

### §8 The real life inversion

Recall from the ~~last~~<sup>first</sup> lecture that for  $f \in S(\mathbb{R}^c)$  [in particular  $f \in C_c^\infty(\mathbb{R}^c)$ , i.e. a smooth function with compact support] we have

$$f(x) = c \int_{\mathbb{S}^1} \int_{-\infty}^{\infty} \frac{\partial_p Rf(\omega, p)}{x \cdot \omega - p} dp d\omega.$$

There are two obvious problems with this inversion formula

- The singularity of the kernel function

$$\frac{1}{x \cdot \omega - p}$$

But note that it is mainly located at  $p = x \cdot \omega$  which is the place where the measurements are done.

- The differentiation  $\partial_p Rf(\omega, p)$ .

Note, that both of those problem come from the multiplication by  $|r|$  in the frequency domain

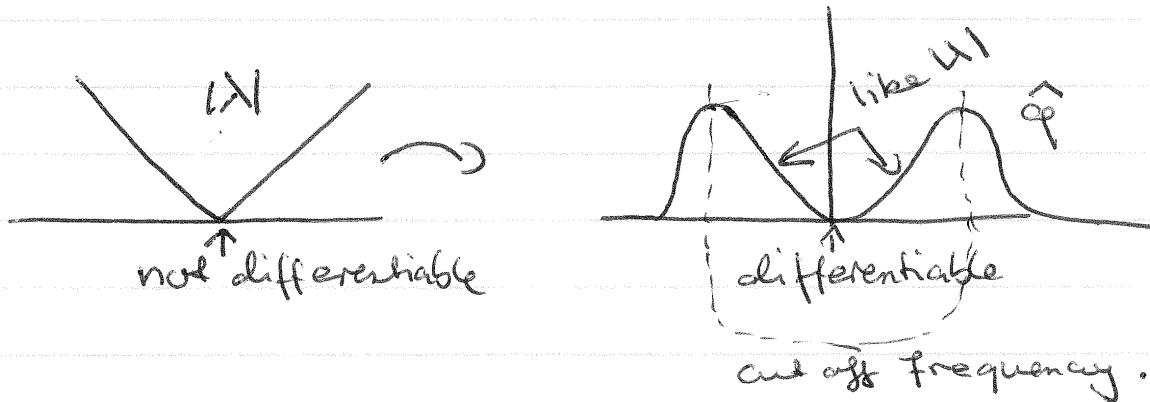
$$gRf(\omega, t) = c \mathcal{F}_t^{-1}( |r| \mathcal{F}_r Rf(\omega, r) ) \text{ at.}$$

The multiplication by  $|r|$  can be

factorized into two parts:

- (1) Multiplication by  $\lambda \Leftrightarrow$  corresponds to the derivative  $\partial_p$  in the inversion formula.
- (2) Multiplication by sign  $\lambda$  which corresponds to the Hilbert transform.

The simplest idea would be to replace  $\lambda \rightarrow |\lambda|$  by a smooth version, that also vanishes at  $\infty$ .



This would replace

$$\int_{-\infty}^{\infty} \frac{\partial_p Rf(\omega, p)}{x \cdot \omega - p} dp$$

by

$$\int_{-\infty}^{\infty} g(x \cdot \omega - p) Rf(\omega, p) dp$$

which can "easily" be discretized using a finite sum approximation to the integral. Another way to look at this is the "filtered backprojection" which is based on the following lemma.

Lemma Let  $f \in S(\mathbb{R}^d)$  and  $g \in S(\Xi)$ .

Then

$$(R^\vee g) * f(x) = R^\vee(g *_{\rho} Rf)(x)$$

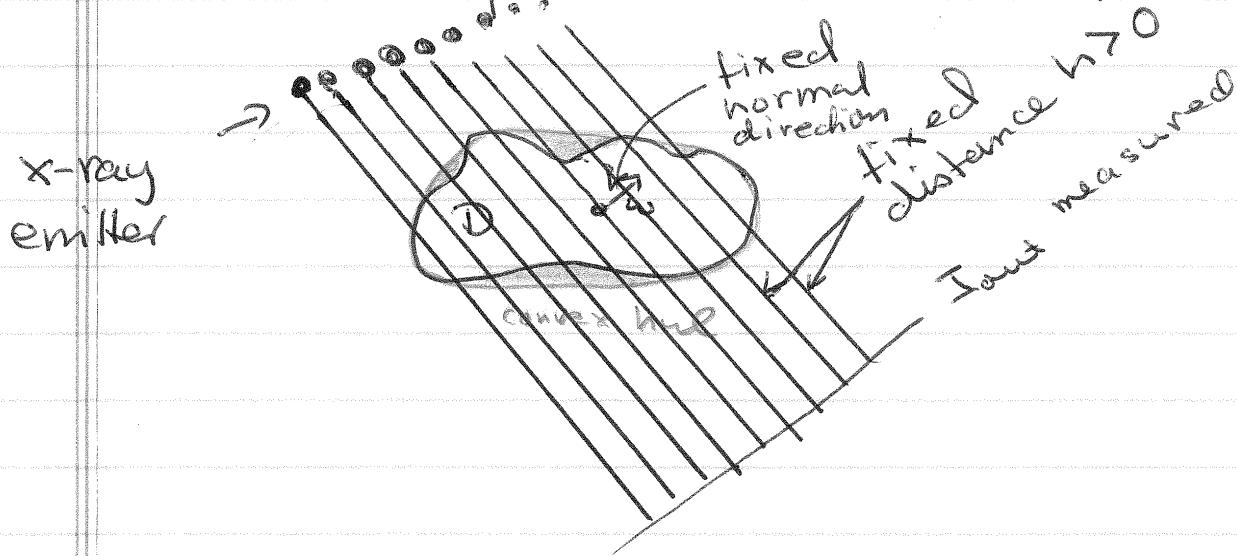
Remark: Here the subscript  $\rho$  in the convolution signs indicates convolution in the  $\rho$ -variable.

$$\begin{aligned} g *_{\rho} f(w, p) &= \int_{-\infty}^{\infty} g(p-t) \hat{f}(qt) dt \\ &= \int_{-\infty}^{\infty} g(t) \hat{f}(w, p-t) dt \quad \square \end{aligned}$$

Proof: We have

$$\begin{aligned} (R^\vee g) * f(x) &= \int_{\mathbb{R}^d} R^\vee g(y) f(x-y) dy \text{ normalized measure on } S^{d-1}. \\ &= \int_{\mathbb{R}^d} \int_{S^{d-1}} g(w, y \cdot w) d\sigma(w) f(x-y) dy \\ &= \int_{S^{d-1}} \int_{\mathbb{R}^d} g(w, y \cdot w) f(x-y) dy d\sigma(w) \\ &= \int_{S^{d-1}} \int_{-\infty}^{\infty} \int_{\omega^\perp} g(w, p) f(x-pw-z) dz dp d\sigma \\ &= \int_{S^{d-1}} \int_{-\infty}^{\infty} \int_{\omega^\perp} g(w, p) f((x-w-p)w-z) dz dp d\sigma \\ &= \int_{S^{d-1}} \int_{-\infty}^{\infty} g(w, p) Rf(w, x \cdot w - p) dp d\sigma(w) \quad \square \end{aligned}$$

This can now be used to approximate the inversion in parallel-beam scan:



other algorithms should be used for other types of scanning.

- If  $R^V g$  would be the  $\delta$ -distribution then we have

$$\begin{aligned} R^V g * f &= \delta * f = f \\ &= R^V(\cancel{R\delta}) \\ &= R^V(g * Rf) \end{aligned}$$

which would give an exact inversion.

- So we choose  $g$  such that
  - $g(w, p) = g(p) = g(-p)$  does not depend on the direction  $w$
  - $R^V g$  is "close" to the  $\delta$ -distribution

Then we get

$$f \sim R^v g * f(x) = \int_{S^{d-1}} g *_p \hat{f}(w, x \cdot w) d\sigma(w)$$

approximate the inner integral by a Riemannian sum

$$\approx \int_{S^{d-1}} h \sum_j g(x \cdot w - jh) \hat{f}(w, jh) dw$$

$\downarrow$  we would really only take finite sum

discretize the outer integral by a finite sum

$$\approx \frac{h}{N} \sum_{k=0}^N \sum_j g(x \cdot w_k - jh) \hat{f}(w_k, jh)$$

"known measurements"

We know this for a finite sequence of inputs  $\{x \cdot w_k - jh\}_{k,j}$  so we can easily program this.

### Remark

(i) The starting point is to fix the "filter"  $g$ . A prior knowledge about  $f$  (the material) can help here (low pass filter etc.)

This way the linear map

$$\int_{S^1} g * p$$

can be chosen arbitrary close to

the inverse of the Radon transform on "good" functions, or so that it preserves, enhances, properties that we are interested in.

- (2) The numbers  $g(x_k \cdot w_e - h_j)$  can be stored and we do not have to redo them each time.
- (3) We then get numbers that are close to  $f(x_k)$ , but what does this "really" tell us about  $f$ , are there error estimates, ... see the book by Natterer.

## § 9 Sampling theory

The final question that we will now discuss is the following: We can only use finitely many  $x$ , so how how exactly do the numbers  $\{f(x_k)\}$  determine the function  $f$ ? How can we reconstruct  $f$  from the discrete values?

Probably the best known theorem is the Whittaker-Shannon-Kotelnikov (WSK) - sampling theorem for band limited functions. From now on we will assume that  $d=1$ .

Def The function  $f \in L^2(\mathbb{R})$  is called band limited if  $\hat{f}$  has compact support, i.e.  $\int_{\mathbb{R}} |\hat{f}(\lambda)|^2 d\lambda = 0 \iff |\lambda| > R$ .

In this case

$$f(x) = (\pi)^{-1/2} \int_{-R}^R \hat{f}(\lambda) e^{i\lambda x} d\lambda$$

which implies that  $f$  extends to a holomorphic function and hence  $f$  is determined if we know  $\{f(x_j)\}_j$  on any sequence  $x_j$ 's which has a limit point. But this does not tell us how to construct  $f$  from  $\{f(x_j)\}_j$ !

Assume for simplicity that  $R=\pi$ , then the functions

$$e_n(x) = \frac{1}{\sqrt{2\pi}} e^{inx}$$

form an orthonormal basis for  $L^2([-π, π])$  and hence

$$\hat{f} = \sum (\hat{f}, e_n) e_n$$

But

$$(\hat{f}, e_n) = \frac{1}{\sqrt{2\pi}} \int_{-\pi}^{\pi} \hat{f}(\lambda) e^{-in\lambda} d\lambda = f(-n)$$

We also note that

$$\begin{aligned}
 \int_{-\pi}^{\pi} e_n(\lambda) e^{i\lambda x} dx &= \frac{1}{\sqrt{2\pi}} \int_{-\pi}^{\pi} e^{i(n+x)\lambda} dx \\
 &= \frac{1}{\sqrt{2\pi}} \left[ \frac{e^{i(n+x)\lambda}}{i(n+x)} \right]_{-\pi}^{\pi} \\
 &= \frac{1}{\sqrt{2\pi}} \frac{\sin(\pi(n+x))}{\pi(n+x)} \\
 &= \sqrt{2\pi} \operatorname{sinc}(\pi(x+n))
 \end{aligned}$$

It follows that

Theorem (WSK). Assume that  $\operatorname{Supp} f \subseteq [-\pi, \pi]$ .

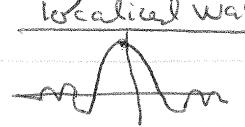
Then

$$f(x) = \sum_{n=-\infty}^{\infty} f(n) \operatorname{sinc}(\pi(x-n)).$$

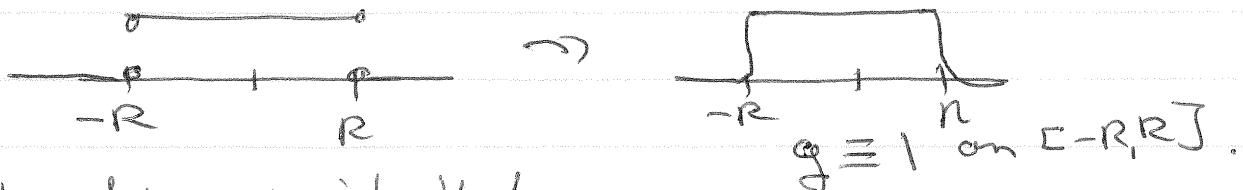
[In general if  $\operatorname{Supp} \hat{f} \subseteq [-R, R]$ , then

$$f(x) = \sum_{n=-\infty}^{\infty} \hat{f}\left(\frac{n\pi}{R}\right) \operatorname{sinc}\left(R(x - \frac{n\pi}{R})\right)$$

The only problem with this formula is that

$$|\operatorname{sinc}(x)| \sim \frac{1}{|x|} \quad |x| \gg 0$$


So the sum converges very slowly. One way is to replace  $\chi_{[-R, R]}$  by a smoother function



Will not go more into that.

Now, if  $g$  has compact support as is the case in the Radon transform of "real" objects.

Then  $g$  is not analytic so  $g$  can not be bandlimited, thus for all  $R$ :

$$g(x) \neq \sum_{n=-\infty}^{\infty} g\left(\frac{n\pi}{R}\right) \operatorname{sinc}\left(R(x - \frac{n\pi}{R})\right)$$

as a function of  $x$ .

"Def." Let  $\varepsilon > 0$ ,  $R > 0$ . we say that  $f \in L^2(\mathbb{R})$  is essentially  $R$ -bandlimited if

$$\int_{|x| \geq R} |\hat{f}(x)| dx \leq \varepsilon$$

Theorem Assume that  $f$  essentially  $R$ -bandlimited  
Then there exists a constant  $C > 0$  s.t.

$$\left| \underbrace{\sum_{n=-\infty}^{\infty} g\left(\frac{n\pi}{R}\right) \operatorname{sinc}\left(R(x - \frac{n\pi}{R})\right)}_{\text{approximation}} - g(x) \right| \leq C \cdot \varepsilon.$$

Note, that this is a bandlimited function that agrees with  $g$  at  $\frac{n\pi}{R}$ ,  $n \in \mathbb{Z}$ , and is close to  $g$  in the  $\infty$ -norm, whereas

$$f \rightarrow \hat{f} \rightarrow \hat{f} \chi_{[-R, R]} \rightarrow \hat{f} (\hat{f} \chi_{[-R, R]})$$

gives a bandlimited function which is close to  $f$  in the  $L^2$ -norm.