

AMERICAN UNIVERSITY OF BEIRUT

FREQUENCY ANALYSIS AND APPLICATIONS

by

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ABSTRACT OF THE THESIS OF

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The goal of this thesis is to study signals that have a regularity property defined in the frequency space, such as a decay on average of the amplitude of their Fourier transform, by using techniques from frequency analysis. Frequency analysis is a set of techniques that involve an analysis in the Fourier domain. We review some of these techniques and some principles. More precisely we will decompose a signal into countable sums of functions of which the Fourier transform is compactly supported in a ball or an annulus by performing a Littlewood–Paley decomposition. We will apply this technique to study the properties of functions having a specific regularity. Over two hundred years ago, Fourier studied problems related to the series expansions of periodic signals using elementary trigonometric polynomials. The theory was extended to non-periodic signals by using the Fourier transform and forms the core of harmonic analysis. Harmonic analysis is used in various fields such as signal processing and partial differential equations (PDEs).

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CHAPTER 1

THE FOURIER TRANSFORM

In this section we recall the definition of the Fourier transform and some of its properties in the classical setting and also in the Lebesgue setting.

1.1 The Fourier transform in the classical setting

In this subsection we follow [6]. The classical Fourier transform is defined for functions that are moderately decreasing or Schwartz functions.

Recall that the set of moderately decreasing functions (denoted by $\mathcal{M}(\mathbb{R}^d)$) is the set of continuous functions such that there exists $A > 0$ such that $|f(x)| \leq \frac{A}{1+|x|^{2d}}$.

Let $f \in \mathcal{M}(\mathbb{R}^d)$. Then \hat{f} denotes the Fourier transform of f , and it is defined for all $\xi \in \mathbb{R}^d$ by

$$\hat{f}(\xi) := \int_{\mathbb{R}^d} f(x) e^{-2i\pi\xi \cdot x} dx \quad (1.1)$$

Let $\mathcal{S}(\mathbb{R}^d)$ be the set of Schwartz functions, i.e the set for functions f such that $\sup_{x \in \mathbb{R}^d} (1 + |x|)^N |\partial^\alpha f(x)| < \infty$ for all $\alpha := (\alpha_1, \dots, \alpha_d) \in \mathbb{N}^d$ and $N \in \mathbb{N}$ (Recall that

$\partial^\alpha := \partial^{\alpha_1} \dots \partial^{\alpha_d}$). Recall (see [2]) that if $f \in \mathcal{S}(\mathbb{R}^d)$ then $\hat{f} \in \mathcal{S}(\mathbb{R}^d)$. Moreover the Fourier inversion formula holds

$$f(x) = \int_{\mathbb{R}^d} \hat{f}(\xi) e^{2i\pi\xi \cdot x} d\xi \quad (1.2)$$

Notice that the proof to establish (1.2) for Schwartz functions also works for functions $(f, \hat{f}) \in \mathcal{M}(\mathbb{R}^d)^2$.

Let $(x_0, \lambda) \in \mathbb{R}^d \times \mathbb{R}^+$. Let T_{x_0} (resp. D_λ) denote the translation of vector x_0 (resp. the dilation operator of factor λ), i.e $T_{x_0}(g)(x) := g(x - x_0)$ (resp. $D_\lambda(g)(x) := g(\lambda x)$). Let $(x_0, \xi_0, \lambda) \in (\mathbb{R}^d)^2 \times \mathbb{R}$. Then elementary changes of variables show that the equalities below hold:

$$\begin{aligned} \widehat{T_{x_0} f}(\xi) &= e^{-2i\pi\xi \cdot x_0} \hat{f}(\xi), \quad \widehat{D_\lambda f}(\xi) = \frac{1}{\lambda^d} \hat{f}\left(\frac{\xi}{\lambda}\right), \quad \text{and} \\ T_{\xi_0} \hat{f}(\xi) &= \widehat{e^{2i\pi\xi_0 \cdot x} f}(\xi). \end{aligned} \tag{1.3}$$

1.2 The Fourier transform in the Lebesgue setting

In this subsection we follow [2]. The Fourier transform in the Lebesgue setting for functions in $L^1(\mathbb{R}^d)$, $L^2(\mathbb{R}^d)$, and $L^1(\mathbb{R}^d) \cap L^2(\mathbb{R}^d)$.

Let $f \in L^1(\mathbb{R}^d)$. Then the Fourier transform is again defined by (1.1). Moreover, assuming also that $\hat{f} \in L^1(\mathbb{R}^d)$, then f is (a.e) equal to a continuous function g , more precisely

$$f(x) = g(x) := \int_{\mathbb{R}^d} \hat{f}(\xi) e^{2i\pi\xi \cdot x} d\xi \tag{1.4}$$

Let $f \in L^2(\mathbb{R}^d)$. Then there exists a sequence $\{f_n\}_{n \geq 1}$ such that f_n and $\hat{f}_n \in L^1(\mathbb{R}^d)$ ¹, and $\hat{f} = \lim_{n \rightarrow \infty} \hat{f}_n$ in $L^2(\mathbb{R}^d)$. Moreover the Plancherel theorem holds, i.e

$$\int_{\mathbb{R}^d} |f(x)|^2 dx = \int_{\mathbb{R}^d} |\hat{f}(\xi)|^2 d\xi \tag{1.5}$$

Assume also that $\hat{f} \in L^1(\mathbb{R}^d)$. Then (1.4) also holds².

Let $f \in L^1(\mathbb{R}^d) \cap L^2(\mathbb{R}^d)$. Then the Fourier transform of f in L^1 is (a.e) equal to its Fourier transform in L^2 . Consequently (1.4) and (1.5) also holds.

¹Observe that this implies that $\hat{f}_n \in L^2(\mathbb{R}^d)$: see [2].

²It suffices to use the same arguments in [2], more precisely in the proof of the Fourier inversion formula for f and \hat{f} in $L^1(\mathbb{R}^d)$, once we have proved that the transfer formula $\int_{\mathbb{R}^d} \hat{f}\phi = \int_{\mathbb{R}^d} f\hat{\phi}$ holds for f , with ϕ a Schwartz function. But the transfer formula clearly holds for all functions h in $L^1(\mathbb{R}^d)$, so it also holds for f by a limit process

CHAPTER 2

PROPERTIES OF A SIGNAL LOCALIZED IN THE FOURIER DOMAIN

In this section we study the properties of a signal that is localized in the Fourier domain, i.e a signal of which the amplitude is essentially concentrated around a point $\xi_0 \in \mathbb{R}^d$. More precisely we say that a function f is localized in the frequency space around a point $\xi_0 \in \mathbb{R}^d$ if and only if for all $N \in \mathbb{N}$ there exists $C_N > 0$ such that

$$\xi \in \mathbb{R}^d : |\hat{f}(\xi)| \leq \frac{C_N}{(1+|\xi-\xi_0|)^N}. \quad (2.1)$$

The proposition below and its proof show that if a function f has a Fourier transform that is localized around some point $\xi_0 \in \mathbb{R}^d$, then f is smooth.

Proposition 1. *Let $\xi_0 \in \mathbb{R}^d$ and $f \in \mathcal{M}(\mathbb{R}^d)$ that is localized at ξ_0 . Then $f \in C^\infty(\mathbb{R}^d)$.*

Proof. We may assume WLOG that $\xi_0 = 0$: indeed, we may use (1.3) to find g such that $\hat{g}(\xi) := \hat{f}(\xi + \xi_0)$ and then apply Proposition 1 to g at $\xi_0 = 0$.

Observe from (2.1) that for all $\alpha \in \mathbb{N}^d$, the map $\xi \rightarrow \partial_x^\alpha \left(\hat{f}(\xi) e^{2i\pi\xi \cdot x} \right)$ is bounded by an L^1 function that does not depend on x . Hence we can use the differentiation rule under the integral sign of the Fourier inversion formula (1.2) to come to the conclusion that $f \in C^\infty(\mathbb{R}^d)$ and, moreover,

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

□

Remark 1. *The analogue of Proposition 1 in Lebesgue spaces is the following:*

Proposition 1'. *Let $\xi_0 \in \mathbb{R}^d$. Assume that $f \in L^1(\mathbb{R}^d)$ and that \hat{f} is localized around ξ_0 . Then f is a.e equal to a $C^\infty(\mathbb{R}^d)$ function.*

Proof. Again we may assume WLOG that $\xi_0 = 0$. We see from (2.1) that $\hat{f} \in L^1(\mathbb{R})$. Hence we get from (1.2) $f = g$ a.e with $g(x) := \int_{\mathbb{R}^d} \hat{f}(\xi) e^{2i\pi\xi \cdot x} d\xi$. Then we replace ‘ $\partial^\alpha f$ ’ with ‘ $\partial^\alpha g$ ’ in the proof of Proposition 1 and we follow verbatim the same steps. □

The next proposition shows that if \hat{f} is very well-localized (in the sense that $\hat{f} \in \mathcal{C}_c(\mathbb{R}^d)$ ¹) then f cannot be equal to zero on a ball centered at x_0 on a radius $R > 0$, even if R is very small. This implies in particular that f cannot be localized (in the sense that $f \notin \mathcal{C}_c(\mathbb{R}^d)$)

Proposition 2. *Let $E := \{ h \text{ of moderate decrease} : \hat{h} \in \mathcal{C}_c(\mathbb{R}^d) \}$. Let $(x_0, R) \in \mathbb{R}^d \times (0, \infty]$. There exists no $f \in E$, $f \neq 0$ such that $f = 0$ on a ball $B(x_0, R)$.*

Proof. Suppose that $f = 0$ on a ball $B(x_0, R)$ where $x_0 \in \mathbb{R}^d$ and $R > 0$. We may assume WLOG that $x_0 = 0$: indeed, if $x_0 \neq 0$, then we can apply (1.3) and Proposition 2 to $x \rightarrow f(x_0 + x)$.

Expanding $e^{2\pi i x \cdot \xi}$ in its Maclaurin series we get²

$$\begin{aligned}
f(x) &= \int_{\mathbb{R}^d} \hat{f}(\xi) e^{2\pi i x \cdot \xi} d\xi \\
&= \int_{\mathbb{R}^d} \hat{f}(\xi) \sum_{k=0}^{\infty} \frac{1}{k!} (2\pi i x \cdot \xi)^k d\xi \\
&= \sum_{k=0}^{\infty} \frac{1}{k!} \int_{\mathbb{R}^d} (2\pi i x \cdot \xi)^k \hat{f}(\xi) d\xi \\
&= \sum_{\alpha \in \mathbb{N}^d} \frac{1}{\alpha!} x^\alpha \int_{\mathbb{R}^d} (2\pi i \xi)^\alpha \hat{f}(\xi) d\xi,
\end{aligned} \tag{2.2}$$

using at the third line the normal convergence of the series in ξ and at the fourth line the multinomial Newton formula. Now, recall from the proof of Proposition 1 that $\int_{\mathbb{R}^d} (2\pi i \xi)^\alpha \hat{f}(\xi) d\xi = \partial^\alpha f(0) = 0$ for all α . Hence $f = 0$, which is impossible. □

Remark 2. *The analogue of Proposition 2 in Lebesgue spaces is the following*

Proposition 2’. *Let $E' := \{ h \in L^1(\mathbb{R}^d) : \hat{h} \in \mathcal{C}_c(\mathbb{R}^d) \}$. Let $(x_0, R) \in \mathbb{R}^d \times (0, \infty)$. There is no $f \neq 0 \in E'$ such that $f = 0$ a.e on $B(x_0, R)$.*

¹Recall that $\mathcal{C}_c(\mathbb{R}^d)$ is the set of continuous and compactly supported functions

²Recall that if $\alpha := (\alpha_1, \dots, \alpha_d) \in \mathbb{N}^d$ then $\alpha! := \alpha_1! \dots \alpha_d!$ and $x^\alpha := x_1^{\alpha_1} \dots x_d^{\alpha_d}$

Proof. Again we assume that the conclusion does not hold. We may assume WLOG the $x_0 = 0$. Let $g(x) := \int_{\mathbb{R}^d} \hat{f}(\xi) e^{2i\pi x \cdot \xi} d\xi$. Then from (2.1) and a standard rule of continuity we see that g is continuous. Hence $g = 0$ on $B(x_0, R)$ and, repeating the same argument as in (2.2), we get $g = 0$. Hence $f = 0$ a.e, which is a contradiction. \square

In fact a stronger result holds:

Proposition 3. *Let $\xi_0 \in \mathbb{R}^d$. Let $\bar{E} = \{ h \text{ of moderate decrease} : \forall N \in \mathbb{N}, \sup_{\xi \in \mathbb{R}^d} (1 + |\xi - \xi_0|)^N |\hat{h}(\xi)| < \infty \}$. Then there does not exist $f \neq 0 \in \bar{E}$ such that $f = 0$ on $B(x_0, R)$.*

Proof. We only prove the proposition for $d = 1$.

Again we may assume WLOG that $\xi_0 = 0$.

Assume that there exists $f \neq 0$ such that $f = 0$ on $B(x_0, R)$. Let $g(z) := \int_{\mathbb{R}} \hat{f}(\xi) e^{2i\pi \xi z} d\xi$.

Claim: g is an analytic function.

Proof:

Indeed, from the fundamental theorem of calculus

$$e^{2i\pi \xi(z+\Delta z)} - e^{2i\pi \xi z} = \Delta z \int_0^1 \frac{\partial e^{2i\pi \xi z}}{\partial z'}(z + t\Delta z) dt$$

Moreover from $e^{z'+z''} = e^{z'} e^{z''}$ and the elementary estimate $|e^{z'} - 1| \lesssim |z'|$ for $|z'| \leq 1$ we get

$$|e^{2i\pi \xi(z+\Delta z)} - e^{2i\pi \xi z}| \lesssim |\xi| |\Delta z|$$

Hence g is analytic and

$$\lim_{\Delta z \rightarrow 0} \frac{g(z+\Delta z) - g(z)}{\Delta z} = \int_{\mathbb{R}} 2i\pi \xi \hat{f}(\xi) e^{2i\pi \xi \cdot z} d\xi.$$

Alternatively one can use the Morera theorem to prove that g is analytic. Let R be a rectangle. Denoting by ∂R its boundary we have, by Fubini theorem

$$\int_{\partial R} g(z) dz = \int_{\mathbb{R}} \hat{f}(\xi) \int_{\partial R} e^{2i\pi \xi z} dz d\xi = 0,$$

where at the last equality we use the Cauchy theorem to the analytic function $z \rightarrow e^{2i\pi \xi z}$.

Moreover from (1.4) we see that g has a zero that is not isolated. Hence $g = 0$ and $f = 0$. \square

2.1 The Heisenberg uncertainty principle

We showed in Proposition 1' and more generally in Proposition 3 that if \hat{f} is localized in the frequency space then f cannot be localized in the physical space. In the next proposition, we prove a quantitative formulation of this phenomenon: the Heisenberg inequality.

Proposition 4. *Let $f \in \mathcal{S}(\mathbb{R}^d)$. Let $(x_0, \xi_0) \in (\mathbb{R}^d)^2$. Then*

$$\left(\int_{\mathbb{R}^d} |x - x_0|^2 |f(x)|^2 dx \right) \left(\int_{\mathbb{R}^d} |\xi - \xi_0|^2 |\hat{f}(\xi)|^2 d\xi \right) \geq \frac{1}{16\pi^2} \left(\int_{\mathbb{R}^d} |f(x)|^2 dx \right)^2 \quad (2.3)$$

The same conclusion holds if $f \in L^2(\mathbb{R}^d)$ satisfies $\int_{\mathbb{R}^d} |x|^2 |f(x)|^2 dx < \infty$, and $\int_{\mathbb{R}^d} |\xi|^2 |\hat{f}(\xi)|^2 d\xi < \infty$.

Proof. First let us assume that $f \in \mathcal{S}(\mathbb{R}^d)$.

By using the translation rules of (1.3) if necessary we may assume WLOG that $\xi_0 = x_0 = 0$.

By using the dilation rule of (1.3) if necessary we may assume WLOG that (*) holds with (*) : $\int_{\mathbb{R}^d} |f(x)|^2 dx = 1$. Indeed an elementary change of variable shows that $\int_{\mathbb{R}^d} |D_\lambda f(x)|^2 dx = 1$ iff $\lambda^d = \|f\|_{L^2}^2$. Applying (2.3) to f_λ we get

$$\left(\int_{\mathbb{R}^d} |x|^2 |f_\lambda(x)|^2 dx \right) \left(\int_{\mathbb{R}^d} |\xi|^2 |\hat{f}_\lambda(\xi)|^2 d\xi \right) \gtrsim 1 \quad (2.4)$$

Hence, an elementary change of variable combined with (1.3) show that

$$\left(\int_{\mathbb{R}^d} |x|^2 |f_\lambda(x)|^2 dx \right) \left(\int_{\mathbb{R}^d} |\xi|^2 |\hat{f}_\lambda(\xi)|^2 d\xi \right) \gtrsim \lambda^{-2d} \gtrsim \|f\|_{L^2}^2$$

We first prove (2.3) for $d = 1$. We follow e.g [6]. Integrating by parts and recalling that f decays fast for $|x|$ large we get

$$\begin{aligned} 1 &= \int_{\mathbb{R}^d} x' |f(x)|^2 dx \\ &= - \int_{\mathbb{R}^d} x \frac{d}{dx} (|f(x)|^2) dx \\ &\lesssim \left(\int_{\mathbb{R}^d} x^2 |f(x)|^2 dx \right)^{\frac{1}{2}} \left(\int_{\mathbb{R}^d} |f'(x)|^2 dx \right)^{\frac{1}{2}} \\ &\lesssim \left(\int_{\mathbb{R}^d} x^2 |f(x)|^2 dx \right)^{\frac{1}{2}} \left(\int_{\mathbb{R}^d} \xi^2 |\hat{f}(\xi)|^2 d\xi \right)^{\frac{1}{2}}, \end{aligned}$$

using the Cauchy-Schwartz inequality at the third row and $\hat{f}'(\xi) = 2i\pi\xi\hat{f}(\xi)$ at the fourth row.

Then we prove (2.3) for $d > 1$. Let $i \in \{1, \dots, d\}$. Then from Fubini theorem

$$\begin{aligned} 1 &= \int_{\mathbb{R}^d} |f(x)|^2 dx_1 \dots dx_d \\ &= \int_{\mathbb{R}^{d-1}} \left(\int_{\mathbb{R}} |f(x)|^2 dx_i \right) dx_1 \dots dx_{i-1} dx_{i+1} \dots dx_d \\ &\lesssim \int_{\mathbb{R}^{d-1}} \left(\int_{\mathbb{R}} x_i^2 |f(x)|^2 dx_i \right)^{\frac{1}{2}} \left(\int_{\mathbb{R}} |\partial_{x_i} f(x)|^2 dx_i \right)^{\frac{1}{2}} dx_1 \dots dx_{i-1} dx_{i+1} \dots dx_d, \\ &\lesssim \left(\int_{\mathbb{R}^d} x_i^2 |f(x)|^2 dx \right)^{\frac{1}{2}} \left(\int_{\mathbb{R}^d} |\partial_{x_i} f(x)|^2 dx \right)^{\frac{1}{2}} \end{aligned}$$

using at the last row $(d - 1)$ - times the Cauchy-Schwarz inequality. By summing over i the inequality above we get, using also the Cauchy-Schwarz inequality for finite sequences

$$\begin{aligned} 1 &\lesssim \left(\int_{\mathbb{R}^d} |x|^2 |f(x)|^2 dx \right) \left(\int_{\mathbb{R}^d} |\nabla f(x)|^2 dx \right) \\ &\lesssim \left(\int_{\mathbb{R}^d} |x|^2 |f(x)|^2 dx \right)^{\frac{1}{2}} \left(\int_{\mathbb{R}^d} |\xi|^2 |\hat{f}(\xi)|^2 d\xi \right)^{\frac{1}{2}} \end{aligned}$$

Now assume that $f \in L^2(\mathbb{R}^d)$ such that $|x|f \in L^2(\mathbb{R}^d)$ and $|\xi|\hat{f} \in L^2(\mathbb{R}^d)$. Let $R_n > 0$ be such that $\int_{|x| \geq R_n} |x|^2 |f(x)|^2 dx \rightarrow 0$ as $n \rightarrow \infty$. We slightly modify an argument in e.g [1]. Let χ be a smooth function such that $\chi(x) = 1$ if $|x| \leq 1$ and $\chi(x) = 0$ if $|x| \geq 2$. Let $f_n := \chi_n(\rho_n * f)$ with $\chi_n(x) := \chi\left(\frac{x}{R_n}\right)$ and $\{\rho_n\}_{n \geq 1}$ a sequence of mollifiers. Recall that $\rho_n * f \rightarrow f$ in $L^2(\mathbb{R}^d)$. Hence using also the decomposition $f_n = \chi_n(\rho_n * f - f) + \chi_n f$ and the dominated convergence theorem, we get $(1 + |x|)f_n \rightarrow (1 + |x|)f$ as $n \rightarrow \infty$. From $\nabla f_n = \nabla \chi_n(\rho_n * f) + \chi_n(\rho_n * \nabla f)$ we see that $\nabla f_n \rightarrow \nabla f$ in $L^2(\mathbb{R}^d)$ as $n \rightarrow \infty$. Hence by Plancherel theorem we have $\int_{\mathbb{R}^d} |\xi|^2 \left| \widehat{f_n - f}(\xi) \right|^2 d\xi \rightarrow 0$ as $n \rightarrow \infty$.

□

A consequence of the inequality (2.3) is the following result that we can relate to Proposition 2' :

Proposition 5. *Let $(R, x_0, \xi_0) \in (0, \infty) \times (\mathbb{R}^d)^2$. There exists a constant $c > 0$ such that there is no function $f \in L^2(\mathbb{R}^d)$ for which $\hat{f} = 0$ outside the ball $B(\xi_0, R)$ and $f = 0$ outside the ball $B(x_0, \frac{c}{R})$.*

Proof. We see from (2.3) and Plancherel theorem that $\int_{\mathbb{R}^d} |x - x_0|^2 |f(x)|^2 dx \leq \frac{c^2}{R^2} \int_{\mathbb{R}^d} |f(x)|^2 dx$, $\int_{\mathbb{R}^d} |\xi - \xi_0|^2 |\hat{f}(\xi)|^2 d\xi \leq R^2 \int_{\mathbb{R}^d} |f(x)|^2 dx$. This contradicts (2.3) for $c > 0$ small enough.

□

2.2 Benedick non-localization principle

We showed in the previous section a simultaneous non localization property for f and \hat{f} . More precisely we proved in Proposition 2' that if $f \neq 0 \in L^1(\mathbb{R})$ has a smooth Fourier transform with bounded support (i.e there exists $\bar{R} > 0$ such that $\{\hat{f} \neq 0\} \subset B(O, \bar{R})$) then for all $(x_0, R) \in \mathbb{R}^d \times (0, \infty)$ we have $|\{f \neq 0\} \cap B(x_0, R)| > 0$ ⁴.

³Here $\nabla g := (\partial_{x_1} g, \dots, \partial_{x_d} g)$, with $\partial_{x_i} g$ the i -th distributional partial derivative (see [2]) of $g \in \{f_n, f\}$.

⁴Here $|E|$ denotes the measure of a set E

We now prove a stronger proposition on simultaneous non-localization due to Benedicks: it says that if a function $f \neq 0 \in L^1(\mathbb{R}^d)$ has its Fourier transform \hat{f} finitely supported (i.e $|\{\hat{f} \neq 0\}| < \infty$) then f cannot be finitely supported (i.e $|\{f \neq 0\}| = \infty$). Notice that $|\{\hat{f} \neq 0\}| < \infty$ (resp. $|\{f \neq 0\}| = \infty$) may be unbounded and that $|\{f \neq 0\} \cap B(x_0, R)| > 0$ does not necessarily imply that $|\{f \neq 0\}| = \infty$.

We now state the Benedick non-localization principle:

Proposition 6. *Assume that $f \in L^1(\mathbb{R}^d)$ satisfies $|\{\hat{f} \neq 0\}| < \infty$ and $|\{f \neq 0\}| < \infty$. Then $f = 0$.*

Proof. We follow the exposition of the proof in [5].

Let $X := \{f \neq 0\}$ and $\hat{X} := \{\hat{f} \neq 0\}$. By using a scaling transformation $x \rightarrow ax$ if necessary, we may assume that $|X| < 1$. Consider the periodization of the indicator function of $\{\hat{f} \neq 0\}$, i.e the function $\xi \rightarrow h(\xi) := \sum_{k \in \mathbb{Z}^d} \mathbb{1}_{\{\hat{f} \neq 0\}}(\xi - k)$.

Claim: $h < \infty$ a.e.

Indeed it suffices to prove that $h < \infty$ a.e on the rectangle $[0, 1] \times \dots [0, 1]$. Given $k := (k_1, \dots, k_d) \in \mathbb{Z}^d$ let R_k denote the rectangle $R_k := [k_1, k_1 + 1] \times \dots \times [k_d, k_d + 1]$. From the monotone convergence theorem and an elementary change of variable we see that

$$\infty > \int_{\mathbb{R}^d} \mathbb{1}_{\{\hat{f} \neq 0\}} d\xi = \sum_{k \in \mathbb{Z}^d} \int_{R_k} \mathbb{1}_{\{\hat{f} \neq 0\}}(\xi) d\xi = \sum_{k \in \mathbb{Z}^d} \int_{R_{(0, \dots, 0)}} \mathbb{1}_{\{\hat{f} \neq 0\}}(\xi - k) d\xi = \int_{R_{(0, \dots, 0)}} h(\xi) d\xi$$

Hence the claim follows.

This implies that the property (P) holds with

(P) : a.e ξ there exists only a finite number of ks such that $\hat{f}(\xi - k) \neq 0$.

Let $f_\xi(x) := e^{2i\pi\xi \cdot x} f(x)$. Let $\tilde{f}_\xi := \sum_{k \in \mathbb{Z}^d} f_\xi(x - k)$ be the periodic extension of f_ξ .

Since $f_\xi \in L^1(\mathbb{R}^d)$, we can apply the Poisson formula and get that the Fourier series of \tilde{f}_ξ (denoted by $FS[\tilde{f}_\xi]$) is equal to

$$FS[\tilde{f}_\xi] = \sum_{k \in \mathbb{Z}^d} \hat{f}(\xi - k) e^{2i\pi k \cdot x}$$

(Here $FS[\tilde{f}_\xi]$ denotes the Fourier series of \tilde{f}_ξ). It is also clear from (P) that $FS[\tilde{f}_\xi]$ converges absolutely. Hence we get a.e

$$\tilde{f}_\xi = FS[\tilde{f}_\xi] = \sum_{k \in \mathbb{Z}^d} \hat{f}(\xi - k) e^{2i\pi k \cdot x} \quad (2.5)$$

Moreover from the definition of \tilde{f}_ξ we see that

$$\{x \in \mathbb{T}^d : \tilde{f}_\xi \neq 0\} \subset \bigcup_{k \in \mathbb{Z}^d} \{x \in \mathbb{T}^d : f(x - k) \neq 0\} \subset \{f \neq 0\}$$

Hence $\left| \left\{ x \in \mathbb{T}^d : \tilde{f}_\xi \neq 0 \right\} \right| < 1$. We also see from (2.5) that \tilde{f}_ξ is a.e equal to a trigonometric polynomial. Hence $\tilde{f}_\xi = 0$ a.e and, since $\hat{f}(\xi - k)$ is the k^{th} - Fourier coefficient of \tilde{f}_ξ we get (for a.e ξ) $\hat{f}(\xi - k) = 0$ for all $k \in \mathbb{Z}^d$. Hence $f = 0$. □

2.3 Hardy uncertainty principle

We showed previously that \hat{f} and f cannot be simultaneously localized. A natural question is: given $f \neq 0$, how fast f and \hat{f} can decay simultaneously? We already know that if f is the Gaussian function (i.e $f(x) := e^{-\pi ax^2}$ with $a > 0$) then $\hat{f}(\xi) = \frac{1}{\sqrt{a}} e^{-\pi \xi^2/a}$. The next question is: can we find a function $f \neq 0$ such that f and \hat{f} decays faster than a gaussian? The Hardy uncertainty principle gives a negative answer. It says that f and \hat{f} can't decay simultaneously faster than a gaussian. More precisely

Proposition 7. *Let C, C' be two nonnegative constants and let a be a positive constant. Suppose that $f : \mathbb{R} \rightarrow \mathbb{C}$ is a (measurable) function such that $|f(x)| \leq C e^{-\pi ax^2}$ and $|\hat{f}(\xi)| \leq C' e^{-\pi \xi^2/a}$. Then there exists $\bar{C} \geq 0$ such that*

$$f(x) = \bar{C} e^{-\pi ax^2}. \quad (2.6)$$

Proof. We follow the exposition of [7]. By using (1.3) if necessary we may assume WLOG that $a = 1$. By considering $g = cf$ with $c > 0$ a small constant if necessary we may also assume WLOG the $C = C' = 1$.

Observe that in view of the decay of the function f as $|x| \rightarrow \infty$, we may extend the definition of the Fourier transform to the complex plane by letting $\hat{f}(z) := \int_{\mathbb{R}} e^{-2i\pi zx} f(x) dx$. Observe that \hat{f} is analytic: this follows again from Morera theorem and Fubini theorem (see also proof of Proposition 3). By completing the square we get

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

Hence if $F(z) := e^{\pi z^2} \hat{f}(z)$ then F is bounded by one on the imaginary axis. Let $0 < \theta < \frac{\pi}{2}$ be close enough to $\frac{\pi}{2}$, $0 < \delta, \epsilon \ll 1$ be small parameters, and $R > 0$ a large number such that all the statements below are correct. Let Γ_θ denote the sector

$$\Gamma_\theta := \{ r e^{i\alpha}, 0 < r \leq R, 0 \leq \alpha \leq \theta \}$$

We would like to apply the maximum principle to the function F on the upper right quadrant $\mathcal{H} \cap B(O, R)$ (Here $B(O, R)$ is the closed ball with radius R and center the

origin, and \mathcal{H} the upper plane) with R that can be arbitrarily large. Nevertheless observe that F is not even bounded on \mathcal{H} . In order to circumvent the difficulty, we proceed as follows. We multiply F by the analytic functions $e^{i\epsilon e^{i\epsilon} z^{2+\epsilon}}$, $e^{i\delta z^2}$ so that the resulting function $G_{\alpha,\epsilon}(z) := e^{i\delta z^2} e^{i\epsilon e^{i\epsilon} z^{2+\epsilon}} F(z)$ is bounded by one on the boundary of the bounded domain $B(O, R) \cap \Gamma_\theta$. Hence by the maximum modulus theorem we see that $G_{\alpha,\epsilon}$ is bounded by one on $\Gamma_{R,\theta}$. By letting $\epsilon \rightarrow 0$, $\theta \rightarrow \frac{\pi}{2}$, $R \rightarrow \infty$, and $\delta \rightarrow 0$, we see that G is bounded by one on \mathcal{H} .

Similar arguments work for the other quadrants.

Hence we may apply Liouville theorem to conclude that F is constant.

□

CHAPTER 3

LOCAL CONSTANCY OF SIGNAL LOCALIZED IN FOURIER DOMAIN AROUND 0 AND CONSEQUENCES

3.1 Local constancy of signal localized in Fourier domain around 0 and consequences

We now explain informally why a signal localized in the Fourier domain around 0 is essentially constant at scales $\ll \frac{1}{R}$.

Indeed let $f \in \mathcal{M}(\mathbb{R}^d)$ such that $\hat{f} = 0$ outside $B(0, R)$. Then from the inversion Fourier formula (1.2) we get for $|x - x_0| \ll \frac{1}{R}$

$$\begin{aligned} f(x) &= \int_{\mathbb{R}^d} \hat{f}(\xi) e^{2i\pi \cdot \xi_0} e^{2i\pi \xi \cdot (x - x_0)} d\xi \\ &\approx \int_{\mathbb{R}^d} \hat{f}(\xi) e^{2i\pi \xi \cdot x_0} d\xi \\ &= f(x_0) \end{aligned}$$

since $|\xi \cdot (x - x_0)| \ll 1$.

We now draw consequences from this informal statement.

3.2 Shannon Sampling Theorem

Since a signal f is essentially constant at scales $\ll \frac{1}{R}$ it is natural to ask oneself whether one can recover a signal localized in the Fourier domain from a sampling of its values at points separated from a distance roughly equal to $\frac{1}{R}$. The Shannon sampling theorem gives a positive answer to this question. It can be formulated as follows:

Theorem 1. *Let $R > 0$ and $f \in L^1(\mathbb{R})$ such that \hat{f} , the Fourier transform of f , is supported on the interval $[-R, R]$.*

Then

$$f(x) = \sum_{n \in \mathbb{Z}} f\left(\frac{n}{2R}\right) \operatorname{sinc}\left(2R\left(x - \frac{n}{2R}\right)\right) \quad (3.1)$$

in the L^∞ - (resp. L^2 sense), i.e the series in the RHS of (3.1) converges to f in $L^\infty(\mathbb{R})$ (resp. $L^2(\mathbb{R})$).

The same conclusions hold if we replace the assumption “ $f \in L^1(\mathbb{R})$ ” with “ $f \in L^2(\mathbb{R})$ ”.

Remark 3. In other words, the sampling theorem, which is often named after Shannon, says that if an absolutely integrable function is band-limited, i.e it contains no frequencies higher than $R > 0$ hertz, then it is completely determined by its samples at a uniform grid spaced at distances $\frac{1}{2R}$ apart via the above formula.

Sampling theory is a tool used for functions to be reconstructed from sampled data, usually from the values of either the functions themselves or some transformations at a discrete set of points.

Remark 4. Let $f \in L^1(\mathbb{R})$ such that $\operatorname{supp}(\hat{f}) \subset [-R, R]$. Then the triangle inequality applied to (1.1) shows that $\hat{f} \in L^\infty(\mathbb{R})$. Hence we see from Hölder inequality that $\hat{f} \in L^2(\mathbb{R}) \cap L^1(\mathbb{R})$. Hence by (1.4) f is almost equal to the continuous function g defined by $g(x) := \int_{\mathbb{R}} \hat{f}(\xi) e^{2i\pi\xi \cdot x} d\xi$. So we may abuse notation in (3.1) by writing “ $f\left(\frac{n}{2R}\right)$ ” for “ $g\left(\frac{n}{2R}\right)$ ”.

Proof. We see from Remark 4 that the Fourier series of \hat{f} converges to \hat{f} in $L^2([-R, R])$: see e.g [6]. We can write

$$\hat{f}(\xi) = \sum_{n \in \mathbb{Z}} c_n e^{\frac{in\pi\xi}{R}},$$

where the equality holds in the $L^2([-R, R])$ - sense. In the expression above c_n is the n - the Fourier coefficient of \hat{f} , i.e

$$c_n := \frac{1}{2R} \int_{-R}^R \hat{f}(\xi) e^{-\frac{in\pi\xi}{R}} d\xi.$$

Hence, using also (1.4), we get $c_n = \frac{1}{2R} f\left(-\frac{n}{2R}\right)$. So

$$\hat{f}(\xi) = \mathbb{1}_{[-R, R]} \sum_{n \in \mathbb{Z}} f\left(\frac{n}{2R}\right) e^{\frac{-in\pi\xi}{R}}, \quad (3.2)$$

where the equality holds in the $L^2(\mathbb{R})$ - sense. Hence we get from (1.4) and the Cauchy-Schwartz inequality

$$\begin{aligned} f(x) &= \frac{1}{2R} \sum_{n \in \mathbb{Z}} f\left(\frac{n}{2R}\right) \int_{-R}^R e^{2i\pi\xi\left(x - \frac{n}{2R}\right)} d\xi \\ &= \sum_{n \in \mathbb{Z}} f\left(\frac{n}{2R}\right) \operatorname{sinc}\left(2R\left(x - \frac{n}{2R}\right)\right), \end{aligned} \quad (3.3)$$

where the equality holds in the L^∞ - sense. Observe that $f \in L^2(\mathbb{R}^d)$ since $f = \hat{h}$ with $h(\xi) = \hat{f}(-\xi)$ and we can use the results of Section 1 to $h \in L^1(\mathbb{R}^d) \cap L^2(\mathbb{R}^d)$.

The equality (3.3) also holds in the L^2 - sense: it follows from the Plancherel theorem, (3.2), and (3.3).

In the case where $f \in L^2(\mathbb{R})$, we have $\hat{f} \in L^1(\mathbb{R}) \cap L^2(\mathbb{R})$ by Hölder inequality. From Section 1 we see that (1.4) holds. Hence, by following the same steps above, we infer that (3.1) holds in the L^∞ - sense and in the L^2 - sense.

□

3.3 Bernstein Inequality

Since f is essentially constant at scales $\ll \frac{1}{R}$, then (formally)

$$f \approx \sum_{\vec{i}=(i_1, \dots, i_d) \in \mathbb{Z}^d} f\left(\frac{i_1}{R}, \dots, \frac{i_d}{R}\right) \mathbb{1}_{Q_{\vec{i}}}$$

with $Q_{\vec{i}} := \left[\frac{i_1}{R}, \frac{i_1+1}{R}\right] \times \dots \times \left[\frac{i_d}{R}, \frac{i_d+1}{R}\right]$, and therefore $\|f\|_{L^\infty(\mathbb{R}^d)} \lesssim R^{\frac{d}{p}} \|f\|_{L^p(\mathbb{R}^d)}$. So it should be possible to control the $L^\infty(\mathbb{R}^d)$ - norm (and more generally the highest $L^q(\mathbb{R}^d)$ - norms) by lower $L^p(\mathbb{R}^d)$ - norms. The Bernstein inequality shows that this is indeed the case:

Proposition 8. *Let $1 \leq p \leq q \leq \infty$, $R > 0$, and $s \in \mathbb{R}$. Let $f \in L^p(\mathbb{R}^d)$. The following hold:*

1. *Assume that \hat{f} is compactly supported on $\{\xi \in \mathbb{R}^d : |\xi| \leq R\}$. Then*

$$\|f\|_{L^q(\mathbb{R}^d)} \lesssim R^{d\left(\frac{1}{p} - \frac{1}{q}\right)} \|f\|_{L^p(\mathbb{R}^d)} \quad (3.4)$$

2. *Assume that \hat{f} is compactly supported on the annulus $\{\xi \in \mathbb{R}^d : \frac{R}{2} \leq |\xi| \leq R\}$.*

Then

$$\|f\|_{L^p(\mathbb{R}^d)} \lesssim R^{-s} \|D^s f\|_{L^p(\mathbb{R}^d)} \quad (3.5)$$

Remark 5. *Observe that (3.4) does not hold for all $f \in L^p(\mathbb{R}^d) \cap L^\infty(\mathbb{R}^d)$. Indeed, arguing by contradiction, let $f_n(x) := 1$ if $|x| \leq \frac{1}{2^n}$ and $f_n(x) := 0$ if $|x| > \frac{1}{2^n}$. Then $\|f_n\|_{L^p(\mathbb{R}^d)} = \frac{1}{2^{\frac{nd}{p}}}$ and $\|f_n\|_{L^\infty(\mathbb{R}^d)} = 1$. This contradicts (3.4) as $n \rightarrow \infty$.*

Proof. Let ϕ be a smooth function such that $\phi(\xi) = 1$ for $|\xi| \leq 1$ and $\phi(\xi) = 0$ for $|\xi| \geq 2$. Let $\tilde{\phi} := |\xi|^{-s} (\phi(\xi) - \phi(4\xi))$. If $h \in \{\phi, \tilde{\phi}\}$ then let $\check{h}(x) := \int_{\mathbb{R}^d} h(\xi) e^{2i\pi\xi \cdot x} d\xi$.

We first prove (3.4). By using the dilation rule of (1.3) if necessary we may assume WLOG that $R = 1$. Indeed let $D_{\frac{1}{R}}f$ satisfies is compactly supported on $\{\xi \in \mathbb{R}^d : |\xi| \leq 1\}$. Hence

$$\|f\|_{L^q(\mathbb{R}^d)} = R^{-\frac{d}{q}} \|D_{\frac{1}{R}}f\|_{L^q(\mathbb{R}^d)} \lesssim R^{\frac{d}{q}} \|D_{\frac{1}{R}}f\|_{L^p(\mathbb{R}^d)} \lesssim R^{d(\frac{1}{p} - \frac{1}{q})} \|f\|_{L^p(\mathbb{R}^d)}$$

Hence $\hat{f} = \phi \hat{f}$, we get $f = f * \check{\phi}$; furthermore, an application of the Young inequality (with \bar{p} such that $\frac{1}{q} + 1 = \frac{1}{p} + \frac{1}{\bar{p}}$) shows that

$$\|f\|_{L^q(\mathbb{R}^d)} \lesssim \|f\|_{L^p(\mathbb{R}^d)} \|\check{\phi}\|_{L^{\bar{p}}(\mathbb{R}^d)} \lesssim \|f\|_{L^p(\mathbb{R}^d)}$$

We then prove (3.5). Again, by using the dilation rule of (1.3) if necessary we may assume WLOG that $R = 1$. Now write $\hat{f} = \tilde{\phi} |\xi|^s \hat{f}$. Hence $f = \check{\phi} * D^s f$ and, taking into account that $\check{\phi} \in \mathcal{S}(\mathbb{R}^d)$, an application of the Young inequality yields

$$\|f\|_{L^p(\mathbb{R}^d)} \lesssim \|D^s f\|_{L^p(\mathbb{R}^d)} \|\check{\phi}\|_{L^1(\mathbb{R}^d)} \lesssim \|D^s f\|_{L^p(\mathbb{R}^d)}$$

□

Remark 6. *Alternatively one can use Hölder inequality to conclude. Indeed from $f = f * \phi$, the definition of the convolution, and Hölder inequality we get $\|f\|_{L^\infty(\mathbb{R}^d)} \lesssim \|f\|_{L^p(\mathbb{R}^d)} \|\check{\phi}\|_{L^{p'}(\mathbb{R}^d)} \lesssim \|f\|_{L^p(\mathbb{R}^d)}$. Hence by interpolation*

$$\|f\|_{L^q(\mathbb{R}^d)} \lesssim \|f\|_{L^p(\mathbb{R}^d)}^{\frac{p}{q}} \|f\|_{L^\infty(\mathbb{R}^d)}^{1-\frac{p}{q}} \lesssim \|f\|_{L^p(\mathbb{R}^d)}.$$

Remark 7. *We claim that the estimates (3.4) and (3.5) are sharp.*

Indeed let h be a smooth function such that $h(\xi) = 1$ for $|\xi| < \frac{3}{4}$ and $h(\xi) = 0$ for $|\xi| \geq 1$. Then $\check{h} \in \mathcal{S}(\mathbb{R}^d)$ (with $\check{h}(x) := \int_{\mathbb{R}^d} h(\xi) e^{2i\pi\xi \cdot x} d\xi$) since $\hat{h}(x) := \hat{h}(-x)$.

Let $q_R(x) := R^d \check{q}(Rx)$. We prove (3.4). Let $q := h$ with ϕ defined in Subsection 4.1.

Then $\widehat{q_R}(\xi) = \phi(\frac{\xi}{R})$ is supported on $|\xi| \leq R$. Moreover an elementary change of

variable shows that for $r \in \{p, q\}$ $\|q_R\|_{L^r(\mathbb{R}^d)} \approx R^{d(1-\frac{1}{r})}$ (Observe that $\check{q} \neq 0$ on a ball with center O and small positive radius by Proposition 2). Hence $\|q_R\|_{L^q(\mathbb{R}^d)} \approx$

$R^{d(\frac{1}{p} - \frac{1}{q})} \|q_R\|_{L^p(\mathbb{R}^d)}$. We now prove (3.5). Let q be such that $q(\xi) := h(\xi) - h(\frac{5\xi}{4})$.

Then we have $D^s q_R(x) = R^{d+s} D^s \check{q}(Rx)$ and $\|D^s q_R\|_{L^p(\mathbb{R}^d)} \approx R^s \|q_R\|_{L^p(\mathbb{R}^d)}$ (again, observe that $D^s \check{q} \neq 0$ on a ball with center O and small and positive radius by Proposition 2). Hence $\|D^s q_R\|_{L^p(\mathbb{R}^d)} \approx R^s \|q_R\|_{L^p(\mathbb{R}^d)}$.

CHAPTER 4

LITTLEWOOD-PALEY DECOMPOSITION

So far we have studied properties of signals that are band-limited, i.e localized in the Fourier domain. It is unfortunate that most of the signals (solutions of PDES, signals in image processing, etc.) are not usually band-limited. So we perform a Littlewood-Paley decomposition. Roughly speaking, a Littlewood-Paley decomposition is a particular way of decomposing the phase plane which takes a function and writes it as a superposition of a countably infinite family of functions of varying frequencies. The Littlewood-Paley decomposition is of interest in multiple areas of mathematics and forms the basis for the so-called Littlewood-Paley theory.

More precisely we first define for $N \in 2^{\mathbb{Z}}$ a Littlewood-Paley projector P_N : this map, when applied to a function f , yields a Littlewood-Paley piece of f , i.e a function $P_N f$ that is localized in the Fourier domain in an annulus $|\xi| \approx N$. By defining appropriately P_N we can prove that we can decompose f into its Littlewood-Paley pieces $P_N f$, i.e $f = \sum_{N \in 2^{\mathbb{Z}}} P_N f (*)$.

Then we establish decay estimates on average in the physical space for $P_N f$.

Finally we use these estimates in the decomposition (*): we get general (by “general” we mean estimates that hold for all functions f and not only functions f that are band-limited) estimates for f by summing over $N \in 2^{\mathbb{Z}}$.

4.1 Setting

Let $\phi(\xi)$ be a radial bump function supported on $\{\xi \in \mathbb{R}^d : |\xi| \leq 2\}$ which equals 1 on $\{\xi \in \mathbb{R}^d : |\xi| \leq 1\}$. Let $\psi(\xi)$ be the function

$$\psi(\xi) := \phi(\xi) - \phi(2\xi).$$

Thus ψ is a bump function supported on the annulus $\{\xi \in \mathbb{R}^d : \frac{1}{2} \leq |\xi| \leq 2\}$. By construction we have

$$\sum_{N \in 2^{\mathbb{Z}}} \psi\left(\frac{\xi}{N}\right) = 1 \tag{4.1}$$

for all $\xi \in \mathbb{R}^d$. Thus we can partition the unity into functions $\psi\left(\frac{\xi}{N}\right)$, with $N \in 2^{\mathbb{Z}}$ (the set of dyadic numbers), each of those is supported on the annulus of the form $|\xi| \sim N$. We now introduce the Littlewood-Paley decomposition operators $P_N, P_{\leq N}$ defined by

$$\widehat{P_N f}(\xi) = \psi\left(\frac{\xi}{N}\right) \hat{f}(\xi)$$

$$\widehat{P_{\leq N} f}(\xi) = \phi\left(\frac{\xi}{N}\right) \hat{f}(\xi)$$

Hence, from the equality $\widehat{p * q} = \hat{p} \hat{q}$ we get

$$P_N f(x) = N^d \int_{\mathbb{R}^d} f(y) \check{\psi}(N(x-y)) dy \text{ and}$$

$$\widehat{P_{\leq N} f}(\xi) = N^d \int_{\mathbb{R}^d} f(y) \check{\phi}(N(x-y)) dy.$$

We see from (4.1) that we have the Littlewood-Paley decomposition

$$f = \sum_{N \in 2^{\mathbb{Z}}} P_N f.$$

This decomposition takes a single function and writes it as a superposition of a countably infinite family of functions $P_N f$, where each one has a frequency of magnitude roughly N . Lower values of N represent lower frequency components of f , while higher values represent high frequency components.

What does $P_N f$ look like? Since $P_N f = P_{\leq 4N} P_N f$, we see from that we have the self-reproducing formula

$$P_N f(x) = (4N)^d \int_{\mathbb{R}^d} P_N f(y) \check{\phi}(4N(x-y)) dy \quad (4.2)$$

Hence $P_N f(x) \approx N^d \int_{N|x-y| \lesssim 1} P_N f(y) dy$ since $\check{\phi}$ is essentially concentrated in a neighborhood of size $o(1)$ around the origin, which means that $P_N f$ is essentially constant at scales $\ll \frac{1}{N}$. On the other hand $P_{\leq \frac{N}{4}} P_N f = 0$. Hence

$$\int_{\mathbb{R}^d} P_N f(y) \check{\phi}\left(\frac{x-y}{4}\right) dy = 0$$

Hence $\int_{N|x-y| \lesssim 1} P_N f(y) dy \approx 0$, which implies that $P_N f$ has essentially a mean equal to zero at scales $\lesssim \frac{1}{N}$; in other words $P_N f$ has $O(1)$ oscillations at these scales.

To reconcile these two properties, we see that on each ball of radius $O(N^{-1})$, the function $P_N f$ is roughly constant at scales $O(N^{-1})$ and it contains about $O(1)$ oscillations.

What does $P_{\leq N} f$ look like? Since $P_{\leq N} f = P_{\leq 4N} P_{\leq N} f$, we see by using similar arguments as before that $P_{\leq N}$ is essentially constant at scales $\ll N^{-1}$.

What does $P_{>N}f$ look like? Since $P \leq \frac{N}{4}$ $P_{>N}f = 0$, we see that $P_{>N}$ has $O(1)$ oscillations at scales $O(N^{-1})$.

4.2 Estimates for Littlewood-Paley projector

In this subsection we prove some quantitative estimates for functions that are localized in the Fourier domain within the annulus $\{\xi \in \mathbb{R}^d : |\xi| \approx N\}$ with $N \in 2^{\mathbb{Z}}$.

Proposition 9. *Let $f \in \mathcal{S}(\mathbb{R}^d)$. Let $N \in 2^{\mathbb{Z}}$. Then the following hold:*

1.

$$\|P_N f\|_{L^p(\mathbb{R}^d)} \lesssim N^{-1} \|\nabla f\|_{L^p(\mathbb{R}^d)}. \quad (4.3)$$

2. Let $s \in \mathbb{R}$. Then

$$\|P_N f\|_{L^p(\mathbb{R}^d)} \lesssim N^{-s} \|D^s f\|_{L^p(\mathbb{R}^d)}. \quad (4.4)$$

Proof. First we prove (4.3).

From the formula $\widehat{\partial_{x_j} f}(\xi) = 2i\pi \xi_j \widehat{f}(\xi)$ we get $P_N f = \frac{1}{N} \sum_{j=1}^d K_{j,N} * \partial_{x_j} f$ with $K_{j,N}(x) := \frac{1}{2i\pi} \int_{\mathbb{R}^d} \frac{\xi_j}{N} \left(\frac{N}{|\xi|}\right)^2 \psi\left(\frac{\xi}{N}\right) e^{2i\pi \xi \cdot x} d\xi$. We claim that (*) holds with (*) : $\|K_{j,N}\|_{L^1(\mathbb{R}^d)} \lesssim 1$. Assuming that the claim holds applying the Young inequality we see that (4.3) also holds. It remains to prove (*). The triangle inequality and an elementary change of variable show that $|K_{j,N}(x)| \lesssim N^d$. Moreover by integration by parts using the formula $e^{2i\pi \xi \cdot x} = \frac{1}{2i\pi} \frac{\nabla(e^{2i\pi \xi \cdot x}) \cdot x}{|x|^2}$ we get $|K_{j,N}(x)| \lesssim \frac{N^d}{|Nx|^{100d}}$. Hence

$$|K_{j,N}(x)| \lesssim N^d \min\left(1, \frac{1}{|Nx|^{100d}}\right). \quad (4.5)$$

Observe that the R.H.S of (4.5) is integrable. Hence (*) holds.

Then we prove (4.4). Let $\bar{\psi}(\xi) := |\xi|^{-s} \psi(\xi)$. From $\widehat{P_N f}(\xi) = N^{-s} \bar{\psi}\left(\frac{\xi}{N}\right) |\xi|^s \widehat{f}(\xi)$ we get

$$P_N f = N^{-s} N^d \int_{\mathbb{R}^d} \check{\bar{\psi}}(N(x-y)) D^s f(y) dy$$

Hence $\|P_N f\|_{L^p(\mathbb{R}^d)} \lesssim N^{-s} \|D^s f\|_{L^p(\mathbb{R}^d)}$. □

Next we state and prove the proposition below stronger for functions such that their Fourier transform is localized on $|\xi| \approx N$.

Proposition 10. *Let $s \in \mathbb{R}$. Let $f \in \mathcal{S}(\mathbb{R}^d)$ such that $\text{supp}(\hat{f}) \subset \{\xi \in \mathbb{R}^d : |\xi| \approx N\}$. Let $p \geq 1$. Then*

$$\|\nabla f\|_{L^p(\mathbb{R}^d)} \approx N\|f\|_{L^p(\mathbb{R}^d)}, \text{ and } \|D^s f\|_{L^p(\mathbb{R}^d)} \approx N^s\|f\|_{L^p(\mathbb{R}^d)}$$

Proof. Since $\text{supp}(\hat{f}) \subset \{\xi \in \mathbb{R}^d : |\xi| \approx N\}$, we can write $f = \tilde{P}_N f$ with \tilde{P}_N a map defined by $\widehat{\tilde{P}_N(f)} := \psi\left(\frac{\xi}{CN}\right)\hat{f}(\xi)$ and ψ defined in Subsection 4.1 and $C \gg 1$ a large positive constant.

We see from Proposition 9 that

$$\|f\|_{L^p(\mathbb{R}^d)} \lesssim N^{-1}\|\nabla f\|_{L^p(\mathbb{R}^d)}, \text{ and } \|f\|_{L^p(\mathbb{R}^d)} \lesssim N^{-s}\|D^s f\|_{L^p(\mathbb{R}^d)}.$$

Let $\tilde{\psi}(\xi) := |\xi|^s \psi\left(\frac{\xi}{C}\right)$. We have

$$D^s f = N^s \int_{\mathbb{R}^d} \check{\psi}(N(x-y)) f(y) dy$$

Hence from Young inequality we get $\|D^s f\|_{L^p(\mathbb{R}^d)} \lesssim N^s\|f\|_{L^p(\mathbb{R}^d)}$.

From the elementary estimate $\|\nabla f\|_{L^p(\mathbb{R}^d)} \lesssim \sum_{j=1}^d \|\partial_{x_j} f\|_{L^p(\mathbb{R}^d)}$, we see that it suffices to estimate $\|\partial_{x_j} f\|_{L^p(\mathbb{R}^d)}$. From $f = \tilde{P}_N f$ we get, after differentiation with respect to x_j

$$\partial_{x_j} f(x) = (CN)^{d+1} \int_{\mathbb{R}^d} \partial_{x_j} \check{\psi}(CN(x-y)) f(y) dy.$$

Hence $\|\partial_{x_j} f\|_{L^p(\mathbb{R}^d)} \lesssim N\|f\|_{L^p(\mathbb{R}^d)}$.

□

CHAPTER 5

APPLICATIONS

In this section we use the Littlewood-Paley decomposition to prove estimates of the form

$$\begin{aligned} \|f\|_{L^r(\mathbb{R}^d)} &\leq A \left(\|f\|_{L^p(\mathbb{R}^d)} + \|\nabla f\|_{L^p(\mathbb{R}^d)} \right), \text{ and more generally} \\ \|f\|_{L^r(\mathbb{R}^d)} &\leq A \left(\|f\|_{L^p(\mathbb{R}^d)} + \|D^s f\|_{L^p(\mathbb{R}^d)} \right), \end{aligned} \quad (5.1)$$

$f \in \mathcal{S}(\mathbb{R}^d)$, (p, r) and $A > 0$ positive constants that do not depend on f and to be determined. The main interest of these estimates is their robustness: they can be applied to a large class of functions. Hence they are useful in

- *signal processing.*

For example, assume that for some $B > 0$ the signals f satisfy the regularity property $\|f\|_{L^p(\mathbb{R}^d)} + \|D^s f\|_{L^p(\mathbb{R}^d)} \leq B$; then we see from (5.1) that they also have the decay property $\|f\|_{L^r(\mathbb{R}^d)} \lesssim B$

- *PDEs or functional analysis.*

The existence of a solution of a PDE (or a function that satisfies some properties) is often related to the existence of a limit of a sequence of functions. For example, let $\{u_n\}_{n \geq 1}$ be a Cauchy sequence of Schwartz functions with respect to the norm $N(f) := \|f\|_{L^p(\mathbb{R}^d)} + \|\nabla f\|_{L^p(\mathbb{R}^d)}$. Then we see from (5.1) that $\{u_n\}_{n \geq 1}$ is also a Cauchy sequence in $L^r(\mathbb{R}^d)$. Since $L^r(\mathbb{R}^d)$ is a Banach space, there exists $u \in L^r(\mathbb{R}^d)$ such that $u_n \rightarrow u$ in $L^r(\mathbb{R}^d)$.

5.1 An application: the non-endpoint of the Sobolev embedding

In this subsection we show the non-endpoint of the Sobolev embedding.

Proposition 11. *Let $f \in \mathcal{S}(\mathbb{R}^d)$. Then the following hold:*

1. *Let $1 \leq p < d$ and \bar{p} such that $\frac{1}{\bar{p}} = \frac{1}{p} - \frac{1}{d}$. Let $q < \bar{p}$. Then*

$$\|f\|_{L^q(\mathbb{R}^d)} \lesssim \|f\|_{L^p(\mathbb{R}^d)} + \|\nabla f\|_{L^p(\mathbb{R}^d)}. \quad (5.2)$$

2. More generally let $s > 0$, $1 \leq p < \frac{d}{s}$, and \bar{p} such that $\frac{1}{\bar{p}} = \frac{1}{p} - \frac{s}{d}$. Let $q < \bar{p}$. Then

$$\|f\|_{L^q(\mathbb{R}^d)} \lesssim \|f\|_{L^p(\mathbb{R}^d)} + \|D^s f\|_{L^p(\mathbb{R}^d)}. \quad (5.3)$$

Remark 8. Proposition 11 says that the estimate (5.2) holds for q lying in the interval $[p, \bar{p})$ that does not contain its endpoint \bar{p} . This is why this estimate is called the non-endpoint of the Sobolev embedding. Later we will see that (5.2) also holds for $q = \bar{p}$: this is the endpoint of (5.2).

Recall from Proposition 15 (see Appendix C) that if $1 < p < \infty$ then $\|Df\|_{L^p(\mathbb{R}^d)} \approx \|\nabla f\|_{L^p(\mathbb{R}^d)}$. Hence, if $1 < p < \frac{d}{s}$ then (5.3) is a generalization of (5.2).

Proof. We only prove (5.2), since the proof of (5.3) is similar and therefore left to the reader.

The conclusion clearly holds if $q = p$. So we may assume WLOG that $q > p$.

We use a Paley-Littlewood decomposition. We have

$$f = \sum_{N \in 2^{\mathbb{Z}}} P_N f = \sum_{N \leq 1} P_N f + \sum_{N > 1} P_N f = I + J$$

We have

$$\|I\|_{L^q(\mathbb{R}^d)} \lesssim \sum_{N \leq 1} \|P_N f\|_{L^q(\mathbb{R}^d)} \lesssim \sum_{N \leq 1} N^{d(\frac{1}{p} - \frac{1}{q})} \|P_N f\|_{L^p(\mathbb{R}^d)} \lesssim \|f\|_{L^p(\mathbb{R}^d)}$$

Hence $\|I\|_{L^q(\mathbb{R}^d)} \lesssim \|f\|_{L^p(\mathbb{R}^d)}$ and

$$\|J\|_{L^q(\mathbb{R}^d)} \lesssim \sum_{N > 1} \|P_N f\|_{L^q(\mathbb{R}^d)} \lesssim \sum_{N > 1} N^{-1+d(\frac{1}{p} - \frac{1}{q})} \|\nabla f\|_{L^p(\mathbb{R}^d)} \lesssim \|\nabla f\|_{L^p(\mathbb{R}^d)}.$$

Hence (5.2) holds. □

5.2 An application: the endpoint of the Sobolev embedding

In this subsection we prove the endpoint of the Sobolev embedding. More precisely

Proposition 12. Let $f \in \mathcal{S}(\mathbb{R}^d)$. Then the following hold

1. Let $1 \leq p < d$ and \bar{p} be such that $\frac{1}{\bar{p}} = \frac{1}{p} - \frac{1}{d}$.

$$\|f\|_{L^{\bar{p}}(\mathbb{R}^d)} \lesssim \|\nabla f\|_{L^p(\mathbb{R}^d)} \quad (5.4)$$

2. More generally let $s > 0$, $1 < p < \frac{d}{s}$, and \bar{p} such that $\frac{1}{\bar{p}} = \frac{1}{p} - \frac{s}{d}$. Then

$$\|f\|_{L^{\bar{p}}(\mathbb{R}^d)} \lesssim \|D^s f\|_{L^p(\mathbb{R}^d)}. \quad (5.5)$$

Remark 9. We see from the arguments used in Remark 8 that if $1 < p < \frac{d}{s}$ then (5.5) is a generalization of (5.4).

We now write three different proofs of these estimates.

5.2.1 The proof of the endpoint Sobolev embedding for $p > 1$

We see from Remark 9 that it suffices to prove (5.5).

We follow an exposition in [8].

By using (1.3) if necessary (and choosing appropriately the value of λ) we may assume WLOG that $\|D^s f\|_{L^p(\mathbb{R}^d)} = 1$.

We first prove the weak-type estimate

$$\|f\|_{L^{\bar{p},\infty}(\mathbb{R}^d)} \lesssim \|D^s f\|_{L^p(\mathbb{R}^d)} \quad (5.6)$$

(Here $\|f\|_{L^{\bar{p},\infty}(\mathbb{R}^d)}^{\bar{p}} := \sup_{\lambda > 0} \lambda^{\bar{p}} | \{ |f| > \lambda \} |$). Let $\lambda > 0$ be a fixed positive number. Let $N \in 2^{\mathbb{Z}}$. From

$$\|P_N f\|_{L^\infty(\mathbb{R}^d)} \lesssim N^{\frac{d}{p}-s} \|D^s f\|_{L^p(\mathbb{R}^d)} \lesssim N^{\frac{d}{p}-s}. \quad (5.7)$$

we see that if N_λ is such that $N_\lambda^{\frac{d}{p}-s} \approx \lambda$ then by summing (5.7) over $N \ll N_\lambda$ we get

$$\|P_{\ll N_\lambda} f\|_{L^\infty} \ll \lambda$$

On the other hand we see from Proposition 9 that

$$\|P_{\gtrsim N_\lambda} f\|_{L^p(\mathbb{R}^d)} \lesssim N_\lambda^{-s} \|D^s f\|_{L^p(\mathbb{R}^d)} \lesssim N_\lambda^{-s}$$

Therefore using also Chebyshev inequality we get

$$\lambda^{\bar{p}} | \{ |f| > \lambda \} | \lesssim \lambda^{\bar{p}} | \{ P_{\gtrsim N_\lambda} f > \lambda \} | \lesssim \lambda^{\bar{p}-p} \|P_{\gtrsim N_\lambda}\|_{L^p(\mathbb{R}^d)}^p \lesssim \lambda^{\bar{p}-p} N_\lambda^{-ps} \lesssim 1,$$

where at the last line we use the definition of \bar{p} . Hence (5.6) holds.

It remains to upgrade the weak-type estimates to strong-type estimates. For his purpose we recall the Marcinkiewicz interpolation theorem (see e.g [2])

Theorem 2. Let $(p_0, p_1, q_0, q_1) \in [1, \infty]^4$. Let $\theta \in (0, 1)$. Let (p, q) such that

$$\frac{1}{p} = \frac{1-\theta}{p_0} + \frac{\theta}{p_1}, \text{ and } \frac{1}{q} = \frac{1-\theta}{q_0} + \frac{\theta}{q_1}.$$

Assume that T is a linear map on L^{p_0} weak-type (p_0, q_0) and (p_1, q_1) i.e

$$\left(\sup_{\lambda > 0} \lambda^{q_0} |Tf| > \lambda \right)^{\frac{1}{q_0}} \lesssim \|f\|_{L^{p_0}(\mathbb{R}^d)}, \text{ and } \left(\sup_{\lambda > 0} \lambda^{q_1} |Tf| > \lambda \right)^{\frac{1}{q_1}} \lesssim \|f\|_{L^{p_1}(\mathbb{R}^d)} \dots$$

Then T is strong-type (p, q) , i.e

$$\|Tf\|_{L^q(\mathbb{R}^d)} \lesssim \|f\|_{L^p(\mathbb{R}^d)}.$$

By applying the Marcinkiewicz interpolation theorem to $T = D^{-s}$ we see from the previous estimates that T is weak type (p, \bar{p}) . Hence T is strong type (p, \bar{p}) and we can conclude that (5.5) holds.

Remark 10. Assume that we want to prove (5.4) without using Remark 9 and by using similar arguments as those above.

Similar arguments show that $(*)$: $\|\Delta^{-1} \nabla \cdot \vec{E}\|_{L^{\bar{p}}(\mathbb{R}^d)} \lesssim \|\vec{E}\|_{L^p(\mathbb{R}^d)}$ holds (Here $\vec{E} := (E_1, \dots, E_d)$, $\nabla \cdot \vec{E} := \sum_{j=1}^d \frac{\partial E_j}{\partial x_j}$ and $\widehat{\Delta^{-1} f}(\xi) := |\xi|^{-2} \hat{f}(\xi)$); then, applying $(*)$ to $\vec{E} := \nabla f$ yields (5.4).

5.2.2 Another proof of the endpoint Sobolev embedding for $p > 1$: fractional integration

In this subsection we write down another proof based upon a physical representation of the fractional integration of a function f .

Observe that (5.5) holds if and only if $\|D^{-s} f\|_{L^{\bar{p}}(\mathbb{R}^d)} \lesssim \|f\|_{L^p(\mathbb{R}^d)}$. From Appendix A (see (10)) we can write $D^{-s} f(x) = \int_{\mathbb{R}^d} f(x-y) |y|^{s-d} dy = K_1(x) + K_2(x)$ with

$$K_1(x) := \int_{|x-y| < R} f(y) |x-y|^{s-d} dy, \text{ and } K_2(x) := \int_{|x-y| \geq R} f(y) |x-y|^{s-d} dy$$

We have

$$|K_1(x)| \lesssim \sum_{M \in 2^{\mathbb{Z}}: M \lesssim R} M^s \frac{1}{|B(x, R)|} \int_{|x-y| \approx M} |f(y)| dy \lesssim R^s M f(x)$$

Moreover the Hölder inequality shows that

$$|K_2(x)| \lesssim \left(\int_{|y| \leq R} |y|^{(s-d)p'} dy \right)^{\frac{1}{p'}} \|f\|_{L^p(\mathbb{R}^d)} \lesssim R^{-\frac{d}{q}} \|f\|_{L^p(\mathbb{R}^d)}$$

Therefore

$$I_s(f)(x) \lesssim R^s M(f)(x) + R^{-\frac{d}{q}} \|f\|_{L^p(\mathbb{R}^d)} \quad (5.8)$$

(Here $M(f)(x) := \sup_{R>0} \int_{B(x,R)} |f(y)| dy$ is the well-known maximal function: see e.g [2]). By choosing R such that $R^s M(f)(x) \approx R^{-\frac{d}{q}} \|f\|_{L^p(\mathbb{R}^d)}$ in order to minimize the R.H.S of (5.8), we get

$$I_s(f) \lesssim (M(f)(x))^{\frac{p}{q}} \|f\|_{L^p(\mathbb{R}^d)}^{1-\frac{1}{q}}$$

Hence, by taking the $L^{\bar{p}}$ norm of $I_s(f)$ and using the well-known estimate $\|M(f)\|_{L^r(\mathbb{R}^d)} \lesssim \|f\|_{L^r(\mathbb{R}^d)}$ for $1 < r \leq \infty$ (see e.g [2]) we get

$$\|I_s(f)\|_{L^{\bar{p}}(\mathbb{R}^d)} \lesssim \|f\|_{L^p(\mathbb{R}^d)}.$$

5.2.3 A proof of (5.4) for $p = 1$

In this subsection we give a proof of (5.4) for $p = 1$ by working in the physical space. The techniques used also yield another proof of (5.4) for $p > 1$.

We first state and prove a technical lemma that we will use later in the proof of (5.4) for $p = 1$.

Lemma 1. *Let $f_1, \dots, f_d \in L^{d-1}(\mathbb{R}^{d-1})$. Given $x \in \mathbb{R}^d$ and $1 \leq i \leq d$, let $\bar{x}_i := (x_1, \dots, x_{i-1}, x_{i+1}, \dots, x_d)$. Let $f(x) := f_1(\bar{x}_1) \dots f_d(\bar{x}_d)$. Then $f \in L^1(\mathbb{R}^d)$ and*

$$\|f\|_{L^1(\mathbb{R}^d)} \lesssim \prod_{i=1}^d \|f_i\|_{L^{d-1}(\mathbb{R}^{d-1})}.$$

Proof. The case $d = 2$ follows from Fubini-Tonelli theorem.

Let $d = 3$. The Cauchy-Schwartz inequality yields

$$\int_{\mathbb{R}} |f(x)| dx_3 \lesssim |f_3(x_1, x_2)| \left(\int_{\mathbb{R}} |f_1(x_2, x_3)|^2 dx_3 \right)^{\frac{1}{2}} \left(\int_{\mathbb{R}} |f_2(x_1, x_3)|^2 dx_3 \right)^{\frac{1}{2}}$$

Integrating the above estimate with respect to x_2 and x_1 and applying the Cauchy-Schwarz inequality again we see that

$$\int_{\mathbb{R}^3} |f(x)| dx \lesssim \left(\int_{\mathbb{R}^3} |f_3(x_1, x_2)|^2 dx_2 dx_1 \right)^{\frac{1}{2}} \left(\int_{\mathbb{R}^2} |f_1(x_2, x_3)|^2 dx_3 dx_2 \right)^{\frac{1}{2}} \left(\int_{\mathbb{R}^2} |f_2(x_1, x_3)|^2 dx_3 dx_1 \right)^{\frac{1}{2}}$$

The general case d is proved by induction. Assuming that Lemma 1 holds for p let us prove it for $p + 1$. The Hölder inequality yields

$$\int_{\mathbb{R}^d} |f(x)| dx \lesssim \|f_{d+1}\|_{L^d(\mathbb{R}^d)} \left(\int_{\mathbb{R}^d} |f_1 \dots, f_d|^{d'} dx_1 \dots dx_d \right)^{\frac{1}{d'}},$$

with d' such that $\frac{1}{d} + \frac{1}{d'} = 1$. Applying the induction assumption to the functions $|f_1|^{d'}, \dots, |f_d|^{d'}$, we obtain

$$\int_{\mathbb{R}^d} |f_1|^{d'} \dots |f_d|^{d'} dx_1 \dots dx_d \lesssim \prod_{j=1}^d \|f_j\|_{L^d(\mathbb{R}^d)}^{d'}$$

Hence

$$\int_{\mathbb{R}^d} |f(x)| dx_1 \dots dx_d \lesssim \|f_{d+1}\|_{L^d(\mathbb{R}^d)} \prod_{j=1}^d \|f_j\|_{L^d(\mathbb{R}^{d-1})}$$

We now integrate the above estimate with respect to x_{d+1} . Applying again the Hölder inequality we get

$$\int_{\mathbb{R}^{d+1}} |f(x)| dx_1 \dots dx_{d+1} \lesssim \prod_{j=1}^{d+1} \|f_j\|_{L^d(\mathbb{R}^d)}$$

□

Let us prove (5.4) for $p = 1$. The fundamental theorem of calculus shows that

$$|u(x_1, \dots, x_d)| = \left| \int_{-\infty}^{x_1} \frac{\partial u}{\partial x_1}(t, x_2, \dots, x_d) dt \right|$$

In fact we have for $1 \leq i \leq d$

$$|u(x_1, \dots, x_d)| \lesssim \int_{-\infty}^{\infty} \left| \frac{\partial u}{\partial x_i}(x_1, \dots, x_{i-1}, t, x_{i+1}, \dots, x_d) \right| dt$$

Let $\bar{x}_i := (x_1, \dots, x_{i-1}, x_{i+1}, \dots, x_d)$. Let $f_i(\bar{x}_i) := \int_{-\infty}^{\infty} \left| \frac{\partial u}{\partial x_i}(x_1, \dots, x_{i-1}, t, x_{i+1}, \dots, x_d) \right| dt$.

The above estimates show that

$$|u(x)|^d \lesssim \prod_{i=1}^d f_i(\bar{x}_i)$$

Hence, recalling that $\bar{p} = \frac{d}{d-1}$, we get from Lemma 1

$$\int_{\mathbb{R}^d} |u(x)|^{\bar{p}} dx \lesssim \prod_{j=1}^d \|f_j\|_{L^1(\mathbb{R}^{d-1})}^{\frac{1}{d-1}} \lesssim \prod_{j=1}^d \left\| \frac{\partial u}{\partial x_j} \right\|_{L^1(\mathbb{R}^d)}^{\frac{1}{d-1}}$$

Hence

$$\|u\|_{L^{\bar{p}}(\mathbb{R}^d)} \lesssim \prod_{j=1}^d \left\| \frac{\partial u}{\partial x_j} \right\|_{L^1(\mathbb{R}^d)}^{\frac{1}{d}} \lesssim \|\nabla u\|_{L^p(\mathbb{R}^d)}.$$

We now turn to the case $1 < p < d$. We show that a modification of the argument for $p = 1$ yields (5.4). Let $m \geq 1$ to be chosen shortly. An application of (5.4) for $p = 1$ to the function $|u|^{m-1}u$ combined with Hölder inequality shows that

$$\begin{aligned} \|u\|_{L^{\frac{md}{d-1}}(\mathbb{R}^d)}^m &\lesssim \prod_{j=1}^d \left\| |u|^{m-1} u \frac{\partial u}{\partial x_j} \right\|_{L^1(\mathbb{R}^d)} \\ &\lesssim \|u\|_{L^{p'(m-1)}(\mathbb{R}^d)}^{m-1} \prod_{j=1}^d \left\| \frac{\partial u}{\partial x_j} \right\|_{L^p(\mathbb{R}^d)} \end{aligned}$$

(Here $\frac{1}{p'} := 1 - \frac{1}{p}$). Choosing m such that $p'(m-1) = \frac{pd}{d-1}$ we get

$$\|u\|_{L^{p^*}(\mathbb{R}^d)} \lesssim \prod_{j=1}^d \left\| \frac{\partial u}{\partial x_j} \right\|_{L^p(\mathbb{R}^d)}^{\frac{1}{d}} \lesssim \|\nabla u\|_{L^p(\mathbb{R}^d)}.$$

5.3 An application: the Gagliardo-Nirenberg inequality

In this subsection we give another application of the Littlewood-Paley decomposition: the Gagliardo-Nirenberg inequality.

Proposition 13. *Let $f \in \mathcal{S}(\mathbb{R}^d)$. Let $\infty > p \geq 1$, $s > 0$ and $\theta \in (0, 1)$. Let $\frac{1}{q} = \frac{1}{p} - \frac{\theta s}{d}$. Then*

$$\|f\|_{L^q(\mathbb{R}^d)} \lesssim \|f\|_{L^p(\mathbb{R}^d)}^\theta \|D^s f\|_{L^p(\mathbb{R}^d)}^{1-\theta} \quad (5.9)$$

Remark 11. *The Gagliardo-Nirenberg inequality is usually better than the non-endpoint of the Sobolev embedding in situations where one controls (additionally) the L^p -norm of the functions f . Indeed if the functions have an L^p -norm that is fixed (i.e there exists $B > 0$ such that $\|f\|_{L^p(\mathbb{R}^d)} = B$)¹ then (5.9) implies that $\|f\|_{L^q(\mathbb{R}^d)} \lesssim B^\theta \|D^s f\|_{L^p(\mathbb{R}^d)}^{1-\theta}$: this estimate is sharper than (5.3) if $\|D^s f\|_{L^p(\mathbb{R}^d)} \gg 1$.*

Proof. We write $f = A + B$ with $A := \sum_{N \leq M} P_N f$ and $B := \sum_{N > M} P_N f$. We have

$$\begin{aligned} \|A\|_{L^q(\mathbb{R}^d)} &\lesssim \sum_{N \leq M} \|P_N f\|_{L^q(\mathbb{R}^d)} \\ &\lesssim \sum_{N \leq M} N^{d(\frac{1}{p} - \frac{1}{q})} \|f\|_{L^p(\mathbb{R}^d)} \\ &\lesssim M^{s\theta} \|f\|_{L^p(\mathbb{R}^d)}. \end{aligned}$$

We have

$$\begin{aligned} \|B\|_{L^q(\mathbb{R}^d)} &\lesssim \sum_{N > M} \|P_N f\|_{L^q(\mathbb{R}^d)} \\ &\lesssim \sum_{N > M} M^{d(\frac{1}{p} - \frac{1}{q}) - s} \|D^s f\|_{L^p(\mathbb{R}^d)} \\ &\lesssim M^{d(\frac{1}{p} - \frac{1}{q}) - s} \|D^s f\|_{L^p(\mathbb{R}^d)} \\ &\lesssim M^{s(\theta-1)} \|D^s f\|_{L^p(\mathbb{R}^d)} \end{aligned}$$

¹This situation happens quite often with $p = 2$ for solutions u of PDEs such as the Schrödinger equations: indeed their mass $\|u(t)\|_{L^2(\mathbb{R}^d)}$ is conserved for all time.

Hence, by choosing M such that $\|f\|_{L^p(\mathbb{R}^d)} \approx M^{-s} \|D^s f\|_{L^p(\mathbb{R}^d)}$ we get (5.9).

□

APPENDIX A

In this subsection we compute the Fourier inverse transform of a particular function.

Proposition 14. *Let $f \in \mathcal{S}(\mathbb{R}^d)$. Let $0 < \alpha < d$. Then*

$$\mathcal{F}^{-1} \left(|\xi|^{-\alpha} \hat{f} \right) = \int_{\mathbb{R}^d} f(x-y) |y|^{\alpha-d} dy \quad (10)$$

with $\mathcal{F}^{-1}(h)(x) := \int_{\mathbb{R}^d} h(\xi) e^{2i\pi\xi \cdot x} d\xi$.

Proof. An elementary change of variable shows that there exists $c_\alpha > 0$ such that

$$c_\alpha |\xi|^{-\alpha} = \int_0^\infty e^{-\pi|\xi|^2 \lambda} \lambda^{\frac{\alpha}{2}-1} d\lambda \quad (11)$$

Hence, using also Fubini theorem, we get

$$\begin{aligned} \mathcal{F}^{-1} \left(|\xi|^{-\alpha} \hat{f}(\xi) \right) (x) &= \int_{\mathbb{R}^d} |\xi|^{-\alpha} \hat{f}(\xi) e^{2i\pi\xi \cdot x} d\xi \\ &= \int_0^\infty \lambda^{\frac{\alpha}{2}-1} \int_{\mathbb{R}^d} e^{-\pi\lambda|\xi|^2} \hat{f}(\xi) e^{2i\pi\xi \cdot x} d\xi d\lambda \\ &= \int_0^\infty \lambda^{\frac{\alpha}{2}-1-\frac{d}{2}} \int_{\mathbb{R}^d} f(y) e^{-\pi\frac{|x-y|^2}{\lambda}} dy \\ &= \int_{\mathbb{R}^d} f(y) \int_0^\infty \lambda^{\frac{\alpha}{2}-1-\frac{d}{2}} e^{-\pi\frac{|x-y|^2}{\lambda}} d\lambda dy \\ &= c_{d-\alpha} I_\alpha(f)(x) \end{aligned}$$

Hence (10) holds. □

APPENDIX B

In this subsection we prove that the quantities $\|Df\|_{L^p(\mathbb{R}^d)}$ and $\|\nabla f\|_{L^p(\mathbb{R}^d)}$ are comparable if $1 < p < \infty$.

Proposition 15. *Let $f \in \mathcal{S}(\mathbb{R}^d)$. Let $1 < p < \infty$. Then*

$$\|Df\|_{L^p(\mathbb{R}^d)} \approx \|\nabla f\|_{L^p(\mathbb{R}^d)}.$$

Proof. Let $1 \leq j \leq d$ and R_j be the j^{th} Riesz transform defined in the Fourier domain by $\widehat{R_j f}(\xi) := \frac{\xi_j}{|\xi|} \hat{f}(\xi)$. Recall that R_j is bounded on $L^p(\mathbb{R}^d)$ (see Appendix C) i.e

$$\|R_j h\|_{L^p(\mathbb{R}^d)} \lesssim \|h\|_{L^p(\mathbb{R}^d)}. \quad (12)$$

Hence

$$\|\nabla f\|_{L^p(\mathbb{R}^d)} \lesssim \sum_{j=1}^d \|\partial_{x_j} f\|_{L^p(\mathbb{R}^d)} \lesssim \sum_{j=1}^d \|R_j Df\|_{L^p(\mathbb{R}^d)} \lesssim \|Df\|_{L^p(\mathbb{R}^d)}.$$

We then show that $\|Df\|_{L^p(\mathbb{R}^d)} \lesssim \|\nabla f\|_{L^p(\mathbb{R}^d)}$. Observe that this estimate is equivalent to show $(*)$: $\|f\|_{L^p(\mathbb{R}^d)} \lesssim \|\nabla D^{-1} f\|_{L^p(\mathbb{R}^d)}$. Observe that $f = -i \sum_{j=1}^d R_j \partial_{x_j} (D^{-1} f)$. Hence, using again (12), we get $(*)$.

□

APPENDIX C

In this subsection we recall the definition of the Riesz transforms R_j . Next we prove their boundedness.

Definition 1. Let $1 \leq j \leq d$. The j^{th} – Riesz transform R_j is the operator acting on functions $\mathcal{S}(\mathbb{R}^d)$ defined by

$$R_j f(x) := \int_{\mathbb{R}^d} \frac{\xi_j}{|\xi|} \hat{f}(\xi) e^{2i\pi\xi \cdot x} d\xi \quad (13)$$

Remark 12. Observe that $R_j f$ is well-defined since $\frac{\xi_j}{|\xi|} \hat{f}(\xi) \in L^1(\mathbb{R}^d)$.

Next we prove that $R_j f$ can be written as the limit of convolution of f with truncated kernels.

Proposition 16. Let $1 \leq j \leq d$. Let $f \in \mathcal{S}(\mathbb{R}^d)$. Then there exists $\tilde{C} \in \mathbb{R}$ such that $R_j f(x) = \tilde{C} \lim_{\epsilon \rightarrow 0} R_{j,\epsilon}^\epsilon f(x)$ with

$$R_{j,\epsilon}^\epsilon f(x) := \tilde{C} \int_{\mathbb{R}^d} k_j^\epsilon(x-y) f(y) dy, \text{ and } k_j^\epsilon(z) := \frac{z_j}{|z|^{d+1}} \mathbf{1}_{|z| \geq \epsilon}.$$

Remark 13. Let $h(y) := \frac{f(y)-f(x)}{y-x}$ if $y \neq x$ and $h(y) := 1$ if $y = x$. Then $h \in L^1(\mathbb{R}^d)$ and $R_{j,\epsilon}^\epsilon f(x) = \int_{|x-y| \geq \epsilon} h(y) dy$. Hence $\lim_{\epsilon \rightarrow 0} R_{j,\epsilon}^\epsilon f(x)$ exists.

Proof. Let $d > \alpha > 0$. Let $\delta > 0$. Let \bar{C} be a positive constant of which the value may change and such that all the statements below are true. Observe from the identity (11) and the Fubini-Tonelli theorem that

$$\begin{aligned} \int_{\mathbb{R}^d} \frac{1}{|x|^{d-\alpha}} e^{-\pi\delta|x|^2} e^{2i\pi\xi \cdot x} dx &= \bar{C} \int_{\mathbb{R}^d} e^{2i\pi\xi \cdot x} \int_0^\infty t^{\frac{d-\alpha}{2}-1} e^{-\pi(t+\delta)|x|^2} dt dx \\ &= \bar{C} \int_0^\infty t^{\frac{d-\alpha}{2}-1} \int_{\mathbb{R}^d} e^{2i\pi\xi \cdot x} e^{-\pi(t+\delta)|x|^2} dx dt \\ &= \bar{C} \int_0^\infty t^{\frac{d-\alpha}{2}-1} (t+\delta)^{-\frac{d}{2}} e^{-\frac{\pi|\xi|^2}{t+\delta}} dt \end{aligned}$$

Let $1 < \alpha < d$. By integrating by parts

$$\int_{\mathbb{R}^d} \partial_{x_j} \left(\frac{1}{|x|^{d-\alpha}} e^{-\pi\delta|x|^2} \right) e^{2i\pi\xi \cdot x} dx = -2\bar{C}i\pi\xi_j \int_0^\infty t^{\frac{d-\alpha}{2}-1} (t+\delta)^{-\frac{d}{2}} e^{-\frac{\pi|\xi|^2}{t+\delta}} dt$$

Hence $I_{\alpha,\delta}(x) + J_{\alpha,\delta}(x) = K_{\alpha,\delta}(x)$ with

$$I_{\alpha,\delta}(\xi) := (\alpha - d) \int_{\mathbb{R}^d} \frac{x_j}{|x|^{d+2-\alpha}} e^{-\pi\delta|x|^2} e^{2i\pi\xi \cdot x} dx,$$

$$J_{\alpha,\delta}(\xi) := -2\pi\delta \int_{\mathbb{R}^d} x_j \frac{1}{|x|^{d-\alpha}} e^{-\pi\delta|x|^2} e^{2i\pi\xi \cdot x} dx, \text{ and}$$

$$K_{\alpha,\delta}(\xi) := -2\bar{C}\pi i \xi_j \int_0^\infty t^{\frac{d-\alpha}{2}-1} (t+\delta)^{-\frac{d}{2}} e^{-\frac{\pi|\xi|^2}{t+\delta}} dt.$$

Elementary considerations show that $\lim_{\delta \rightarrow 0} J_{\alpha,\delta}(\xi) = 0$. Using again (11) we get

$$\lim_{\delta \rightarrow 0} I_{\alpha,\delta}(\xi) = \bar{C}i \xi_j \int_0^\infty t^{-\frac{\alpha}{2}-1} e^{-\frac{\pi|\xi|^2}{t}} dt = \frac{\bar{C}\xi_j}{|\xi|^\alpha}.$$

Hence from the dominated convergence theorem and the equality $\widehat{h_1 * h_2}(\xi) = \widehat{h_1}(\xi)\widehat{h_2}(\xi)$ we get

$$\begin{aligned} R_j f(x) &= \bar{C} \lim_{\substack{\alpha \rightarrow 1 \\ \alpha > 1}} \lim_{\delta \rightarrow 0} (d - \alpha) \int_{\mathbb{R}^d} I_{\alpha,\delta}(x - y) f(y) dy \\ &= \bar{C} \lim_{\epsilon \rightarrow 0} \int_{\mathbb{R}^d} k_j^\epsilon(x - y) f(y) dy. \end{aligned} \quad (14)$$

Indeed, from $\int_{R_2 \geq |x-y| \geq R_1} \frac{x_j - y_j}{|x-y|^\alpha} dy = 0$, we get $\int_{|x-y| \geq \epsilon} \frac{x_j - y_j}{|x-y|^{d+2-\alpha}} f(y) dy - \int_{|x-y| \geq \epsilon} \frac{x_j - y_j}{|x-y|^{d+1}} f(y) dy = A_1 + A_2$ with

$$\begin{aligned} A_1 &:= \int_{1 \geq |x-y| \geq \epsilon} \left(\frac{x_j - y_j}{|x-y|^{d+2-\alpha}} - \frac{x_j - y_j}{|x-y|^{d+1}} \right) (f(y) - f(x)) dy, \text{ and} \\ A_2 &:= \int_{|x-y| \geq 1} \left(\frac{x_j - y_j}{|x-y|^{d+2-\alpha}} - \frac{x_j - y_j}{|x-y|^{d+1}} \right) f(y) dy. \end{aligned}$$

The mean value theorem shows that

$$\begin{aligned} |A_1| &\lesssim (\alpha - 1) \int_{1 \geq |x-y| \geq \epsilon} \frac{|\ln(|x-y|)|}{|x-y|^{d-1}} dy, \text{ and} \\ |A_2| &\lesssim (\alpha - 1) \int_{|x-y| \geq 1} |f(y)| dy. \end{aligned}$$

We also have $\lim_{\epsilon \rightarrow 0} \int_{|x-y| \geq \epsilon} \frac{x_j - y_j}{|x-y|^{d+1-\alpha}} f(y) dy = 0$. This explains the last equality in (14). \square

Next we turn to the boundedness of the Riesz transforms. To this end we first recall the Calderon-Zygmund kernel (see e.g [3]):

Definition 2. K is called a Calderon-Zygmund kernel if $K \in L_{loc}^1(\mathbb{R}^d - \{0\})$ and if there exists $B > 0$ such that K satisfies the following conclusions

- Decay property:

$$|K(x)| \leq B|x|^{-d} \quad (15)$$

- *Regularity property: for all $y \in \mathbb{R}^d$*

$$\int_{|x| \geq 2|y|} |K(x+y) - K(x)| dx \leq B. \quad (16)$$

- *Cancellation property: for all $R_1 \leq R_2$ we have*

$$\int_{R_1 \leq |x| \leq R_2} K(y) dy = 0 \cdot \cdot \quad (17)$$

Remark 14. *Observe that if there exists $C > 0$ such that*

$$x \neq 0 : |\nabla K(x)| \leq \frac{C}{|x|^{d+1}}, \quad (18)$$

then (16) holds for some $B := B(C)$.

Indeed the fundamental theorem of calculus yields

$$\begin{aligned} \int_{|x| \geq 2|y|} |K(x+y) - K(x)| dx &\leq \int_{|x| \geq 2|y|} \int_0^1 |\nabla K(x+ty)| |y| dt dx \\ &\lesssim |y| \int_0^1 \int_{|x| \geq 2|y|} \frac{1}{|x+ty|^{d+1}} dx dt \\ &\lesssim |y| \int_0^1 \int_{|x| \geq 2|y|} \frac{1}{\left(\frac{|x|}{2}\right)^{d+1}} dx dt \\ &\lesssim |y| \int_{2|y|}^\infty \frac{1}{r^2} dr \\ &\lesssim 1 \end{aligned}$$

We then prove that a kernel related to the j^{th} –Riesz transform in a Calderon-Zygmund kernel.

Proposition 17. *Let $1 \leq j \leq d$. Let K_j defined by $K_j(x) := \frac{x_j}{|x|}$. Then K_j is a Calderon-Zygmund kernel.*

Proof. The proof is short. It is clear that the decay property (15) and the cancellation property (17) hold. The regularity property (16) follows from (18). □

Next we prove the boundedness of the limit of operators associated to truncated Calderon-Zygmund kernels in $L^p(\mathbb{R}^d)$, $1 < p < \infty$.

Proposition 18. *Suppose that K is Calderon-Zygmund kernel. Let $1 < p < \infty$. Given $\epsilon > 0$ let*

$$T_\epsilon(f)(x) := \int_{|y| \geq \epsilon} f(x-y)K(y) dy.$$

Then there exists $A_p > 0$ such that

$$\|T_\epsilon f\|_{L^p(\mathbb{R}^d)} \leq A_p \|f\|_{L^p(\mathbb{R}^d)}. \quad (19)$$

Moreover $\lim_{\epsilon \rightarrow 0} T_\epsilon(f)$ exists in the sense of the L^p norm, and if $Tf := \lim_{\epsilon \rightarrow 0} T_\epsilon(f)$ then

$$\|Tf\|_{L^p(\mathbb{R}^d)} \leq A_p \|f\|_{L^p(\mathbb{R}^d)}. \quad (20)$$

Proof. We prove that $\lim_{\epsilon \rightarrow 0} T_\epsilon(f)$ exists in L^p . We have

$$\begin{aligned} T_\epsilon(f)(x) - T_\eta(f)(x) &= \int_{|y| \geq \epsilon} K(y) f(x-y) dy - \int_{|y| \geq \eta} K(y) f(x-y) dy \\ &= \operatorname{sgn}(\eta - \epsilon) \int_{\min(\epsilon, \eta) \leq |y| \leq \max(\epsilon, \eta)} K(y) (f(x-y) - f(x)) dy : \end{aligned}$$

this comes from the cancellation property (17). Hence the fundamental theorem of calculus and the Mikowski inequality show that

$$\begin{aligned} \|T_\epsilon(f) - T_\eta(f)\|_{L^p(\mathbb{R}^d)} &\lesssim \left\| \int_{\min(\epsilon, \eta) \leq |y| \leq \max(\epsilon, \eta)} |K(y)| |\nabla f(x-y)| |y| dy \right\|_{L^p(\mathbb{R}^d)} \\ &\lesssim \int_{\min(\epsilon, \eta) \leq |y| \leq \max(\epsilon, \eta)} |K(y)| \|\nabla f(x-y)\|_{L^p(\mathbb{R}^d)} |y| dy \\ &\lesssim \int_{\min(\epsilon, \eta) \leq |y| \leq \max(\epsilon, \eta)} |y|^{-d+1} dy \\ &\lesssim \int_{\min(\xi, \eta)}^{\max(\xi, \eta)} dr \\ &\lesssim |\eta - \epsilon| \end{aligned}$$

Hence, if $x_n \rightarrow 0$ as $n \rightarrow \infty$ then $(T_{x_n})_{n \geq 1}$ is a Cauchy sequence so it converges. So $T(f) := \lim_{\epsilon \rightarrow 0} T_\epsilon(f)$ exists.

The proof of (19) relies upon Calderon-Zygmund theory and can be found in e.g [3, 9].

By letting $\epsilon \rightarrow 0$ in (19) we get (20). □

Proposition 19. *Let $1 \leq j \leq d$ and $1 < p < \infty$. Then R_j defined by (13) is bounded on $L^p(\mathbb{R}^d)$, i.e there exists $A_p > 0$ such that for all $f \in \mathcal{S}(\mathbb{R}^d)$*

$$\|R_j f\|_{L^p(\mathbb{R}^d)} \leq A_p \|f\|_{L^p(\mathbb{R}^d)}.$$

Proof. It follows from Proposition 16, Proposition 17, and Proposition 18. □

BIBLIOGRAPHY

- [1] Brezis, *Functional Analysis, Sobolev Spaces and Partial Differential Equations*, Universitext, Springer, 2011
- [2] Folland, *Real Analysis: Modern Techniques and Their Applications 2nd Edition*, Wiley, 2007
- [3] Hao, [Lecture Notes On Harmonic Analysis, 2020](#)
- [4] Lerman, [The Shannon Sampling Theorem and Its Implications](#)
- [5] Pinsky, *Introduction to Fourier analysis and wavelets*. Graduate Studies in Mathematics, American Mathematical Society, 2009
- [6] Stein, Shakarshi, *Fourier Analysis: An Introduction (Princeton Lectures in Analysis)*, Princeton University Press, 2003
- [7] Terence Tao, [Hardy's uncertainty principle, 2009](#)
- [8] Terence Tao, [Terence Tao's lecture notes for 254A, Winter 2001](#)
- [9] Stein, *Singular Integrals and Differentiability Properties of Functions*, Princeton University Press, 1970