p}-1} \left| \frac{g(x)}{x} \right|^q dx \right)^{\frac{1}{q}}$$

Proof. Duality $(L^q(X))^* = L^{q'}(X)$ i.e. $\|G\|_{L^q(X)} = \sup_{\|\varphi\|_{L^{q'}(X)}=1} |\int G\varphi|$, $X = \mathbb{R}_+$, $d\mu(t) =$

$$t^{\frac{q}{p}-1}dt$$

$$\begin{aligned}
 \left(\int_0^\infty t^{\frac{q}{p}-1} \left| \frac{1}{t} \int h(s) ds \right|^q dt \right)^{\frac{1}{q}} &= \left\| \frac{1}{t} \int_0^t h(s) ds \right\|_{L^q(X, d\mu)} = \sup_{\varphi_{L^{q'}}=1} \left| \int_0^\infty \frac{1}{t} \int_0^t h(s) ds \varphi(t) dt \right| \\
 &= \left| \int_0^\infty \frac{1}{t} \int_0^t h(s) ds \varphi(t) d\mu(t) \right| \\
 &= \left| \int_0^\infty \int_0^1 h(t\xi) d\xi \varphi(t) d\mu(t) \right| \\
 &\leq \int_0^1 \|h(\xi \cdot)\|_{L^q(X, d\mu)} \|\varphi\|_{L^{q'}(X, d\mu)} d\xi \\
 &\leq \int_0^1 \|h\|_{L^q(X)} \xi^{\frac{-1}{p}} d\xi \\
 &= \|h\|_{L^q(X)} \left[\frac{\xi^{1-\frac{1}{p}}}{1-\frac{1}{p}} \right]_0^1 \\
 &= p' \|h\|_{L^q(X)}
 \end{aligned}$$

□

proof of the normality of $L^{p,q}$ for $p > 1$.

Define $f_{**}(t) = \frac{1}{t} \int_0^t f_*(s) ds$ and

$$\|f\| = \begin{cases} \left(\int_0^\infty \lambda^{\frac{q}{p}-1} f_{**}(\lambda)^q d\lambda \right)^{\frac{1}{q}} & \text{if } q < \infty \\ \left\| \lambda^{\frac{1}{p}} f_{**}(\lambda) \right\|_{L^\infty} & \text{if } q = \infty \end{cases}$$

Then we claim that $\|f\|$ is a norm in $L^{p,q}$ and

$$\|f\|_{L^{p,q}} \leq ||| \cdot ||| \leq p' \|f\|_{L^{p,q}}$$

Step 1: (triangle inequality)

$$(f + g)_{**}(\lambda) \leq f_{**}(\lambda) + g_{**}(\lambda), \quad \forall \lambda > 0$$

Remark: We only have $(f + g)_*(t) \leq f_*(\frac{t}{2}) + g_*(\frac{t}{2})$ (exercise)

From lemma 4.14:

$$\begin{aligned}
 (f + g)_{**} &= \frac{1}{t} \int_0^t (f + g)(s) ds = \sup \left\{ \int_E |f + g| : |E| \leq t \right\} \\
 &\leq \sup \left\{ \int_E |f| + \int_E |g| : \dots \right\} \\
 &\leq \sup \left\{ \int_E |f| : |E| \leq t \right\} + \sup \left\{ \int_E |g| : |E| \leq t \right\}
 \end{aligned}$$

As a consequence:

$$\begin{aligned}
|||f + g||| &= \left(\int_0^\infty \lambda^{\frac{q}{p}-1} (f + g)_{**}(\lambda)^q d\lambda \right)^{\frac{1}{q}} \\
&= \left\| \lambda^{\frac{1}{p}} (f + g)_{**}(\lambda) \right\|_{L^q(\mathbb{R}_+, \frac{d\lambda}{\lambda})} \\
&\leq \left\| \lambda^{\frac{1}{p}} (f_{**}(\lambda) + g_{**}(\lambda)) \right\|_{L^q(\mathbb{R}_+, \frac{d\lambda}{\lambda})} \\
&\leq \left\| \lambda^{\frac{1}{p}} f_{**}(\lambda) \right\|_{L^q(\mathbb{R}_+, \frac{d\lambda}{\lambda})} + \left\| \lambda^{\frac{1}{p}} g_{**}(\lambda) \right\|_{L^q(\mathbb{R}_+, \frac{d\lambda}{\lambda})}
\end{aligned}$$

Step 2: We still have to prove that

$$\|f\|_{L^{p,q}} \leq |||f||| \leq p' \|f\|_{L^{p,q}}$$

It follows directly from the definition that $|||f||| \geq \|f\|_{L^{p,q}}$.

The other side follows from Hardy's inequality:

$$\begin{aligned}
g(x) &= \int_0^x h(t) dt, \quad g(0) = 0, \quad g'(x) = h(x) \\
\Rightarrow p' \left(\int_0^\infty t^{\frac{p}{q}-1} |h(t)|^q \right)^{\frac{1}{q}} &\geq \left(\int_0^\infty t^{\frac{p}{q}-1} \left| \frac{1}{t} \int_0^t h(s) ds \right|^q \right)^{\frac{1}{q}} \\
&\stackrel{h \rightsquigarrow f_*}{\Rightarrow} p' \|f\|_{L^{p,q}} \geq |||f|||
\end{aligned}$$

□

4.3 Some functional inequalities for Lorentz norms

Theorem 4.16 (Monotonicity). If $q < r$, then $\exists C > 0$ s.t. $\forall f \in L^{p,q}$:

$$C \|f\|_{L^{p,q}} \geq \|f\|_{L^{p,r}}$$

Remark. If $|\Omega| < \infty$, then $L^{p_1} \subset L^{p_2}$ if $p_1 > p_2$. We get the same for all q, r with $L^{p_1,q} \subset L^{p_2,r}$ (exercise)

Proof.

$r = \infty$:

$$\begin{aligned}
\|f\|_{L^{p,q}} &= \left(\int_0^\infty r^{\frac{q}{p}-1} f_*(t)^q dt \right)^{\frac{1}{q}} \\
&\geq \left(\int_0^\lambda r^{\frac{q}{p}-1} \underbrace{f_*(t)}_{\geq f_*(\lambda)}^q dt \right)^{\frac{1}{q}} \\
&\geq \left(\int_0^\lambda t^{\frac{q}{p}-1} dt \right)^{\frac{1}{q}} f_*(\lambda) \\
&= \left(\frac{p}{q} \lambda^{\frac{q}{p}} \right)^{\frac{1}{q}} f_*(\lambda) \sim \lambda^{\frac{1}{p}} f_*(\lambda), \quad \forall \lambda > 0 \\
\Rightarrow \|f\|_{L^{p,q}} &\geq \sup_{\lambda > 0} \lambda^{\frac{1}{p}} f_*(\lambda) = \|f\|_{L^{p,\infty}}
\end{aligned}$$

$r < \infty$:

$$\begin{aligned}
\|f\|_{L^{p,r}} &= \left(\int_0^\infty r^{\frac{r}{p}-1} f_*(t)^r dt \right)^{\frac{1}{r}} \\
&\leq \left(\int_0^\infty t^{\frac{q}{p}-1} f_*^q(t) dt \left(\sup_{\lambda > 0} \lambda^{\frac{r-q}{p}} f_*^{r-q}(\lambda) \right) \right)^{\frac{1}{r}} \\
&= (\|f\|_{L^{p,q}}^q \|f\|_{L^{p,\infty}}^{r-q})^{\frac{1}{r}} \lesssim \|f\|_{L^{p,q}}
\end{aligned}$$

(We used $\|f\|_{L^{p,\infty}} \lesssim \|f\|_{L^{p,q}}$ version $r = \infty$)

□

Theorem 4.17 (Hölder). If $1 < p_1, p_2, p < \infty$, $1 \leq q_1, q_2, q \leq \infty$ s.t.

$$\frac{1}{p_1} + \frac{1}{p_2} = 1, \quad \frac{1}{q_1} + \frac{1}{q_2} = \frac{1}{q}$$

then:

$$\|fg\|_{L^{p,q}} \lesssim \|f\|_{L^{p_1,q_1}} \|g\|_{L^{p_2,q_2}}$$

Remark. 1. We can replace $\frac{1}{q_1} + \frac{1}{q_2} = \frac{1}{q}$ by $\frac{1}{q_1} + \frac{1}{q_2} \geq \frac{1}{q}$

$$\begin{aligned}
2. p = q = 1 \Rightarrow \|fg\|_{L^1} &\lesssim \|f\|_{L^{p_1,q_1}} \|g\|_{L^{p_2,q_2}} \\
\Rightarrow (L^{p_1,q_1})^* &= L^{p_2,q_2}, \quad \forall 1 < p_1 < \infty, 1 \leq q_1 < \infty
\end{aligned}$$

Proof. Claim: $(fg)_*(t_1 + t_2) \leq f_*(t_1)g_*(t_2)$. To see this, we use the density relation:

$$\begin{aligned}
|\{|f| > \lambda\}| > t &\iff f_*(t) > \lambda \\
\rightsquigarrow |\{|f| > \lambda\}| \leq t &\iff f_*(t) \leq \lambda \\
\rightsquigarrow |\{|f| > f_*(t)\}| \leq t, \quad \forall t > 0
\end{aligned}$$

Consequentially:

$$\begin{aligned} |\{|fg| > f(t_1)g(t_2)\}| &\leq |\{|f| > f_*(t_1)\}| + |\{|g| > g_*(t_2)\}| \leq t_1 + t_2 \\ \Rightarrow (fg)_*(t_1 + t_2) &\leq f_*(t_1)g_*(t_2) \end{aligned}$$

To conclude:

$$\begin{aligned} \|fg\|_{L^{p,q}} &= \left(\int_0^\infty t^{\frac{q}{p}-1} (fg)_*^q \left(\underbrace{t}_{=\frac{t}{2}+\frac{t}{2}} \right) dt \right)^{\frac{1}{q}} \\ &\leq \left(\int_0^\infty t^{\frac{q}{p}-1} f_*^q \left(\frac{t}{2} \right) g_*^q \left(\frac{t}{2} \right) dt \right)^{\frac{1}{q}} \\ &\leq \left(\left(\int_0^\infty t^{\frac{q_1}{p_1}-1} f_*^{q_1} \left(\frac{t}{2} \right) dt \right)^{\frac{q}{q_1}} \left(\int_0^\infty t^{\frac{q_2}{p_2}-1} g_*^{q_2} \left(\frac{t}{2} \right) dt \right)^{\frac{q}{q_2}} \right)^{\frac{1}{q}} \\ &\stackrel{\frac{t}{2} \rightsquigarrow t}{\leq} \|f\|_{L^{p_1, q_1}} \|g\|_{L^{p_2, q_2}} \end{aligned}$$

□

Theorem 4.18 (Improved Sobolev inequality).

If $1 \leq p < d$, $p^* = \frac{dp}{d-p}$, then

$$\|f\|_{L^{p^*, p}} \lesssim \|\nabla f\|_{L^p}$$

Remark. Since $p < p^*$, this is an improvement of the standard Sobolev inequality. This can be proven in two ways. 1) Weak-Young inequality 2) Dyadic decomposition. We will have a look at 1) first. The proof of the improved Sobolev inequality with weak-young is motivated by the proof of the standard Sobolev inequality with weak young.

Weak-Young: $\|g * h\|_{L^p} \lesssim \|g\|_{L^{q,\infty}} \|h\|_{L^r}$ if $\frac{1}{q} + \frac{1}{r} = 1 + \frac{1}{p}$ for $1 < p, q, r < \infty$

proof of standard Sobolev inequality by weak-young. Note that the equation $-\Delta f = g$ in \mathbb{R}^d can be written as

$$f(x) = (-\Delta)^{-1}g(x) = (G(x) * g)(x) = \int_{\mathbb{R}^d} G(x-y)g(y)dy$$

and the Green function

$$\begin{cases} \text{const.} \ln(x), \text{ if } d = 2 \\ \text{const.} \frac{1}{|x|^{d-2}}, \text{ if } d \geq 3 \end{cases}$$

(Formally: $(-\Delta)^{-1}g(k) = \frac{1}{|2\pi k|^2} \hat{g}(k) \stackrel{(?)}{=} \widehat{G * g}(k) = \hat{G}(k)\hat{g}(k) \Rightarrow \hat{G}(g) = \frac{1}{|2\pi k|^2} \Rightarrow G(x)$ above) We will discuss the Fourier transform later to make sense of all of this.

Put differently, $f = G \star (-\Delta f) = (-\Delta)(G \star f) = (-\nabla G) \star (\nabla f)$ i.e.

$$f(x) = - \sum_{i=1}^d (\partial_{x_i} G \star \partial_{x_i} f)(x)$$

if f regular enough! (e.g. $f \in C_c^\infty$)

Hence:

$$\|f\|_{L^{p^*}(\mathbb{R}^d)} = \|(-\nabla G) \star (\nabla f)\|_{L^{p^*}(\mathbb{R}^d)} \stackrel{\text{weak-young}}{\lesssim} \|\nabla G\|_{L^{q,\infty}} \|\nabla f\|_{L^p}$$

provided that: $\frac{1}{q} + \frac{1}{p} = 1 + \frac{1}{p^*}$

Here $|\nabla G(x)| = \text{const.} \frac{1}{|x|^{d-1}}$ for $d \geq 2$

$$\Rightarrow \|\nabla G\|_{L^{1,\infty}} < \infty \text{ with } q = \frac{d}{d-1}$$

$$\Leftrightarrow \frac{1}{p^*} = \frac{1}{q} + \frac{1}{p} - 1 = \frac{1}{\frac{d}{d-1}} + \frac{1}{p} - 1 = \frac{d-1}{d} + \frac{1}{p} - 1 = \frac{1}{p} - \frac{1}{d}$$

$$\Leftrightarrow p^* = \frac{dp}{d-p}$$

□

Inspired by this, we want to prove a Young inequality for Lorentz spaces s.t. we can conclude the improved Sobolev inequality. For that we first need the following lemma.

Lemma 4.19. If $f, g : \mathbb{R}^d \rightarrow [0, \infty]$, then:

$$(f \star g)_{**}(t) \leq t f_{**}(t) g_{**}(t) + \int_0^\infty f_*(s) g_*(s) ds \quad \forall t > 0$$

Proof. Step 1: $f = \mathbb{1}_A$, $A \subset \mathbb{R}^d$, $|A| < \infty$.

Easy to check:

$$\begin{aligned} f_*(s) &= \mathbb{1}(s \leq |A|) = \mathbb{1}_{(0,|A|]}(s) \\ f_{**} &= \frac{1}{t} \int_0^t f_*(s) ds = \frac{1}{t} \int_0^t \mathbb{1}_{\{s \leq |A|\}} ds = \frac{1}{t} \min(|A|, t) \end{aligned}$$

RHS:

$$t f_{**}(t) g_{**}(t) + \int_t^\infty f_*(s) g_*(s) ds = \min(|A|, t) g_{**}(t) + \int_t^\infty \mathbb{1}_{\{s \leq |A|\}} g_{**}(s) ds$$

$$\begin{cases} \stackrel{|A| \geq t}{=} \int_0^t g_*(s) ds + \int_t^{|A|} g_*(s) ds = \int_0^{|A|} g_*(s) ds = |A| g_{**}(|A|) = |A| g_{**}(\max(t, |A|)) \\ \stackrel{|A| < t}{=} |A| g_{**}(t) + 0 = |A| \min(g_{**}(t), g_{**}(|A|)) \end{cases}$$

Why:

$$(\mathbb{1}_A \star g)_{**}(t) \leq |A| \min(g_{**}(t), g_{**}(|A|))$$

Recall:

$$h_{**}(t) = \frac{1}{t} \sup \left\{ \int_B |h| : \forall B \subset \mathbb{R}^d \text{ s.t. } |B| = t \right\}$$

(we used this in the proof of theorem 4.7)

Take any $B \subset \mathbb{R}^d$, $|B| = t$. Then:

$$\begin{aligned}\frac{1}{t} \int_B |\mathbb{1}_A \star g| &\leq \frac{1}{t} \int_B \left(\int_A |g(x-y)| dy \right) dx \\ &= \frac{1}{t} \int_B \left(\int_{\substack{X-A \\ |X-A|=A}} |g(y)| dy \right) dx \\ &\leq \frac{1}{t} \int_B (|A| g_{**}(|A|)) dx \\ &= |A| g_{**}(|A|)\end{aligned}$$

and

$$\begin{aligned}\frac{1}{t} \int_B |\mathbb{1}_A \star g| &\stackrel{Fubini}{\leq} \frac{1}{t} \int_A \left(\int_B |g(x-y)| dx \right) dy = \frac{1}{t} \int_A \left(\int_{\substack{B-y \\ |B-y|=|B|=t}} |g(x)| dx \right) dy \leq \frac{1}{t} \int_A (tg_{**}(t)) dy = |A| g_{**}(t) \\ &\Rightarrow \frac{1}{t} \int_B |\mathbb{1}_A \star g| \leq |A| \min(g_{**}(|A|), g_{**}(t)) \Rightarrow \text{optimizing over } B\end{aligned}$$

Step 2: f as step function:

$$f(x) = \sum_{i=1}^N \alpha_i \mathbb{1}_{A_i}(x), \quad \forall x \in \mathbb{R}^d$$

with $A_1 \supsetneq A_2 \supsetneq \dots \supsetneq A_N$

Why can we write it like that?

Assume $f = \alpha_1 \mathbb{1}_{B_1} + \alpha_2 \mathbb{1}_{B_2}(x)$, $B_1 \cap B_2 =$

In this case, we can write $f(x) = \alpha_1 B_1 \cup B_2(x) + (\alpha_2 - \alpha_1) \mathbb{1}_{B_2}(x)$ with $A_1 = B_1 \cup B_2$ and $A_1 = B_2$

Claim: If $f = \sum_{i=1}^N \alpha_i \mathbb{1}_{A_i}$, $\alpha_i \geq 0$, $\mathbb{R}^d = A_0 \supsetneq A_1 \supsetneq A_2 \supsetneq \dots \supsetneq A_N \supsetneq A_{N+1} =$

Then:

$$f_*(s) = \sum_{i \leq j} \alpha_i \mathbb{1}_{(|A_{j+1}|, |A_j|)}(s), \quad s > 0 \quad (\text{exercise})$$

Come back to the lemma with this f:

$$(f + g)_{**}(t) = \left(\sum_{i=1}^N \alpha_i (\mathbb{1}_{A_i} \star g) \right)_{**}(t) = \sum_{i=1}^N \alpha_i (\mathbb{1}_{A_i} \star g)_{**}(t) = \sum_{i=1}^N \alpha_i |A_i| g(\max(t, |A_i|))$$

Assume $\exists i_0$ s.t. $|A_{i_0}| = t$ (we can take α_{i_0} if we want)

$$\Rightarrow (f \star g)_{**}(t) \leq \underbrace{\sum_{i < i_0} \alpha_i |A_i| g(|A_i|)}_{(I)} + \underbrace{\sum_{i \geq i_0} \alpha_i |A_i| g(|A_i|)}_{(II)}$$

$$\begin{aligned}
 (I) &= \sum_{i \leq i_0} \alpha_i \int_0^{|A_i|} g_*(s) ds = \sum_{i < i_0} \alpha_i \sum_{j \geq i} \int_{|A_{j+1}|}^{|A_j|} g_*(s) ds \\
 &= \underbrace{\sum_{i < i_0 \leq j} \alpha_i \int_{|A_{j+1}|}^{|A_j|} g_*(s) ds}_{(I')} + \underbrace{\sum_{i \leq j < i_0} \alpha_i \int_{|A_{j+1}|}^{|A_j|} g_*(s) ds}_{(I'')}
 \end{aligned}$$

$$\begin{aligned}
 (I') &= \sum_{i < i_0} \sum_{j \geq i_0} \alpha_i \int_{|A_{j+1}|}^{|A_j|} g_*(s) ds = \sum_{i < i_0} \sum_{j \geq i_0} \sum_{j \geq i} \int_0^\infty \alpha_i \mathbb{1}_{[|A_{j+1}|, |A_j|]} g_*(s) ds \\
 &= \sum_{i < i_0} \int_0^\infty \alpha_i \mathbb{1}_{[0, t]} g_*(s) ds \\
 &= \sum_{i < i_0} \alpha_i \int_0^t g_*(s) ds \\
 &= \sum_{i < i_0} \alpha_i t g_{**}(t)
 \end{aligned}$$

$$\begin{aligned}
 (I'') &= \sum_{i \leq j < i_0} \alpha_i \int_{|A_{j+1}|}^{|A_j|} g_*(s) ds = \sum_{j < i_0} j < i_0 \int_{|A_{j+1}|}^{|A_j|} (\sum_{i \leq j} \alpha_i) g_*(s) ds \\
 &= \sum_{j < i_0} \int_{|A_{j+1}|}^{|A_j|} f_*(s) g_*(s) ds \\
 &= \int_t^\infty f_*(s) g_*(s)
 \end{aligned}$$

$$\Rightarrow (I') + (II) = (\sum_{i < i_0} + \sum_{i \geq i_0}) \alpha_i \min(|A_i|, t) g_{**}(t) = \sum_{i=1}^N \alpha_i \min(|A_i|, t) g_{**}(t) = t f_{**}(t) g_{**}(t)$$

□

Theorem 4.20 (Young inequality for Lorentz Spaces).

$$\|g \star h\|_{L^{p,q}} \lesssim \|g\|_{L^{p_1, q_1}} \|h\|_{L^{p_2, q_2}}$$

with

$$\begin{cases} \frac{1}{p_1} + \frac{1}{p_2} = 1 + \frac{1}{p} & 1 < p_1, p_2, p < \infty \\ \frac{1}{q_1} + \frac{1}{q_2} = \frac{1}{q} & 1 < q_1, q_2, q < \infty \end{cases}$$

Remark. 1. The previous weak Young inequality corresponds to the case $p_1 = q$, $q_1 = \infty$, $p_2 = q_2 = r$

$$\begin{aligned}
 & \Rightarrow \frac{1}{p_1} + \frac{1}{p_2} = \frac{1}{q} + \frac{1}{r} = 1 + \frac{1}{p} \text{ and} \\
 & \frac{1}{q} = \frac{1}{q_1} + \frac{1}{q_2} = \frac{1}{\infty} + \frac{1}{r} = \frac{1}{r} \\
 & \Rightarrow q = r \xrightarrow{Y-Lorentz} \|g\|_{L^{q,\infty}} \|h\|_{L^{r,r}} \gtrsim \|g \star h\|_{L^{p,p}} \gtrsim \|g \star h\|_{L^{p,p}} \text{ since } r \leq p \Leftrightarrow \frac{1}{r} > \frac{1}{p} \\
 & \text{since } \underbrace{\frac{1}{q} + \frac{1}{r}}_{<1} = 1 + \frac{1}{p}
 \end{aligned}$$

2. Stronger result: $\|g\|_{L^{p_1}} \|h\|_{L^{p_2}} \gtrsim \|g \star h\|_{L^{p,q}}$ where
 $\frac{1}{p_1} + \frac{1}{p_2} = 1 + \frac{1}{p}$, $\frac{1}{q} \frac{1}{p_1} + \frac{1}{p_2} = 1 + \frac{1}{p}$
 $\Rightarrow q < p \Rightarrow \|g \star h\|_{L^{p,q}} \gtrsim \|g \star h\|_{L^p}$
3. The improved Sobolev inequality follows easily from the Y-Lorentz spaces:

$$\|f\|_{L^{p^*,p}} = \|\nabla G \star \nabla f\|_{L^{p^*,p}} \lesssim \|\nabla G\|_{L^{\frac{d}{d-1}-\infty}} \|\nabla f\|_{L^{p,p}} \lesssim \|\nabla f\|_{L^p}$$

$$\begin{aligned}
 & \text{i.e. } p_1 = \frac{d}{d-1}, \quad q_1 = \infty, \quad p_2 = q_2 = p \\
 & \Rightarrow \frac{1}{p^*} + 1 = \frac{1}{p_1} + \frac{1}{p_2}, \quad \frac{1}{p} = \frac{1}{q_1} + \frac{1}{q_2}
 \end{aligned}$$

proof of the Y-Lorentz inequality (by Niel). Recall

$$\|f\|_{L^{p,q}} = \left(\int_0^\infty \lambda^{\frac{q}{p}-1} f_*(\lambda)^q d\lambda \right)^{\frac{1}{q}} \sim \left(\int_0^\infty \lambda^{\frac{q}{p}-1} f_{**}(\lambda)^q d\lambda \right)^{\frac{1}{q}}$$

with f_* decreasing rearrangement, $f_{**}(t) = \frac{1}{t} \int_0^t f_*(s) ds \geq f_*(t)$

$$\begin{aligned}
 \|g \star h\|_{L^{p,q}} & \sim \left(\int_0^\infty t^{\frac{q}{p}-1} (g \star h)_{**}^q(t) dt \right)^{\frac{1}{q}} \\
 & = \left\| t^{\frac{1}{p}-\frac{1}{q}} (g \star h)(t) \right\|_{L^1(\mathbb{R}_+)} \\
 & \stackrel{\text{lemma 4.19}}{\leq} \left\| t^{\frac{1}{p}-\frac{1}{q}+1} g_{**}(t) h_{**}(t) + t^{\frac{1}{p}+\frac{1}{q}} \int_t^\infty g_*(s) h_*(s) ds \right\|_{L^q(\mathbb{R}_+, dt)} \\
 & \leq \left\| t^{\frac{1}{p}-\frac{1}{q}+1} g_{**}(t) h_{**}(t) \right\| + \left\| t^{\frac{1}{p}+\frac{1}{q}} \int_t^\infty g_*(s) h_*(s) ds \right\|_{L^q(\mathbb{R}_+, dt)} = (I) + (II)
 \end{aligned}$$

$$\begin{aligned}
 (I) & = \left\| t^{\frac{1}{p}-\frac{1}{q}+1} g_{**}(t) h_{**}(t) \right\|_{L^q(\mathbb{R}_+)} \stackrel{\text{Hölder}}{\leq} \left\| t^{\frac{1}{p_1}-\frac{1}{q_1}} g_{**}(t) \right\|_{L^{q_1}(\mathbb{R}_+)} \left\| t^{\frac{1}{p_2}-\frac{1}{q_2}} h_{**}(t) \right\|_{L^{q_2}(\mathbb{R}_+)} \\
 & \leq \|g\|_{L^{p_1,q_1}} \|h\|_{L^{p_2,q_2}}
 \end{aligned}$$

Here for Hölder:

$$\begin{aligned}
 & \left(\int_0^\infty t^{\frac{q}{p}-1+q} g_{**}^q(t) h_{**}^q(t) dt \right) \leq \left(\int_0^\infty t^{\frac{q_1}{p_1}-1} g_{**}^{q_1}(t) dt \right)^{\frac{q}{q_1}} \left(\int_0^\infty t^{\frac{q_2}{p_2}-1} h_{**}^{q_2}(t) dt \right)^{\frac{q}{q_2}} \\
 & RHS \geq \left(\int_0^\infty t^{\frac{q_1}{p_1}-1} g_{**}^{q_1}(t) dt \right)^{\frac{q}{q_1}} \left(\int_0^\infty t^{\frac{q_2}{p_2}-1} h_{**}^{q_2}(t) dt \right)^{\frac{q}{q_2}} = \int_0^\infty t^{\frac{q}{p_1}-\frac{q}{q_1}+\frac{q_2}{p_2}-\frac{q}{q_2}} g_{**}^q(t) h_{**}^q(t) dt
 \end{aligned}$$

So we get

$$\frac{q}{p} - 1 + q = \frac{q}{p_1} - \frac{q}{q_1} + \frac{q_2}{p_2} - \frac{q}{q_2} = q\left(1 + \frac{1}{p}\right) - 1$$

(II)

$$\begin{aligned}
& \left\| t^{\frac{1}{q}-1} \int_t^\infty g_*(s) h_*(s) ds \right\|_{L^q(\mathbb{R}_+)} = \left(\int_0^\infty t^{\frac{q}{p}-1} \left(\int_t^\infty g_*(s) h_*(s) ds \right)^q dt \right)^{\frac{1}{q}} \\
& \stackrel{t=\frac{1}{u}, s=\frac{1}{v}}{=} \left(\int_0^\infty \left(\frac{1}{u} \right)^{\frac{q}{p}-1} \left(\int_0^u g_*(\frac{1}{v}) h_*(\frac{1}{v}) \frac{dv}{v^2} \right)^q \frac{du}{u^2} \right)^{\frac{1}{q}} \\
& = \left(\int_0^\infty \left(\frac{1}{u} \right)^{\frac{q}{p}+1} \underbrace{\left(\int_0^u g_*(\frac{1}{v}) h_*(\frac{1}{v}) \frac{dv}{v^2} \right)^q du}_{f(u):=} \right)^{\frac{1}{q}} \\
& = \left(\int_0^\infty \frac{|f(u)|^q}{u^{\frac{q}{p}+1}} du \right)^{\frac{1}{q}} \\
& = \left(\int_0^\infty u^{\frac{q}{p'}-1} \left| \frac{f(u)}{u} \right|^q du \right)^{\frac{1}{q}} \\
& \stackrel{Hardy}{\lesssim} \left(\int_0^\infty u^{\frac{q}{p'}-1} |f'(u)|^q du \right)^{\frac{1}{q}} \\
& = \left(\int_0^\infty u^{\frac{q}{p'}-1} \left| g_*(\frac{1}{u}) h_*(\frac{1}{u}) \frac{1}{u^2} \right|^q du \right)^{\frac{1}{q}} \\
& = \left(\int_0^\infty \frac{1}{u^{\frac{q}{p}+1+q}} |g_*(\frac{1}{u}) h_*(\frac{1}{u})|^q du \right)^{\frac{1}{q}} \\
& \stackrel{t=\frac{1}{u}}{=} \left(\int_0^\infty t^{\frac{q}{p}+1+q} |g_*(t) h_*(t)|^q \frac{dt}{t^2} \right)^{\frac{1}{q}} \\
& = \left(\int_0^\infty t^{\frac{q}{p}-1+q} |g_*(t) h_*(t)|^q dt \right)^{\frac{1}{q}} \\
& = \left\| t^{\frac{1}{p}+1-\frac{1}{q}} g_*(t) h_*(t) \right\|_{L^q(\mathbb{R}_+, dt)} \\
& \leq \|g\|_{L^{p_1, q_1}} \|h\|_{L^{p_2, q_2}}
\end{aligned}$$

The last inequality works the same as in (I) except that we use $g_* h_*$ instead of $g_{**} h_{**}$

□

4.4 Dyadic decomposition

We will have a look at another proof of the improved Sobolev inequality. For this we will use the dyadic decomposition. The basic idea comes from deviding the positive real

number line in a smart, disjoint way, i.e. $(0, \infty) = \bigcup_{j \in \mathbb{Z}} [2^j, 2^{j+1})$. We can use this for the values of functions.

Example 4.21. $f \in L^p$, $\|f\|_{L^p}^p = p \int \lambda^{p-1} |\{|f| > \lambda\}| d\lambda$
 We can decompose $f = \sum_i \underbrace{f \mathbf{1}_{\{|f| \in [2^j, 2^{j+1})\}}}_{f_i} = \sum_i f_i \Rightarrow$ all $\{f_i\}$ disjoint support and
 $\text{supp } f_i \subset \{|f| \in [2^j, 2^{j+1})\}$

$$\Rightarrow \|f\|_{L^p}^p = \sum_i \|f_i\|_{L^p}^p \text{ as } |\sum_i f_i|^p = \sum_i |f_i|^p$$

We will extend this to a version with smooth cut-off s.t. f smooth gives us a set of $\{f_i\}$ smooth functions.

Lemma 4.22. $\exists \varphi \in C^\infty(\mathbb{R}, \mathbb{R})$ s.t. $\varphi(x) = 0$ if $|x| \notin [\frac{1}{2}, 1]$ s.t.

$$\sum_{j \in \mathbb{Z}} \varphi(2^j t) = 1, \quad \forall t \neq 0$$

Proof. $\varphi(t) = \Psi(t) - \Psi(2t)$

$$\Rightarrow \sum_{j \in \mathbb{Z}} \varphi(2^j t) = \sum_j (\Psi(2^j t) - \Psi(2^{j+1} t)) = \Psi(0) = 1$$

if $\Psi \in C^\infty$ s.t. $\Psi(x) = 1$ if $|x| \leq \frac{1}{2}$, $\Psi(x) = 0$ if $|x| \geq 1$ □

Lemma 4.23. For any f , we decompose

$$f = \sum_{j \in \mathbb{Z}} f_j \text{ where } f_j(x) = f(x) \varphi(2^{-j} |f|)$$

In particular, we have $\text{supp } f_j \subset \{2^{j-1} \leq |f| \leq 2^j\}$.

Then:

$$\|f\|_{L^{p,q}} \lesssim \left(\sum_j \underbrace{\|f_j\|_{L^p}^q}_{\|\|f_j\|_{L^p}\|_{l^q(\mathbb{Z})}} \right)^{\frac{1}{q}} \lesssim \|f\|_{L^{p,q}} \quad \forall q \leq p$$

Remark.

$$\left(\sum_{j \in \mathbb{Z}} \|f_j + g_j\|_{L^p}^q\right)^{\frac{1}{q}} \leq \left(\sum_{j \in \mathbb{Z}} \|f_j\|_{L^p}^q\right)^{\frac{1}{q}} \left(\sum_{j \in \mathbb{Z}} \|g_j\|_{L^p}^q\right)^{\frac{1}{q}}$$

but $f \mapsto \left(\sum_{j \in \mathbb{Z}} \|f_j\|_{L^p}^q\right)^{\frac{1}{q}}$ is not a norm, since it is non linear.

Proof. Note that

$$f_j \sim 2^j \mathbb{1}_{\{|f| \in [2^{j-1}, 2^j]\}}$$

Thus:

$$\begin{aligned} \sum_{j \in \mathbb{Z}} \|f_j\|_{L^p}^q &\sim \sum_{j \in \mathbb{Z}} 2^{jq} |\{|f| \in [2^{j-1}, 2^j]\}|^{\frac{q}{p}} \leq \sum_j 2^{jq} |\{|f| > 2^{j-1}\}|^{\frac{q}{p}} \\ &\leq \int_0^\infty \lambda^{q-1} |\{|f| > \lambda\}|^{\frac{q}{p}} d\lambda = \|f\|_{L^{p,q}}^q \end{aligned}$$

On the other hand:

$$\begin{aligned} \|f\|_{L^{p,q}}^q &= \int_0^\infty \lambda^{q-1} |\{|f| > \lambda\}|^{\frac{q}{p}} d\lambda = \sum_j \int_{2^j}^{2^{j+1}} \underbrace{\lambda^{q-1}}_{\lambda \sim 2^j, \leq |\{|f| \geq 2^j\}|} \underbrace{|\{|f| > \lambda\}|}_{\lambda \sim 2^j, \leq |\{|f| \geq 2^j\}|}^{\frac{q}{p}} d\lambda \\ &\leq \sum_j \int_{2^j}^{2^{j+1}} \lambda^{q-1} \left(\sum_{k \geq j} |\{2^k < |f| < 2^{k+1}\}| \right)^{\frac{q}{p}} d\lambda \\ &\leq \sum_j \sum_k \int_{2^j}^{2^{j+1}} \lambda^{q-1} |\{2^k < |f| < 2^{k+1}\}|^{\frac{q}{p}} d\lambda \\ &\sim \sum_j \sum_k 2^{j2^{j(q-1)}} |\{2^k < |f| < 2^{k+1}\}|^{\frac{q}{p}} \\ &= \sum_k \underbrace{\sum_{\substack{j \leq k \\ 2^{q(k+1)}}} |\{2^k < |f| < 2^{k+1}\}|^{\frac{q}{p}}}_{2^{q(k+1)}} \\ &\sim \sum_k \underbrace{2^{qk} |\{2^k < |f| < 2^{k+1}\}|^{\frac{q}{p}}}_{\|f_k\| + \|f_{k+1}\|_{L^p}^q \lesssim \|f_k\|_{L^p}^q + \|f_{k+1}\|_{L^p}^q} \end{aligned}$$

The last step used that

$$|f_k| + |f_{k+1}| \sim 2^k \mathbb{1}_{\{2^k < |f| < 2^{k+1}\}}$$

□

Another proof of the improved Sobolev inequality using the standard Sobolev inequality.

$$\begin{aligned} \|f\|_{L^{p,q}}^p &\lesssim \sum_j \|f_j\|_{L^{p^*}}^p \stackrel{\text{Standard Sobolev}}{\lesssim} \sum_j \|\nabla f_j\|_{L^p}^p \\ &= \int_{\mathbb{R}^d} \sum_j |\nabla f_j|^p \lesssim \int_{\mathbb{R}^d} |\nabla f|^p = \|\nabla f\|_{L^p}^p \end{aligned}$$

$$\lesssim \int_{\mathbb{R}^d} (|\nabla f|^p |f|^p) \leq \|\nabla f\|_{L^p}^p + \|f\|_{L^p}^p = \|f\|_{W^{1,p}}^p$$

We used the pointwise bound:

$$\left| \underbrace{\nabla f_j(x)}_{=\nabla(f\varphi(2^{-j}f))=\nabla f\cdot\varphi+f\nabla\varphi} \right| \lesssim |\nabla f(x)| \mathbb{1}_{\{2^{j-1} \leq |f| \leq 2^{j+1}\}}$$

By Scaling argument: If $\|f\|_{L^{p^*,p}} \lesssim \|\nabla f\|_{L^p} + \|f\|_{L^p}$, $\forall f$

$$\Rightarrow \|f\|_{L^{p^*,p}} \lesssim \|\nabla f\|_{L^p}$$

□

Remark. Scaling argument in 3D:

$$(1) \|f\|_{L^6} \lesssim (\int |\nabla f|^2 + \int |f|^2)^{\frac{1}{2}}, \quad \forall f$$

$$(2) \Rightarrow \|f\|_{L^6} \leq (\int |\nabla f|^2)^{\frac{1}{2}}$$

Use (1) $f_\lambda(x) = \lambda^{\frac{3}{2}} f(\lambda x) \Rightarrow \int_{\mathbb{R}^3} |f_\lambda|^2 = \int_{\mathbb{R}^3} |f|^2$

$$\int |\nabla f_\lambda|^2 = \lambda^2 \int |\nabla f|^2, \quad \|f\|_{L^6}^2 = \lambda^2 \|f\|_{L^6}^2$$

Chapter 5

Fourier Transform

We will define the Fourier transform for tempered distributions, which can be viewed as a generalization of L^p functions. The space of tempered distributions is constructed as the dual space of the Schwartz class $C_c^\infty(\mathbb{R}^d) \subset C^\infty$.

First, let's start with the Fourier transform in L^p .

Definition 5.1 (Fourier Transform in L^1). Let $f \in L^1(\mathbb{R}^d)$, then the Fourier transform of f is defined as

$$\hat{f} = \mathcal{F}f(\xi) = \int_{\mathbb{R}^d} f(x)e^{-2\pi i \xi x} dx$$

Remark. 1. $\|\hat{f}\|_{L^\infty} \leq \|f\|_{L^1}$

2. \hat{f} is uniformly continuous: follows from

$$\lim_{\xi \rightarrow 0} \int_{\mathbb{R}^d} |f(x)|e^{2\pi i \xi x} dx$$

by dominated convergence

3. $\phi(x) = e^{-\pi|x|^2} \Rightarrow \hat{\phi}(\xi) = e^{-\pi|\xi|^2}$. To see this, observe

$$\begin{aligned} \partial_{\xi_n} \hat{\phi}(\xi) &= \int_{\mathbb{R}^d} -2\pi i x_n e^{-\pi|x|^2} e^{-2\pi i \xi x} dx = i \int_{\mathbb{R}^d} \partial_{x_n} (e^{-\pi|x|^2}) e^{-2\pi i \xi x} dx \\ &= -i \int_{\mathbb{R}^d} e^{-\pi|x|} (\partial_{x_n} e^{-2\pi \xi x}) dx = -2\pi \xi_n \hat{\phi}(\xi) \end{aligned}$$

So $\partial_{\xi_n} \hat{\phi}(\xi) = -2\pi \xi_n \hat{\phi}(\xi)$ and the claim follows from solving this ODE.

4. $g(x) = f(\lambda x) \Rightarrow \hat{g}(x) = \frac{1}{\lambda^d} \hat{f}(\frac{\xi}{\lambda})$

5. $\tau_a f(x) = f(x - a) \Rightarrow \widehat{\tau_a f}(k) = e^{-2\pi i \xi a} \hat{f}(\xi)$

Lemma 5.2 (Riemann-Lebesgue). If $f \in L^1(\mathbb{R}^d)$, then

$$\lim_{\xi \rightarrow \infty} \hat{f}(\xi) = 0$$

Proof.

$$\begin{aligned}\hat{f}(\xi) &= \frac{1}{2}(f(\xi) + e^{2\pi\xi a} \widehat{\tau_a f}(\xi)) \\ &= \frac{1}{2} \int_{\mathbb{R}^d} (f(x) - \tau_a f(x)) e^{-2\pi i \xi x} dx \\ \Rightarrow |\hat{f}(\xi)| &\leq \frac{1}{2} \int_{\mathbb{R}^d} |f(x) - f(x-a)| dx\end{aligned}$$

and now take $a(\xi) = \frac{\xi}{2|\xi|^2}$.

If f is continuous and compactly supported in $B(0, R)$, then for every $\varepsilon > 0$ we can choose $\delta > 0$ s.t.

$$\|f - \tau_a f\|_{L^1} \leq \varepsilon$$

if $|a| < \delta$ and $\text{supp } \tau_a f \subset B(0, R)$

$\Rightarrow |\hat{f}(\xi)| \leq \frac{\varepsilon}{2}$. Now conclude with approximation argument. \square

Lemma 5.3. Let $f, g \in L^1(\mathbb{R}^d)$, then

$$\widehat{f \star g}(\xi) = \hat{f}(\xi) \hat{g}(\xi)$$

Proof.

$$\begin{aligned}\widehat{f \star g}(\xi) &= \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} t_{\mathbb{R}^d} f(x-y) g(y) e^{-2\pi i \xi x} dy dx \\ &\stackrel{\text{Fubini and change of variable}}{=} \left(\int_{\mathbb{R}^d} f(x-y) e^{-2\pi i \xi(x-y)} dx \right) \left(\int_{\mathbb{R}^d} g(y) e^{-2\pi i \xi(y)} dy \right) = \hat{f}(\xi) \hat{g}(\xi)\end{aligned}$$

\square

Theorem 5.4. If $f \in L^1(\mathbb{R}^d)$ and $\hat{f} \in L^1$, then

$$f(x) = \int_{\mathbb{R}^d} \hat{f}(\xi) e^{2\pi i x \cdot \xi} d\xi$$

Proof. The problem is that we want to consider

$$\int \int f(y) e^{2\pi i x \xi} d\xi dy$$

BUT $(\xi, y) \mapsto f(y) e^{2\pi i \xi(x-y)} \notin L^1(\mathbb{R}^d \times \mathbb{R}^d)$.

So we define

$$A_\varepsilon(x) = \int_{\mathbb{R}^d} \hat{f}(\xi) e^{2\pi i x \xi} e^{-\varepsilon |x|^2} d\xi$$

and observe that

$$\lim_{\varepsilon \rightarrow 0} A_\varepsilon = \mathcal{F}^{-1}(f)(x)$$

again by DCT and obsereve

$$A_\varepsilon(x) = \int \int \underbrace{f(y) e^{2\pi i \xi(x-y)} e^{\varepsilon |x|^2}}_{\in L^1(\mathbb{R}^d \times \mathbb{R}^d)} d\xi dy = \int f(y) \Phi_\varepsilon(x-y) dy$$

where $\Phi_\varepsilon(y) = (\frac{\pi}{\varepsilon})^{\frac{d}{2}} e^{-\frac{\pi^2}{\varepsilon} |y|^2}$.

$$\begin{aligned} (f \star \Phi_\varepsilon)(x) - f(x) &= \int (f(x-y) - f(x)) \Phi_\varepsilon(y) dy \\ \Rightarrow \| (f \star \Phi_\varepsilon) - f \|_{L^1} &\leq \int \| f(x-\sqrt{\varepsilon}y) - f(x) \|_{L^1} \Phi_1(y) dy \xrightarrow{\varepsilon \rightarrow 0} 0 \end{aligned}$$

by DCT

so we can take a subsequence $\varepsilon_k \rightarrow 0$ s.t. $A_{\varepsilon_k} \rightarrow 0$ s.t. $A_{\varepsilon_k}(x) \rightarrow f(x)$ pointwise \square

Lemma 5.5 (Uncertainty Principle). Let $f \in L^1$. If both f, \hat{f} have compact support, then $f = 0$ a.e.

Proof. (sketch) $\hat{f}(\xi) = \int f(x) e^{-2\pi i \xi x} dx$. Consider it as a function if $\xi \in \mathbb{C}^d$. f compact support $\Rightarrow \hat{f}$ analytic and since it has compact support, then $\hat{f} = 0$ a.e. and thus $f = 0$ a.e. \square

Lemma 5.6. Let $f \in L^1 \cap L^2$. Then $\hat{f} \in L^1 \cap L^2$ and $\| \hat{f} \|_{L^2} = \| f \|_{L^2}$

Proof. First one proves that

$$\int f(x) \hat{g}(x) dx = \int \int f(x) g(y) e^{-2\pi i xy} dy dx = \int \hat{f}(y) g(y) dy$$

and just take $g = \mathcal{F}^{-1}(\bar{f}) = \overline{\mathcal{F}^{-1}(f)}$. Thus $\| \hat{f} \|_{L^2} = \| f \|_{L^2}$ \square

Now we have $\|\mathcal{F}\|_{L^1 \rightarrow L^\infty} \leq 1$ and $\|\mathcal{F}\|_{L^2 \rightarrow L^2} = 1$. We can use interpolation!

Theorem 5.7 (Hausdorff-Young). Let $[1, 2]$, then $\|\mathcal{F}\|_{L^p \rightarrow L^{p'}} \leq 1$, i.e. $\|\mathcal{F}f\|_{L^{p'}} \leq \|f\|_{L^p}$ where $\frac{1}{p} + \frac{1}{p'} = 1$

Proof. This is a direct application of Riesz-Thorin with $T : L^{p_0} + L^{p_1} \rightarrow L^{q_0} + L^{q_1}$ where we choose $T = \mathcal{F}$, $p_0 = 1$, $q_0 = \infty$, $p_1 = q_1 = 2$ \square

5.1 Schwarz class

Definition 5.8. Let $\varphi \in C^\infty(\mathbb{R}^d)$ and $\alpha, \beta \in \mathbb{N}_0^d$. Define

$$\rho_{\alpha, \beta}(\varphi) = \sup_{x \in \mathbb{R}^d} |x^\alpha \partial^\beta \varphi|$$

The Schwarz class is then defined as

$$S(\mathbb{R}^d) = \{\varphi \in C^\infty : \forall \alpha, \beta \in \mathbb{N}_0^d : \rho_{\alpha, \beta}(\varphi) < \infty\}$$

- Remark.**
1. $C_c^\infty(\mathbb{R}^d) \subset S(\mathbb{R}^d)$
 2. $\varphi \in S \Leftrightarrow \forall \alpha \in \mathbb{N}_0^d, N \in \mathbb{N} \exists C_{\alpha, N}$ s.t. $|\partial^\alpha f(x)| \leq \frac{C_{\alpha, N}}{(1+|x|^2)^N}$
 3. $S() \subset L^p$ for every $p \geq 1$
 4. $S()$ is closed under differentiation and multiplication by polynomials

Proposition 5.9. Let $f \in S()$ then

1. $\|\hat{f}\|_{L^\infty} \leq \|f\|_{L^1}$
2. $\|\hat{f}\|_{L^2} = \|f\|_{L^2}$
3. $\widehat{f \star g} = \hat{f} \hat{g}$
4. $\partial^\alpha \hat{f}(\xi) = ((-2\pi x)^\alpha \widehat{f(x)})(\xi)$
5. $\widehat{\partial^\alpha f}(\xi) = (2\pi i \xi)^\alpha \hat{f}(\xi)$
6. $\hat{f} \in S(\mathbb{R}^d)$

Proof. (1)-(3) already done.

(4)-(5) just use DCT: for example $\partial_{\xi_k} \hat{f}(\xi) = \partial_{\xi_k} \int f(x) e^{-2\pi i \xi x} dx = -2\pi i \xi_k \hat{f}(\xi)$

(6) To show that $\hat{f} \in S$, observe that

$$\rho_{\alpha,\beta}(\hat{f}) = \left\| \xi^\alpha \partial^\beta \hat{f} \right\|_{L^\infty} = \frac{(2\pi)^{|\beta|}}{(2\pi)^{|\alpha|}} \left\| \widehat{x \partial^\beta f} \right\|_{L^\infty} \leq (2\pi)^{|\beta|-|\alpha|} \|x^\alpha \partial^\beta f\|_{L^1} < \infty$$

□

Definition 5.10. Let X be a vector space. A function $\rho : X \rightarrow [0, \infty)$ is called a seminorm, if

1. $\rho(\lambda x) = |\lambda| \rho(x) \quad \forall \lambda \in \mathbb{C}, x \in X$
2. $\rho(x+y) \leq \rho(x) + \rho(y) \quad \forall x, y \in X$

Definition 5.11. Let X be a vector space with seminorms $\{\rho_n\}_{n \in \mathbb{N}}$ s.t.

$$\rho_n(x) = 0 \quad \forall n \in \mathbb{N} \Leftrightarrow x = 0$$

We call X a Fréchet space if (X, d) is a complete metric space for

$$d(x, y) := \sum_{n \in \mathbb{N}} \frac{1}{2^n} \frac{\rho_n(x-y)}{1 + \rho_n(x-y)}$$

Proposition 5.12. Let X be a Fréchet space, then

1. $f_n \xrightarrow{d} f \Leftrightarrow \forall m \in \mathbb{N} \rho_m(f_n - f) \xrightarrow{n \rightarrow \infty} 0$
2. $\{f_n\}_{n \in \mathbb{N}}$ is Cauchy in d iff it is Cauchy in each ρ_m
3. ρ_m are continuous wrt d

Proof. (1) $f_n \rightarrow f$ in $d \Rightarrow \rho_m(f_n - f) \rightarrow 0$ for each $m \in \mathbb{N}$.

Now assume $\rho_m(f_n - f) \rightarrow 0$ for each m , then we can choose $M \in \mathbb{N}$ s.t.

$$\sum_{n=M+1}^{\infty} \frac{1}{2^n} < \frac{\varepsilon}{2}$$

and choose $N \in \mathbb{N}$ s.t. $\rho_m(f_n - f) < \frac{\varepsilon}{2} \forall m = 1, \dots, M, n \geq N$

$$\Rightarrow d(f_n, f) \leq \sum_{n=0}^M \frac{\rho_m(f_n - f)}{1 + \rho_m(f_n - f)} \frac{1}{2^m} + \frac{\varepsilon}{2} < \varepsilon$$

(2) and (3) are similar \square

Theorem 5.13. $S(\mathbb{R}^d)$ with the seminorms $\{\rho_{\alpha,\beta}\}_{\alpha,\beta \in \mathbb{N}_0^d}$ is a Fréchet space.

Proof. (exercise) \square

Proposition 5.14. $\mathcal{F} : S(\mathbb{R}^d) \rightarrow S(\mathbb{R}^d)$ is a linear homeomorphism, i.e. it is continuous and has a continuous inverse

Proof.

$$\rho_{\alpha,\beta}(\hat{\phi}) = \left\| \xi^\alpha \partial^\beta \hat{\phi} \right\|_{L^\infty} = \frac{(2\pi)^{|\beta|}}{(2\pi)^{|\alpha|}} \left\| \widehat{\partial^\alpha x^\beta \phi} \right\|_{L^\infty} \leq C_{\alpha,\beta} \left\| \partial^\alpha x^\beta \phi \right\|_{L^1}$$

And using Leibnitz rule:

$$\partial^\alpha(fg) = \sum_{\gamma \leq \alpha} \binom{\alpha}{\gamma} \partial^{\alpha-\gamma} f \partial^\gamma g$$

we get

$$\rho_{\alpha,\beta}(\hat{\phi}) \leq C_{\alpha,\beta} \sum_{\gamma \leq \alpha} c_{\alpha,\beta,\gamma} \left\| x^{\beta-\alpha+\gamma} \check{\phi} \right\|_{L^1}$$

and observe that

$$\left\| \rho \right\|_{L^1} = \int \frac{(1+|x|^{2d})}{(1+|x|^{2d})} |\phi| dx \leq C_d \left\| (1+|x|^d) \phi \right\|_{L^\infty}$$

$$\rho_{\alpha,\beta}(\hat{\phi}) \leq \check{C}_{\alpha,\beta,\gamma,d} \sum_{|\nu| \leq N, |\kappa| \leq N} \rho_{\nu,\kappa}(\phi)$$

where $N = \max(|\alpha|, |\beta|) + 2d$. Thus if $\phi_n \rightarrow 0$ in $S(\mathbb{R}^d)$ we also get $\mathcal{F}(\phi_n) \rightarrow 0$ in $S(\mathbb{R}^d)$ \square

5.2 Tempered Distributions

Definition 5.15. The space $S'(\mathbb{R}^d)$ of tempered distributions is the set of linear continuous functionals $T : S(\mathbb{R}^d) \rightarrow \mathbb{C}$.

Notation: $T \in S'(\mathbb{R}^d)$ and $\phi \in S(\mathbb{R}^d) \Rightarrow T(\phi) = \langle T, \phi \rangle$

Proposition 5.16. A linear map $T : S(\mathbb{R}^d) \rightarrow \mathbb{C}$ belongs to $S'(\mathbb{R}^d)$ iff $\exists c < \infty$ and $N \in \mathbb{N}_0$ s.t.

$$|\langle T, \phi \rangle| \leq c \sum_{|\alpha| \leq N, |\beta| \leq N} \rho_{\alpha, \beta}(\phi) \quad \forall \phi \in S(\mathbb{R}^d)$$

Proof. " \Leftarrow " is clear, since $\phi_n \rightarrow 0$ in $S(\mathbb{R}^d) \Rightarrow \langle T, \phi_n \rangle \rightarrow 0$ in \mathbb{C}

" \Rightarrow " By way of contradiction. Suppose for every $N \in \mathbb{N}_0$ we can find $\varphi_n \in S$ s.t. $\langle T, \varphi_n \rangle \geq N \sum_{|\alpha| \leq N, |\beta| \leq N} \rho_{\alpha, \beta}(\varphi_n) =: N \mathcal{N}_N(\varphi_N)$ and define $\Psi_N := \frac{\varphi_N}{\sqrt{N} \sum_{|\alpha| \leq N, |\beta| \leq N} \rho_{\alpha, \beta}(\varphi_N)}$ \Rightarrow for every $|\alpha|, |\beta| \leq K$: $\rho_{\alpha, \beta}(\Psi_N) \leq \frac{\rho_{\alpha, \beta}(\varphi_N)}{\sqrt{N} \sum_{|\alpha| \leq N, |\beta| \leq N} \rho_{\alpha, \beta}(\varphi_N)} \leq \frac{1}{\sqrt{N}}$

i.e. $\Psi_N \rightarrow 0$ in S . Thus by continuity, $\langle T, \Psi_N \rangle \rightarrow 0$ in \mathbb{C} BUT $|\langle T, \Psi_N \rangle| \geq \frac{|\langle T, \varphi_N \rangle|}{\sqrt{N} \sum_{|\alpha| \leq N, |\beta| \leq N} \rho_{\alpha, \beta}(\varphi_N) \geq \sqrt{N}}$ which is a contradiction. \square

Example 5.17. 1. $f \in L^p(\mathbb{R}^d)$ for some $p \in [1, \infty]$

$$T_f : \phi \mapsto \int_{\mathbb{R}^d} f \phi$$

is a tempered distribution. Indeed,

$$\begin{aligned} |\langle T_f, \phi \rangle| &\leq \|f\|_{L^p} \|\phi\|_{L^{p'}} = \|f\|_{L^p} \left\| |\phi|^{\frac{1}{p'}} |\phi|^{1-\frac{1}{p'}} \right\| \leq \|f\|_{L^p} \|\phi\|_{L^1}^{\frac{1}{p'}} \|\phi\|_{L^\infty}^{1-\frac{1}{p'}} \\ &\leq \|f\|_{L^p} \sum_{|\alpha|, |\beta| \leq K} \rho_{\alpha, \beta}(\phi) \end{aligned}$$

in the last step we used $a^\theta b^{1-\theta} \leq \theta a + (1-\theta)b$

2. Polynomials are in $S'(\mathbb{R}^d)$
3. Dirac-delta in $S'(\mathbb{R}^d)$: Define $\langle \delta_a, \phi \rangle = \phi(a)$, then clearly $|\langle \delta_a, \phi \rangle| \leq \rho_{0,0}(\phi)$ and one usually writes $\phi \mapsto \int \phi d\mu_a = \phi(a) = \int \phi(x) \delta(x-a) dx$
4. Given a finite measure μ on \mathbb{R}^d , then $\phi \mapsto \int_{\mathbb{R}^d} \phi d\mu \in S'$ since $|\int_{\mathbb{R}^d} \phi d\mu| \leq \|\phi\|_{L^\infty} \mu(\mathbb{R}^d)$

Definition 5.18. $\{T_n\}_{n \in \mathbb{N}} \subset S'(\mathbb{R}^d)$ converges to $T \in S'(\mathbb{R}^d)$ if

$$\forall \phi \in S(\mathbb{R}^d) : \quad \langle T_n, \phi \rangle \rightarrow \langle T, \phi \rangle \text{ in } \mathbb{C}$$

One usually writes $T_n \rightharpoonup T$

Remark. Given $g \in C_c^\infty(\mathbb{R}^d)$ with $\|g\|_{L^1} = 1$, define $g_k(x) = 2^{-kd}g(2^kx)$. Then $g_k = T_{g_k} \rightharpoonup \delta_0$ in S'

Observe that for $\phi, \psi \in S$ we have the integration by parts formula:

$$\int_{\mathbb{R}^d} (\partial^\alpha \phi) \psi = (-1)^{|\alpha|} \int_{\mathbb{R}^d} \phi (\partial^\alpha \psi)$$

Definition 5.19. For $T \in S'$ and $\alpha \in \mathbb{N}_0^d$, we define $\partial^\alpha T \in S'$ as

$$\partial^\alpha T(\phi) = (-1)^{|\alpha|} T(\partial^\alpha \phi) \quad \forall \phi \in S$$

Remark. To show that $\partial^\alpha T \in S'$ we have to check $\exists C, K$ s.t. $|\langle \partial^\alpha T, \phi \rangle| \leq C \mathcal{N}_K(\phi)$ $\forall \phi \in S$ and this is true, since $|\langle \partial^\alpha T, \phi \rangle| = |T, \partial^\alpha \phi| \leq C \mathcal{N}_{K+|\alpha|}(\phi)$

Example 5.20. 1. $\partial^\alpha \delta_a = (-1)^{|\alpha|} \partial^\alpha \phi(0)$ by definition

2. In $d = 1$ define $g(x) = \begin{cases} x, & \text{if } x \geq 0 \\ 0, & \text{else} \end{cases}$ so that $|\int g(x)\phi(x)| \leq \|x\phi\|_{L^1} \leq C \mathcal{N}_K(\phi)$ and $\langle \partial g, \phi \rangle = - \int_0^\infty x \phi'(x) dx = \int_0^\infty \phi(x) dx = \langle \Theta, \phi \rangle$ where $\partial g = \Theta = \begin{cases} 1, & \text{if } x \geq 0 \\ 0, & \text{else} \end{cases}$ the Heavyside function

$$\langle \partial \Theta, \phi \rangle = - \int_0^\infty \phi'(x) dx = \phi(0) = \delta_0(\phi) \Rightarrow \partial \Theta = \delta_0$$

Definition 5.21. The space of slowly increasing functions is given by

$$\mathcal{O}(\mathbb{R}^d) := \{\phi \in C^\infty(\mathbb{R}^d) : \forall \alpha \in \mathbb{N}_0^d \exists C_\alpha, m_\alpha \in \mathbb{N} \text{ s.t. } |\partial^\alpha \phi| \leq C_\alpha (1 + |x|)^{m_\alpha}\}$$

Remark. 1. $\mathcal{O} \hookrightarrow S'$ in the sense of integration

2. Given $\psi \in \mathcal{O}$: $M_\psi : \phi \in S \mapsto \psi\phi \in S$ is continuous

Definition 5.22. Given $T \in S'$ and $g \in \mathcal{O}$ we can define $\langle gT, \phi \rangle = \langle T, g\phi \rangle \quad \forall \phi \in S$

Definition 5.23. Let $T \in S'$. Define its Fourier transform $\hat{T} = \mathcal{F}(T)$ as

$$\langle \hat{T}, \phi \rangle = \langle T, \hat{\phi} \rangle, \quad \forall \phi \in S$$

Remark. $\hat{T} \in S'$ because $\mathcal{F} : S \rightarrow S$ is a linear homeomorphism. This also allows us to define the inverse Fourier transform for tempered distributions:

$$\langle \check{T}, \phi \rangle = \langle T, \check{\phi} \rangle$$

Example 5.24. 1. $\langle \hat{\delta}_a, \phi \rangle = \langle \delta_a, \hat{\phi} \rangle = \hat{\phi}(a) = \int \phi(x) e^{-2\pi i ax} dx$ i.e. $\hat{\delta}_a = e^{-2\pi i ax}$ and in particular $\hat{\delta}_0 = 1$

2. Similarly $\langle \hat{1}, \phi \rangle = \langle 1, \hat{\phi} \rangle = \int \hat{\phi}(\xi) d\xi = \phi(0)$, i.e. $\hat{1} = \delta_0$

3. $\langle \widehat{\partial^\alpha \delta_a}, \phi \rangle = \langle \partial^\alpha \delta_a, \hat{\phi} \rangle = (-1)^{|\alpha|} \langle \delta_a, \partial^\alpha \hat{\phi} \rangle = (-1)^{|\alpha|} \langle \delta_a, \widehat{(2\pi i x)^\alpha \phi} \rangle = (2\pi i x)^\alpha e^{-2\pi i ax}$

Proposition 5.25. Let $T \in S'$ and $\alpha \in \mathbb{N}_0^d$, then we have

$$\partial^\alpha \hat{T} = \widehat{(-2\pi i x)^\alpha T}$$

and

$$\widehat{\partial^\alpha T} = \widehat{(-2\pi i \xi)^\alpha \hat{T}}$$

in the sense of distributions

Proof. (exercise) □

Definition 5.26. Let $T \in S'$ and $\Omega \subset \mathbb{R}^d$ open. We say that T vanishes in Ω if $\forall \phi \in S$ with $\text{supp}(\phi) \subset \Omega$ we have

$$\langle T, \phi \rangle = 0$$

The support of T is the complement of the largest open set in which T vanishes.

Proposition 5.27. Suppose $T \in S'$ has a support at the origin. Then $\exists K \in \mathbb{N}$ s.t.

$$T = \sum_{|\alpha| \leq K} c_\alpha \partial^\alpha \delta_0$$

Proof. Since $T \in S'$, we can find $c > 0$ and $K \in \mathbb{N}$ s.t.

$$|\langle T, \phi \rangle| \leq c \sum_{|\alpha|, |\beta| \leq K} \rho_{\alpha, \beta}(\phi) = c \mathcal{N}_K(\phi)$$

We use Taylor's formula for ϕ at the origin:

$$\phi(x) = \sum_{|\gamma| \leq K} a_\gamma \partial^\gamma \phi(0) x^\gamma + R_K(x)$$

where $R_K(x) = \mathcal{O}(|x|^\gamma)$ as $|x| \rightarrow 0$ and thus $\lim_{|x| \rightarrow 0} \frac{|\partial^\gamma R_K(x)|}{|x|^{K-|\gamma|}} = 0$ for $|\gamma| \leq K$.

Let $\chi \in \mathbb{C}_c^\infty$ with support $B(0, 1)$ and $\chi = 1$ in $B(0, \frac{1}{2})$ s.t.

$$\langle T, \phi \rangle = \langle T, \chi\phi + (1-\chi)\phi \rangle = \langle T, \chi\phi \rangle = \sum_{|\gamma| \leq K} a_\gamma \partial^\gamma \phi(0) \langle T, \chi x^\gamma \rangle + \langle T, \chi R_K(x) \rangle.$$

So it is enough to show that the second term is equal to 0.

Let $\varepsilon < 1$. Define $\chi_\varepsilon := \chi(\frac{x}{\varepsilon})$ s.t.

$$\langle T, \chi(x) R_K(x) \rangle = \langle T, \chi(x) R_K(x) \chi_\varepsilon + \chi(x) R_K(x)(1-\chi_\varepsilon) \rangle = \langle T, \chi R_K(x) \chi_\varepsilon(x) \rangle = \langle T, \chi_\varepsilon R_K(x) \rangle$$

and by continuity of T

$$\begin{aligned} |\langle T, \chi_\varepsilon R_K \rangle| &\leq C \mathcal{N}_K(\chi_\varepsilon R_K) \leq C \sum_{|\alpha|, |\beta| \leq K} \sum_{\gamma \leq \beta} c_{\gamma, \beta} \|x^\alpha \partial^{\beta-\gamma} R_K(x) \partial^\gamma \chi_\varepsilon\|_{L^\infty} \\ &\leq \tilde{C} \sum_{|\beta| \leq K} \|\partial^{\beta-\gamma} R_K(x) \partial^\gamma \chi_\varepsilon(x)\|_{L^\infty} \end{aligned}$$

Now observe that $|\partial^\gamma \chi_\varepsilon(x)| \leq \frac{C_\gamma}{|\varepsilon|^\gamma} \chi_{B(0, \varepsilon)}(x)$ and $\|\partial^{\beta-\gamma} R_K(x) \chi_{B(0, \varepsilon)}(x)\|_{L^\infty} = \mathcal{O}(\varepsilon^{K-|\beta-\gamma|})$ by Taylor.

Thus all in all: $|\langle T, \chi_\varepsilon R_K \rangle| = \mathcal{O}\left(\sum_{|\alpha|, |\beta| \leq K} \sum_{\gamma \leq \beta} \varepsilon^{K-|\beta-\gamma|-|\gamma|}\right)$ as $\varepsilon \rightarrow 0$ and by our constraints: $K - |\beta - \gamma| - |\gamma| \geq 0$ \square

Corollary 5.28. Let $T \in S'$ be s.t. $\Delta T = 0$, i.e. $\sum_{i=1}^d \partial_{x_i}^2 T = 0$. Then T is a polynomial.

Proof. $0 = \langle \Delta T, \hat{\phi} \rangle = \langle T, \Delta \hat{\phi} \rangle = - \langle T, 4\pi^2 |x|^2 \phi \rangle$, i.e. $4\pi^2 \xi^2 \hat{T} = 0 \Rightarrow \hat{T}$ has support at the origin.

$$\Rightarrow \hat{T} = \sum_{|\gamma| \leq K} c_\gamma \partial^\gamma \delta_0$$

and $T = \sum_{|\gamma| \leq K} c_\gamma \underbrace{\mathcal{F}^{-1}(\partial^\gamma \delta_0)}_{\text{is polynomial from examples}}$

□

5.3 Remarks on Distributions vs. Tempered distributions

Definition 5.29. Let $\Omega \subset \mathbb{R}^d$ open. Define the space of distributions $D'(\Omega)$ as the dual space of $D(\Omega)$.

Here $D(\Omega) = C_c^\infty(\Omega)$, with the topology

$$\varphi_n \rightarrow \varphi \text{ in } D(\Omega) \Leftrightarrow \begin{cases} \|D^\alpha \varphi_n - D^\alpha \varphi\|_{L^\infty} \rightarrow 0 \ \forall \alpha \\ \text{supp } (\phi_n) \subset K \text{ compact} \subset \Omega \end{cases}$$

Heuristic idea:

$$D(\Omega) \subset L^2 \Rightarrow (D(\Omega))^* = D'(\Omega) \supset (L^2)^* = L^2$$

since "smaller space \Rightarrow larger dual"

Example 5.30. Any function $f \in L^1_{loc}(\Omega)$ is a distribution $T_f(\varphi) = \int_\Omega f \phi \ \forall \phi \in D(\Omega)$

Lemma 5.31 (Fundamental Lemma of Calculus of Variations). If $f \in L^1_{loc}(\Omega)$ and $\int_\Omega f \varphi = 0, \ \forall \varphi \in C_c^\infty(\Omega)$, then $f = 0$ a.e. $x \in \Omega$

Proof. Step 1 Case $\Omega = \mathbb{R}^d$. Let $f \in L^1_{loc}$ and $\int_{\mathbb{R}^d} f \varphi = 0, \ \forall \varphi \in C_c^\infty(\Omega)$

Take $\varphi_0 \in C_c^\infty$ and define $\varphi_{n,y}(x) = n^d \varphi(\frac{y-x}{n}) \in C_c^\infty(\mathbb{R}^d, dx)$. Also choose $1 = \int_\Omega \varphi_0 = \int_\Omega \varphi_n$.

Then: $0 = \int_{\Omega} f(x) \varphi_{n,y}(x) dx = \int_{\Omega} n^d f(x) \varphi(\frac{y-x}{n}) dx \xrightarrow{n \rightarrow \infty} f(y)$, i.e. $f = 0$ a.e.

Step 2: Consider the general case $\Omega \subset \mathbb{R}^d$, $f \in L^1_{loc}(\Omega)$. Take $\chi \in C_c^\infty(\Omega)$. Then $\int_{\Omega} f \underbrace{\chi \varphi}_{\in C_c^\infty}$,

$$C_c^\infty(\Omega) = \int_{\Omega} g \varphi \text{ where } g = f \chi \in L^1_c(\Omega) \subset L^1_c(\mathbb{R}^d)$$

By step 1, $g = f \chi = 0$ a.e. $\Rightarrow f = 0$ a.e. \square

Comparing to the tempered distributions:

Note that D' is really bigger than the space of tempered distributions!

E.g. Take $f(x) = e^x$ in \mathbb{R} is a distribution but not a tempered distribution!

For proving that, consider the Schwartz function $\varphi(x) = e^{-\sqrt{1+x^2}}$ and show that $\int_{\mathbb{R}} \varphi e^x = \infty$ (exercise)

Derivatives: For all $T \in D$ we can define $\forall \alpha$

$$(D^\alpha T) \in D' \text{ by } (D^\alpha T)(\varphi) = (-1)^{|\alpha|} T(D^\alpha \varphi), \forall \varphi \in D$$

Q: Why is the space of distributions too big for the Fourier transform?

If we want to define the Fourier transform of a distribution $T \in D'$, then we want to do something like

$$(\mathcal{F}T)(\varphi) = \langle \mathcal{F}T, \varphi \rangle = \langle T, \mathcal{F}^* \varphi \rangle$$

BUT this does not work in general in D' , since $\mathcal{F} : C_c^\infty(\mathbb{R}^d) \not\rightarrow C_c^\infty(\mathbb{R}^d)$

This is related to the "uncertainty principle", i.e. if f is more localized $\Rightarrow \hat{f}$ is less localized.

Theorem 5.32 (Hardy Uncertainty Principle). If $f : \mathbb{R}^d \rightarrow \mathbb{C}$ satisfies

$$|f(x)| \leq C e^{-\pi a|x|^2}$$

and

$$|\hat{f}(x)| \leq C e^{-\pi b|\xi|^2}$$

for parameters $a, b > 0$, then

$$ab \leq 1$$

Moreover, if $ab = 1$, then $f(x) = \text{const} e^{-\pi a|x|^2}$ and $\hat{f}(x) = \text{const} e^{-\frac{\pi}{a}|\xi|^2}$

It is easy to prove, that if $f \in C_c^\infty$, then \hat{f} is not compactly supported.

Let $d = 1$. Consider $\hat{f}(z) = \int_{\mathbb{R}} f(x) e^{-2\pi i zx} dx$, $z \in \mathbb{C}$.

Since $f \in C_c^\infty(\mathbb{R}) \Rightarrow \hat{f}(z)$ is entire. (This is sufficient, if $f \in L^1_c$)

Here, if $\hat{f}(\xi)$, $\xi \in \mathbb{R}$ is compactly supported, then $\hat{f}(\xi) = 0$ is an open integral in \mathbb{R} .

$\Rightarrow \hat{f} = 0$ in $\mathbb{C} \Rightarrow \hat{f} = 0$ in $\mathbb{R} \Rightarrow f = 0$ in \mathbb{R}

proof of Hardy Uncertainty Principle. We only consider $d = 1$. The argument can be extended to higher dimensions. We prove that if $\begin{cases} |f(x)| \leq Ce^{-\pi x^2}, \\ |\hat{f}(x)| \leq Ce^{-\pi \xi^2} \end{cases}$ for all $x, \xi \in \mathbb{R}$,

then $f = ce^{-\pi x^2}$.

Define $\hat{f}(z) = \int_{\mathbb{R}} f(x)e^{-2\pi izx}dx$, $z \in \mathbb{C}$, $z = \xi + i\nu$, $\xi, \nu \in \mathbb{R}$

$\Rightarrow \partial_z \hat{f}(z) = \int_{\mathbb{R}} f(x)(-2\pi ix)e^{-2\pi ixz}dx \Rightarrow \hat{f}(z)$ is entire.

We have the following easy upper bound:

$$\begin{aligned} |\hat{f}(z)| &= \left| \int_{\mathbb{R}} f(x)e^{-2\pi i\xi x + 2\pi \nu x} dx \right| \\ &\leq \int_{\mathbb{R}} |f(x)| e^{2\pi \nu x} dx \\ &\leq \int_{\mathbb{R}} e^{-\pi x^2} e^{2\pi \nu x} dx \\ &= \int_{\mathbb{R}} e^{-\pi|x-\nu|^2} e^{\pi \nu^2} dx \\ &= (\int_{\mathbb{R}} e^{-\pi y^2} dy) e^{\pi \nu^2} = e^{\pi \nu^2} \end{aligned}$$

1.try: $F(z) = e^{\pi z^2} \hat{f}(z)$ entire. $|F(\xi)| = |e^{\pi \xi^2} \hat{f}(\xi)| \leq C$ for all $\xi \in \mathbb{R}$.

We want to somehow show $|F(z)| \leq C$ for all $z \in \mathbb{C}$.

$\Rightarrow F(z) = \text{const}$ in $\mathbb{C} \Rightarrow \hat{f}(z) = \text{const} e^{-\pi z^2} \forall z \in \mathbb{C}$

The problem here is that $|F(z)| = |F(\xi + i\nu)| = e^{\pi^2(\xi^2 - \nu^2)} \hat{f}(\xi + i\nu) \leq e^{\pi(\xi^2 - \nu^2)} e^{\pi \nu^2} \leq e^{\pi \xi^2}$

This is not enough to get the bound for F .

2.try: Take $\varepsilon > 0$ small, $\theta > 0$ small, $\delta > 0$ small

$$F_\varepsilon(z) = e^{(i\varepsilon e^{i\varepsilon} z^{2+\varepsilon})} e^{i\delta z^2} e^{\pi z^2} \hat{f}(z)$$

and $\Gamma_\theta = \{z = re^{i\alpha}, 0 \leq \alpha \leq \theta\}$

Claim: $|F_\varepsilon(z)| \leq C$ as the bound of Γ_θ and $|F_\varepsilon(z)| \rightarrow 0$ as $z \in \Gamma_\theta$ and $|z| \rightarrow \infty$

$$|F_\varepsilon(z)| = \underbrace{|e^{(i\varepsilon(\cos(\varepsilon) + i\sin(\varepsilon))r^{2+\varepsilon})} e^{i\delta z^2} e^{\pi z^2} \hat{f}(z)|}_{\substack{-\varepsilon \sin(\varepsilon) r^{2+\varepsilon} \\ \geq 0}} \rightarrow 0 \text{ as } r \rightarrow \infty$$

$$e^{cr^2}$$

Thus, by the maximum principle: $|F_\varepsilon(z)| \leq C$ for all $z \in \Gamma_\theta$ with C independent of ε (need to verify!)

Take $\varepsilon \rightarrow 0$ and $\delta \rightarrow 0 \Rightarrow |e^{\pi z^2} \hat{f}(z)| \leq C$ for all $z \in \Gamma_\theta$, C independent of θ if θ is small and fixed.

By rotations we can increase $\theta \Rightarrow |e^{\pi z^2} \hat{f}(z)| \leq C$

□

Remark. The Sobolev inequality is also a form of the uncertainty principle!

Operators defined by integral kernel:

$$(Kf)(x) = \int_{\mathbb{R}^d} K(x, y) f(y) dy$$

If we take $K = f(p)g(x)$, then this defines a kernel.

$$K\varphi(x) = f(k)(g\varphi)(x) = \mathcal{F}^{-1}(f(k)\widehat{g\varphi}(k))$$

Consequence: If $f, g \in L^2(\mathbb{R}^d)$, then $\int_{\mathbb{R}^d \times \mathbb{R}^d} |K(x, y)|^2 dx dy = \int_{\mathbb{R}^d} |\check{f}(x - y)|^2 |g(y)|^2 dx dy = \|f\|_{L^2}^2 \|g\|_{L^2}^2 < \infty$

$\Rightarrow K$ is a Hilbert Schmidt operator $\Rightarrow K$ is compact, i.e. if $\varphi_n \rightharpoonup \varphi$ weakly in L^2 , then $K\varphi_n \rightarrow K\varphi$ strongly in L^2 .

Theorem 5.33. If $f, g : \mathbb{R}^d \rightarrow \mathbb{C}$ and $f, g \rightarrow 0$ at ∞ , then the operator $K = f(p)g(x)$ is a compact operator in $L^2(\mathbb{R}^d)$

Remark. This implies the Sobolev embedding, i.e.

$$H^1(\mathbb{R}^d) \subset L^{2^*}(\mathbb{R}^d)$$

where $2^* = \frac{2d}{d-2}$

$$\|u\|_{L^{2^*}} \lesssim \|u\|_{H_1}$$

Sobolev compact embedding:

$$\mathbb{1}_\Omega H^1(\mathbb{R}^d) \underset{\text{compact}}{\subset} \mathbb{1}_\Omega L^p(\mathbb{R}^d) \quad \forall \Omega \text{ bd. in } \mathbb{R}^d \quad \forall p < 2^*$$

i.e. $u_n \rightharpoonup u$ weakly in H^1 , then $\mathbb{1}_\Omega u_n \rightarrow \mathbb{1}_\Omega u$ strongly in L^p for $p < 2^*$

Why? $\mathbb{1}_\Omega u_n = (\underbrace{\mathbb{1}_\Omega(x) \frac{1}{\sqrt{p^2+1}}}_{\text{compact operator}}) (\underbrace{\sqrt{p^2+1} u_n}_{\text{bd. } -\sqrt{p^2+1} u}) \rightarrow \mathbb{1}_\Omega u$ in L^2

Example 5.34. For \mathbb{R}^d , let $r \in (0, d]$, then $\frac{1}{,|x|^r} \in S'(\mathbb{R}^d)$

$$|\langle \frac{1}{|x|^r}, \varphi \rangle| \leq \int_{\mathbb{R}^d} \frac{|\varphi(x)|}{|x|^r} dx \leq \int_{\mathbb{B}(0,1)} \frac{|\varphi(x)|}{|x|^r} dx + \int_{\mathbb{R}^d \setminus \mathbb{B}(0,1)} |\varphi(x)| dx$$

Now fix $r = 1, d = 1$. Let $\varphi \in C_c^\infty(\mathbb{R}^d)$. We define the cauchy principal value in the sense of distributions as

$$(p.v. \left(\frac{1}{x} \right))(\varphi) := \lim_{\varepsilon \rightarrow 0} \int_{x>\varepsilon} \frac{\varphi(x) - \varphi(-x)}{x} dx = \int_{\mathbb{R}_+} \frac{\varphi(x) - \varphi(-x)}{x} dx$$

This is well defined, since

$$|\frac{1}{x}(\varphi(x) - \varphi(-x))| \leq \frac{1}{x} \int_{-x}^x |\varphi'(y)| dy \leq 2 \|\varphi\|_{L^\infty}$$

$$\Rightarrow | \langle p.v. \frac{1}{x}, \varphi \rangle | \leq 2 \int_0^1 \|\varphi'\|_{L^\infty} + \int_{\mathbb{R}} |\varphi| \Rightarrow p.v. \in S'(\mathbb{R})$$

We have

$$\begin{aligned} xp.v. \frac{1}{x} &= 1 \\ \widehat{xp.v. \frac{1}{x}} &= \hat{1} = \delta_0 \\ \widehat{\partial p.v. \frac{1}{x}} &= -2\pi i x \widehat{p.v. \frac{1}{x}} = -2\pi i \delta_0 \end{aligned}$$

Further, we know that $\partial H = \delta_0$, where $H = \begin{cases} 1, & x > 0 \\ 0, & \text{else} \end{cases}$ is the Heaviside function. Thus :

$$\widehat{\partial(p.v. \frac{1}{x} + 2\pi i H)} = 0$$

Lemma 5.35. If $T \in S'(\mathbb{R})$, then

$$\partial T = 0 \Rightarrow T = \text{const.}$$

Proof. Let $\varphi \in S$ s.t. $\int_{\mathbb{R}} \varphi = 0$. Then $x \mapsto \int_{-\infty}^x \varphi(y) dy \in S(\mathbb{R})$

$$\Rightarrow 0 = \langle T, \varphi \rangle = \langle T, \partial \int_{-\infty}^x \varphi(y) dy \rangle = - \langle \partial, \int_{-\infty}^x \varphi(y) dy \rangle$$

Let $\psi \in S$

$$\begin{aligned} 0 &= \langle T, \psi - \frac{\int e^{-x^2} \int \psi(y) dy}{\int e^{-y^2} dy} \rangle = \langle T, \psi \rangle - \frac{\int \psi(y) dy}{\int e^{-y^2} dy} \langle T, e^{-x^2} \rangle \\ &\Rightarrow \langle T, \psi \rangle = \frac{\int \langle T, e^{-x^2} \rangle \psi(y) dy}{\int e^{-y^2} dy} \end{aligned}$$

□

$$\widehat{p.v. \frac{1}{x}} = -\partial \pi i H + C$$

Let φ be even $\langle \widehat{p.v. \frac{1}{x}}, \check{\varphi} \rangle = \langle p.v. \frac{1}{x}, \varphi \rangle = 0$

If ψ is even, then $\check{\psi}$ and $\hat{\psi}$ are even,

$$\langle \widehat{p.v. \frac{1}{x}}, \psi \rangle = \langle p.v. \frac{1}{x}, \check{\psi} \rangle = 0 = \int_{\mathbb{R}} (-2\pi i H + C) \varphi$$

and $-2\pi i H + C$ is odd, where $C = \pi i$

$$\widehat{p.v. \frac{1}{x}} = -i\pi sign(\xi)$$

Remark. This is important for defining the Hilbert transform.

$$H\varphi = \varphi \star p.v. \frac{1}{x}$$

5.4 Convolutions

We want to define convolution for distributions. Let $\nu, \varphi, \psi \in S$

$$\begin{aligned} \int (\psi \star \nu)(x) \varphi(x) dx &= \int \int \psi(x-y) \nu(y) dy \varphi(x) dx = \int \int \psi^\#(y-x) \nu(y) dy \varphi(x) dx \\ &= \int \nu(y) \psi^\# \star \varphi(y) dy \end{aligned}$$

where $\psi^\#(x) = \psi(-x)$. If $\alpha \in \mathbb{N}_0^d$, $\partial^\alpha(\varphi \star \psi) = \partial^\alpha \int \varphi(x-y) \psi(y) dy = (\partial^\alpha \varphi) \star \psi = \varphi \star (\partial^\alpha \psi)$

Remark. If $\varphi \in S$, then $\psi \mapsto \varphi \star \psi$ is continuous and linear in $S \rightarrow S$

Definition 5.36. Let $T \in S'(\mathbb{R}^d)$ and $\varphi \in S(\mathbb{R}^d)$, then

$$\langle \varphi \star T, \psi \rangle := \langle T, \varphi^\# \star \psi \rangle$$

for all $\psi \in S$, where $\varphi^\# = \varphi(-x)$

Proposition 5.37. Let $T \in S'$, $\varphi \in S$

1. $\partial^\alpha(\varphi \star T) = \partial^\alpha(\varphi \star T) = \partial^\alpha \varphi \star T = \varphi \star \partial^\alpha T$
2. $\widehat{\varphi \star T} = \hat{\varphi} \hat{T}$

Proof. 1.

$$\begin{aligned} \langle \partial^\alpha(\varphi \star T), \psi \rangle &= (-1)^{|\alpha|} \langle \varphi \star T, \partial^\alpha \psi \rangle \\ &= (-1)^{|\alpha|} \langle T, \varphi^\# \star \partial^\alpha \psi \rangle \\ &= (-1)^{|\alpha|} \langle T, \partial^\alpha \varphi^\# \star \psi \rangle \\ &= (-1)^{|\alpha|} \langle \partial^\alpha \varphi \star T, \psi \rangle \end{aligned}$$

The other equality works similarly.

2.

$$\langle \widehat{\varphi \star T}, \psi \rangle = \langle \varphi \star T, \hat{\psi} \rangle = \langle T, \varphi^\# \star \hat{\psi} \rangle$$

$$\begin{aligned} (\varphi^\# \star \hat{\psi})(x) &= \int \varphi(y-x) \hat{\psi}(y) dy \\ &= \int \varphi(y-x) \int \psi(z) e^{-2\pi i yz} dz dy \\ &= \int \int \varphi(y) \psi(z) e^{-2\pi i yz} e^{-2\pi i xz} dz dy \\ &= \int \hat{\varphi}(z) \psi(z) e^{-2\pi i xz} dz \end{aligned}$$

$$\langle T, \mathcal{F}(\hat{\varphi}\psi) \rangle = \langle \hat{T}, \hat{\varphi}\psi \rangle$$

□

Theorem 5.38. Let $T \in S'(\mathbb{R}^d)$, $\varphi \in S(\mathbb{R}^d)$, then

$$\varphi \star T \in \mathcal{O}(\mathbb{R}^d)$$

where

$$\mathcal{O}(\mathbb{R}^d) = \{g \in C^\infty : \forall \beta \in \mathbb{N}_0^d \ C_\beta(1+|x|)^{m_\beta}\}$$

Remark.

$$\tau_y : S(\mathbb{R}^d) \rightarrow S(\mathbb{R}^d)$$

is clearly a homeomorphism of S , so we can use duality to find

$$\langle \tau_y T, \phi \rangle = \langle T, \tau_y \phi \rangle$$

for $T \in S'$

Proof.

$$\begin{aligned} \langle \varphi \star T, \phi \rangle &= \langle T, \varphi^\# \star \psi(x) \rangle \\ &= \langle T, \int \varphi^\#(x-y) \psi(y) dy \rangle \\ &= \langle T, \int \tau_y \varphi^\#(x) \psi(y) dy \rangle \\ &= \int \langle T, \tau_y \varphi^\# \rangle \psi(y) dy \\ &= \int \langle T, \tau_y \varphi^\# \rangle \psi(y) dy \end{aligned}$$

So we identify $(\varphi \star T)(x) = \langle T, \tau_x \varphi^\# \rangle$

$\varphi \star T \in C^\infty$: Let $e_k = (0, \dots, 0, \underbrace{1}_{k-th}, 0, \dots, 0)$

$$\frac{\tau_{-he_k}(\varphi \star T)(x) - (\varphi \star T)(x)}{h} = \langle T, \frac{\tau_{-he_k}(\tau_x \varphi^\#) - (\tau_x \varphi^\#)}{h} \rangle$$

As $h \rightarrow 0$ we get $\langle T, \tau_x \partial^{e_k} \varphi^\# \rangle$. Repeat for higher order. $\Rightarrow \varphi \star T \in C^\infty$

$$\begin{aligned} |\partial^\gamma(\varphi \star T)(x)| &\leq C \sum_{|\alpha|, |\beta| \leq K} \|y^\alpha \tau_x \partial^{\beta+\gamma} \varphi^\# \|_{L^\infty} \\ &= C \sum_{|\alpha|, |\beta| \leq K} \||x+y|^\alpha \partial^{\beta+\gamma} \varphi^\# \|_{L^\infty} \\ &\leq \tilde{C} \sum_{|\beta| \leq K} \|(1+|x|^K + |y|^K) \partial^{\beta+\gamma} \varphi^\# \|_{L^\infty} \end{aligned}$$

So $\varphi \star T \in \mathcal{O}(\mathbb{R}^d)$

□

Corollary 5.39. If $T \in \mathcal{E}'(\mathbb{R}^d) = \{U \in S'(\mathbb{R}^d) : \text{supp } U \text{ is compact}\}$, then for any $\varphi \in S$: $\varphi \star T \in S(\mathbb{R}^d)$

Proof. Let $\varphi \in S$ be 1 in $\text{supp } T$, then $\varphi T = T$ (exercise).

Thus $\hat{T} = \widehat{\varphi T} = \widehat{\varphi} \star \hat{T} \mathcal{O}(\mathbb{R}^d)$ by thm 5.38. Since for all $\psi \in S$ and $g \in \mathcal{O}(\mathbb{R}^d)$ we have $g\psi \in S$ and $\widehat{\psi \star T} = \widehat{\psi} \hat{T} \in S(\mathbb{R}^d)$

□

Definition 5.40. Let $T \in S'(\mathbb{R}^d)$, $U \in \mathcal{E}'$.

$$\langle U \star T, \varphi \rangle := \langle T, \varphi U^\# \rangle$$

where $\langle U^\#, \varphi \rangle := \langle U, \varphi^\# \rangle$

Exercise. Let $T \in S'$, $U \in \mathcal{E}'$, then

$$\widehat{U \star T} = \hat{U} \hat{T}$$

Chapter 6

Applications of Fourier transform

6.1 Distribution Theory and PDEs

Let $p(\partial) = \sum_{|\alpha| \leq K} c_\alpha \partial^\alpha$, then we call $T \in S'$ a fundamental solution, if $p(\partial)T = \delta_0$. If we want to solve for $f \in S$

$$p(\partial)u = f \quad (*)$$

then

$$u = f \star T$$

solves (*):

$$p(\partial)(f \star T) = f \star (p(\partial)T) = f \star \delta_0 = f$$

In particular, $m = \frac{1}{4\pi^2|x|^2} \in S'$.

Let $T = \mathcal{F}^{-1}m$

$$\begin{aligned} \langle -\widehat{\Delta T}, \varphi \rangle &= \langle 4\pi|\xi|^2\hat{T}, \varphi \rangle \\ &= \langle \hat{T}, 4\pi^2|\xi|^2\varphi(\xi) \rangle \\ &= \int \frac{4\pi^2|\xi|^2\varphi(\xi)}{4\pi^2|\xi|^2} d\xi \\ &= \int \varphi = \langle 1, \varphi \rangle \end{aligned}$$

Thus, $-\widehat{\Delta T} = 1$, so $-\Delta T = \delta_0$

6.1.1 Poisson equation

$$\begin{aligned} -\Delta u &= f \text{ in } \mathbb{R}^d \\ \Rightarrow u(x) &= (G \star f)(x) \end{aligned}$$

$$\text{with } G = \begin{cases} \frac{c}{|x|^{d-2}}, & \text{if } d \geq 3 \\ c \log |x|, & \text{if } d = 2 \\ c|x|, & \text{if } d = 1 \end{cases}$$

Remark. It is not trivial to see how to transfer the regularity of f to the regularity of u , i.e. "Which properties should f have, s.t. we get that u is a classical solution?"

Theorem 6.1 (Regularity of Poisson equation). Assume f is compactly supported in \mathbb{R}^d .

1. If $f \in L^p(\mathbb{R}^d)$ with $p > \frac{d}{2}$, then $u \in C^{0,\alpha}(\mathbb{R}^d)$, i.e. u is Hölder continuous:

$$|u(x) - u(y)| \leq C|x - y|^\alpha, \quad \forall x, y \in \mathbb{R}^d$$

2. If $f \in L^p$, $p > d$, then $u \in C^{1,\alpha}$ with $0 < \alpha < 1 - \frac{d}{p}$.

3. If $f \in C^{k,\alpha}$ for $k \geq 0$ and $0 < \alpha < 1$, then $u \in C^{k+2,\alpha}$

Remark. $f \in C^{k,\alpha}$, iff $\|D^\beta f\|_\infty < \infty$ for all $|\beta| \leq k$ and $\|D^\beta f\|_{C^{0,\alpha}} < \infty$ for all $|\beta| = k$

Proof. 1) $f \in L^p$, $p > \frac{d}{2}$, $u = G \star f$, $d \geq 3$, $G \sim \frac{1}{|x|^{d-2}}$

$$|u(x) - u(y)| = \left| \int_{\mathbb{R}^d} (G(x-y) - G(z-y))f(y)dy \right| \lesssim \int_{\mathbb{R}^d} \left(\frac{1}{|x-y|^{d-2}} - \frac{1}{|z-y|^{d-2}} \right) |f(y)| dy$$

By the triangle inequality:

$$\begin{aligned} \left| \frac{1}{|x-y|^{d-2}} - \frac{1}{|z-y|^{d-2}} \right| &\leq C \frac{\||x-y|^{d-2} - |z-y|^{d-2}\|}{|x-y|^{d-2}|z-y|^{d-2}} \\ &\leq C \frac{\||x-y| - |z-y|\|}{|x-y|^{d-2}|z-y|^{d-2}} (|x-y|^{d-3} - |z-y|^{d-3}) \leq C \frac{|x-z|}{|x-y|^{d-2}|z-y|^{d-2}} (|x-y|^{d-3} - |z-y|^{d-3}) \end{aligned}$$

First try:

$$\begin{aligned} |u(x - u(z))| &\leq C|x - z| \int_{\mathbb{R}^d} \left(\frac{1}{|x-y||z-y|^{d-2}} + \frac{1}{|x-y|^{d-2}|z-y|} \right) |f(y)| dy \\ &\leq C|x - z| \int_{\mathbb{R}^d} \left(\frac{1}{|x-y|^{d-1}} + \frac{1}{|z-y|^{d-1}} \right) |f(y)| dy \end{aligned}$$

By Hölder:

$$\int_{\mathbb{R}^d} \frac{1}{|x-y|^{d-1}|f(y)|} dy \leq \left(\int_{\mathbb{R}^d} |f(y)|^p dy \right)^{\frac{1}{p}} \left(\int_{\text{supp } f} \frac{1}{|x-y|^{(d-1)p'}} dy \right)^{\frac{1}{p'}} < \infty$$

if $(d-1)p' < d$, i.e. $1 - \frac{1}{d} < \frac{1}{p'} = 1 - \frac{1}{p} \Leftrightarrow p > d$ with $\frac{1}{p} + \frac{1}{p'} = 1$. But this is not enough!

Second try:

$$||x-y| - |z-y|| \leq \min(|x-z|, |x-y| + |z-y|) \leq |x-z|^\alpha (|x-y|^{1-\alpha} + |z-y|^{1-\alpha})$$

Then:

$$\begin{aligned} |u(x) - u(y)| &\leq C|x - z|^\alpha \int_{\mathbb{R}^d} \underbrace{\left(\frac{|x - y|^{d-2-\alpha} + |z - y|^{d-2-\alpha}}{|x - y|^{d-2}|z - y|^{d-2}} \right)}_{\leq C(\frac{1}{|x-y|^{d-2+\alpha}} + \frac{1}{|z-y|^{d-2+\alpha}})} |f(y)| dy \end{aligned}$$

By Hölder:

$$\int_{\mathbb{R}^d} \frac{1}{|x - y|^{d-2+\alpha}} |f(y)| dy \leq \left(\int |f|^p \right)^{\frac{1}{p}} \left(\int_{\text{supp } f} \frac{1}{|x - y|^{(d-2+\alpha)p'}} dy \right)^{\frac{1}{p'}}$$

where $\frac{1}{p} + \frac{1}{p'} = 1$

The last integral is finite if $(d - 2 + \alpha)p' < d$

$$\Leftrightarrow 1 - \frac{2 - \alpha}{d} = \frac{d - 2 + \alpha}{d} < \frac{1}{p'} = 1 - \frac{1}{p} \Leftrightarrow \frac{2 - \alpha}{d} > \frac{1}{p}$$

Remark: By the same analysis, we can obtain the Sobolev embedding $\underbrace{H^s(\mathbb{R}^d)}_{(1-\Delta)^{\frac{s}{2}} f \in L^2} \subset C^{0,\alpha}(\mathbb{R}^d)$ if $s > \frac{d}{2}$

2) If $f \in L^p$ with $p > d$ using $u = G \star f$

$$\begin{aligned} \Rightarrow \partial_i u(x) &= (\partial_i G) \star f(x) = \int_{\mathbb{R}^d} \underbrace{\partial_i G(x - y)}_{\frac{(x-y)_i}{|x-y|^d}} f(y) dy \\ \Rightarrow |\partial_i u(x) - \partial_i u(z)| &\leq \int_{\mathbb{R}^d} |\partial_i G(x - y) - \partial_i G(z - y)| |f(y)| dy \end{aligned}$$

By the triangle inequality:

$$\begin{aligned} |\partial_i G(x - y) - \partial_i G(z - y)| &\leq \left| \frac{(x - y)_i}{|x - y|^d} - \frac{(z - y)_i}{|z - y|^d} \right| \\ &\leq \left| \frac{x - y}{|x - y|^d} - \frac{z - y}{|z - y|^d} \right| \leq c|x - z|^\alpha \left(\frac{1}{|x - y|^{d-1+\alpha}} - \frac{1}{|z - y|^{d-1+\alpha}} \right) \\ \Rightarrow |\partial_i u(x) - \partial_i u(z)| &\leq c|x - z|^\alpha \int_{\mathbb{R}^d} \left(\frac{1}{|x - y|^{d-1+\alpha}} - \frac{1}{|z - y|^{d-1+\alpha}} \right) |f(y)| dy \end{aligned}$$

Hölder:

$$\int_{\mathbb{R}^d} \frac{1}{|x - y|^{d-1+\alpha}} |f(y)| dy \leq \|f\|_{L^p} \left(\int_{\text{supp } f} \frac{1}{|x - y|^{(d-1+\alpha)p'}} dy \right)^{\frac{1}{p'}}$$

where $\frac{1}{p} + \frac{1}{p'} = 1$. We need $(d - 1 + \alpha)p' < d \Leftrightarrow \alpha < 1 - \frac{d}{p}$

3) Let $f \in C^{0,\alpha}$, then $u = G \star f \in C^{1,\alpha}$ (this only needs $f \in L^p$, $p > d$)
 We will compute $\partial_{i,j}^2 u$. Formally $\partial_{i,j}^2 G \star f$ is ill defined:

$$\begin{aligned}\partial_i G(x) &= \partial\left(\frac{1}{|x|^{d-2}}\right) = -(d-2)\underbrace{\frac{1}{|x|^{d-1}}}_{\frac{x_i}{|x|}} \partial|x| = \frac{-(d-2)x_i}{|x|^d} \\ \partial_{i,j}^2 G(x) &\sim \partial_j\left(\frac{x_i}{|x|^d}\right) = \delta_{i,j}\frac{1}{|x|^d} + x_i \partial_j\left(\frac{1}{|x|^d}\right) = \delta_{i,j}\frac{1}{|x|^d} - \frac{x_i x_j}{|x|^d} = \frac{1}{|x|^d}(\delta_{i,j} - d\frac{x_i}{|x|}\frac{x_j}{|x|}) \\ &\Rightarrow |\partial_{i,j}^2 G(x)| \leq \frac{c}{|x|^d} \notin L^1_{loc}(\mathbb{R}^d)\end{aligned}$$

Take $\varphi \in C_c^\infty(\mathbb{R}^d)$. Then:

$$\begin{aligned}(\partial_{i,j} u)(\varphi) &= - \int_{\mathbb{R}^d} (\partial_i u)(x) \partial_j \varphi(x) dx \sim \int_{\mathbb{R}^d \times \mathbb{R}^d} (\partial_i G)(x-y) f(y) \partial_j \varphi(x) dx dy \\ &= \int_{\mathbb{R}^d} f(y) \int_{\mathbb{R}^d} \partial_j \varphi(x) \partial_i G(x-y) dx = \lim_{\varepsilon \rightarrow 0} \int_{\mathbb{R}^d} f(y) \underbrace{\int_{|x-y| \geq \varepsilon} \partial_j \varphi(x) \partial_i G(x-y) dx}_{(I)} \\ &\quad - \underbrace{\int_{|x-y|=\varepsilon} (\varphi \partial_i G) \nu_j dx}_{(II)} - \underbrace{\int_{|x-y| \geq \varepsilon} (\varphi \partial_{i,j}^2 G) dx}_{(III)} \\ (I) &= c \int_{|x-y|=\varepsilon} \varphi \frac{(x-y)_i}{|x-y|^{d-i}} \frac{(x-y)_j}{|x-y|} = c \int_{|x-y|=\varepsilon} \varphi(x) (x-y)_i (x-y)_j (x-y)_i \varepsilon^{-(d+1)} dx \\ &\quad \rightarrow c \delta_{i,j} \varphi(y) \text{ as } \varepsilon \rightarrow 0 \\ (II) &= \int_{|x-y| \geq 1} \varphi \partial_{i,j}^2 G dx + \int_{1 > |x-y| \geq \varepsilon} (\varphi \partial_{i,j}^2 G) dx\end{aligned}$$

We don't need to worry about the first integral, since $\partial_{i,j} G(x-y)$ is smooth in the $|x-y| \geq 1$ domain.

$$\int_{1 > |x-y| \geq \varepsilon} \varphi \partial_{i,j}^2 G dx = \int_{1 > |x-y| \geq \varepsilon} \underbrace{(\varphi(x) - \varphi(y))}_{c|x-y|} \underbrace{\partial_{i,j}^2 G(x-y)}_{\frac{c}{|x-y|^d}} dx \rightarrow \text{good limit}$$

Thus:

$$\begin{aligned}(\partial_{i,j} u)(\varphi) &= - \int_{\mathbb{R}^d} (\partial_i u)(x) \partial_j \varphi(x) dx \\ &= c \delta_{i,j} \int_{\mathbb{R}^d} f(y) \varphi(y) dy + \int_{|x-y| \geq 1} f(y) \varphi(x) \partial_{i,j}^2 G(x-y) dx dy \\ &\quad + \int_{|x-y| < 1} \underbrace{f(y)(\varphi(x) - \varphi(y))}_{(f(x)) - f(y)} \partial_{i,j}^2 G(x-y) dx dy\end{aligned}$$

$$\Rightarrow (\partial_{i,j} u)(x) = c\delta_{i,j}f(x) + \int_{|x-y|\geq 1} f(y)\partial_{i,j}^2 G(x-y)dy + \int_{|x-y|<1} (f(x)-f(y))\partial_{i,j}^2 G(x-y)dy$$

Now we want to show, that the RHS is $C^{0,\alpha}(dx)$. It is already clear that $f \in C^{0,\alpha}$ and $f \star (\partial_{i,j}^2 G \mathbf{1}_{|x|\geq 1}) \in C^{0,\alpha}$. The main difficulty is in the last term.

To simplify the notation $z = 0$:

$$\begin{aligned} &\rightsquigarrow \int_{|y|<1} \underbrace{\partial_{i,j} G(y)}_{|\cdot|\leq \frac{c}{|y|^d}} \underbrace{(f(x+y) - f(y))}_{|\cdot|\leq C|x|^\alpha} dy \\ &= \int_{4|x|<|y|<1} \partial_{i,j} G(y)(f(x+y) - f(y))dy + \int_{|y|<4|x|} \partial_{i,j} G(y)(f(x+y) - f(y))dy \end{aligned}$$

Second term:

$$\begin{aligned} \left| \int_{|y|<4|x|} \partial_{i,j} G(y)(f(x+y) - f(y))dy \right| &= \left| \int_{|y|<4|x|} \partial_{i,j} G(y)(f(x+y) + f(x) - f(y) + f(0))dy \right| \\ &\leq C \int_{|y|<4|x|} \frac{|y|^\alpha}{|y|^d} dy = C|x|^\alpha \end{aligned}$$

First term:

$$\begin{aligned} &\int_{4|x|<|y|<1} \partial_{i,j} G(y)(f(x+y) - f(y))dy \\ &= \underbrace{\int_{4|x|<|y-x|<1} \partial_{i,j} G(y-x) \underbrace{f(y)}_{f(y)-f(0)} dy}_{A} - \underbrace{\int_{4|x|<|y-x|<1} \partial_{i,j} G(y) \underbrace{f(y)}_{f(y)-f(0)} dy}_{B} \\ &= \int_{A \cap B} (\partial_{i,j} G(y-x) - \partial_{i,j} G(y))(f(y) - f(0))dy \\ &\quad + \int_{A \setminus B} \partial_{i,j} G(y)(f(y) - f(0))dy - \int_{B \setminus A} \partial_{i,j} G(y)(f(y) - f(0))dy \end{aligned}$$

For these three integrals we have:

$$\begin{aligned} \left| \int_{A \cap B} \partial_{i,j} G(y)(f(y) - f(0))dy \right| &\leq \int_{A \cap B} \underbrace{|\partial_{i,j} G(y)|}_{c|x|(\frac{1}{|y-x|^{d+1}} + \frac{1}{|y|^{d+1}})} \underbrace{|f(y) - f(0)|}_{\leq c|y|^\alpha} dy \\ &\leq c|x| \int_{1>|y|>4|x|} \frac{|y|^\alpha}{|y|^{d+1}} dy \leq c|x|(|x|^{\alpha-1}) = c|x|^\alpha \end{aligned}$$

The second:

$$\begin{aligned} \left| \int_{A \setminus B} \partial_{i,j} G(y)(f(y) - f(0))dy \right| &\leq \int_{4|x|<|y-x|<1, |y|\leq 4|x|} |\partial_{i,j} G(y-x)| |f(y) - f(0)| dy \\ &\leq \int_{|y|\leq 4|x|} \frac{|y|^\alpha}{|x|^d} dy \leq C|x|^\alpha \end{aligned}$$

The third:

$$\begin{aligned}
& \left| \int_{B \setminus A} \dots \right| \leq \int_{4|x| < |y| < 1, |x-y| \leq 4|x|} \dots + \int_{4|x| < |y| < 1, |x-y| \leq 4|x|} \dots \\
& \leq \int_{4|x| < |y| < 5|x|} |\partial_{i,j} G(y)| |f(y) - f(0)| dy + \int_{1-|x| < |y| < 1} |\partial_{i,j} G(y)| |f(y) - f(0)| dy \\
& \leq \int_{4|x| < |y| < 5|x|} \frac{|y|^\alpha}{|y|^\alpha} dy + c \int_{[1-|x|, 1]} \frac{r^\alpha}{r^d} r^{d-1} dr \\
& \leq c|x|^\alpha + c'|x|^\alpha \sim |x|^\alpha
\end{aligned}$$

For general k :, e.g $f \in C^{1,\alpha}$ we have

$$\begin{aligned}
-\Delta u = f \Rightarrow -\Delta(\partial_i u) &= \underbrace{(\partial_i f)}_{C^{0,\alpha}} \\
\stackrel{k=0}{\Rightarrow} \partial_i u &\in C^{2,\alpha} \quad \forall i \Rightarrow u \in C^{3,\alpha}
\end{aligned}$$

□

Remark. We need $D^\alpha D^\beta T = D^\beta D^\alpha T$ for all distributions T . This can be proved by
1) seeing that this is true for $\varphi \in C_c^\infty$ instead of a distribution T (Schwarz's theorem)
2) $(D^\alpha D^\beta T)(\varphi) = T(D^\alpha D^\beta \varphi)(-1)^{|\alpha|+|\beta|} = (D^\beta D^\alpha T)(\varphi)$

6.1.2 Heat equation

$$\partial_t u(t, x) = \Delta_x u(t, x), \quad t \geq 0, x \in \mathbb{R}^d$$

By Fourier transform in x :

$$\begin{aligned}
\partial_t \hat{u}(t, x) &= -|2\pi k|^2 \hat{u}(t, k) \Rightarrow \hat{u}(x, t) = e^{-t|2\pi k|^2} \hat{u}(0, k) = e^{-t|2\pi k|^2} \hat{f}(k) \\
u(t, x) &= \int_{\mathbb{R}^d} e^{-t|2\pi k|^2} \hat{f}(k) e^{2\pi i k x} dx = (G_t \star f)(x) \\
&= \frac{1}{(4\pi)^{\frac{d}{2}}} \int_{\mathbb{R}^d} e^{-\frac{|x-y|^2}{4t}} f(y) dy
\end{aligned}$$

where $G_t(x) = \frac{e^{-\frac{|x|^2}{4t}}}{(4\pi t)^{\frac{d}{2}}}$ is the heat kernel $\hat{G}_t(k) = e^{-t|2\pi k|^2}$

Theorem 6.2. If $f \in L^2(\mathbb{R}^d)$, then the solution of the heat equation $u(t, x) = (e^{t\Delta} f)(x) = (G_t \star f)(x)$ satisfies

1. $\lim_{t \rightarrow 0} u(t, x) = f(x)$ in $L^2(\mathbb{R}^d)$
2. $\lim_{t \rightarrow \infty} u(t, x) = 0$ in $L^2(\mathbb{R}^d)$

6.1.3 Schrödinger equation

$$\begin{cases} i\partial_t u(t, x) = \Delta_x u(t, x), & t \geq 0, x \in \mathbb{R}^d \\ u(0, x) = f(x), & x \in \mathbb{R}^d \end{cases}$$

By following the heat equation formally, $t \mapsto it$ (imaginary time)

$$u(t, x) = \frac{1}{(4\pi it)^{\frac{d}{2}}} \int_{\mathbb{R}^d} e^{-\frac{|x-y|^2}{4it}} f(y) dy \quad (*)$$

Theorem 6.3.

1. If $f \in L^1 \cap L^2$, then $u(t, x) = (e^{it\Delta} f)(x)$ given by $(*)$
2. $\forall f \in L^2$, we can define $u(t, x)$ by a limiting argument. We have:
 $\hat{u}(t, k) = e^{-it|2\pi k|^2} \hat{f}(k)$

Proof. First take $f \in L^1 \cap L^2$, then use $(*)$

$$\begin{aligned} \Rightarrow \hat{u}(t, k) &= e^{-t|2\pi k|^2} \hat{f}(k) \\ \Rightarrow \|u(t, \cdot)\|_{L^2} &= \|\hat{u}(t, \cdot)\|_{L^2} = \left\| \hat{f} \right\|_{L^2} = \|f\|_{L^2} \end{aligned}$$

For any $f \in L^2$, find $f_1 \in L_1 \cap L^2$, $f_n \rightarrow f$, then $u_n = e^{-it\Delta} f_n$ well-defined and a cauchy sequence in L^2

$$\|u_m(t) - u_n(t)\|_{L^2(dx)} = \|e^{-it\Delta} (f_n - f_m)\|_{L^2(dx)} = \|f_n - f_m\|_{L^2} \rightarrow 0$$

So $\exists u(t, x) = \lim_{n \rightarrow \infty} u_n(t, x)$ in L^2 \square

Theorem 6.4. Take $f \in L^2(\mathbb{R}^d)$ and $u(t, x) = e^{-it\Delta} f(x)$. Then:

1. $\lim_{t \rightarrow 0} u(t, x) = f(x)$ in L^2
2. $\lim_{t \rightarrow \infty} \int_B |u(t, x)|^2 dx = 0$, for all $B \subset \mathbb{R}^d$ bounded (scattering theory)

Proof. Consider the case $f \in L^1 \cap L^2$. Then:

$$|u(t, x)| = \frac{1}{(4\pi t)^{\frac{d}{2}}} \|f\|_{L^1} \rightarrow 0 \text{ as } t \rightarrow \infty$$

If B bounded, then:

$$\int_B |u(t, x)|^2 dx \leq |B| \|u(t)\|_{L^\infty} \leq \frac{|B| \|f\|_{L^1}}{t^{\frac{d}{2}}} \rightarrow 0 \text{ as } t \rightarrow \infty$$

General case $f \in L^2$: Take $f_n \rightarrow f$ in L^2 , $f_n \in L^1 \cap L^2$.

Define $u_n = e^{-it\Delta} f_n$, then $u_n(t, x) \rightarrow u(t, x)$ in L^2 .

$$\int_B |u(t, x)|^2 dx \leq 2 \int_B |u_n(t, x)|^2 dx + 2 \int_B |(u - u_n)(t, x)|^2 dx \leq \frac{|B| \|f\|_{L^1}}{t^{\frac{d}{2}}} + C \|f_n - f\|_{L^2}$$

Then take $t \rightarrow \infty$ and $n \rightarrow \infty$ \square

6.1.4 Well posedness of PDEs

Stability: small errors of data \rightsquigarrow a small error of equation

Schrödinger equation:

$$\begin{cases} i\partial_t u = \Delta_x u \\ u(0, x) = f(x) \end{cases} \Rightarrow \left\| \underbrace{e^{-it\Delta} f}_u - e^{-it\Delta} f_\varepsilon \right\|_{L^2} = \|f - f_\varepsilon\|$$

which is exactly what we want.

Heat equation:

$$\begin{cases} \partial_t u = \Delta_x u \\ u(0, x) = f(x) \end{cases} \Rightarrow \|e^{t\Delta} f - e^{t\Delta} f_\varepsilon\|_{L^2} = \|f - f_\varepsilon\| \quad (\text{exercise})$$

Heat equation:

$$\begin{cases} \partial_t u = \Delta_x u \\ u(T, x) = f(x) \end{cases} \Rightarrow \left\| \underbrace{u(0, x)}_{e^{-T\Delta} f} - e^{t\Delta} f_\varepsilon \right\|_{L^2} \xrightarrow{\text{may}} \infty$$

even if $\|f_\varepsilon - f\| \rightarrow 0$. This is because

$$\left\| \underbrace{u(0, x)}_{e^{-T\Delta} f} - e^{t\Delta} f_\varepsilon \right\|_{L^2} = \int_{\mathbb{R}^d} e^{2T|2\pi k|^2} |\hat{f}(k) - \hat{f}_\varepsilon(k)|^2 >> \int |\hat{f}(k) - \hat{f}_\varepsilon(k)|^2 = \|f_\varepsilon - f\|_{L^2}^2$$

Theorem 6.5. Let $f \in L^2$ s.t. $e^{-T\Delta} f \in L^2$, i.e. $e^{T|2\pi k|^2} \hat{f}(k) \in L^2$ for $T > 0$. Let u be the solution of the backward heat equation. Given $f_\varepsilon \in L^2$, $\|f_\varepsilon - f\|_{L^2} \leq \varepsilon$. Define $\hat{u}_\varepsilon(0, k) = e^{T|2\pi k|^2} \hat{f}_\varepsilon(k) \mathbf{1}_{|k| \leq \delta_\varepsilon}$ with $\delta_\varepsilon := |\log \varepsilon|^{\frac{1}{4}}$. Then:

$$\|u_\varepsilon(0, x) - u(0, x)\|_{L^2} \xrightarrow{\varepsilon \rightarrow 0} 0$$

6.2 Oscillatory Integrals

Definition 6.6.

$$I(t) = \int_{\mathbb{R}^d} e^{it\varphi(x)} f(x) dx$$

is called an oscillatory integral, where $f \in L^1(\mathbb{R}^d)$ complex valued and φ real valued.

Q: How does $I(t)$ decay as $t \rightarrow \infty$?

Fourier $\rightsquigarrow t\varphi(x) = 2\pi kx$, $k \in \mathbb{R}^d$. But here we consider a general phase.

Theorem 6.7 (φ has no critical point). Assume $f \in C_c^\infty(\mathbb{R}^d)$, $\varphi \in C^\infty$, $\nabla\varphi(x) \neq 0 \forall x \in \text{supp } f$, then

$$|I(t)| = \left| \int_{\mathbb{R}^d} e^{it\varphi(x)} f(x) dx \right| \leq \frac{C_N}{|t|^N} \text{ as } |t| \rightarrow \infty$$

for all $N \geq 1$

Proof. Step 1: In 1D: $\varphi'(x) \neq 0$ for all $x \in \text{supp } f \Rightarrow \varphi$ is monotone in $\text{supp } f$. We can change the variable in $u = \varphi(x)$.

$$\begin{aligned} & \Rightarrow dx = d(\varphi^{-1}(\varphi(x))) = (\varphi^{-1})'(u)du \\ I(t) &= \int_{\mathbb{R}^d} e^{it\varphi(x)} f(x) dx = \int_{\mathbb{R}^d} \underbrace{f(\varphi^{-1}(u))(\varphi^{-1})'(u)}_{\in C_c^\infty} du \end{aligned}$$

\rightsquigarrow usual Fourier transform.

$$\Rightarrow |I(t)| \leq \frac{C_N}{|t|^N} \forall N \geq 1$$

as $I(t)$ is Schwartz.

Step 2: ($d \geq 1$) Assume $\partial_{x_i}\varphi(x) \neq 0$ for all $x \in \text{supp } f$, then

$$I(t) = \int_{\mathbb{R}^d} e^{it\varphi(x)} f(x) dx = \int_{\mathbb{R}^{d-1}} \underbrace{\left(\int_{\mathbb{R}} e^{it\varphi(x)} f(x) dx_1 \right) dx_2 \dots dx_d}_{|\cdot| \leq \frac{C_N(x_2, \dots, x_d)}{|t|^N}}$$

since $C_N(x_2, \dots, x_d)$ is nice enough:

$$\Rightarrow |I(t)| \leq \frac{\tilde{C}_N}{|t|^N}$$

In general, we only assume $\nabla\varphi(x) \neq 0 \Leftrightarrow (\partial_{x_1}\varphi, \dots, \partial_{x_d}\varphi) \neq 0$. We can find a partition of unity $1 = \sum_{k=1}^d \chi_k$, $\chi_k \in C^\infty$, $0 \leq \chi_k \leq 1$ s.t. $\partial_{x_k}\varphi \neq 0$ on $\text{supp } f\chi_k$

$$\Rightarrow |I(t)| \leq \sum_k = 1^d \left| \int e^{it\varphi(x)} f(x) \chi_k(x) dx \right| \leq \frac{C_N}{|t|^N} \quad \forall N \geq 1$$

□

Remark (geometric method). Assume $\Omega_1, \dots, \Omega_l \subset \mathbb{R}^d$ open, $\bigcup \Omega_j = \mathbb{R}^d$, then for all Ω_j we can find a function

$$a_i : \mathbb{R}^d \rightarrow [0, 1], \text{ s.t. } a_i \simeq \mathbb{1}_{\Omega_i} \begin{cases} a_i(x) = 1 \text{ if } x \in \Omega_i \\ a_i(x) = 0 \text{ if } \text{dist}(x, \Omega_i) \geq \varepsilon \end{cases} \quad (\text{Urgushen lemma})$$

$$\rightsquigarrow \chi_i = \frac{a_i}{\sum_j a_j}$$

$$\Rightarrow \sum \chi_i = 1, \text{ supp } \chi_i = \text{supp } a_i, \quad \sum_j a_j > 0 \text{ since } \bigcup_i \Omega_i = \mathbb{R}^d$$

Problem: We will see the oscillatory integral with φ having critical points!

Remark. 1D: If we bound $|\int_a^b e^{it\varphi(x)} dx| \lesssim \frac{c}{|t|^\alpha}$, then

$$\left| \int_a^b \underbrace{e^{it\varphi(x)}}_{F'} f(x) dx \right| \stackrel{\text{Ibp}}{=} |[Ff]_a^b - \int_a^b F(x) f'(x) dx| \leq \|F\|_{L^\infty} (\|f\|_{L^\infty} + \int_a^b |f'|)$$

Theorem 6.8. (1D, Van der Corput)

(a) Take $a < b$ real $\varphi \in C^k$ for some $k \geq 2$, $|\varphi^{(k)}(t)| \geq 1$. Then

$$\left| \int_a^b e^{it\varphi(x)} dx \right| \leq \frac{Ck}{|t|^{\frac{1}{k}}}$$

(b) If $\varphi \in C^1$, φ' monotone and $|\varphi'| \geq 1$, then

$$\left| \int_a^b e^{it\varphi(x)} dx \right| \leq \frac{C}{|t|}$$

Example 6.9 (Bessel function).

$$\hat{1}_{[-1,1]}(t) \rightsquigarrow \left| \int_a^b e^{it\cos(x)} \cos(x) dx \right| \leq \frac{c}{|t|^{\frac{1}{2}}} \text{ as } t \rightarrow \infty$$

Here $\varphi(x) = \cos(x)$. $\varphi'(x) = -\sin(x)$, $\varphi''(x) = -\cos(x)$. We can divide $[0, \pi]$ to sub-intervals where either $|\varphi'| \geq \frac{1}{2}$ and φ' monotone or $|\varphi''| \geq \frac{1}{2}$. In our proof of the Gauss circle problem bound we used this decay of the Bessel function.

Proof. (b) Assume $\varphi' \geq 1$ in $[a, b]$ and φ' monotone.

$$\begin{aligned} \left| \int_a^b e^{it\varphi(x)} dx \right| &= \left| \int_a^b \frac{(e^{it\varphi(x)})'}{it\varphi'(x)} dx \right| = \frac{1}{t} \left| - \int_a^b \left(e^{it\varphi} \left(\frac{1}{\varphi'} \right)' + \left[\frac{e^{it\varphi}}{\varphi'} \right]_a^b \right) dx \right| \\ &\leq \frac{1}{t} \left| \frac{1}{\varphi'(a)} + \frac{1}{\varphi'(b)} \right| + \underbrace{\int_a^b \left| \left(\frac{1}{\varphi'} \right)' \right| dx}_{\stackrel{(*)}{=} \int_a^b \left(\frac{1}{\varphi'} \right)' dx = \left| \frac{1}{\varphi'(b)} - \frac{1}{\varphi'(a)} \right|} \leq \frac{4}{t} \end{aligned}$$

In $(*)$ we used that φ' monotone $\Rightarrow \frac{1}{\varphi'}$ monotone $\Rightarrow (\frac{1}{\varphi'})'$ either ≥ 0 on $[a, b]$ or ≤ 0 on $[a, b]$

For (a) we need some new tools! We will prove (a) later. □

Remark. Without the assumption φ' monotone, this fails!

Lemma 6.10. $\forall k \geq 1, \forall \{a_m\}_{m=0}^k \subset [a, b]$ distinct, $\forall f \in C^k([a, b])$ real valued. Then $\exists y \in [a, b]$ s.t. $f^{(k)}(y) = \sum_{m=0}^k c_m f(a_m)$ where $c_m = (-1)^k k! \prod_{l \in \{0, \dots, k\} \setminus \{m\}} \frac{1}{a_l - a_m}$

Proof. Induction:

$k=1$: $f \in C^1, a_0 < a_1, \exists y \in (a_0, a_1)$ s.t.

$$f'(y) = \frac{f(a_1) - f(a_0)}{a_1 - a_0}, \quad c_0 = \frac{-1}{a_1 - a_0}, \quad c_1 = \frac{1}{a_1 - a_0}$$

\rightsquigarrow Rolle's theorem

$k \geq 1$: We will try to apply Rolle's theorem many times. More precisely, we find a polynomial $p(x) = \sum_{m=0}^k b_m x^m$ s.t. the function $F(x) = f(x) - p(x)$ has zeros $\{a_m\}_{m=0}^k$, i.e. $F(a_0) = F(a_1) = \dots = F(a_k) = 0$.

$\exists a_0 < b_0 < a_1 < b_1 < \dots < b_{k-1} < a_k$ s.t. $F'(b_0) = F'(b_1) = \dots = F'(b_{k-1}) = 0$

$$\Rightarrow \exists y \in [a, b] \text{ s.t. } F^{(k)}(y) = 0$$

$$\Rightarrow f^{(k)}(y) - k!b_k = 0$$

The claim follows from the fact that $k!b_k = \sum_{m=0}^k c_m f(a_m)$. How can we find $p(x) = \sum_{m=0}^k b_m x^m$? We want $\underbrace{p(a_l)}_{\sum_{m=0}^k b_m a_l^m} = f(a_l)$ for all $l = 0, \dots, k$.

$$\Leftrightarrow \begin{pmatrix} 1 & a_0^1 & \dots & a_0^k \\ \vdots & \ddots & \vdots \\ 1 & a_k^1 & \dots & a_k^k \end{pmatrix} \begin{pmatrix} b_0 \\ \vdots \\ b_k \end{pmatrix} = \begin{pmatrix} f(a_0) \\ \vdots \\ f(a_k) \end{pmatrix}$$

determine the $\{b_l\}$ uniquely.

$$\det(\cdot) = \text{Van der monde determinant} = \prod_{l=0}^{k-1} \prod_{j=l+1}^k (a_l - a_j) \neq 0$$

$$\Rightarrow b_m = \sum_{m=0}^k (-1)^m f(a_m) \prod_{l \in \{0, \dots, k\} \setminus \{m\}} \frac{1}{a_l - a_m} (-1)^{k-m}$$

□

Lemma 6.11. Let $E \subset \mathbb{R}$ measurable, $0 < |E| < \infty$, then $\forall k \geq 1$, $\exists \{a_m\}_{m=0}^k \subset E$ s.t.

$$\prod_{l \neq j} |a_l - a_j| \geq \left(\frac{|E|}{2e} \right)^k, \quad e = 2.7\dots$$

Proof. (exercise) □

Lemma 6.12 (Controlling of sub-level set). $\forall k \geq 1$, $\forall \varphi \in C^k$ real-valued, $|\varphi^{(k)}(t)| \geq 1$. Then

$$|\{t : |\varphi(t)| \leq \alpha\}| \leq Ck\alpha^{\frac{1}{k}}$$

Proof. By the previous lemma $E = \{|\varphi| \leq \alpha\}$, $\exists \{a_m\}_{m=0}^k \subset E$ s.t.

$$\frac{|E|^k}{(2\alpha)^k} \leq \prod_{l \neq j} |a_l - a_j|$$

By lemma 6.10, $\exists y : \varphi^{(k)}(y) = (-1)^k k! \sum_{m=0}^k \varphi(a_m) \prod_{l \neq j} |a_l - a_j|$

$$\Rightarrow 1 \leq |\varphi^{(k)}| \leq k! \alpha \prod_{l \neq j} |a_l - a_j| \leq k! \alpha \frac{(2e)^k}{|E|^k}$$

$$\Rightarrow |E| \leq (k! \alpha (2e)^k)^{\frac{1}{k}} \leq C k \alpha^{\frac{1}{k}}$$

□

proof of theorem 6.8 (a). $|\varphi^{(k)}(t)| \geq 1$. We consider

$$R_1 = \{t : |\varphi'(t)| \geq \alpha\}$$

$$R_2 = \{t : |\varphi'(t)| \geq \alpha\}$$

For R_2 : $|R_2| \leq C k \alpha^{\frac{1}{k-1}}$ by previous sub-level set lemma.

For R_1 : $\varphi^{(k)} \neq 0$ on $(a, b) \Rightarrow \varphi^{(2)}$ has at most $2k$ zeros.

$\Rightarrow R_1 =$ union of at most k intervals where $\varphi^{(2)} \neq 0$ in each interval $\Rightarrow \varphi'$ is monotone in each interval.

Consider 1 interval (c, d) , $|\varphi'| \geq 0$ on (c, d) and φ' is monotone in (c, d) . (b) $\Rightarrow |\int_c^d e^{it\varphi(x)} dx| \leq \frac{c}{|t|\alpha}$

$$\stackrel{\text{triangle inequality}}{\Rightarrow} |\int_{R_1} e^{it\varphi(x)} dx| \leq \frac{Ck}{|t|\alpha}.$$

Conclusion:

$$\begin{aligned} |\int_a^b e^{it\varphi(x)} dx| &\leq |R_2| + |\int_{R_1} e^{it\varphi(x)} dx| \\ C k \alpha^{\frac{1}{k-1}} + \frac{Ck}{|t|\alpha} &= C k (\alpha^{\frac{1}{k-1}} + \frac{1}{|t|\alpha}) \quad \forall \alpha \\ \stackrel{\text{optimize}}{\Rightarrow} \min_{\alpha>0} (\alpha^{\frac{1}{k-1}} + \frac{1}{|t|\alpha}) &\lesssim \frac{1}{|t|^{\frac{1}{k}}} \end{aligned}$$

□

6.3 Number Theory

Recall that for the Gauss circle problem we used:

Lemma 6.13 (Poisson summation formula). Let $f \in S(\mathbb{R}^d)$, then

$$\sum_{x \in \mathbb{Z}^d} f(x) = \sum_{k \in \mathbb{Z}^d} \hat{f}(k)$$

Lemma 6.14 (Bessel).

$$|\int_0^\pi e^{it \cos(\theta)} \cos(\theta) d\theta| \lesssim \frac{1}{|t|^{\frac{1}{2}}} \text{ as } t \rightarrow \infty$$

Question: What is the distribution of the first digit of 2^m ?

$$2^0 = \underline{1}, \quad 2^1 = \underline{2}, \quad 2^2 = \underline{4}, \quad 2^3 = \underline{8}, \quad 2^4 = \underline{16}, \quad 2^5 = \underline{32}, \quad \dots$$

One would expect that all digits are equally distributed. In reality, e.g. 1 appears more than any other digit!

Theorem 6.15. Define $\{1, \dots, 9\} \ni \alpha(n) = \text{first digit of } n$, then

$$\frac{\#\{1 \leq k \leq N : \alpha(2^k) = m\}}{N} \rightarrow \log_{10}(1 + \frac{1}{m})$$

for all $m \in \{1, \dots, 9\}$

Theorem 6.16. If $\alpha \in \mathbb{R}$ is irrational, then $\{k\alpha - \lfloor k\alpha \rfloor\}_{k=1}^\infty$ is equidistributed in $[0, 1)$, i.e.

$$\frac{1}{N} \#\{1 \leq k \leq N : k\alpha - \lfloor k\alpha \rfloor \in Q\} \rightarrow |Q|$$

for all $Q \subset [0, 1)$ measurable.

Theorem 6.15 is a consequence of this:

Proof. $\alpha(2^k) = m \Leftrightarrow m10^s \leq 2^k < (m+1)10^s$ for some $s \in \mathbb{N}$.

$$\begin{aligned} 2^k &= \underbrace{m \dots}_{s \text{ digits}} \Leftrightarrow s + \log_{10}(m) \leq k \log_{10} 2 < s + \log_{10}(m+1) \\ &\Rightarrow \log_{10}(m) \leq k \log_{10}(2) - \lfloor k \log_{10}(2) \rfloor \end{aligned}$$

$\forall m \in \{1, \dots, 9\}$, we have that $\alpha = \log_{10} 2$ is irrational and

$$\frac{1}{N} \#\{1 \leq k \leq N : k\alpha - \lfloor k\alpha \rfloor \in [\log_{10}(m), \log_{10}(m+1)]\} \rightarrow \log_{10}(1 + \frac{1}{m})$$

□

Theorem 6.17. Take $\{a_k\}_{k=1}^\infty \subset \underbrace{\mathbb{T}^d}_{\text{torus}}$, d fixed. Then TFAE:

(a) Take $\{a_k\}$ is equidistributed, namely

$$\frac{1}{N} \#\{k \leq N : a_k \in Q\} \rightarrow |Q|, \quad \forall Q \subset \mathbb{T}^d \text{ measurable}$$

(b) $\forall f$ smooth $\frac{1}{N} \sum_{k=1}^N f(a_k) \rightarrow \int_{\mathbb{T}^d} f(x) dx$

$$(c) \forall m \in \mathbb{Z}^d \setminus \{0\}, \frac{1}{N} \sum_{k=1}^N e^{2\pi i m a_k} \rightarrow 0$$

Remark. To us the torus is simply $\mathbb{T}^d = [0, 1]^d$

Proof. (a) \Rightarrow (b): We can conclude $(a) \Rightarrow (b)$ for smooth f by approximating it with step functions and then using a density argument.

(b) \Rightarrow (a): We will conclude this by using the idea from the Gauss circle problem. We can find 2 smooth functions g, h s.t. $0 \leq g(x) \leq \mathbb{1}_Q(x) \leq 1$ of every x . Then:

$$\underbrace{\frac{1}{N} \sum_{k=1}^N g(a_k)}_{\rightarrow \int_{\mathbb{T}^d} g(x) dx} \leq \underbrace{\frac{1}{N} \sum_{k=1}^N \mathbb{1}_Q(a_k)}_{\rightarrow \int_{\mathbb{T}^d} f(x) dx} \leq \underbrace{\frac{1}{N} \sum_{k=1}^N h(a_k)}_{\rightarrow \int_{\mathbb{T}^d} h(x) dx}$$

(b) \Rightarrow (c): Use (b) for $x \mapsto e^{2\pi i mx} = f(x)$

$$\frac{1}{N} \sum_{k=1}^N e^{2\pi i m a_k} = \frac{1}{N} \sum_{k=1}^N f(a_k) \rightarrow \int f(x) dx = \int_{\mathbb{T}^d} e^{2\pi i mx} dx = \begin{cases} 1, & \text{if } m = 0 \\ 0, & \text{if } m \in \mathbb{Z}^d \setminus \{0\} \end{cases}$$

(c) \Rightarrow (b):

$$\begin{aligned} \frac{1}{N} \sum_{k=1}^N f(a_k) &= \frac{1}{N} \sum_{k=1}^N \left(\sum_{l \in \mathbb{Z}^d} \hat{f}(l) e^{2\pi i l a_k} \right) = \frac{1}{N} \sum_l \sum_k = \hat{f}(0) + \sum_{l \neq 0} \underbrace{\left(\frac{1}{N} \sum_{k=1}^N e^{2\pi i l a_k} \right)}_{\rightarrow 0, \text{ as } N \rightarrow \infty} \hat{f}(l) \\ &\rightarrow \hat{f}(0) = \int f(x) dx \end{aligned}$$

□

Exercise. Use this thm to prove $k\alpha - \lfloor k\alpha \rfloor$ is equidistributed.

Riemann Zeta function:

$$\zeta(s) := \sum_{k=1}^{\infty} \frac{1}{k^s}, \quad \forall s > 1$$

Basel problem: $\zeta(s)$ is transcendental $\forall s \in \mathbb{N}, s > 1$, i.e. $\zeta(s)$ is not a root of any polynomial of rational coefficients.

Theorem 6.18 (Euler). The Basel problem is true if $s \in \mathbb{N}$ is even.

Proof. Consider B-polynomial $\{B_m\}_{m=0}^{\infty}$, $B_0 = 1$, $B'_m(x) = mB_{m-1}(x)$, $\int_0^1 B_m(x)dx = 0$.
 Actually: $B_m(x) = -m! \sum_{k \in \mathbb{Z} \setminus \{0\}} \frac{e^{2\pi i k x}}{(2\pi i k)^m}$, $\forall m \geq 2$ the series converges absolutely.

$$B'_m(x) = -m! \sum_{k \in \mathbb{Z} \setminus \{0\}} \frac{e^{2\pi i k x}}{(2\pi i k)^{m-1}} = mB_{m-1}(x)$$

(for $m = 1$ the series converges conditionally)

Application: If m is even, $x = 0$, then:

$$B_m(0) = \left(\underbrace{c_m}_{\in \mathbb{Q}} \sum_{k=1}^{\infty} \frac{\cos(2\pi k x)}{(2\pi k)^m} \right) \Big|_{x=0} = c_m \sum_{k=1}^{\infty} \frac{1}{(2\pi k)^m} = \frac{c_m}{(2\pi)^m} \zeta(m)$$

Conclusion follows from $B_m(0)$ being rational! Since $B'_m(x) = mB_{m-1}(x)$, $\int B_m = 0$
 $\Rightarrow B_m(x)$ has rational coefficients $\forall m \Rightarrow B_m \in \mathbb{Q}$ \square

Chapter 7

Decomposition Methods

7.1 Caldéron-Zugmung decomposition

Idea: For a constant $\alpha > 0$ decompose the domain of $0 \geq f \in L^1$ into cubes s.t. restricted to one of the cubes either $f \leq \alpha$ or $f \sim \alpha$.

Theorem 7.1. Let $0 \geq f \in L^1(\mathbb{R}^d)$. Then $\forall \alpha > 0 \exists$ countable family of disjoint cubes $\{Q_k\}_{k=1}^\infty$ s.t.

1. If $x \notin \Omega = \bigcup_k Q_k$, then $f(x) \leq \alpha$ a.e.
2. $\alpha < \int_{Q_k} f \leq 2^d \alpha, \forall k$

Proof. The proof uses a stopping time argument. Assume f is compactly supported, then we cover $\text{supp } f$ by a big cube Q_0 . We can assume that $\int_{Q_0} f = \frac{1}{|Q_0|} \int_{\mathbb{R}^d} f < \alpha$. Now we divide Q_0 into 2^d sub-cubes of half-length side. For a sub cube Q of Q_0 , if

$$\int_Q f > \alpha \rightarrow \text{bad cube} \rightarrow \text{stop} \rightarrow \text{add to collection}$$

Otherwise, if $\int_Q f \leq \alpha \rightarrow$ good cube \rightarrow divide Q again in 2^d sub-cubes and repeat. By induction, we obtain a countable collection $\{Q_k\}_{k=1}^\infty$ of sub-cubes s.t. all Q_k are bad cubes.

Clearly, $\{Q_k\}$ are disjoint. Now we check the properties 1) and 2).

1) If $x \notin \Omega = \bigcup_k Q_k \Rightarrow \exists$ a sequence of good cubes $\{g_n\}_{n=1}^\infty$ s.t. $x \in g_n, \forall n$, and $|g_n| \rightarrow 0$ as $n \rightarrow \infty$. Then:

$$\int_{g_n} f \leq \alpha, \forall n \Rightarrow f(x) = \lim_{n \rightarrow \infty} \int_{g_n} f \leq \alpha.$$

for a.e. x by Lebesgue differentiation theorem.

2) \forall bad cubes Q_k , then by the definition $\int_{Q_k} f > \alpha$. From the construction, Q_k is a sub-cube of a good cube \tilde{Q} s.t. $|Q_k| = 2^{-d}|\tilde{Q}|$ and $\int_{\tilde{Q}} f \leq \alpha$
Consequently:

$$\int_{Q_k} f = \frac{1}{|Q_k|} \int_{Q_k} f \leq \frac{1}{2^{-d}|\tilde{Q}|} \int_{\tilde{Q}} f = 2^d \int_{\tilde{Q}} f \leq 2^d \alpha$$

□

Remark. From the above construction, we have:

$$|\Omega| = \sum_k |Q_k| < \sum_k \frac{1}{\alpha} \int_{Q_k} f \leq \frac{1}{\alpha} \int_{\mathbb{R}^d} f$$

Note that if $x \in \Omega^c \Rightarrow f(x) \leq \alpha$. This means $|\Omega| \rightsquigarrow |\{f > \alpha\}|$ related.
 \rightarrow recall that $|\{f > \alpha\}| \lesssim \frac{1}{\alpha} \int_{\mathbb{R}^d} |f|$

Lemma 7.2 (Reverse maximal inequality). Let $0 \leq f \in L^1(\mathbb{R}^d)$, $\alpha > 0$, then

$$|\{Mf > c\alpha\}| \geq \frac{1}{2^d \alpha} \int_{\{f > \alpha\}} f$$

for some constant $c = c_d > 0$.

Proof. Let $\{Q_k\}_{k=1}^\infty$ be the collection of cubes obtained by the Caldéron-Zugmund decomposition. Then for $\Omega = \bigcup_k Q_k$:

$$\begin{aligned} \int_{\{f > \alpha\}} f &\stackrel{1)}{\leq} \int_\Omega f = \sum_k \int_{Q_k} f \stackrel{2)}{\leq} \sum_k 2^d \alpha |Q_k| = 2^d \alpha |\Omega| \\ \forall k : \quad \int_{Q_k} f &> \alpha \Rightarrow \{Mf > c_d \alpha\} \supset \Omega \end{aligned}$$

We conclude that

$$|\{Mf > c_d \alpha\}| \geq |\Omega| \geq \frac{1}{2^d \alpha} \int_{\{f > \alpha\}} f$$

□

Definition 7.3.

$$\|f\|_{L^1 loc L^1(B)} = \|f\|_{L^1} + \int_B |f(y)| \log\left(\frac{|f(y)|}{\|f\|_{L^1}}\right) dy$$

We can apply the lemma from above to obtain:

Theorem 7.4. Let $f \in L^1(B)$ for $B \subset \mathbb{R}^d$ ball. Then:

$$Mf \in L^1(B) \Leftrightarrow f \in L^1 \log L^1(B)$$

Remark. Recall that $f \in L^p \Leftrightarrow Mf \in L^q$, $\forall p > 1$.

Proof. Step 1: (Assume B unit ball and $f \geq 0$ for simplicity) We prove that

$$\int_B |Mf| \lesssim \|f\|_{L^1 \log L^1(B)}$$

By layer-cake:

$$\begin{aligned} \int_B |Mf| &= \int_0^\infty |\{x \in B : Mf(x) > \lambda\}| d\lambda \\ &\leq \int_0^\delta \dots + \int_\delta^\infty \dots \\ &\leq \int_0^\delta |B| d\lambda + \int_\delta^\infty \frac{1}{\lambda} \int_{\{f > \lambda\}} f d\lambda \\ &= |B|\delta + \int_{\mathbb{R}^d} f(y) (\int_\delta^{f(y)} \frac{1}{\lambda} d\lambda) dy \\ &= |B|\delta + \int_{\mathbb{R}^d} f(y) \log(\frac{f(y)}{\delta}) dy \stackrel{\delta = \|f\|_{L^1}}{\leq} \|f\|_{L^1 \log L^1(B)} \end{aligned}$$

We used the weak L^1 inequality: $|\{Mf > \lambda\}| \leq \frac{1}{\lambda} \|f\|_{L^1} \rightsquigarrow \underbrace{|\{Mf > \lambda\}|}_{\leq |\{Mf \mathbf{1}_{|f| \geq \frac{\lambda}{2}} > \frac{\lambda}{2}\}|} \lesssim \frac{1}{\lambda} \left\| f \mathbf{1}_{|f| > \frac{\lambda}{2}} \right\|_{L^1}$

Step 2: We prove

$$\int_B |Mf| \gtrsim \|f\|_{L^1 \log L^1(B)}$$

Recall lemma $|\{Mf > \lambda\}| \gtrsim \frac{1}{\lambda} \int_{\{f > \lambda\}} f$ (We proved this by CZ decomposition)

By layer cake:

$$\int_B |Mf| = \int_0^\infty |\{x \in B : Mf > \lambda\}| dy \gtrsim \int_0^\infty \left(\frac{1}{\lambda} \int_{\{f > \lambda\}} f \right) d\lambda = \int_{\mathbb{R}^d} f(y) \left(\int_{\lambda_0}^{f(y)} \frac{1}{\lambda} d\lambda \right) dy \sim \|f\|_{L^1 \log L^1(B)}$$

Here $\lambda_0 \geq c_d \|f\|_{L^1}$

$$\Rightarrow |\{x \in B : Mf > \lambda\}| \gtrsim \frac{1}{\lambda} \int_{\{f > \lambda\}} f$$

□

Theorem 7.5 (A variant of CZ decomposition). Let $0 \geq f \in L^1(\mathbb{R}^d)$, assume $\int_{\mathbb{R}^d} f \geq c_d$. Then $\forall 0 < \Lambda \leq c_d$, $\forall \alpha_d > 0$, we can find a cover \mathbb{R}^d of disjoint cubes $\{Q_k\}_{k=1}^N$ s.t.

1. $\int_{Q_k} f \leq \Lambda, \forall k$
2. $\sum_Q \frac{1}{|Q_k|^\alpha} (\int_{Q_k} f - \alpha_d \Lambda) \geq 0$

Proof. We use again the stopping time argument. Assume $\text{supp } f \subset Q_0$ big cube. We divide Q_0 into 2^d sub-cubes. For all Q sub-cubes of Q_0 one of the following two holds:

- $\int_Q f \leq \Lambda$, then stop and add Q to collection
- $\int_Q f > \Lambda$, then divide again in 2^d sub-cubes and repeat!

After finitely many steps, we stop and get a collection $\{Q_k\}$. Now we have to check 1) and 2). 1) is obvious.

2) By writing the cubes as a tree, we can divide $\{Q\} \rightarrow \bigcup_F \{Q\}_{Q \in F}$ where Q are the nodes of the tree and F are the branches of the tree. So we get a union of disjoint collections $\{\{Q\}_{Q \in F}\}_F$ s.t. for all F we have:

$\forall m, \exists$ at most 2^d cubes $Q \subset F : |Q| = m$. Now we prove that $\forall F : \sum_{Q \in F} \frac{1}{|Q|^\alpha} (\int_Q f - c_d \Lambda) \geq 0$.

Take $m = \text{smallest volume of cubes in } F$. Then $\forall Q \in F \Rightarrow |Q| = m 2^{dn}$, then

$$\sum_{Q \in F} \frac{1}{|Q|^\alpha} \leq \sum_{n=0}^{\infty} \frac{2^d}{(m 2^{dn})^\alpha} \lesssim c_d \frac{1}{m^\alpha}$$

On the other hand:

$$\sum_{Q \in F} \frac{1}{|Q|^\alpha} \left(\int_Q f \right) \geq \sum_{Q \in F, |Q|=m} \frac{1}{m^\alpha} \int_Q f = \frac{1}{m^\alpha} \int_{\tilde{Q}} f \geq \frac{\Lambda}{m^\alpha}$$

where $\int_{\tilde{Q}} f > 1$

$$\Rightarrow \sum_{Q \in F} \frac{1}{|Q|^\alpha} \left(\int_{Q_k} f - \alpha_d \Lambda \right) \geq \frac{\Lambda}{m^\alpha} - \underbrace{c_d \tilde{c}_d}_{<1} \frac{\Lambda}{m^\alpha} \geq 0$$

□

Theorem 7.6. $\forall d \geq 1$, if $\{u_n\}_{n=1}^N$ ONF, in $L^2(\mathbb{R}^d)$, then

$$\sum_{n=1}^N \int_{\mathbb{R}^d} |\nabla u_n|^2 \geq K_d \int_{\mathbb{R}^d} \rho^{1+\frac{2}{d}}$$

where $\rho(x) = \sum_{n=1}^N |u_n(x)|^2$ density of the system. Here K_d is independent of N . In particular, we can take $N \rightarrow \infty$.

Remark. (d= 3) We have the Sobolev inequality:

$$\int_{\mathbb{R}^3} |\nabla u|^2 \geq C \left(\int_{\mathbb{R}^3} |u|^6 \right)^{\frac{1}{2}} \stackrel{\text{H\"older}}{\geq} C \frac{\int_{\mathbb{R}^3} |u|^{\frac{10}{3}}}{\left(\int_{\mathbb{R}^3} |u|^2 \right)^{\frac{2}{3}}} = c \int_{\mathbb{R}^3} \rho^{\frac{5}{3}}$$

$\rho = |u|^2$, $\int_{\mathbb{R}^3} |u|^2 = 1$. This is the LT inequality for $N = 1$.

Lemma 7.7. For any normalized functions $\{u_n\}_{n=1}^N \subset L^2(\mathbb{R}^d)$ we have:

$$(*) \quad \sum_{n=1}^N \int_{\mathbb{R}^d} |\nabla u_n|^2 \geq \frac{c_d}{N^{\frac{2}{d}}} \int_{\mathbb{R}^d} \rho^{1+\frac{2}{d}} \quad (c_d > 0 \text{ independent of } N)$$

Proof. If $u_n = u \Rightarrow (*)$ becomes the LT inequality for $N = 1$, namely

$$\text{LHS of } (*) = N \int_{\mathbb{R}^d} |\nabla u|^2 = \int_{\mathbb{R}^d} |\nabla \sqrt{\rho}|^2 \leq \frac{c \int \rho^{1+\frac{2}{d}}}{\left(\int \rho \right)^{\frac{2}{d}}} = \frac{c}{N^{\frac{2}{d}}} \int \rho^{1+\frac{2}{d}}$$

$$\rho = N|u|^2$$

For general $\{u_n\}_{n=1}^N$ normalized in L^2 , then for $\rho(x) = \sum_{n=1}^N |u_n(x)|^2$

$$|\nabla \sqrt{\rho}|^2 |\nabla \sqrt{\sum_n |u_n|^2}|^2 \leq \left| \frac{\sum_n |u_n| |\nabla u_n|}{\sqrt{\sum_n |u_n|^2}} \right|^2 \leq \sum_n |\nabla u_n|^2$$

(conevextiy of the gradient. Brezis/Hoffman-Ostenhof) $\Rightarrow \text{LHS}(*) = \int_{\mathbb{R}^d} \sum_n |\nabla u_n|^2 \geq \int_{\mathbb{R}^d} |\nabla \sqrt{\rho}|^2 \geq \frac{c_d}{N^{\frac{2}{d}}} \int_{\mathbb{R}^d} \rho^{1+\frac{2}{d}}$ by the previous estimate. \square

History of LT:

- 1969 Dyson-Lenard: proof of Stability of matter. Key observation: For kinetic energy of fermions in a cube $\varphi: \infty > c > \frac{1}{|\varphi|^{\frac{2}{d}}}(N-1)$ (Pauli-exclusion priciple)
- 1975 Lieb-Thirring: Nice proof of stability of matter using their inequality.
LT = uncertainty + exclusion principle
In 1975, LT proved their inequality in the dual form

$$|\sum_{\lambda < 0} \lambda(-\Delta + V(x))| \leq c_d \int_{\mathbb{R}^d} |v|^{1+\frac{d}{2}}$$

Lemma 7.8 (Dyson-Lenard).

$$\sum_{n=1}^N \int_Q |\nabla u_n|^2 =: T_Q \geq \frac{(2\pi)^2}{|Q|^{\frac{2}{d}}} (\int_Q \rho - 1), \quad \rho(x) = \sum_{n=1}^N |u_n(x)|^2$$

Proof.

$$\begin{aligned}
T_Q &= \sum_{n=1}^N \\
&\stackrel{(*)}{\lesssim} \frac{1}{|Q|^{\frac{2}{d}}} \sum_{n=1}^N \int_Q |u_n - \bar{u}_n^Q|^2 \\
&\stackrel{(**)}{\lesssim} \frac{1}{|Q|^{\frac{2}{d}}} \sum_{n=1}^N \int_Q (|u_n|^2 - 2|\bar{u}_n^Q|^2) \\
&= \frac{1}{|Q|^{\frac{2}{d}}} \int_Q \rho - 2|Q| \sum_{n=1}^N |\int_Q u_n|^2
\end{aligned}$$

Here : $\sum_{n=1}^N |\int_Q u_n|^2 = \sum_{n=1}^N |\langle 1, u_n \rangle_{L^2(Q)}|^2 \stackrel{ONF}{\leq} \|1\|_{L^2}^2 = |Q|$

Conclusion: $T_Q \gtrsim \frac{1}{|Q|^{\frac{2}{d}}} (\int_Q \rho - 2)$ The improved one $\frac{(2\pi)^2}{|Q|^{\frac{2}{d}}} (\int_Q \rho - 1)$ needs a bit more careful analysis.

(**) : We used $|a - b|^2 \geq \frac{1}{2}|a|^2 - 2|b|^2$

(*) : In d=3: Sobolev in $L^2(\mathbb{R}^3)$

$$\int_{\mathbb{R}^3} |\nabla u|^2 \gtrsim (\int_{\mathbb{R}^3} |u|^6)^{\frac{1}{3}}$$

Poincaré: $\underbrace{\int_Q |\nabla u|^2}_{=\nabla(u-c)} \gtrsim (\int_Q |u - \bar{u}_n^Q|^6)^{\frac{1}{6}} \gtrsim \frac{1}{|Q|^{\frac{2}{3}}} |u - \bar{u}_n^Q|^2, c = \bar{u}_n^Q = \int_Q u$ □

proof of LT. Denote $T = \sum_{n=1}^N \int_{\mathbb{R}^d} |\nabla u_n|^2$ and $T_Q = \sum_{n=1}^N \int_Q |\nabla u_n|^2$. Take $\{Q\}$ a collection of disjoint cubes in \mathbb{R}^d . Then: $T \geq \sum_Q T_Q$

For $d \geq 3$:

- We proved $T_Q \gtrsim \frac{1}{|Q|^{\frac{2}{d}}} (\int_Q \rho - 1), \rho = \sum_n |u_n|^2$

- By the convexity of gradient/H-O inequality:

$$\begin{aligned}
T_Q &\geq \int_Q |\nabla \sqrt{\rho}|^2 \stackrel{\text{Poincaré}}{\gtrsim} (\int_Q |\rho - \bar{\rho}^Q|^p)^{\frac{2}{p}}, p = 2^* = \frac{2d}{d-2} \\
&\geq \frac{\int_Q |\sqrt{\rho} - \bar{\sqrt{\rho}}^Q|^{2(1+\frac{2}{d})}}{(\int_Q |\sqrt{\rho} - \bar{\sqrt{\rho}}^Q|^2)^{\frac{2}{d}}} \gtrsim \frac{\int_Q \rho^{1+\frac{2}{d}}}{(\int_Q \rho)^{\frac{2}{d}}} - \frac{c_d}{|Q|^{\frac{2}{d}}} \int_Q \rho
\end{aligned}$$

$$(\overline{\sqrt{\rho}}^Q = \frac{1}{|Q|} \int_Q |\sqrt{\rho}| \leq \frac{1}{|Q|^{\frac{1}{2}}} (\int_Q \rho)^{\frac{1}{2}})$$

We can combine the two bounds:

$$\begin{aligned} (1 + \varepsilon)T_Q &\gtrsim \sum_Q \left(\frac{\varepsilon \int_Q \rho^{1+\frac{2}{d}}}{(\int_Q \rho)^{\frac{2}{d}}} - \frac{\varepsilon c_d}{|Q|^{\frac{2}{d}}} \int_Q \rho + \frac{1}{|Q|^{\frac{2}{d}}} (\int_Q \rho - 1) \right) \\ &\geq \sum_Q \frac{\varepsilon \int_Q \rho^{1+\frac{2}{d}}}{(\int_Q \rho)^{\frac{2}{d}}} + \sum_Q \frac{1}{|Q|^{\frac{2}{d}}} ((1 - \varepsilon c_d) \int_Q \rho - 1) \\ &\stackrel{?}{\geq} \int_{\mathbb{R}^d} \rho^{1+\frac{2}{d}} \end{aligned}$$

To get the conclusion, we need to choose $\{Q\}$ nicely. By our variant of CZ decomposition, we can choose $\{Q\}$ disjoint s.t. $\bigcup_Q Q = \mathbb{R}^d$ and

1. $\int_Q \rho \leq \Lambda$
2. $\sum_Q \frac{1}{|Q|^\alpha} (\int_Q \rho - \alpha_d \Lambda) \geq 0, \alpha = \frac{2}{d}$

Then:

$$(1 + \varepsilon)T \geq \sum_Q \frac{\varepsilon \int_Q \rho^{1+\frac{2}{d}}}{\Lambda^{\frac{2}{d}}} + \underbrace{(1 - \varepsilon c_d)}_{>0} \sum_Q \frac{1}{|Q|^{\frac{2}{d}}} (\int_Q \rho - \frac{1}{1 - \varepsilon c_d}) = \varepsilon \int_{\mathbb{R}^d} \rho^{1+\frac{2}{d}}$$

Here we can take $\varepsilon > 0$ small s.t. $(1 - \varepsilon c_d) > 0$. Then we choose $\Lambda > 0$ large s.t. $\alpha_d \Lambda > \frac{1}{1 - \varepsilon c_d}$

$$\Rightarrow T \geq \left(\frac{\varepsilon}{1 + \varepsilon} \frac{1}{\Lambda^{\frac{2}{d}}} \right) \int_{\mathbb{R}^d} \rho^{1+\frac{2}{d}}$$

(good for $\int_{\mathbb{R}^d} \rho \geq \Lambda$)

Actually if $\int_{\mathbb{R}^d} \rho = N \leq \Lambda$, then we can use the simple lemma:

$$T \geq \frac{c_d}{N^{\frac{2}{d}}} \int_{\mathbb{R}^d} \rho^{1+\frac{2}{d}} \geq \frac{c_d}{\Lambda^{\frac{2}{d}}} \int_{\mathbb{R}^d} \rho^{1+\frac{2}{d}}$$

□

7.2 Littlewood-Paley decomposition

We write $f = \sum_{n \in \mathbb{Z}} f_n$, $\hat{f}_n(k) \mathbb{1}_{\{2^{n-1} \leq |k| \leq 2^n\}}$. (Similar to dyadic decomposition)

Theorem 7.9 (Bernstein inequality/Reverse Poincaré). $\text{supp } \hat{f}_n \subset \underbrace{\{2^{n-1} \leq |k| \leq 2^n\}}_{\Omega}$,
 $\forall d \geq 1, \forall p \in [1, \infty]$:

$$\tilde{c}_{d,p} \|f\|_{L^p(\mathbb{R}^d)} \leq 2^{-n} \|\nabla f_n\|_{L^p(\mathbb{R}^d)} \leq c_{d,p} \|f\|_{L^p(\mathbb{R}^d)}$$

Proof. (d = 2, Sketch)

$$\begin{aligned}\|\nabla f_n\|_{L^2}^2 &= \int_{\mathbb{R}^d} |k|^2 \mathbb{1}_\Omega |\hat{f}_n(k)|^2 dk \leq \int_{\mathbb{R}^d} |k|^2 \chi(k) |\hat{f}_n(k)|^2 dk = \|\nabla(G \star f_n)\|_{L^2}^2 \\ &= \|(\nabla G) \star f_n\|_{L^2}^2 \stackrel{\text{Young}}{\leq} \|\nabla G\|_{L^1}^2 \|f_n\|_{L^2}^2\end{aligned}$$

for $\hat{G}(k) = \chi(k) \simeq \mathbb{1}_{2^{n-1} \leq |k| \leq 2^n}$, $\chi, G \in C^\infty$

□

Smooth dyadic decomposition:

$$\begin{aligned}\text{Take } \psi \in C_c^\infty(\mathbb{R}^d), \quad &\begin{cases} \psi(k) = 1, \text{ if } |k| \leq 1 \\ \psi(k) = 0, \text{ if } |k| \geq 2 \end{cases} \\ \varphi(k) = \psi(k) - \psi(2k), \quad &\varphi_n(x) = \psi(2^{-n}k) - \psi(2^{-n+1}k), \quad \forall n \in \mathbb{N}, \quad \varphi_0 = \psi(x)\end{aligned}$$

$$\begin{aligned}\Rightarrow \sum_{n=0}^N \varphi_n(k) &= \psi(2^{-N}k) \rightarrow 1, \text{ as } N \rightarrow \infty \\ \Rightarrow 1 &= \sum_{n=0}^{\infty} \varphi_n\end{aligned}$$

Definition 7.10. $f = \sum_{n=0}^{\infty} f_n$, $\hat{f}_n(k) = \varphi_n(k) \hat{f}(k)$

Remark. If $f \in S(\mathbb{R}^d) \Rightarrow f_n \in S(\mathbb{R}^d)$, $\forall n$.

$$\|D^\alpha f_n\|_{L^p(\mathbb{R}^d)} \sim 2^{n|\alpha|} \|f_n\|_{L^p}$$

Lemma 7.11. Let $0 \leq f \in L^1(\mathbb{R}^d)$, $\alpha > 0$, then we can find disjoint cubes $\{Q\} \subset \mathbb{R}^d$ s.t. $f = g_\alpha + b_\alpha = g_\alpha + \sum_Q (b_\alpha)|_Q$, where

1. $\text{supp}(g_\alpha) \subset \{\bigcup_Q Q\}^c$, $|g_\alpha| \lesssim \alpha$
2. $\forall Q : \int_Q b_\alpha = 0$, $\int_Q |b_\alpha| \lesssim \alpha$
3. $\sum_Q |Q| \lesssim \frac{c}{\alpha} \int_{\mathbb{R}^d} |f|$

Proof. (exercise)

□

Lemma 7.12 (Hörmander). Let φ and φ_n be as above. Define $K_n(x) := \check{\varphi}_n(x) = 2^{nd} \check{\varphi}(2^n x)$, then

$$\begin{aligned}\int_{|x| > 2|y|} \underbrace{\|K_n(x-y) - K_n(x)\|_{l^2(n)}}_{=\sqrt{\sum_n |K_n(x-y) - K_n(x)|^2}} dx\end{aligned}$$

Proof. We will use Minkowski's inequality: $\forall 1 \leq p \leq \infty :$

$$\left(\int_{\Omega_2} \left| \int_{\Omega_1} F(x, y) d\mu_1(x) \right|^p d\mu_2 \right)^{\frac{1}{p}} \leq \int_{\Omega_1} \left(\int_{\Omega_2} |F(x, y)|^p d\mu_2(x) \right)^{\frac{1}{p}} d\mu_1$$

with μ_1, μ_2 sigma finite.

Now consider

$$|K_n(x - y) - K_n(x)| \leq \int_0^{|y|} |\nabla K_n(x - te)| dt$$

where $e = \frac{y}{|y|}$

$$\begin{aligned} & \Rightarrow \int_{|x| > 2|y|} \sqrt{\sum_n |K_n(x - y) - K_n(x)|^2} dx \leq \int_{|x| > 2|y|} \sqrt{\sum_n \int_0^{|y|} |\nabla K_n(x - te)|^2 dt} dx \\ & \stackrel{\text{Minkowski}}{\leq} \int_{|x| > 2|y|} \int_0^{|y|} \sqrt{\sum_n |\nabla K_n(\underbrace{x - te}_z)|^2} dt dx \quad ((d\mu_2, \Omega_2) \rightsquigarrow \text{counting on } \mathbb{N}) \\ & \leq \int_{|z| > |y|} \int_0^{|y|} \sqrt{\sum_n |\nabla K_n(z)|^2} dt dz \\ & = |y| \int_{|z| > |y|} \sqrt{\sum_n |\nabla K_n(z)|^2} dz \end{aligned}$$

Recall: $K_n(z) = 2^{nd} \check{\varphi}(2^n z)$, $\varphi \in C_c^\infty \subset S(\mathbb{R}^d) \Rightarrow \check{\varphi} \in S(\mathbb{R}^d)$

$$\Rightarrow |\nabla K_n(z)| \leq 2^{n(d+1)} |\nabla \check{\varphi}(2^n z)| \leq 2^{n(d+1)} \min(1, |2^n z|^{-2(d+2)})$$

$$\Rightarrow \sum_n |\nabla K_n(z)|^2 \leq \sum_n \min(2^{n(d+1)}, 2^{-2n} |z|^{-2(d+2)})$$

$$\leq \sum_{|n| \leq L} 2^{2n(d+1)} + \sum_{n \geq L} 2^{-2n} |z|^{-2(d+2)}$$

$$\lesssim 2^{2L(d+1)} + 2^{-2L} |z|^{-2L} |z|^{-2(d+2)} \stackrel{\text{opt over } L}{\rightsquigarrow} |z|^{-2(d+1)}$$

$$\Rightarrow (\text{LHS lemma}) \leq |y| \int_{|z| \geq |y|} |z|^{-(d+1)} dz \lesssim C$$

□

Theorem 7.13 (L^p theory of Littlewood-Paley decomposition). Let $f = \sum_{n=0}^{\infty} f_n$, then

$$C^{-1} \|f\|_{L^p(\mathbb{R}^d)} \leq \left\| \sqrt{\sum_{n=0}^{\infty} |f_n|^2} \right\|_{L^p(\mathbb{R}^d)} \leq C \|f\|_{L^p(\mathbb{R}^d)}$$

Proof. $p = 2$:

$$\left\| \sqrt{\sum_n |f_n|^2} \right\|_{L^2}^2 = \int \sum_n |f_n|^2 = \int \sum_n |\hat{f}_n|^2 = \int \sum_n |\varphi_n(k)|^2 |\hat{f}(k)|^2 dk \sim \int |\hat{f}|^2 = \|f\|_{L^2}$$

The main part of the proof is to extend the bound to all $p \in (1, \infty)$.

Define $S(f)(x) = \sqrt{\sum_{n=0}^{\infty} |f_n(x)|^2}$, $f = \sum_n f_n$. We need to prove that $S : L^p \rightarrow L^p$ is bounded. We will prove this by real-interpolation:

- $|S(f + g)(x)| \leq S(f)(x) + S(g)(x)$ (Sub-additivity) ✓
- $S : L^2 \rightarrow L^2$ bounded ✓
- $S : L^1 \rightarrow L^{1,\infty}$ bounded (difficult)

From those, we find that $S : L^p \rightarrow L^p$ bounded $\forall p \in (1, 2]$. To get the result for $\infty > p > 2$, we use a duality argument. $\infty > p > 2 \Leftrightarrow 1 < p' < 2$

$$\begin{aligned} \|S(f)\|_{L^p} &\sim \sup \left| \sum_n \int_{\mathbb{R}^d} S(f)(x) h_n(x) dx \right| \\ &= \sup_{\sum_n \|h_n\|_{L^{p'}}^2 \leq 1} \left| \int_{\mathbb{R}^d} f(x) S^*(h_n) dx \right| \\ &\leq \sup_{\sum_n \|S^* h_n\|_{L^{p'}}^2 \leq 1} \sim \|f\|_{L^p} \end{aligned}$$

We used $S^* : L^{p'} \rightarrow L^{p'}$ bounded.

Let us prove the weak $(1, 1)$ property.

$$|\{x : S(f)(x) > \alpha\}| \leq \frac{c}{\alpha} \|f\|_{L^1(\mathbb{R}^d)}, \quad c = c_d \text{ (universal)}$$

We use the variant of Calderon-Zugmund from above. Using the sub-additivity $S(f) = S(g_\alpha + b_\alpha) \leq S(g_\alpha) + S(b_\alpha)$

$$\Rightarrow |\{x : S(f)(x) > \alpha\}| \leq |\{x : S(g_\alpha) > \frac{\alpha}{2}\}| + |\{x : S(b_\alpha) > \frac{\alpha}{2}\}|$$

good one: We use $S : L^2 \rightarrow L^2$ bounded and $|g_\alpha| \lesssim \alpha$

$$\begin{aligned} |\{x : S(g_\alpha)(x) > \frac{\alpha}{2}\}| &\lesssim \int_{\mathbb{R}^d} \frac{|S(g_\alpha)(x)|^2}{\alpha^2} dx \\ &\lesssim \int_{\mathbb{R}^d} \frac{|g_\alpha(x)|^2}{\alpha^2} dx \lesssim \int_{\mathbb{R}^d} \frac{|g_\alpha(x)|}{\alpha} dx \leq \frac{1}{\alpha} \int_{\mathbb{R}^d} |f| \end{aligned}$$

bad one:

$b_\alpha = \sum_Q (b_\alpha)|_Q$, c_Q = center of Q . Define $\tilde{Q} = \text{const. } Q$ s.t. $c_{\tilde{Q}} = c_Q$ and $|x - c_Q| \geq 2|y - c_Q|, \forall x \in \tilde{Q}, y \in Q$. Then:

$$|\bigcup_Q \tilde{Q}| \leq \sum_Q |\tilde{Q}| \lesssim \sum_Q |Q| \lesssim \frac{1}{\alpha} \int_{\mathbb{R}^d} |f|$$

Here, it suffices to bound:

$$|\{x : S(b_\alpha)(x) > \frac{\alpha}{2}\} \cap (\bigcup_Q \tilde{Q})^c| \lesssim \frac{1}{\alpha} \int_{(\bigcup_Q \tilde{Q})^c} S(b_\alpha)(x) dx$$

We need to prove $\int_{(\bigcup_Q \tilde{Q})^c} S(b_\alpha)(x) dx \lesssim \|f\|_{L^1}$.

We conclude the bound with Hörmander's lemma from above. So let $K_n(x) = \check{\varphi}_n(x) = 2^{nd} \check{\varphi}(2^n x)$ as in the lemma. Then: $\hat{f}_n(k) = \varphi_n(k) \hat{f}(k) \Rightarrow f_n(x) = (\check{\varphi}_n \star f)(x) = (K_n \star f)(x)$. We prove $\int_{(\bigcup_Q \tilde{Q})^c} S(b_\alpha)(x) dx \lesssim \|f\|_{L^1}$.

With Hörmander's lemma and Minkowski we get:

$$\begin{aligned} \int_{(\bigcup_Q \tilde{Q})^c} S(b_\alpha)(x) dx &= \int_{(\bigcup_Q \tilde{Q})^c} \sqrt{\sum_n |K_n \star b_\alpha|^2} dx \\ &= \int_{(\bigcup_Q \tilde{Q})^c} \sqrt{\sum_n \left| \sum_Q (K_n \star b_{\alpha,Q})(x) \right|^2} dx, \quad b_\alpha = \sum_Q b_{\alpha,Q} \end{aligned}$$

Here note that $\int_Q b_{\alpha,Q} = 0$, hence:

$$(K_n \star b_{\alpha,Q})(x) = \int_{y \in Q} K_n(x-y) b_\alpha(y) dy = \int_{y \in Q} (K_n(x-c_Q - (y-c_Q)) - K_n(x-c_Q)) b_\alpha(y) dy$$

Thus:

$$\begin{aligned} \int_{(\bigcup_Q \tilde{Q})^c} S(b_\alpha)(x) dx &\stackrel{\text{C.S.ineq, Minkowski}}{\leq} \sum_Q \int_{(\bigcup_Q \tilde{Q})^c} dx \int_Q |b_\alpha(y)| \sqrt{\sum_n |K_n(x-y) - K(x-c_Q)|^2} dy \\ &\stackrel{\text{Fubini}}{\leq} \sum_Q \int_Q |b_{\alpha,Q}| \underbrace{\int_{|x-c_Q| > 2|y-c_Q|} dx \sqrt{\sum_n |K_n(x-y) - K(x-c_Q)|^2} dy}_{\leq C < \infty \text{ (H lemma)}} \lesssim \sum_Q \int_Q |b_\alpha(y)| dy \lesssim \|f\|_{L^1} \end{aligned}$$

Thus:

$$\begin{aligned} |\{x : S(b_\alpha) > \frac{\alpha}{2}\}| &\leq |\{x : S(b_\alpha) > \frac{\alpha}{2}\} \cap (\bigcup_Q \tilde{Q})| + |(\bigcup_Q \tilde{Q})| \lesssim \frac{1}{\alpha} \|f\|_{L^1} \\ &\Rightarrow \|S(f)\|_{L^{1,\infty}} \lesssim \|f\|_{L^1} \end{aligned}$$

By interpolation: $\|S(f)\|_{L^p} \lesssim \|f\|_{L^p}, \forall p \in (1, 2]$

Duality:

$$\|(f_n)_n\|_{L^p, l^2(n)} = \left\| \sqrt{\sum_n |f_n(x)|^2} \right\|_{L^p(x)} = \sup_{\|h_n\|_{L^{p'}, l^2(n)} \leq 1} \left| \sum_n \int (\bar{f}_n)(h_n) \right|$$

If $p > 2$:

$$\begin{aligned} \|S(f)\|_{L^p} &= \|(f_n)\|_{L^p, l^2(n)} = \sup_{\|h_n\|_{L^{p'}, l^2(n)} \leq 1} \left| \sum_n \int (\bar{f}_n)(h_n) \right| = \sup_{\|h_n\|_{L^{p'}, l^2(n)} \leq 1} \left| \int f \sum_n (\varphi_n \hat{h}_n)^v \right| \\ &\stackrel{\text{Hölder}}{\leq} \|f\|_{L^p} \sup \left\| \sum_n (\varphi_n \hat{h}_n)^v \right\|_{L^{p'}} \lesssim f_{L^p} \end{aligned}$$

□

Remark. L^p theory for a function can be extended to orthonormal family to prove the Lieb Thirring inequality (Julien Sabin 2015)