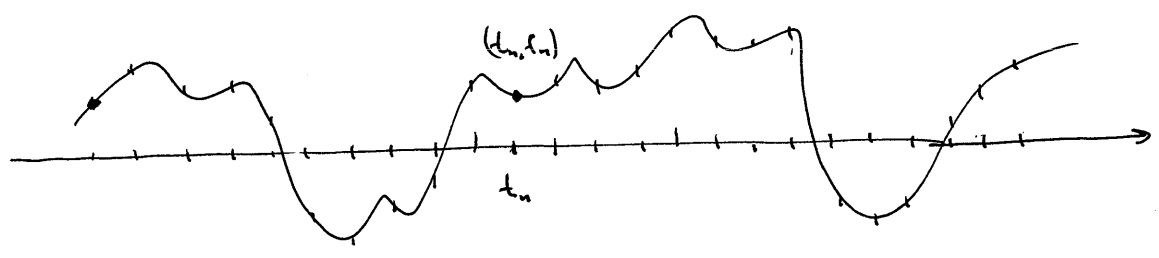


The discrete wavelet transform

Questions: how to compute the scaling and wavelet coefficients?

$$f_j(t) = \sum_{k=-\infty}^{\infty} w_{j,k} \psi_{j,k}(t) + \sum_k s_{j_0} \varphi_{j_0,k}(t)$$

First: we have to decide a finest scale, j_0 and determine $s_{j,k} = \langle f, \varphi_{j,k} \rangle$.



Usually the signal $f(t)$ is sampled to give a discrete signal $(f_n)_{n=-\infty}^{\infty}$. In the case of the Haar wavelet the $s_{j,k}$ are simply the f_n , but for higher order wavelets the calculation is more involved. Note that the signal must not contain too high frequencies.

Once the $s_{j,k}$ are known, we may proceed recursively:

Assume that $s_{j_{r+1},k} = \langle f, \varphi_{j_{r+1},k} \rangle$ are known.

We may write $f_{j_r} = \sum \langle f, \varphi_{j_r,k} \rangle \varphi_{j_r,k}$ (n.b. notation!)

and we know that $f_{j_{r+1}} = f_{j_r} + d_{j_r}$.

hence

$$f_{j,l}(t) = \sum_k s_{j,l,k} \varphi_{j,l,k}(t) \\ = \sum_k s_{j,k} \varphi_{j,k}(t) + \sum_k w_{j,k} \varphi_{j,l}(t) = f_{j,k}(t) + d_j(t).$$

We may compute the scalar product with $\varphi_{j,l}$, to get

$$\sum_k s_{j,l,k} \langle \varphi_{j,l,k}, \varphi_{j,l} \rangle = \sum_k s_{j,k} \underbrace{\langle \varphi_{j,k}, \varphi_{j,l} \rangle}_{= \delta_{k,l}} + \sum_k w_{j,k} \underbrace{\langle \varphi_{j,k}, \varphi_{j,l} \rangle}_{= 0}$$

Because $\varphi_{j,k} = \sqrt{2} \sum_m h_m \varphi_{j,l,m+2k}$

we check this:

$$\varphi_0 = \varphi(t) = \sum_m h_m \varphi(2t-m).$$

$$\text{Then } \varphi_{0,k} = \varphi(t-k) = \sum_m h_m \sqrt{2} \sqrt{2} \varphi(2t-m-2k) \\ = \sum_m \sqrt{2} h_m \varphi_{1,2k+m}(t)$$

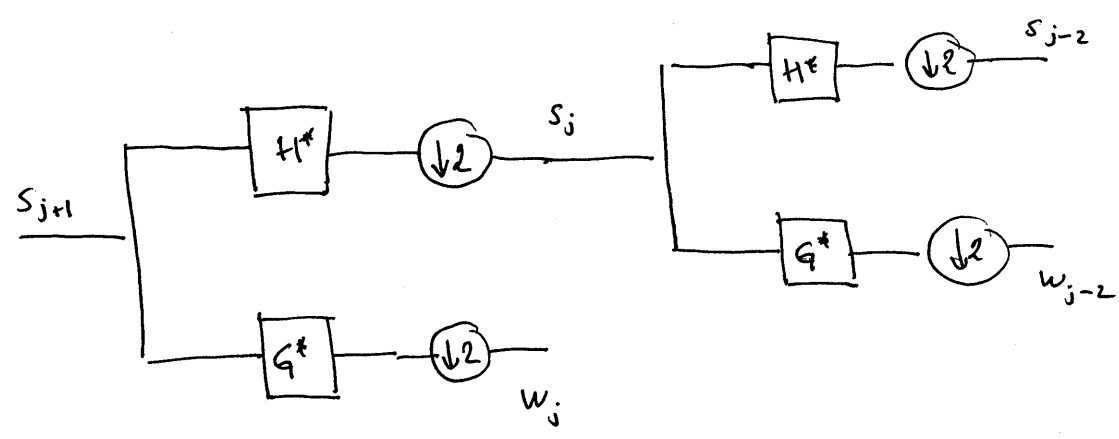
$$\text{so } \langle \varphi_{j,l,k}, \varphi_{j,l} \rangle = \sqrt{2} \sum_m h_m \underbrace{\langle \varphi_{j,l,k}, \varphi_{j,l,m+2l} \rangle}_{= \delta_{k,2l+m}} = \sqrt{2} h_{k-2l}$$

A similar calculation holds for $\langle f_{j,l}, \varphi_{j,l} \rangle$, and

in summary

$$s_{j,k} = \sqrt{2} \sum_l h_{l-2k} s_{j,l} \\ w_{j,k} = \sqrt{2} \sum_l g_{l-2k} s_{j,l}$$

This may be expressed with a filter bank:



The inverse calculation is carried out similarly:

write

$$f_{j+1} = \sum_l s_{j,l} \psi_{j,l} + \sum_l w_{j,l} \psi_{j,l}$$

$$= \sqrt{2} \sum_l \sum_m s_{j,l} h_m \psi_{j+1, m+2l} + \sqrt{2} \sum_l \sum_m w_{j,l} g_m \psi_{j+1, m+2l}$$

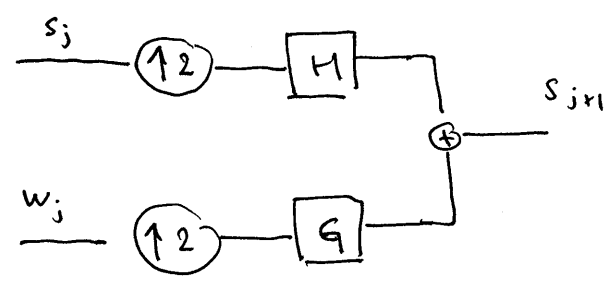
and

$$s_{j+1, k} = \langle f_{j+1}, \psi_{j+1, k} \rangle =$$

$$= \frac{1}{\sqrt{2}} \sum_l \sum_m (s_{j,l} h_m + w_{j,l} g_m) \langle \psi_{j+1, m+2l}, \psi_{j+1, k} \rangle$$

$$= \sqrt{2} \sum_l (s_{j,l} h_{k-2l} + w_{j,l} g_{k-2l})$$

This corresponds to the following filter bank:



(the factors $\sqrt{2}$ may be included in the filters)

Biorthogonal systems

Let $\{V_j\}$ be an MRA, and let $\{\tilde{V}_j\}$ be a dual MRA.

The scaling function and mother wavelets for \tilde{V}_j satisfy

$$\tilde{\varphi}(t) = \sum_k \tilde{h}_k \tilde{\varphi}(2t-k)$$

$$\tilde{\psi}(t) = \sum_k \tilde{g}_k \tilde{\varphi}(2t-k)$$

The biorthogonality conditions are

$$\langle \varphi_{j,k}, \tilde{\varphi}_{j,l} \rangle = \delta_{k,l}$$

$$\langle \psi_{j,k}, \psi_{j,l} \rangle = \delta_{k,l}$$

$$\langle \varphi_{j,k}, \tilde{\psi}_{j,l} \rangle = 0$$

$$\langle \tilde{\varphi}_{j,k}, \psi_{j,l} \rangle = 0$$

At the finest resolution we may then write

$$f_j(t) = \sum_k \langle f, \tilde{\varphi}_{j,k} \rangle \varphi_{j,k}(t) = \sum_k s_{j,k} \varphi_{j,k} + \sum_k w_{j,k} \psi_{j,k}$$

$$= \sum_k s_{j,k} \varphi_{j,k}(t) + \sum_k w_{j,k} \psi_{j,k}$$

and just as in the orthogonal case,

$$f(t) = \sum_{j,k} \langle f, \tilde{\psi}_{j,k} \rangle \psi_{j,k}$$

Note that in this formula we have both $\tilde{\psi}_{j,k}$ and $\psi_{j,k}$.

Approximation

Assume that the scaling function is such that it reproduces polynomials of order $N-1$.

That means that for each $\alpha = 0, \dots, N-1$,

$$t^\alpha = \sum_k c_k^\alpha \varphi(t-k), \quad \text{for some coefficients } \{c_k^\alpha\}.$$

It is a standard result from interpolation theory, that if f is differentiable α times, then it can be well approximated by polynomials, and it is possible to deduce

$$\|f - p_j f\| \leq C 2^{-j\alpha} \|D^\alpha f\|.$$

Note that if $t^\alpha \in V_j$, then $t^\alpha \perp \tilde{W}_j$, and therefore

$$\langle t^\alpha, \tilde{\Psi}_{j,k} \rangle = 0 \quad \text{for every wavelet } \tilde{\Psi}_{j,k}, \text{ i.e.,}$$

$$\int t^\alpha \tilde{\Psi}_{j,k}(t) dt = 0, \quad \alpha = 0, \dots, N-1.$$

We say that the dual wavelets have N vanishing moments (including 0). But that means in turn that

$$D^\alpha \mathcal{F}(\tilde{\Psi})(\omega) = 0 \quad \text{for } \alpha = 0, \dots, N-1.$$

And because

$$\mathcal{F}(\tilde{\Psi})(d\omega) = \mathcal{F}(\tilde{\varphi})(\omega) \tilde{G}(\omega) \quad \text{and } \mathcal{F}(\tilde{\varphi})(\omega) = 1,$$

we ~~must~~ must have $\tilde{G}(\omega)$ has a zero of order N at $\omega = 0$, which in turn implies that

$$H(\omega) = \left(\frac{e^{-i\omega} + 1}{2} \right)^N Q(\omega), \quad \text{where } Q \text{ is } 2\pi\text{-periodic.}$$

Two dimensional signal processing

$$\mathbb{Z}^2 = \{ (k_x, k_y), \quad k_x = \dots, -1, 0, 1, \dots, \quad k_y = \dots, -1, 0, 1, \dots \}$$

A filter in two dimensions is a map that takes a signal (image), and transforms it:

$$H: f \mapsto g$$

$$\text{where } f: \mathbb{Z}^2 \rightarrow \mathbb{R} \quad (\text{or } \mathbb{C})$$

$$g: \mathbb{Z}^2 \rightarrow \mathbb{R} \quad (\text{or } \mathbb{C})$$

The "shift operator", S^n , $n \in \mathbb{Z}^2$, is defined

$$\text{by } g = S^n f, \quad g_k = f_{k-n} \quad \forall k \in \mathbb{Z}^2.$$

A Filter is "shift invariant" if

$$H(S^n f) = S^n(Hf).$$

The impulse response of a filter is

$$h = H\delta,$$

$$\text{where } \delta_k = \begin{cases} 1 & \text{when } k = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \\ 0 & \text{otherwise} \end{cases}.$$

If $f = (f_n)_{n \in \mathbb{Z}^2}$, then $g = Hf$ may be

written

$$g = \sum f_n S^n h = h * f$$

\uparrow
def.

The 2-dimensional (discrete) Fourier transform is

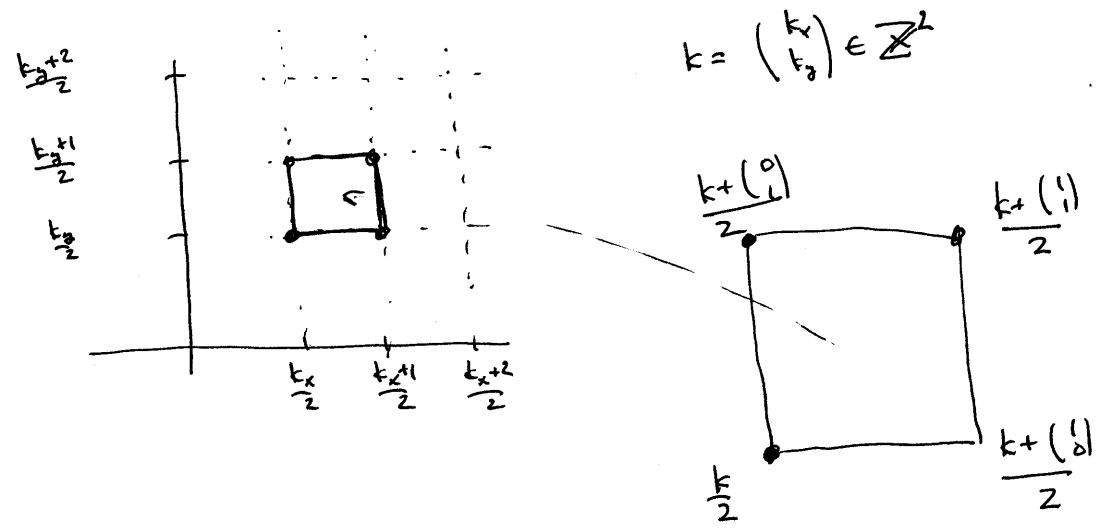
$$F(\xi, \eta) = \sum_{k \in \mathbb{Z}^2} f_k e^{-i(k_x \xi + k_y \eta)}$$

and the convolution theorem is

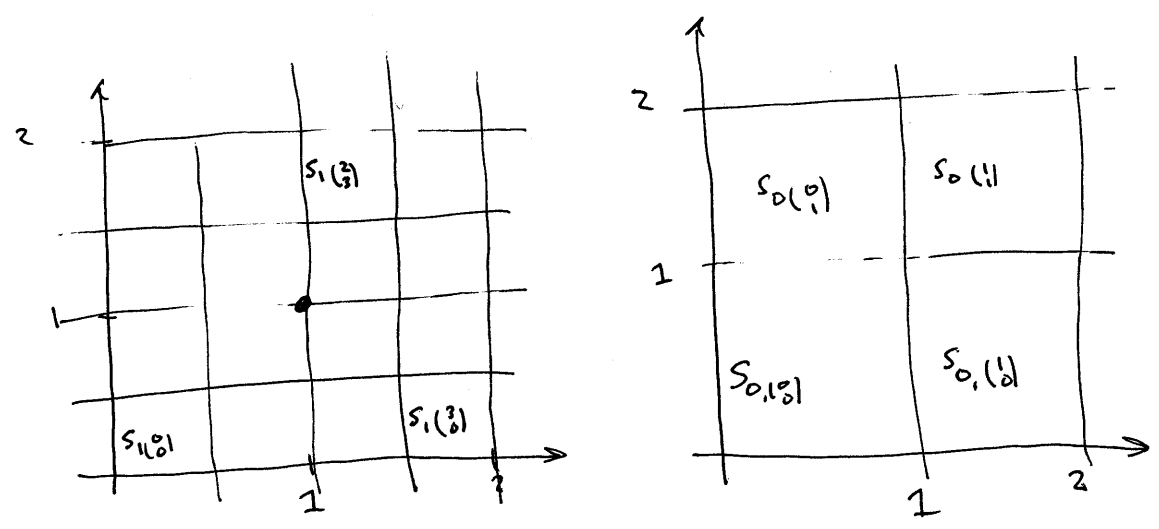
$$g = h * f \iff G(\xi, \eta) = H(\xi, \eta) F(\xi, \eta)$$

Wavelets in higher dimension

The 2-dimensional Haar system



Notation



We construct a function

$f_i(x,y)$ that is constant on the indicated squares, with values $s_{i,k}$. We can then reduce the resolution by computing averages:

$$s_{0,k} = \frac{1}{4} (s_{i,2k} + s_{i,2k+(b)} + s_{i,2k+(p)} + s_{i,2k+(l)})$$

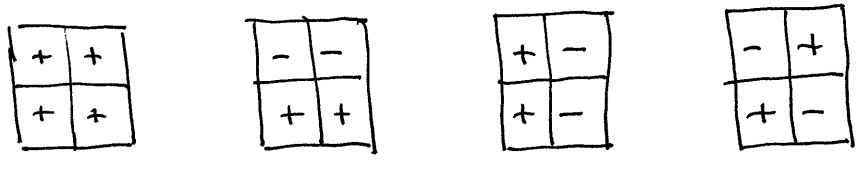
and similarly we may compute differences (but now there are three different ones):

$$w_{0,k}^H = \frac{1}{4} (s_{i,2k} + s_{i,2k+(b)} - s_{i,2k+(p)} - s_{i,2k+(l)})$$

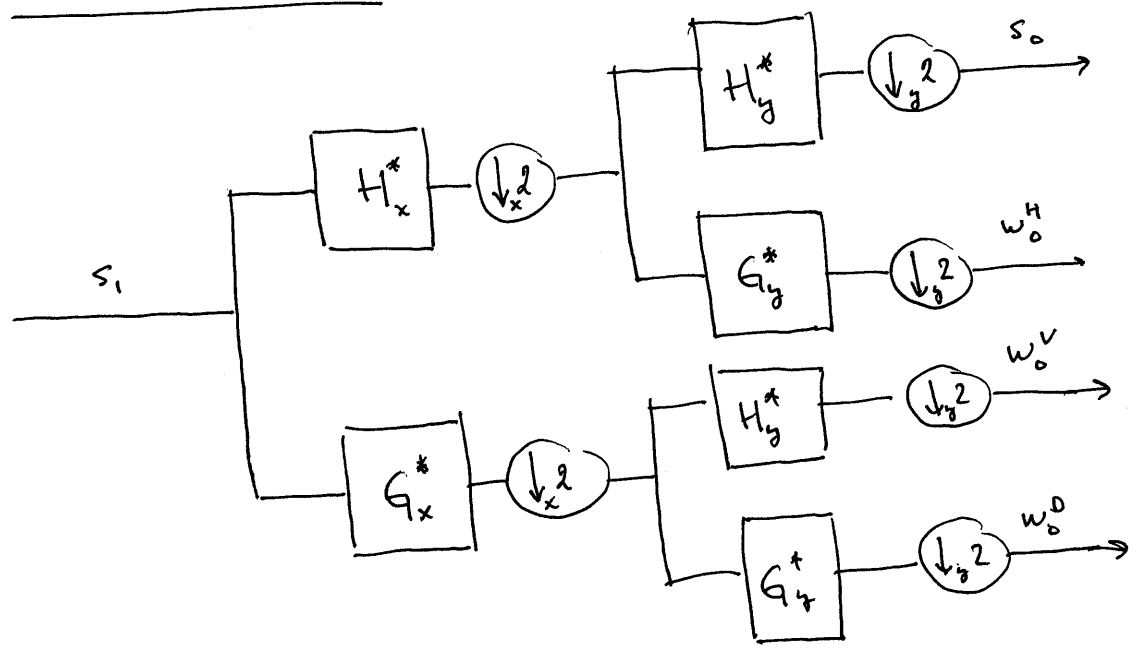
$$w_{0,k}^V = \frac{1}{4} (s_{i,2k} - s_{i,2k+(b)} + s_{i,2k+(p)} - s_{i,2k+(l)})$$

$$w_{0,k}^D = \frac{1}{4} (s_{i,2k} - s_{i,2k+(b)} - s_{i,2k+(p)} + s_{i,2k+(l)})$$

Here H, V and D stand for "horizontal", "vertical" and "diagonal", and they may be remembered by



With filter banks:



Separable scaling functions

For the Haar system,

$$\Phi(x, y) = \phi(x) \phi(y)$$

$$\Psi^H(x, y) = \phi(x) \psi(y)$$

$$\Psi^V(x, y) = \psi(x) \phi(y)$$

$$\Psi^D(x, y) = \psi(x) \psi(y).$$

Any 1-dimensional wavelet system may be used to construct a 2-dimensional system in this way.

The scaling equation in this case is

$$\Phi(x, y) = 4 \sum_k h_k \phi(2x - k_x, 2y - k_y)$$

$$D = D, V, H.$$

$$\Psi^D(x, y) = 4 \sum_k g_k^D \phi(2x - k_x, 2y - k_y)$$

The MRA is as before:

$$V_j \subset V_{j+1} \quad W_j^D \subset V_{j+1},$$

and

$$f_{j+1} = f_j + d_j^H + d_j^V + d_j^D$$

Multi dimensional Fourier transforms

Notation: $x = (x_1, \dots, x_n)$
 $\xi = (\xi_1, \dots, \xi_n)$

If $k = (k_1, \dots, k_n) \in \mathbb{Z}^n$ ($k_i \geq 0$),
 let $|k| = k_1 + \dots + k_n$

We define $x^k = x_1^{k_1} \dots x_n^{k_n}$

$$\frac{\partial^{1k}}{\partial x^k} = \frac{\partial^{|k|}}{\partial x_1^{k_1} \dots \partial x_n^{k_n}}$$

Def Let $f: \mathbb{R}^n \rightarrow \mathbb{C}$; then

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

Prop The inverse transform is given by

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

Almost all rules from the Fourier transform in \mathbb{R} apply:

$$f(x, y) \rightarrow \hat{f}(\xi, \eta)$$

[note that here we write f as a function of two arguments, $x, y \in \mathbb{R}$]

$$f(ax, by) \rightarrow \frac{1}{|ab|} \hat{f}\left(\frac{\xi}{a}, \frac{\eta}{b}\right)$$

(for $a, b \in \mathbb{R}$)

$$f(x-a, y-b) \rightarrow e^{-2\pi i(a\xi + b\eta)} \hat{f}(\xi, \eta)$$

$$\left(\frac{\partial}{\partial x}\right)^m \left(\frac{\partial}{\partial y}\right)^n f(x, y) \rightarrow (2\pi i \xi)^m (2\pi i \eta)^n \hat{f}(\xi, \eta)$$

$$\left(\frac{\partial^2}{\partial x^2} + \frac{\partial^2}{\partial y^2}\right) f(x, y) \rightarrow -4\pi^2 (\xi^2 + \eta^2) \hat{f}(\xi, \eta)$$

Def $S(\mathbb{R}^n) \subset C^\infty(\mathbb{R}^n)$ is defined by the family of seminorms $\varphi \mapsto \sup_{x \in \mathbb{R}^n} \left| x^\alpha \frac{\partial^\beta}{\partial x^\beta} \varphi(x) \right|$, $\alpha \in \mathbb{N}^n$, $\beta \in \mathbb{N}^n$.

Def $\varphi_n \rightarrow 0$ in $S(\mathbb{R}^n)$ if for all $\alpha, \beta \in \mathbb{N}^n$

$$\sup_{x \in \mathbb{R}^n} \left| x^\alpha \frac{\partial^\beta}{\partial x^\beta} \varphi_n \right| \rightarrow 0 \text{ when } n \rightarrow \infty.$$

Def A tempered distribution in \mathbb{R}^n is a complex valued linear functional T ,

$$T: S(\mathbb{R}^n) \rightarrow \mathbb{C}, \text{ such that}$$

$T(\varphi_n) \rightarrow 0$ when $n \rightarrow \infty$ for all sequences $\varphi_n \rightarrow 0$ in $S(\mathbb{R}^n)$.

Ex $\delta \otimes \delta$ (this could be written $\delta(x)\delta(y)$)

$$\langle \delta \otimes \delta, \varphi \rangle = \varphi(0,0).$$

Note this is common notation: if $f: \mathbb{R} \rightarrow \mathbb{R}$ and $g: \mathbb{R} \rightarrow \mathbb{R}$ the function $h(x,y) = f(x)g(y)$ is often denoted $f \otimes g$.

Observe that $f \otimes g \neq g \otimes f$.

One can define $T_1 \otimes T_2 \in S'(\mathbb{R}^2)$ for any pair $T_1, T_2 \in S'(\mathbb{R})$ by their action on $\varphi = \varphi_1 \otimes \varphi_2 \in S(\mathbb{R}^2)$:

$$\langle T_1 \otimes T_2, \varphi_1 \otimes \varphi_2 \rangle = T_1(\varphi_1) T_2(\varphi_2).$$

Problem: is it enough to define $T_1 \otimes T_2$ on the subset of $S(\mathbb{R}^2)$ given by $\varphi = \varphi_1 \otimes \varphi_2$?

Solution One can approximate any $\varphi \in S(\mathbb{R}^2)$

by a sum of tensor products:

$$\varphi(x,y) = \sum_{k=1}^N \varphi_{1k}(x) \varphi_{2k}(y) + O(\varepsilon),$$

which means that $\varphi_N(x_1, x_2) = \sum_{k=1}^N \varphi_{1k}(x) \varphi_{2k}(y) \rightarrow \varphi(x,y)$
in $S(\mathbb{R}^2)$ when $N \rightarrow \infty$.

$$\begin{aligned} \text{Then } \langle T_1 \otimes T_2, \varphi \rangle &= \lim_{N \rightarrow \infty} \langle T_1 \otimes T_2, \sum_{k=1}^N \varphi_{1k} \otimes \varphi_{2k} \rangle \\ &= \lim_{N \rightarrow \infty} \sum_{k=1}^N T_1(\varphi_{1k}) T_2(\varphi_{2k}) \end{aligned}$$

The Fourier transform of tensor products

Let $\varphi = \varphi_1 \otimes \varphi_2$, i.e., $\varphi(x,y) = \varphi_1(x) \varphi_2(y)$.

Then $\hat{\varphi}(\xi, \eta) = \hat{\varphi}_1(\xi) \hat{\varphi}_2(\eta)$ (check that!).

This makes it easy to compute the Fourier transform of many $\varphi: \mathbb{R}^2 \rightarrow \mathbb{C}$ and tempered distributions

$$T \in S'(\mathbb{R}^2).$$

$$\begin{aligned} \text{Ex: } \frac{\partial^{|\mathbf{k}|}}{\partial x_1^{k_1} \partial x_2^{k_2}} \varphi_1 \otimes \varphi_2 &= \left(\frac{\partial^{k_1}}{\partial x_1^{k_1}} \varphi_1 \right) \otimes \left(\frac{\partial^{k_2}}{\partial x_2^{k_2}} \varphi_2 \right) \\ &= (2\pi i \xi_1)^{k_1} \hat{\varphi}_1(\xi_1) (2\pi i \xi_2)^{k_2} \hat{\varphi}_2(\xi_2). \end{aligned}$$

The Hankel transform and relatives

Many variations of the Fourier transform are derived from the usual Fourier transform in \mathbb{R}^n , by using certain symmetries. - The Hankel transform is one example.

Suppose that $\varphi(x, y) = \varphi(r)$ where $r = \sqrt{x^2 + y^2}$:
 $\varphi(x, y)$ is invariant under rotations in the plane \mathbb{R}^2 , or in other words, rotationally symmetric.

Then

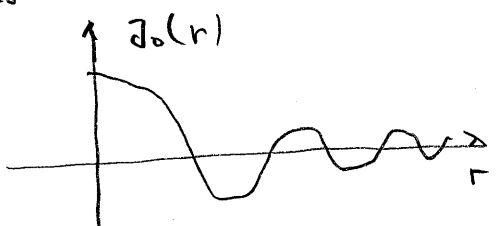
$$\begin{aligned} \hat{\varphi}(\xi, \eta) &= \iint_{\mathbb{R}^2} e^{-2\pi i(x\xi + y\eta)} \varphi(x, y) dx dy = \text{(use polar coordinates)} \\ &= \int_0^\infty \int_0^{2\pi} e^{-2\pi i r(\xi \cos \theta + \eta \sin \theta)} \varphi(r) r dr d\theta \\ &= \int_0^\infty \int_0^{2\pi} e^{-2\pi i r(\xi \cos \theta + \eta \sin \theta)} d\theta \varphi(r) r dr. \end{aligned}$$

But $\xi \cos \theta + \eta \sin \theta = \sqrt{\xi^2 + \eta^2} \cos(\theta + \lambda)$ for some λ .

Let $\rho = \sqrt{\xi^2 + \eta^2}$. Then

$$\begin{aligned} \frac{1}{2\pi} \int_0^{2\pi} e^{-2\pi i r(\xi \cos \theta + \eta \sin \theta)} d\theta &= \frac{1}{2\pi} \int_0^{2\pi} e^{-2\pi i r \rho \cos(\theta + \lambda)} d\theta \\ &= \frac{1}{2\pi} \int_0^{2\pi} e^{-2\pi i r \rho \cos \theta} d\theta := J_0(2\pi r \rho) \end{aligned}$$

This is the Bessel function of order 0.



It solves the equation

$$x^2 f''(x) + x f'(x) + x^2 f(x) = 0.$$

Therefore $\hat{\psi}(\xi, \eta) = \int_0^\infty 2\pi J_0(2\pi r \rho) \psi(r) r dr = \tilde{\psi}(\rho)$.

Def $F(\eta) = 2\pi \int_0^\infty J_0(2\pi r \eta) f(r) r dr$ is the (zeroth order) Hankel transform of f .

Note The inverse transform is $f(r) = 2\pi \int_0^\infty F(\eta) J_0(2\pi r \eta) \eta d\eta$.

In general the Fourier transform of a rotationally symmetric function $f: \mathbb{R}^n \rightarrow \mathbb{C}$ is computed in a similar way: if $f(x_1, \dots, x_n) = g(\sqrt{x_1^2 + \dots + x_n^2}) = g(r)$,

$$\hat{f}(\xi_1, \dots, \xi_n) = \int_{\mathbb{R}^n} e^{-2\pi i(x_1 \xi_1 + \dots + x_n \xi_n)} g(r) dx_1 \dots dx_n.$$

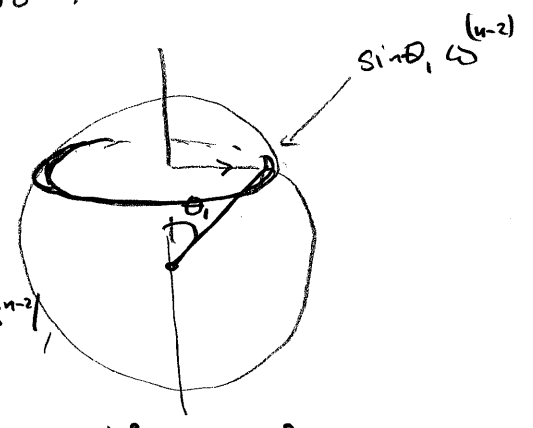
In polar coordinates,

$$(x_1, \dots, x_n) = r (\cos \theta_1, \sin \theta_1, \cos \theta_2, \dots) \equiv r \omega \text{ where } |\omega| = 1.$$

$$\text{Then } \hat{f}(\xi_1, \dots, \xi_n) = \int_0^\infty \int_{S^{n-1}} e^{-2\pi i r \xi \cdot \omega} d\omega^{n-1} g(r) r^{n-1} dr$$

Here $\int_{S^{n-1}} e^{-2\pi i \xi \cdot \omega} d\omega^{n-1}$ only depends on the scalar product between ξ and ω , and therefore we may choose coordinates on S^{n-1} so that $\xi \cdot \omega = (\xi_1, \dots, \xi_n) \cdot (\omega_1, \dots, \omega_n) = \xi_1 \omega_1$.

$$\text{Also, } d\omega^{n-1} = d\theta_1 (\sin \theta_1)^{n-2} d\omega^{n-2}$$



$$\Rightarrow \int_{S^{n-1}} e^{-2\pi i \xi \cdot \omega} d\omega^{n-1} = \int_0^\pi e^{-2\pi i \xi_1 \cos \theta} (\sin \theta)^{n-2} d\theta \cdot |S^{n-2}|$$

where $|S^{n-2}|$ is the "area" of $S^{n-2} = \{x \in \mathbb{R}^{n-2}, |x|=1\}$

Homogeneous functions and distributions

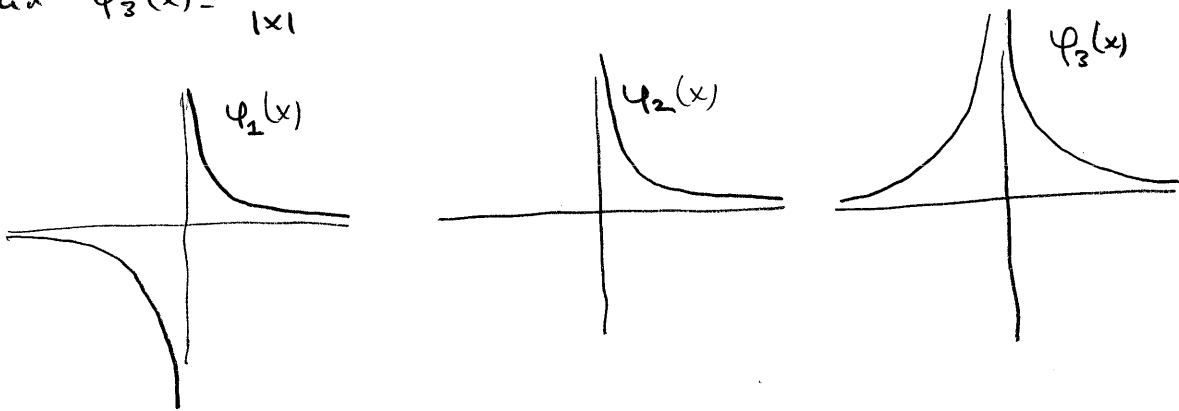
Def A function $f: \mathbb{R}^n \setminus \{0\} \rightarrow \mathbb{C}$ is called homogeneous of degree k if

$$f(tx) = t^k f(x) \quad \text{for all } x \neq 0.$$

Example $\varphi_1(x) = \frac{1}{x}$ is homogeneous of degree -1 .

But that is true also for $\varphi_2(x) = \begin{cases} \frac{1}{x} & \text{when } x > 0 \\ 0 & \text{otherwise} \end{cases}$

and $\varphi_3(x) = \frac{1}{|x|}$



Note that if f is homogeneous of degree k ,

$$\begin{aligned} \langle f, \varphi \rangle &= \int_{\mathbb{R}^n} f(x) \varphi(x) dx = \int_{\mathbb{R}^n} f(\alpha x) \alpha^{-k} \varphi(x) dx = \\ &\quad \text{from the homogeneity} \\ &= \int_{\mathbb{R}^n} f(y) \alpha^{-k} \varphi\left(\frac{y}{\alpha}\right) \frac{dy}{\alpha^n} = \langle f, \frac{1}{\alpha^{n+k}} \varphi\left(\frac{\cdot}{\alpha}\right) \rangle \end{aligned}$$

A distribution $T \in \mathcal{S}'(\mathbb{R}^n)$ is homogeneous of degree k

if $\langle T, \varphi \rangle = \langle T, \frac{1}{\alpha^{n+k}} \varphi\left(\frac{\cdot}{\alpha}\right) \rangle$ for all $\varphi \in \mathcal{S}(\mathbb{R}^n)$.

Ex $\langle \delta, \varphi \rangle = \varphi(0) = \varphi\left(\frac{0}{\alpha}\right) \frac{1}{\alpha^{n+n}}$

$\Rightarrow \delta$ is homogeneous of degree $-n$.

The Fourier transform of homogeneous functions and distributions

Formally, if $f: \mathbb{R} \rightarrow \mathbb{R}$ is homogeneous of degree k , and

$$\hat{f}(\xi) = \int_{\mathbb{R}} e^{-2\pi i x \xi} f(x) dx, \text{ then}$$

$$\begin{aligned} \hat{f}(t\xi) &= \int_{\mathbb{R}} e^{-2\pi i x t\xi} f(x) dx = \int_{\mathbb{R}} e^{-2\pi i y \xi} f\left(\frac{y}{t}\right) \frac{dy}{t} = \\ & \quad \uparrow \text{by a change of variables} \qquad \qquad \qquad \uparrow \text{by the homogeneity of } f \\ &= \int_{\mathbb{R}} e^{-2\pi i y \xi} f(y) \frac{dy}{t^{k+1}} = \frac{1}{t^{k+1}} \hat{f}(\xi) \end{aligned}$$

So, if f is homogeneous of degree k , then \hat{f} is homogeneous of degree $-k-1$. But note that this may not be well defined!

However if $T \in \mathcal{S}'(\mathbb{R})$ is homogeneous of degree λ , then

$$\langle \hat{T}, \varphi \rangle = \langle T, \hat{\varphi} \rangle = \langle T, \frac{1}{t^{1+\lambda}} \hat{\varphi}\left(\frac{\cdot}{t}\right) \rangle, \text{ and}$$

because

$$\frac{1}{t^{1+\lambda}} \hat{\varphi}\left(\frac{\xi}{t}\right) = \frac{1}{t^{1+\lambda}} \int_{\mathbb{R}} e^{-2\pi i x \frac{\xi}{t}} \varphi(x) dx = \int_{\mathbb{R}} e^{-2\pi i \xi y} \varphi(ty) \frac{1}{t^\lambda} dy$$

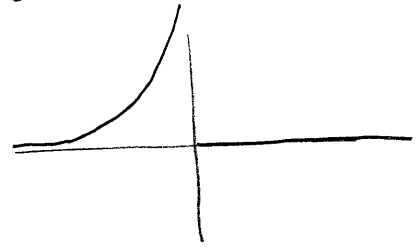
$$\Rightarrow \langle \hat{T}, \varphi \rangle = \langle T, \varphi(t \cdot) \frac{1}{t^\lambda} \rangle = \langle \hat{T}, t^{-\lambda} \varphi(t \cdot) \rangle$$

and therefore \hat{T} is homogeneous of degree $-\lambda-1$.

(recall $\langle \hat{T}, \varphi \rangle = \langle \hat{T}, t^{\alpha+1} \varphi(t \cdot) \rangle$ so we need to set $\alpha = -\lambda-1$).

Example Compute the Fourier transform

of $f(x) = \begin{cases} (-x)^{-1/2} & (x < 0) \\ 0 & \text{otherwise} \end{cases}$



Solution:

$$\text{Let } f(x) = \frac{1}{2} \left(\frac{1}{|x|^{1/2}} - \operatorname{sgn} x \frac{1}{|x|^{1/2}} \right)$$

$f_1(x) = |x|^{-1/2}$ is even, real and homogeneous of degree $-1/2$

$\Rightarrow \hat{f}_1(\xi)$ is even, real and homogeneous of degree $-(-1/2) - 1/2$

Therefore $\hat{f}_1(\xi) = \pm C |\xi|^{-1/2}$.

In the usual way, we have

$$\langle \hat{f}_1, e^{-\pi \cdot^2} \rangle = \langle f_1, e^{-\pi \cdot^2} \rangle > 0 \Rightarrow \hat{f}_1(\xi) = |\xi|^{-1/2}.$$

$f_2(x) = \operatorname{sgn}(x) \frac{1}{|x|^{1/2}}$ is odd, real and homogeneous of degree $-1/2$, so

$\hat{f}_2(\xi)$ is odd, imaginary and homogeneous of degree $-1/2$.

Therefore $\hat{f}_2(\xi) = i c \operatorname{sgn} \xi |\xi|^{-1/2}$.

But what is $c \in \mathbb{R}$?

We have $2\pi i \xi \hat{f}_2(\xi) = \mathcal{F}(f_2')$, so

$$\langle 2\pi i \xi \hat{f}_2, \varphi \rangle = \langle \mathcal{F}(f_2'), \varphi \rangle = \langle f_2', \hat{\varphi} \rangle = -\langle f_2, \hat{\varphi}' \rangle.$$

Then

$$\begin{aligned} \langle -2\pi c |\cdot|^{1/2}, e^{-\pi \cdot^2} \rangle &= -\langle \operatorname{sgn}(\cdot) |\cdot|^{-1/2}, 2\pi(\cdot) e^{-\pi \cdot^2} \rangle \\ &= -2\pi \langle |\cdot|^{1/2}, e^{-\pi \cdot^2} \rangle \end{aligned}$$

so we must have $c=1$.

Finally

$$\hat{f}(z) = \frac{1}{2} (\hat{f}_1(z) - \hat{f}_2(z)) = \frac{1}{2} (|z|^{-1/2} - i \operatorname{sgn} z |z|^{-1/2}) \quad \text{for } z \in \mathbb{R}.$$

For $z \in \mathbb{R}, z > 0$,

$\hat{f}(z) = \frac{1-i}{2} z^{-1/2}$. This can be extended to an analytical function

$$\hat{f}(z) = \frac{1}{(2iz)^{1/2}} \quad \text{in } z \in \mathbb{C} \setminus \{z \in \mathbb{C} : \operatorname{Im} z \leq 0, \operatorname{Re} z = 0\}.$$



The Abel transform

Let $k(r, x) = \begin{cases} 2r(r^2 - x^2)^{-1/2} & (r > x) \\ 0 & \text{otherwise} \end{cases}$

and let $f: [0, \infty) \rightarrow \mathbb{C}$.

Def The Abel transform of f is given by

$$f_A(x) = \int_0^\infty k(r, x) f(r) dr.$$

Inversion of the Abel transform

Let $z = x^2$ and $p = r^2$ and define $F_A(z) = f_A(x)$
 $F(p) = f(r)$.

Then

$$F_A(z) = \int_0^\infty K(z-p) F(p) dp = K * F$$

where $K(z) = \begin{cases} \frac{1}{(-z)^{1/2}} & (z < 0) \\ 0 & \text{otherwise} \end{cases}$

The map $F \mapsto F_A$ is called the "modified Abel transform".

$$F_A(\bar{z}) = K * F(\bar{z})$$

implies that

$$\mathcal{F}(F_A) = \mathcal{F}(K) \mathcal{F}(F),$$

and recall: $\mathcal{F}(K) = \frac{1}{(-2i\bar{z})^{1/2}}$. Then we may see that

$$\mathcal{F}(F) = (-2i\bar{z})^{1/2} \mathcal{F}(F_A)(\bar{z})$$

$$= -\frac{1}{\pi} \frac{1}{(-2i\bar{z})^{1/2}} 2\pi i\bar{z} \mathcal{F}(F_A)(\bar{z})$$

$$= -\frac{1}{\pi} \frac{1}{(-2i\bar{z})^{1/2}} \mathcal{F}(F'_A) = -\frac{1}{\pi} \mathcal{F}(F) \mathcal{F}(F'_A).$$

We then see that

$$F(p) = -\frac{1}{\pi} K * F'_A(p).$$

In the original variables

$$f(r) = -\frac{1}{\pi} \int_r^\infty \frac{f'_A(x)}{\sqrt{x^2 - r^2}} dx$$

Note that

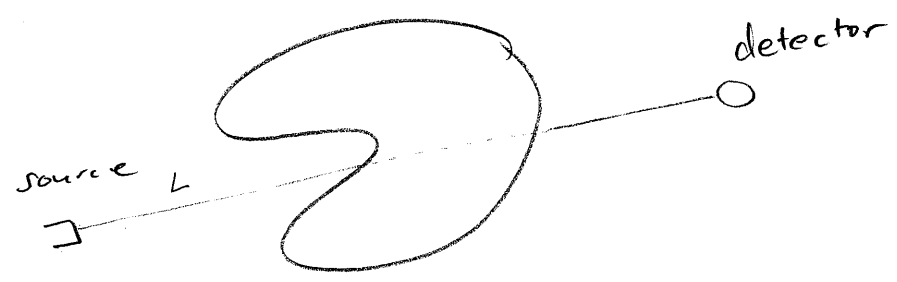
$$K * K * F' = -\pi F$$

so the operator

$$F \mapsto \frac{K}{\sqrt{\pi}} * F \quad \text{corresponds to taking a}$$

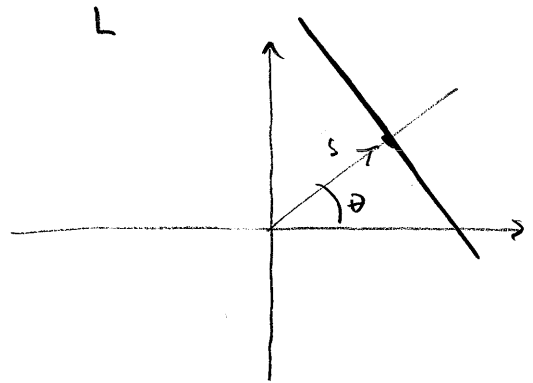
"half order integral" of F .

The Radon transform



Let $f: \mathbb{R}^2 \rightarrow \mathbb{R}$ denote the absorption per unit length of a ray passing through the object:

$$\int_L f(x(t)) dt = \text{absorption along the line } L$$



We identify the angle θ and the unit vector θ .

Let $x = (x_1, x_2)$.

Def The Radon transform of $f \in S(\mathbb{R}^2)$

is

$$R_\theta f(s) = \int_{x \cdot \theta = s} f(x) dx = \int_{\mathbb{R}^2} f(x_1, x_2) \delta(s - x \cdot \theta) dx_1 dx_2$$

Note that if f is a radial function, then

$$R_\theta f(s) = Af(s)$$

i.e. the Abel transform.

(this is an exercise to prove).

Theorem if $f \in S(\mathbb{R}^2)$ and $R_\theta f(s) = \int_{x \cdot \theta = s} f(x) dx$,

then $(R_\theta f)^\wedge(\sigma) = \hat{f}(\sigma\theta)$,

where $(R_\theta f)^\wedge(\sigma)$ is the usual Fourier transform with respect to s .

Proof Note first that the natural domain of definition for $R_\theta f(s)$ is $\mathbb{R} \times S^1$. But from the definition it is also clear that $R_{-\theta} f(-s) = R_\theta f(s)$.

We compute $(R_\theta f)^\wedge(\sigma)$:

$$(R_\theta f)^\wedge(\sigma) = \int_{-\infty}^{\infty} e^{-2\pi i \sigma s} R_\theta f(s) ds$$

$$= \int_{-\infty}^{\infty} e^{-2\pi i \sigma s} \int_{\mathbb{R}^2} f(x) \delta(x \cdot \theta - s) dx ds$$

(first compute the ds -integral)

$$= \int_{\mathbb{R}^2} f(x) e^{-2\pi i \sigma x \cdot \theta} dx = \hat{f}(\sigma\theta), \text{ where}$$

$\hat{f}(\xi)$ is the usual Fourier transform in \mathbb{R}^2 .

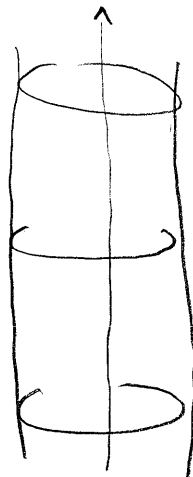
Corollary: If $f \in S(\mathbb{R}^2)$ and $R_\theta f(s) \in S(\mathbb{R} \times S^1)$,

then f can be recovered from $R_\theta f(s)$

by first obtaining $\hat{f}(\xi)$ by the theorem above,

and then computing the 2-dimensional inverse Fourier transform.

This is not efficient when discretizing in polar coordinates.



A better inversion formula for the Radon transform

Let $f \in \mathcal{S}(\mathbb{R}^2)$. Then

$$\begin{aligned}
 f(x) &= \int_{\mathbb{R}^2} e^{2\pi i x \cdot \vec{z}} \hat{f}(\vec{z}) d\vec{z} = \\
 &= \int_0^\infty \int_{S^1} e^{2\pi i x \cdot \sigma} \hat{f}(\sigma) \sigma d\sigma d\theta \quad (\text{polar coordinates}) \\
 &= \frac{1}{2} \int_{S^1} \int_{\mathbb{R}} e^{2\pi i \sigma \theta \cdot x} \hat{f}(\sigma) \underbrace{\sigma \operatorname{sgn}(\sigma)}_{=|\sigma|} d\sigma d\theta \\
 &= \frac{1}{2} \int_{S^1} \int_{\mathbb{R}} e^{2\pi i \sigma \theta \cdot x} (R_\theta f)^\wedge(\sigma) \sigma \operatorname{sgn} \sigma d\sigma d\theta \quad (*)
 \end{aligned}$$

Recall that $2\pi i \sigma \hat{f}(\sigma) = \mathcal{F}(f')(\sigma)$

$$\operatorname{sign} \sigma = \mathcal{F}\left(\frac{i}{\pi(\cdot)}\right) \quad \leftarrow \text{Fourier transform}$$

Therefore

$$(R_\theta f)^\wedge(\sigma) \sigma \operatorname{sgn} \sigma = \frac{1}{2\pi i} \mathcal{F}\left(\frac{d}{dt} R_\theta f(\cdot) * \frac{i}{\pi(\cdot)}\right)(\sigma)$$

$$\begin{aligned}
 \Rightarrow (*) &= \frac{1}{4\pi^2} \int_{S^1} \underbrace{\left((R_\theta f)' * \frac{1}{(\cdot)} \right) (\theta x)}_{=} d\theta \\
 &= \int_{\mathbb{R}} \frac{(R_\theta f)'(u)}{u-s} du
 \end{aligned}$$

Note The map $\mathcal{S}(\mathbb{R} \times S^1) \longrightarrow \mathcal{S}(\mathbb{R})$

$$g(s, \theta) \longmapsto \int_{S^1} g(x \cdot \theta, \theta) d\theta$$

is "the dual" of the Radon transform, and it is denoted R^*

The reason for the name "dual" is the following.

Take $\varphi \in S(\mathbb{R} \times S^1)$. We define

$$\langle g, \varphi \rangle = \int_{S^1} \int_{\mathbb{R}} g(s, \theta) \varphi(s, \theta) ds d\theta$$

Then

$$\langle Rf, \varphi \rangle = \int_{S^1} \int_{\mathbb{R}} R_\theta f(s) \varphi(s, \theta) ds d\theta$$

$$= \int_{S^1} \int_{\mathbb{R}} \int_{\mathbb{R}^2} f(x) \delta(x \cdot \theta - s) \varphi(s, \theta) dx ds d\theta$$

$$= \int_{\mathbb{R}^2} f(x) \int_{S^1} \int_{\mathbb{R}} \delta(x \cdot \theta - s) \varphi(s, \theta) ds d\theta dx$$

$$= \int_{\mathbb{R}^2} f(x) \underbrace{\int_{S^1} \varphi(x \cdot \theta, \theta) d\theta}_{\in S(\mathbb{R}^2)} dx = \langle f, R^* \varphi \rangle$$

↑ this is the usual $\langle \cdot, \cdot \rangle$ in \mathbb{R}^2 .

The Hilbert transform

Def $F_{H_i}(x) = \frac{1}{\pi} \int_{-\infty}^{\infty} \frac{f(y)}{y-x} dy = \left(\frac{1}{\pi(-\cdot)} * f \right)(x)$

Recall that $\mathcal{F} \left(\frac{1}{\pi(-\cdot)} \right)(s) = i \operatorname{sgn} s$

Hence

$$\left(\frac{1}{\pi(-\cdot)} * \frac{1}{\pi(-\cdot)} * f \right)^\wedge(s) = -(\operatorname{sgn} s)^2 \hat{f} = -\hat{f}$$

and then

$$\frac{1}{\pi(-\cdot)} * \frac{1}{\pi(-\cdot)} * f = -f$$

Therefore $f(x) = -\frac{1}{\pi(-\cdot)} * F_{H_i}(x)$

The analytical signal

Let $f \in S(\mathbb{R})$ be real valued and let

$$g(x) = f(x) - i F_{Hi}(x).$$

This is called "the analytical signal". The Fourier transform of

$$\begin{aligned} \hat{g}(s) &= \hat{f}(s) - i \operatorname{sgn} s \hat{f}(s) \\ &= \hat{f}(s) + \operatorname{sgn} s \hat{f}(s). \end{aligned}$$

Then $\hat{g}(s) = 0$ for $s \leq 0$.

Why is this called "the analytical signal"?

$$\text{Let } \varphi(x, y) = \frac{1}{\pi} \int_{-\infty}^{\infty} \frac{y}{(x-\xi)^2 + y^2} f(\xi) d\xi$$

$$\text{Then } \begin{cases} \Delta \varphi = 0 & \text{in } y > 0 \\ \varphi(x, 0) = f \end{cases}$$

There is an analytic function $\Phi(z)$ so that

$$\varphi(x, y) = \operatorname{Re} \Phi(z) \quad (z = x + iy)$$

Φ is analytic in $\operatorname{Im} z > 0$ and

$$\lim_{y \rightarrow 0^+} \operatorname{Im} \Phi(x + iy) = F_{Hi}(x).$$