

FOURIER ANALYSIS–PART 2

XU WANG

This is the second part of the Fourier analysis notes for TMA4170.

CONTENTS

1. More Fourier analysis	1
1.1. Eigentheory of the Fourier transform	1
1.2. Central limit theorem	4
2. Wavelet analysis	7
2.1. Haar wavelets	7
2.2. Multiresolution analysis	8
2.3. Wavelet transform	11
References	13

1. MORE FOURIER ANALYSIS

1.1. **Eigentheory of the Fourier transform.** The starting point is the following:

$$\widehat{f'} = i \cdot 2\pi x \widehat{f}, \quad \widehat{2\pi x f} = i \cdot (\widehat{f})'.$$

Consider the following operators A, B on the Schwartz space defined by

$$Af := f', \quad Bf := 2\pi x f.$$

Then

$$\widehat{Af} = iB\widehat{f}, \quad \widehat{Bf} = iA\widehat{f}.$$

Think of the Fourier transform as an operator

$$\mathcal{F}(f) := \widehat{f}.$$

Then we have

$$\mathcal{F}((A + B)f) = i \cdot (A + B)\mathcal{F}(f),$$

and

$$\mathcal{F}((A - B)f) = -i \cdot (A - B)\mathcal{F}(f).$$

Now suppose that f is a fixed point of the Fourier transform, i.e. $\mathcal{F}(f) = f$, then

$$\mathcal{F}((A - B)f) = -i \cdot (A - B)\mathcal{F}(f) = -i \cdot (A - B)f.$$

Let us inductively define

$$(A - B)^k f := (A - B)((A - B)^{k-1} f), \quad (A - B)^1 f := (A - B)f.$$

Then we have

$$\mathcal{F}((A - B)^k f) = (-i)^k (A - B)^k f, \quad \mathcal{F}((A + B)^k f) = i^k (A + B)^k f.$$

Now suppose that $f = e^{-\pi x^2}$, a direct computation gives

$$(A + B)f = 0, \quad (A - B)f = -4\pi x f = -2Bf.$$

Definition 1. We call

$$h_k := (A - B)^k e^{-\pi x^2}, \quad k = 0, 1, \dots$$

the k -th Hermite function associated to the Fourier transform \mathcal{F} .

Remark 1. We have

$$h_0(x) = e^{-\pi x^2}, \quad h_1(x) = -4\pi x e^{-\pi x^2}, \quad h_2(x) = ((-4\pi x)^2 - 4\pi) e^{-\pi x^2},$$

In general, we can write

$$h_k(x) = P_k(x) \cdot e^{-\pi x^2},$$

where

$$P_k(x) = (-4\pi x)^k + \text{lower order terms}.$$

We call the above degree k polynomial the k -th Hermite polynomial. From the definition, we have

$$(h_k, h_l) = \int_{\mathbb{R}} P_k(x) P_l(x) e^{-2\pi x^2} dx.$$

The right hand side is called the Segal–Bargmann–Fock inner product of P_k and P_l .

The main theorem in this section is the following:

Theorem 1. $\{h_k\}_{k \geq 0}$ defines an orthogonal basis of the Schwartz space \mathcal{S} , moreover, we have

$$\mathcal{F}(h_k) = (-i)^k h_k, \quad k = 0, 1, 2, \dots$$

Since $A - B$ is an eigen-operator of \mathcal{F} , we have $\mathcal{F}(h_k) = (-i)^k h_k$. Now let us prove that $\{h_k\}_{k \geq 0}$ is an orthogonal system. It is enough to find an operator \mathfrak{f} such that

$$\mathfrak{f}(h_k) = \lambda_k h_k$$

and $\lambda_k \neq \lambda_j$ if $j \neq k$. The main idea is to find a fixed operator of the Fourier transform. Since

$$\mathcal{F}((A - B)(A + B)f) = i(-i)(A - B)(A + B)\mathcal{F}(f) = (A - B)(A + B)\mathcal{F}(f),$$

we know that $(A - B)(A + B)$ is a fixed operator of \mathcal{F} . A direct computation gives

$$(A - B)(A + B) = 4\pi \mathfrak{f} + 2\pi = (A + B)(A - B) + 4\pi,$$

where

$$\mathfrak{f}(g) := \frac{g''}{4\pi} - \pi x^2 g.$$

We know that \mathfrak{f} is also a fixed operator of \mathcal{F} . Now $(A + B)h_0 = 0$ gives

$$0 = (A - B)(A + B)h_0 = 4\pi\mathfrak{f}(h_0) + 2\pi h_0.$$

Thus

$$\mathfrak{f}(h_0) = -\frac{1}{2}h_0.$$

Assume that

$$\mathfrak{f}(h_k) = -\left(k + \frac{1}{2}\right)h_k.$$

Then $(A - B)(A + B)h_{k+1} = (A - B)(A + B)(A - B)h_k$ can be written as

$$(A - B)(4\pi\mathfrak{f} - 2\pi)h_k = 4\pi(-k - 1)h_{k+1}.$$

Thus we have

$$(4\pi\mathfrak{f} + 2\pi)h_{k+1} = 4\pi(-k - 1)h_{k+1},$$

which gives

$$\mathfrak{f}(h_{k+1}) = -\left(k + 1 + \frac{1}{2}\right)h_{k+1}.$$

Thus we have

Lemma 1. *Each h_k is an eigenfunction of \mathfrak{f} with eigenvalue $-k - \frac{1}{2}$, in particular, it implies that $\{h_k\}$ is an orthogonal system.*

Now it suffices to show that $\{h_k\}$ is a basis of the Schwartz space \mathcal{S} in order to prove our main theorem. The idea is consider the *generating function* of $\{h_k\}$ defined by

$$H(t, x) := \sum_{k=0}^{\infty} h_k(x) \frac{t^k}{k!}.$$

Lemma 2. $H(t, x) = e^{-\pi(x+2t)^2} \cdot e^{2\pi t^2}$.

Proof. Notice that

$$(A - B)f = e^{\pi x^2} \frac{d}{dx} (e^{-\pi x^2} f),$$

gives

$$(A - B)^k f = e^{\pi x^2} \left(\frac{d}{dx}\right)^k (e^{-\pi x^2} f).$$

Thus we have

$$H(t, x) = e^{\pi x^2} \sum_{k=0}^{\infty} \left(\frac{d}{dx}\right)^k e^{-2\pi x^2} \frac{t^k}{k!}.$$

Notice that

$$\left(\frac{d}{dx}\right)^k e^{-2\pi x^2} = \left(\left(\frac{d}{dt}\right)^k e^{-2\pi(x+t)^2}\right) \Big|_{t=0}.$$

Thus the Taylor expansion of $e^{-2\pi(x+t)^2}$ gives

$$H(t, x) = e^{\pi x^2 - 2\pi(x+t)^2} = e^{-\pi(x+2t)^2} \cdot e^{2\pi t^2}.$$

□

Now we can prove the following lemma.

Lemma 3. $\{h_k\}$ is a basis of the Schwartz space \mathcal{S} ,

Proof. Assume that $f \in \mathcal{S}$ and $(f, h_k) = 0$ for $k = 0, 1, \dots$. Then we have $(f, H(t, \cdot)) = 0$ for every $t \in \mathbb{R}$, which implies that $f \star e^{-\pi x^2} \equiv 0$. Taking the Fourier transform, we get $\hat{f}(y)e^{-\pi y^2} \equiv 0$. Thus $\hat{f} = 0$ and we know that $f = \check{f} = 0$. □

Remark 2 ().** Up to a constant, \mathfrak{f} is equal to the Hermite operator (see Week 8 exercise). Since

$$\mathcal{F}(h_k) = (-i)^k h_k, \quad \mathfrak{f}(h_k) = -\left(k + \frac{1}{2}\right) h_k,$$

and

$$e^{\frac{i\pi}{2}(-k-\frac{1}{2})} = (-i)^k e^{-\frac{i\pi}{4}}.$$

We get the following crucial identity

$$e^{\frac{i\pi}{2}\mathfrak{f}} = \mathcal{F} \cdot e^{-\frac{i\pi}{4}}.$$

Recall that we have

$$\mathfrak{f} = L + \Lambda, \quad L := -\pi x^2, \quad \Lambda := \left(\frac{d}{dx}\right)^2 / 4\pi.$$

Put $N := [L, \Lambda] = x \frac{d}{dx} + \frac{1}{2}$, then we have the following sl_2 -identities

$$[N, L] = 2L, \quad [N, \Lambda] = -2\Lambda, \quad [L, \Lambda] = N.$$

Usually we call

$$e^{\frac{i\pi}{2}(L+\Lambda)} = e^{\frac{i\pi}{2}\mathfrak{f}}$$

the sl_2 -star operator (up to a constant, the famous Hodge star operator is a special sl_2 -star operator). Thus the Fourier transform is essentially the star operator in sl_2 -representation.

1.2. Central limit theorem. From page 114–116 in [3]. The mathematical content of the "central limit theorem" of probability theory is the following

Theorem 2. Let f be a non-negative function in \mathcal{S} with $\int f = 1$, $\int x f = 0$ and $\int x^2 f = 1$, denote by f_n the n -fold convolution defined inductively by

$$f_1 := f, \quad f_2 := f * f, \quad f_{n+1} := f * f_n.$$

Then $\sqrt{n}f_n(\sqrt{n}x)$ converges weakly to $(2\pi)^{-\frac{1}{2}}e^{-\frac{x^2}{2}}$.

Proof. Since $\hat{\mathcal{S}} = \mathcal{S}$, it is enough to check that

$$\lim_{n \rightarrow \infty} \int_{\mathbb{R}} \sqrt{n} f_n(\sqrt{n}x) \hat{g}(x) dx = \int_{\mathbb{R}} (2\pi)^{-\frac{1}{2}} e^{-\frac{x^2}{2}} \hat{g}(x) dx, \quad \forall g \in \mathcal{S}.$$

Notice that

$$\sqrt{n} \widehat{f_n(\sqrt{n}x)} = \left(\hat{f}(\gamma/\sqrt{n}) \right)^n$$

gives

$$\int_{\mathbb{R}} \sqrt{n} f_n(\sqrt{n}x) \hat{g}(x) dx = \int_{\mathbb{R}} \left(\hat{f}(\gamma/\sqrt{n}) \right)^n g(\gamma) d\gamma.$$

By definition, we have

$$\hat{f}(\gamma/\sqrt{n}) = \int_{\mathbb{R}} e^{-2\pi i \gamma x / \sqrt{n}} f(x) dx.$$

Since $f \in \mathcal{S}$, we know $c_l := \sup_{x \in \mathbb{R}} |f(x)x^{2l+3}| < \infty$ and

$$\left| \int_{|x| \geq n^{1/l}} e^{-2\pi i \gamma x / \sqrt{n}} f(x) dx \right| \leq c_l \int_{|x| \geq n^{1/l}} x^{-2l-1} dx = c_l / (\ln^2).$$

When $|x| \leq n^{1/l}$ and $l > 6$, then $d := 3(1/2 - 1/l) > 1$ and Taylor expansion of e^x gives

$$\left| e^{-2\pi i \gamma x / \sqrt{n}} - 1 + 2\pi i \gamma x / \sqrt{n} - 2\pi^2 \gamma^2 x^2 / n \right| \leq \sum_{k \geq 3} (2\pi \gamma)^k / (n^{(k/2 - k/l)k!}) \leq n^{-d} e^{2\pi|\gamma|}.$$

Thus

$$\left| \hat{f}(\gamma/\sqrt{n}) - \int_{|x| \leq n^{1/l}} (1 - 2\pi i \gamma x / \sqrt{n} - 2\pi^2 \gamma^2 x^2 / n) f(x) dx \right| \leq c_l / (\ln^2) + n^{-d} e^{2\pi|\gamma|}.$$

Moreover, we have

$$\left| \int_{|x| \geq n^{1/l}} (1 - 2\pi i \gamma x / \sqrt{n} - 2\pi^2 \gamma^2 x^2 / n) f(x) dx \right| \leq c_l (1 + 2\pi \gamma / \sqrt{n} + 2\pi^2 \gamma^2 / n) / (\ln^2).$$

Now we have

$$\left| \hat{f}(\gamma/\sqrt{n}) - \int_{\mathbb{R}} (1 - 2\pi i \gamma x / \sqrt{n} - 2\pi^2 \gamma^2 x^2 / n) f(x) dx \right| \leq c(\gamma) n^{-d}, \quad d > 1.$$

But our assumption gives

$$\int_{\mathbb{R}} (1 - 2\pi i \gamma x / \sqrt{n} - 2\pi^2 \gamma^2 x^2 / n) f(x) dx = 1 - 2\pi^2 \gamma^2 / n.$$

Thus we get

$$\left| \hat{f}(\gamma/\sqrt{n}) - (1 - 2\pi^2 \gamma^2 / n) \right| \leq c(\gamma) n^{-d},$$

which gives

$$\lim_{n \rightarrow \infty} \left(\hat{f}(\gamma/\sqrt{n}) \right)^n = \lim_{n \rightarrow \infty} (1 - 2\pi^2 \gamma^2 / n)^n = e^{-2\pi^2 \gamma^2}.$$

Notice that $|\hat{f}(\gamma/\sqrt{n})| \leq \int f = 1$, the above formula gives (by the dominated convergence theorem, see wikipedia page)

$$\lim_{n \rightarrow \infty} \int_{\mathbb{R}} \left(\hat{f}(\gamma/\sqrt{n}) \right)^n g(\gamma) d\gamma = \int_{\mathbb{R}} e^{-2\pi^2 \gamma^2} g(\gamma) d\gamma.$$

Now it is enough to show

$$\int_{\mathbb{R}} (2\pi)^{-\frac{1}{2}} e^{-\frac{x^2}{2}} \hat{g}(x) dx = \int_{\mathbb{R}} e^{-2\pi^2 \gamma^2} g(\gamma) d\gamma,$$

which follows from

$$\widehat{(2\pi)^{-\frac{1}{2}} e^{-\frac{x^2}{2}}} = e^{-2\pi^2 \gamma^2}.$$

The proof is complete. \square

Remark 3. By approximating the indicator function of $[a, b]$ above and below by functions in \mathcal{D} , the above theorem implies that

$$\lim_{n \rightarrow \infty} \int_a^b \sqrt{n} f_n(\sqrt{n}x) dx = \int_{a\sqrt{n}}^{b\sqrt{n}} f_n(x) dx = \int_a^b \frac{1}{(2\pi)^{\frac{1}{2}}} e^{-\frac{x^2}{2}} dx$$

for any $a < b$.

The **probabilistic content** of the central limit theorem is as follows: Think of an infinite number of statistically independent copies $u_1, u_2, \dots, u_n, \dots$ of a statistical quantity u distributed according to the rule

$$P(a \leq u < b) = \int_a^b f(x) dx,$$

in which the letter P stands for the probability of the indicated event. The adjective "independent" means that probabilities multiply:

$$P(a_j \leq u_j < b_j, j = 1, 2, \dots) = \int_{a_1}^{b_1} f(x) dx \cdot \int_{a_2}^{b_2} f(x) dx \cdots$$

and you infer that the sum $s_n = u_1 + \dots + u_n$ is a distribution according to the rule

$$P(a \leq s_n < b) = \int_{a \leq x_1 + \dots + x_n < b} f(x_1) \cdots f(x_n) dx_1 \cdots dx_n = \int_a^b f_n(x) dx.$$

The content of the central limit theorem is now seen to be that the scale sum s_n/\sqrt{n} is nearly Gaussian distributed for large n :

$$P(a \leq s_n/\sqrt{n} < b) = \sqrt{n} \int_a^b f_n(\sqrt{n}x) dx$$

is approximately

$$\int_a^b (2\pi)^{-\frac{1}{2}} e^{-\frac{x^2}{2}} dx.$$

The fact goes back to Moivre and Laplace in the 18th century; for additional information, see [4] (Vol. 1, pp. 174–195, Vol. 2, pp. 252–259).

2. WAVELET ANALYSIS

This part is a list of results from Chapter 4-7 in [1].

2.1. Haar wavelets.

Definition 2 (Haar scaling function). $\phi(x) = 1$ if $0 \leq x < 1$; $\phi(x) = 0$ otherwise.

Definition 3. The space of step functions at level j , $j = 0, 1, \dots$, is defined to be the \mathbb{R} -linear space spanned by $\{\phi(2^j x + k)\}_{k \in \mathbb{Z}}$.

Fact 1: $V_0 \subset V_1 \subset \dots \subset V_j \subset V_{j+1} \subset \dots$;

Fact 2: $f(x) \in V_0$ iff $f(2^j x) \in V_j$, $f(x) \in V_j$ iff $f(2^{-j} x) \in V_0$;

Fact 3: $\{2^{j/2} \phi(2^j x + k)\}_{k \in \mathbb{Z}}$ is an orthonormal basis of V_j .

The so called Haar wavelet is a function that can be used to express the *orthogonal complement* of V_j in V_{j+1} .

Definition 4. The Haar wavelet is the function $\psi(x) := \phi(2x) - \phi(2x - 1)$.

Fact 4: $V_{j+1} = V_j \oplus W_j$, where W_j is spanned by $\{\psi(2^j x + k)\}_{k \in \mathbb{Z}}$;

Fact 5: $L^2(\mathbb{R}) = \overline{V_0 \oplus W_0 \oplus W_1 \oplus \dots}$.

Remark: The so called Haar decomposition and reconstruction algorithms are essentially based on the following orthogonal decomposition

$$V_j = V_0 \oplus W_1 \oplus \dots \oplus W_{j-1}.$$

Step 1: Sample Approximated the original signal by a step function of the following form

$$f_j(x) = \sum a_k \phi(2^j x - k) \in V_j;$$

Step 2: Decompose Use $\phi(2x) = (\phi(x) + \psi(x))/2$, $\phi(2x - 1) = (\phi(x) - \psi(x))/2$ with x replaced by $2^{j-1}x - k$ to write

$$f_j = f_{j-1} + w_{j-1}, \quad f_{j-1} \in V_{j-1}, \quad w_{j-1} \in W_{j-1};$$

Repeating the process gives

$$f_j = f_0 + w_0 + \dots + w_{j-1}, \quad f_0 \in V_0, \quad w_k \in W_k;$$

Step 3: Surgery Remove unwanted w_k (filter out noise or data compression);

Step 4: Recover: Use $\phi(x) = \phi(2x) + \phi(2x - 1)$, $\psi(x) = \phi(2x) - \phi(2x - 1)$ with x replaced by $2^{j-1}x - k$ to inductively write $f_0 + w_0 + \dots + w_{j-1}$ as $\sum a_k \phi(2^j x - k)$.

2.2. Multiresolution analysis. The Haar wavelet theory is based on the Haar scaling function ϕ . One drawback is that ϕ is not continuous. The wavelet theory based on a general scaling function (can be continuous) is called a multiresolution analysis.

Definition 5 (Multiresolution analysis). A sequence of linear spaces $\{V_j\}_{j \in \mathbb{Z}}$ in $L^2(\mathbb{R})$ is called a multiresolution analysis with scaling function ϕ if

- 1) $\{\phi(x - k)\}_{k \in \mathbb{Z}}$ is an orthonormal basis of V_0 ;
- 2) $V_j = \{f : f(2^{-j}x) \in V_0\}$, $j \in \mathbb{Z}$;
- 3) $V_j \subset V_{j+1}$;
- 4) $\bigcap_{j \in \mathbb{Z}} V_j = \{0\}$;
- 4) $\overline{\bigcup_{j \in \mathbb{Z}} V_j} = L^2(\mathbb{R})$.

It is clear that the Haar scaling function defines a multiresolution analysis. Another interesting scaling function is the **Shannon scaling function** $\phi(x) = \text{sinc}(x)$, where

$$\text{sinc}(x) := \frac{\sin \pi x}{\pi x}$$

on \mathbb{R} (with the understanding that $\text{sinc}(0) = 1$). Notice that

$$\text{sinc} = 1_{\widehat{[-1/2, 1/2]}}$$

Thus the Shannon scaling function is equal to the Fourier transform of the Haar scaling function up to a translation by $1/2$.

By a linear change of variable, assumption 1) in the definition of multiresolution analysis gives

Theorem 3. For every $j \in \mathbb{Z}$, $\{\phi_k^j := 2^{j/2} \phi(2^j x - k)\}_{k \in \mathbb{Z}}$ defines an orthonormal basis for V_j .

The orthogonal decomposition gives the following

$$\phi_l^{j-1} = \sum_{k \in \mathbb{Z}} (\phi_l^{j-1}, \phi_k^j) \phi_k^j.$$

Notice that (replace $2^{j-1}x$ by x)

$$(\phi_l^{j-1}, \phi_k^j) = 2^{-1/2} 2^j \int_{\mathbb{R}} \phi(2^{j-1}x - l) \overline{\phi(2^j x - k)} dx = 2^{1/2} \int_{\mathbb{R}} \phi(x - l) \overline{\phi(2x - k)} dx.$$

Replace $x - l$ by x , we get

$$(\phi_l^{j-1}, \phi_k^j) = 2^{1/2} \int_{\mathbb{R}} \phi(x) \overline{\phi(2x + 2l - k)} dx = (\phi, \phi_{k-2l}^1) := p_{k-2l},$$

which gives the following theorem:

Theorem 4 (Scaling relation theorem).

$$\phi_l^{j-1} = \sum_{k \in \mathbb{Z}} p_{k-2l} \phi_k^j.$$

Example: For the Haar scaling function ϕ , we have

$$p_0 = p_1 = 2^{-1/2}, \quad p_k = 0, \quad \forall k \neq 0, 1.$$

The following results contains identities for the p_k which will be important later.

Theorem 5. *We have the following formula*

$$\sum_{k \in \mathbb{Z}} |p_k|^2 = 1$$

and

$$\sum_{k \in \mathbb{Z}} p_{k-2l} \overline{p_k} = 0, \quad \forall l \neq 0.$$

Assume further that ϕ has compact support and $\int_{\mathbb{R}} \phi(x) dx \neq 0$. Then

$$\sum_{k \in \mathbb{Z}} p_{2k} = \sum_{k \in \mathbb{Z}} p_{2k+1} = \frac{1}{2} \sum_{k \in \mathbb{Z}} p_k = 2^{-1/2}.$$

Proof. The first two formulas follows from the orthogonal decomposition. For the remaining ones, notice that if ϕ has compact support then only a finite number of p_k is not zero. Thus we have integrate $\phi = \sum p_k \phi_k^1$ termwise, which gives

$$\int_{\mathbb{R}} \phi(x) dx = 2^{1/2} \sum_{k \in \mathbb{Z}} p_k \int_{\mathbb{R}} \phi(2x - k) dx = 2^{-1/2} \sum_{k \in \mathbb{Z}} p_k \int_{\mathbb{R}} \phi(x) dx,$$

which gives (here we use $\int_{\mathbb{R}} \phi(x) dx \neq 0$) $\sum_{k \in \mathbb{Z}} p_k = 2^{1/2}$. Also notice that

$$\sum_{l \in \mathbb{Z}} \sum_{k \in \mathbb{Z}} p_{k-2l} \overline{p_k} = 1.$$

Divide the sum over k into even and odd terms gives

$$1 = |a|^2 + |b|^2, \quad a := \sum_{k \in \mathbb{Z}} p_{2k}, \quad b := \sum_{k \in \mathbb{Z}} p_{2k+1}.$$

Since $a + b = \sqrt{2}$, we have

$$\sqrt{2} = a + b = \sqrt{2(|a|^2 + |b|^2)},$$

which gives $a = b = \sqrt{2}/2$ by the Schwarz inequality. \square

The so called *wavelet function* in multiresolution analysis is the following single function

$$\psi := \sum (-1)^k \overline{p_{1-k}} \phi_k^1.$$

that can be used to decode the orthogonal complement of V_j in V_{j+1} . Notice that $\phi = \sum p_k \phi_k^1$ gives

$$(\phi, \psi) = \sum (-1)^k p_k p_{1-k} = \cdots - p_{-1} p_2 + p_0 p_1 - p_1 p_0 + p_2 p_{-1} + \cdots = 0.$$

Put

$$\psi_k^j := 2^{j/2} \psi(2^j x - k)$$

and defines

$$W_j := \text{Span}\{\psi_k^j\}_{k \in \mathbb{Z}},$$

then similar to the Haar wavelet case, we have

Theorem 6. Each W_j is the orthogonal complement of V_j in V_{j+1} and

$$L^2(\mathbb{R}) = \bigoplus_{j \in \mathbb{Z}} W_j$$

has an orthonormal basis $\{\psi_k^j\}_{j,k \in \mathbb{Z}}$.

The above theorem gives the decomposition and reconstruction algorithms in multiresolution analysis (see section 5.1 and 5.2 in [1] for details). The following theorem (see Appendix A 2.1 in [1] for the proof) provides an equivalent formulation of multiresolution analysis in terms of ϕ .

Theorem 7. Let ϕ be a continuous function with compact support such that $\{\phi(x - k)\}_{k \in \mathbb{Z}}$ is an orthogonal system. Put

$$V_j := \text{Span}\{\phi(2^j x - k)\}_{k \in \mathbb{Z}}.$$

Then $\bigcap_{j \in \mathbb{Z}} V_j = \{0\}$. Assume further that

$$\int_{\mathbb{R}} \phi(x) dx = 1$$

and ϕ is a finite combination of $\phi(2x - k)$, then $L^2(\mathbb{R}) = \overline{\bigcup_{j \in \mathbb{Z}} V_j}$.

How is a scaling function ϕ constructed? The idea is to first assume a sufficiently regular ϕ exists and see what properties must be satisfied, then we reverse the process. The main idea is to use *Fourier transform*. Recall that

$$\hat{\phi}(\xi) := \int_{\mathbb{R}} \phi(x) e^{-2\pi i x \xi} dx.$$

Thus $\int \phi = 1$ iff $\hat{\phi}(0) = 1$, moreover notice that

$$\widehat{\phi(x - k)}(\xi) = \hat{\phi}(\xi) e^{-2\pi i k \xi}.$$

Thus by the Plancherel identity, $\{\phi(x - k)\}_{k \in \mathbb{Z}}$ is an orthogonal system iff $\{\hat{\phi}(\xi) e^{-2\pi i k \xi}\}_{k \in \mathbb{Z}}$ is an orthogonal system, i.e.

$$\|\hat{\phi}\| = 1 \text{ and } (\hat{\phi}, \hat{\phi} e^{-2\pi i k \xi}) = 0, \quad \forall k \neq 0.$$

The above identities are equivalent to that

$$|\widehat{|\hat{\phi}|^2}(0)| = 1 \text{ and } |\widehat{|\hat{\phi}|^2}(k)| = 1, \quad \forall k \neq 0.$$

Thus the Poisson summation formula gives

$$\sum_{k \in \mathbb{Z}} |\widehat{|\hat{\phi}|^2}(\xi + k)|^2 = \sum_{k \in \mathbb{Z}} |\widehat{|\hat{\phi}|^2}(k)|^2 e^{2\pi i k \xi} \equiv 1.$$

Now let us look at the final assumption in the above theorem: ϕ is a finite combination of $\phi(2x - k)$. As before we write $\phi_k^j := 2^{j/2} \phi(2^j x - k)$ and assume that $\phi = \sum p_k \phi_k^1$. Taking the Fourier transform, we get

$$\hat{\phi}(\xi) = \sum p_k 2^{-1/2} \hat{\phi}(\xi/2) e^{-\pi i k \xi}.$$

Thus

$$\hat{\phi}(2\xi) = \hat{\phi}(\xi) P(z), \quad P(z) := 2^{-1/2} \sum p_k z^k, \quad z = e^{-2\pi i \xi}.$$

Now

$$1 = \sum_{k \in \mathbb{Z}} |\hat{\phi}(2\xi + k)|^2 = \sum_{k \in \mathbb{Z}} |\hat{\phi}(\xi + k/2)P((-1)^k z)|^2 = |P(z)|^2 + |P(-z)|^2.$$

Also notice that $\hat{\phi}(0) = 1$ implies that $P(1) = 1$. The following theorem (Theorem 5.23 in [1]) says that one may get ϕ from P with one extra condition: $P(z) \neq 0$ if $\operatorname{Re} z \geq 0$, $|z| = 1$.

Theorem 8. *Let $P(z) := 2^{-1/2} \sum p_k z^k$ be a polynomial satisfies*

$$P(1) = 1, \quad |P(z)|^2 + |P(-z)|^2 = 1, \quad \forall |z| = 1,$$

and $P(z) \neq 0$ if $\operatorname{Re} z \geq 0$, $|z| = 1$. Let ϕ_0 be the Haar scaling function, define ϕ_n inductively by

$$\phi_n(x) = \sum p_k \cdot 2^{1/2} \phi_{n-1}(2x - k), \quad n \geq 1.$$

Then $\phi := \lim_{n \rightarrow \infty} \phi_n$ satisfies the assumption in Theorem 7.

Remark: Put

$$P(z) := \frac{1}{8} \left((1 + \sqrt{3}) + (3 + \sqrt{3})z + (3 - \sqrt{3})z^2 + (1 - \sqrt{3})z^3 \right).$$

One may check that $P(z)$ satisfies the assumptions above. The associated wavelet is a special [Daubechies wavelet](#). For another approach of constructing ϕ and ψ , see Theorem 42.5.3 and Theorem 42.5.8 in [5].

2.3. Wavelet transform. Let $\psi \in L^2(\mathbb{R}) \cap L^1(\mathbb{R})$. Assume that

$$\int_{\mathbb{R}} \psi(x) dx = 0.$$

Then we know that $\hat{\psi}$ is an L^2 continuous function such that $\hat{\psi}(0) = 0$. We need an extra condition to define the wavelet transform

$$C_\psi := \int_{\mathbb{R}} \frac{|\hat{\psi}(\xi)|^2}{|\xi|} d\xi < \infty.$$

Notice that the above inequality implies $\hat{\psi}(0) = 0$, which is equivalent to $\int \psi = 0$. In all practical cases $\int \psi = 0$ also implies $C_\psi < \infty$ (in fact assume further that $x\psi \in L^1(\mathbb{R})$ then $\hat{\psi} \in C^1(\mathbb{R})$ and $C_\psi < \infty$). The [wavelet transform](#) is defined by

$$(Tf)(a, b) := (f, \psi^{a,b}),$$

where

$$\psi^{a,b}(x) := |a|^{-1/2} \psi \left(\frac{x-b}{a} \right), \quad a, b \in \mathbb{R}, \quad a \neq 0.$$

Since $\|\psi^{a,b}\| = \|\psi\|$, the Cauchy–Schwarz inequality gives $|(Tf)(a, b)| \leq \|f\| \cdot \|\psi\| < \infty$, thus Tf is bounded. A function f can also be recovered from Tf (see page 24 in [2]):

Theorem 9. Assume that $\psi \in L^2(\mathbb{R}) \cap L^1(\mathbb{R})$ and $C_\psi = 1$. Then for all $f, g \in L^2(\mathbb{R})$,

$$(Tf, Tg) = (f, g),$$

where $(f, g) := \int f \bar{g}$ and $(Tf, Tg) = \int Tf \cdot \overline{Tg} \frac{dadb}{a^2}$.

Proof. The main idea of the proof is to use the Plancherel identity for Fourier transform:

$$Tf = (f, \psi^{a,b}) = (\hat{f}, \widehat{\psi^{a,b}}).$$

Notice that

$$\widehat{\psi^{a,b}}(\xi) = \int_{\mathbb{R}} |a|^{-1/2} \psi \left(\frac{x-b}{a} \right) e^{-2\pi i x \cdot \xi} dx,$$

change of variable $x - b = ay$ gives

$$\widehat{\psi^{a,b}}(\xi) = |a|^{1/2} \hat{\psi}(a\xi) e^{-2\pi i b \xi}.$$

Thus

$$Tf = (\hat{f}, \widehat{\psi^{a,b}}) = \int_{\mathbb{R}} \hat{f}(\xi) |a|^{1/2} \overline{\hat{\psi}(a\xi)} e^{2\pi i b \xi} d\xi = \mathcal{F}^{-1} \left(\hat{f}(\xi) |a|^{1/2} \overline{\hat{\psi}(a\xi)} \right),$$

where \mathcal{F}^{-1} denotes the inverse Fourier transform. Apply the Plancherel identity again, we get

$$\int_{\mathbb{R}} Tf \overline{Tg} db = \int_{\mathbb{R}} \hat{f}(\xi) \overline{\hat{g}(\xi)} \cdot |a| \cdot |\hat{\psi}(a\xi)|^2 d\xi,$$

which implies

$$(Tf, Tg) = \int_{\mathbb{R}} \left(\int_{\mathbb{R}} \frac{|a| \cdot |\hat{\psi}(a\xi)|^2}{a^2} da \right) \hat{f}(\xi) \overline{\hat{g}(\xi)} d\xi.$$

Notice that the change of variable $a\xi = \eta$ gives

$$\int_{\mathbb{R}} \frac{|a| \cdot |\hat{\psi}(a\xi)|^2}{a^2} da = \int_{\mathbb{R}} \frac{|\hat{\psi}(\eta)|^2}{|\eta|} d\eta = 1.$$

Thus we get

$$(Tf, Tg) = (\hat{f}, \hat{g})$$

and Theorem 9 follows from the Plancherel identity (third time). \square

Remark: Think of T as an operator, formally we have

$$(Tf, Tg) = (T^*Tf, g),$$

where T^* denotes the adjoint T . Thus the above theorem implies the following **inversion formula**:

$$f = T^*Tf.$$

Now let us find a formula for T^* :

$$(F, Tg) = \int_{\mathbb{R}^2} F(a, b) \overline{(g, \psi^{a,b})} \frac{dadb}{a^2} = \int_{\mathbb{R}^3} F(a, b) \psi^{a,b}(x) \overline{g(x)} \frac{dadbdx}{a^2},$$

which gives

$$(T^*F)(x) = \int_{\mathbb{R}^2} F(a, b) \psi^{a,b}(x) \frac{dadb}{a^2}, \quad \forall F \in T(L^2(\mathbb{R})).$$

Of course the above formula is well defined only if ψ is sufficiently regular, but if we think of T^*F which maps $g \in L^2(\mathbb{R})$ to

$$\int_{\mathbb{R}^2} F(a, b) \left(\int_{\mathbb{R}} \psi^{a,b}(x) \overline{g(x)} dx \right) \frac{dadb}{a^2},$$

then T^*f is well defined and satisfies $(T^*F, g) = (F, Tg)$.

Compare to windowed Fourier transform (or short time Fourier transform): The main idea of the windowed Fourier transform is the following: given $w(t)$ such that

$$\int_{\mathbb{R}} w(t) dt = 1,$$

we call w a **window function**. Notice that a change of variable gives $\int w(t-b) db = 1$ for every $t \in \mathbb{R}$ thus

$$f(t) = f(t) \cdot \int_{\mathbb{R}} w(t-b) db = \int_{\mathbb{R}} f(t)w(t-b) db, \quad \forall b \in \mathbb{R}.$$

Now we can write the Fourier transform $\hat{f}(\lambda) = \int f(t)e^{-2\pi it\lambda} dt$ as

$$\hat{f}(\lambda) = \int_{\mathbb{R}} \left(\int_{\mathbb{R}} f(t)w(t-b) db \right) e^{-2\pi it\lambda} dt = \int_{\mathbb{R}} \left(\int_{\mathbb{R}} f(t)w(t-b)e^{-2\pi it\lambda} dt \right) db.$$

We call

$$(Wf)(\lambda, b) := \int_{\mathbb{R}} f(t)w(t-b)e^{-2\pi it\lambda} dt$$

the **windowed Fourier transform**, which gives another expression for the Fourier transform:

$$\hat{f}(\lambda) = \int_{\mathbb{R}} (Wf)(\lambda, b) db.$$

The Fourier inversion formula gives

$$f(t) = \int_{\mathbb{R}} \hat{f}(\lambda)e^{2\pi it\lambda} d\lambda = \int_{\mathbb{R}} \left(\int_{\mathbb{R}} (Wf)(\lambda, b) db \right) e^{2\pi it\lambda} d\lambda.$$

For more on windowed Fourier transform (usually we consider the case that $w = |\tilde{w}|^2$, $\|\tilde{w}\| = 1$ and define the windowed Fourier transform using \tilde{w}), see section 2.7–2.8 in [2].

REFERENCES

- [1] A. Boggess and F. Narcowich, *A first course in Wavelets with Fourier Analysis*, Wiley, 2nd Edition, 2009.
- [2] I. Daubechies, *Ten lectures on wavelets*, Philadelphia, PA, 1992.
- [3] H. Dym and H. P. McKean, *Fourier Series and Integrals*, Academic Press, 1972.
- [4] W. Feller, *An Introduction to Probability Theory and its Applications*, Volume I, II around 1968.
- [5] C. Gasquest and P. Witomski, *Fourier analysis and applications*, TAM 30.
- [6] E. Stein and R. Shakarchi, *Fourier analysis*, Princeton lectures in analysis.

DEPARTMENT OF MATHEMATICAL SCIENCES, NORWEGIAN UNIVERSITY OF SCIENCE AND TECHNOLOGY,
NO-7491 TRONDHEIM, NORWAY

E-mail address: xu.wang@ntnu.no