

# On inverse problems of Fourier synthesis

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# Outline

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- Introduction
- A reminder on ill-posed problems
- Fourier synthesis
- Asymptotic analysis
- A dual algorithm

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# Fourier Synthesis

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Recover a function from a partial and approximate knowledge of its Fourier transform.

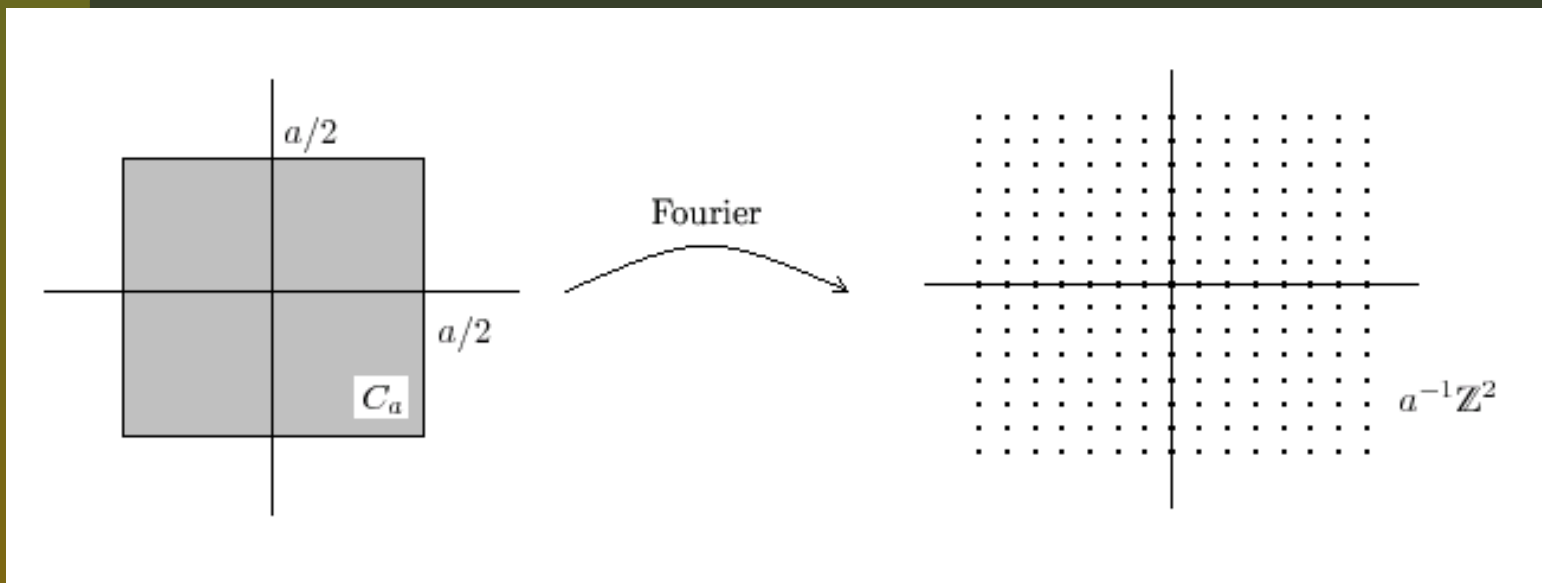
# Example 1: Fourier series

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$$f \in L^2(C_a) \quad \text{where} \quad C_a := [-a/2, a/2]^d$$

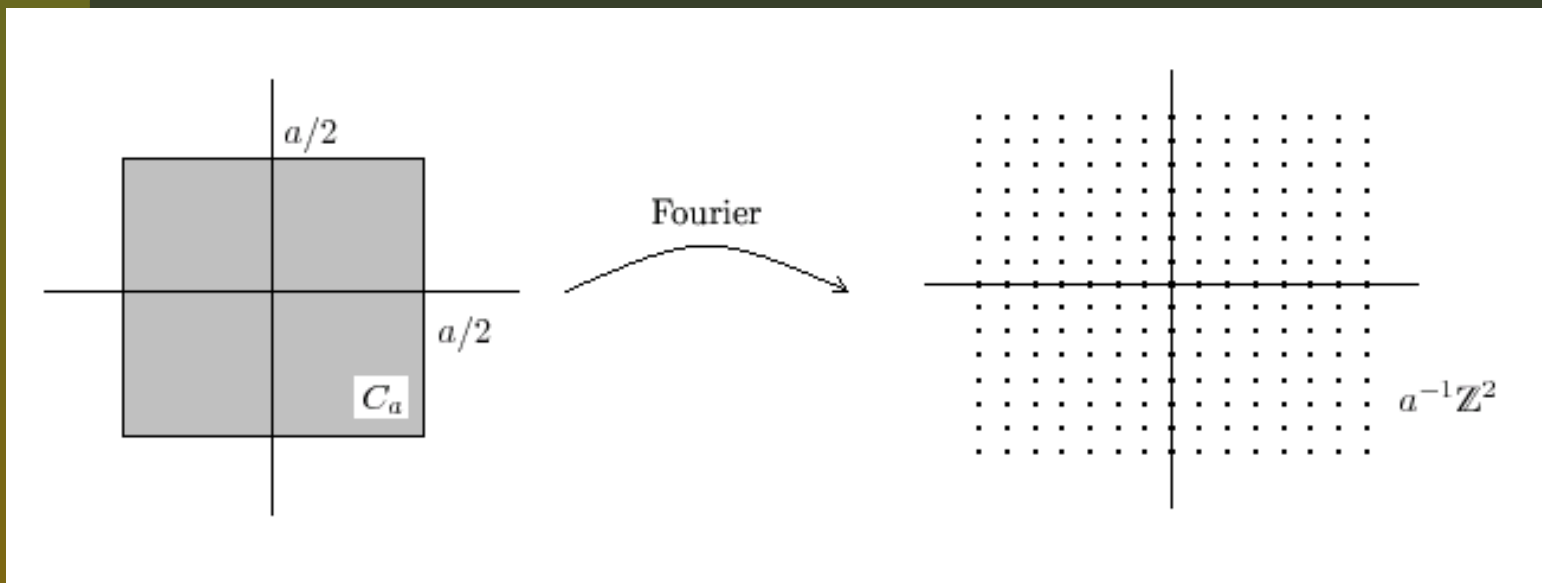
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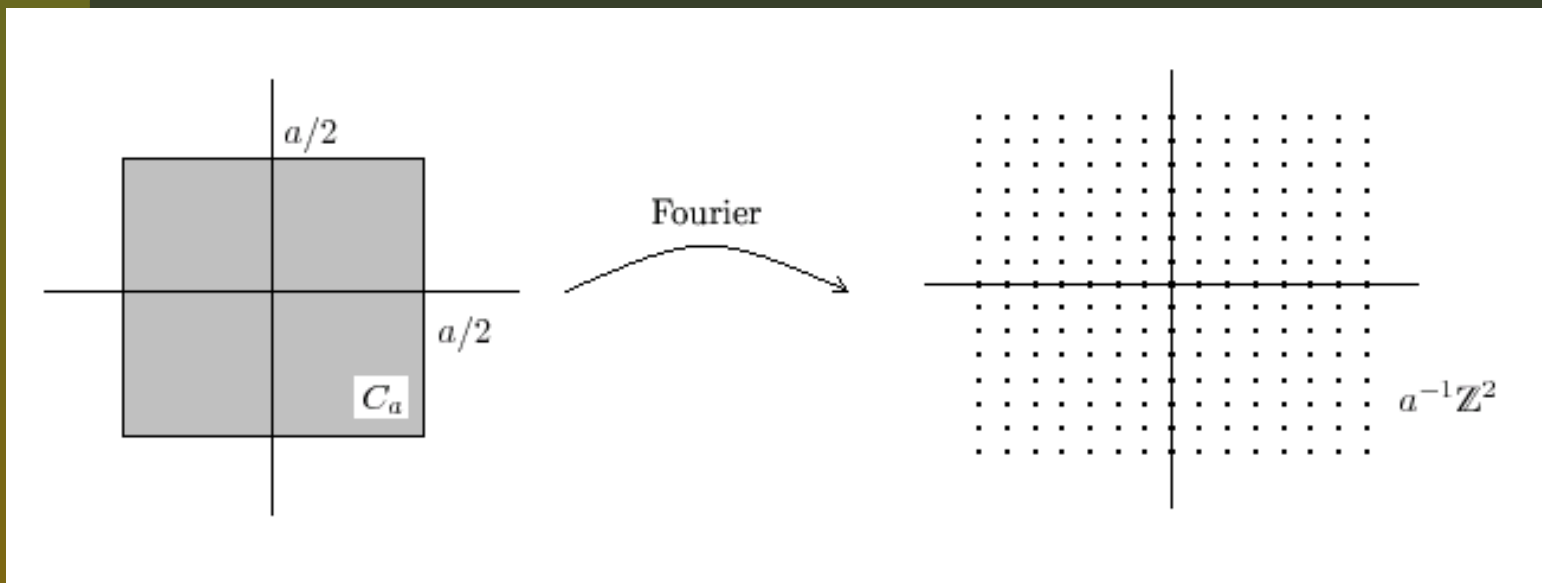
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$$f(\mathbf{x}) = \frac{1}{a^n} \sum_{\mathbf{k} \in \mathbb{Z}^n} \hat{f}\left(\frac{\mathbf{k}}{a}\right) \exp\left[2i\pi \left\langle \frac{\mathbf{k}}{a}, \mathbf{x} \right\rangle\right] \mathbf{1}_{C_a}(\mathbf{x})$$

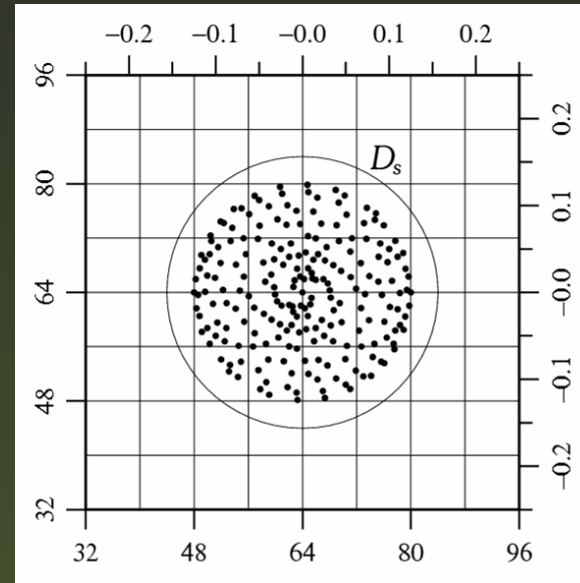
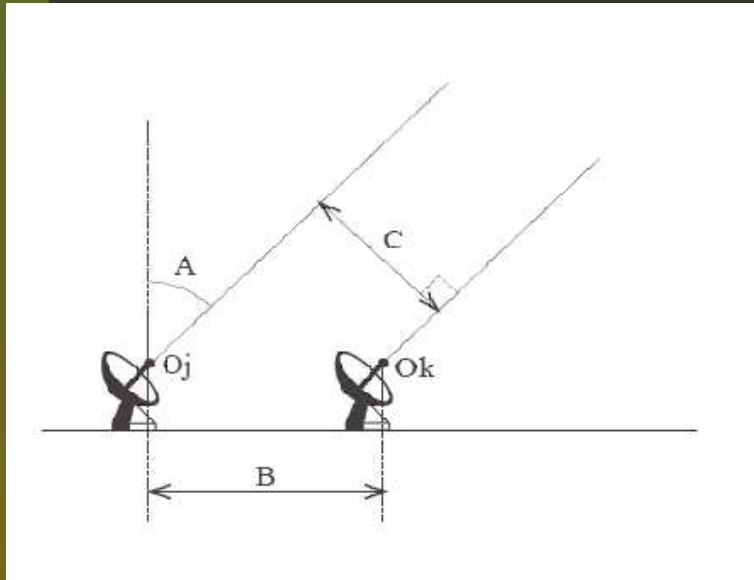
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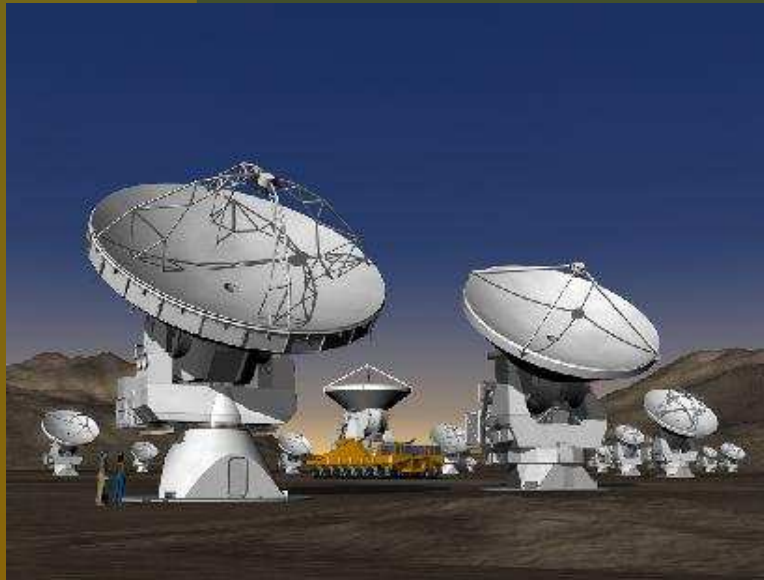
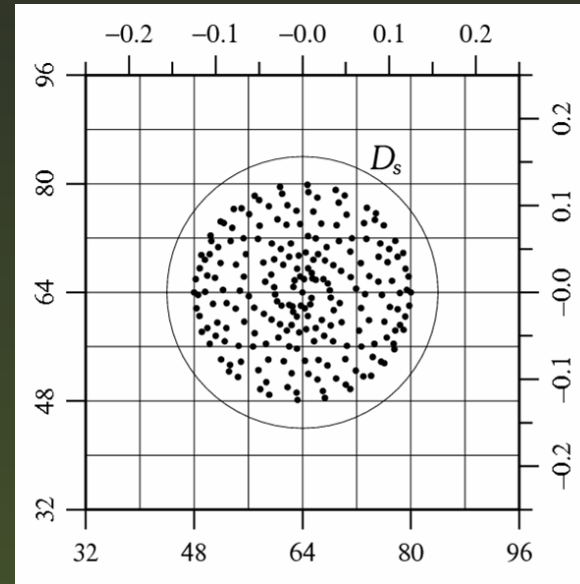
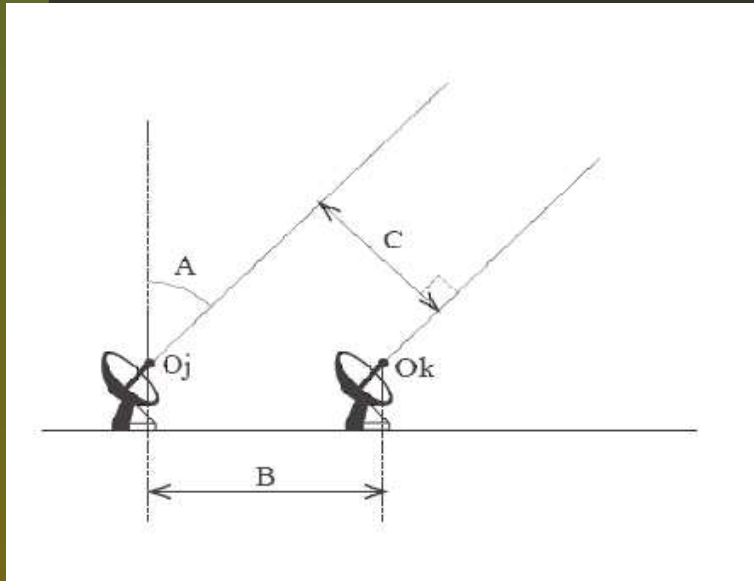


$$\hat{f}(\xi) = \sum_{\mathbf{k} \in \mathbb{Z}^n} \hat{f}\left(\frac{\mathbf{k}}{a}\right) \operatorname{sinc} \pi a \left(\xi - \frac{\mathbf{k}}{a}\right)$$

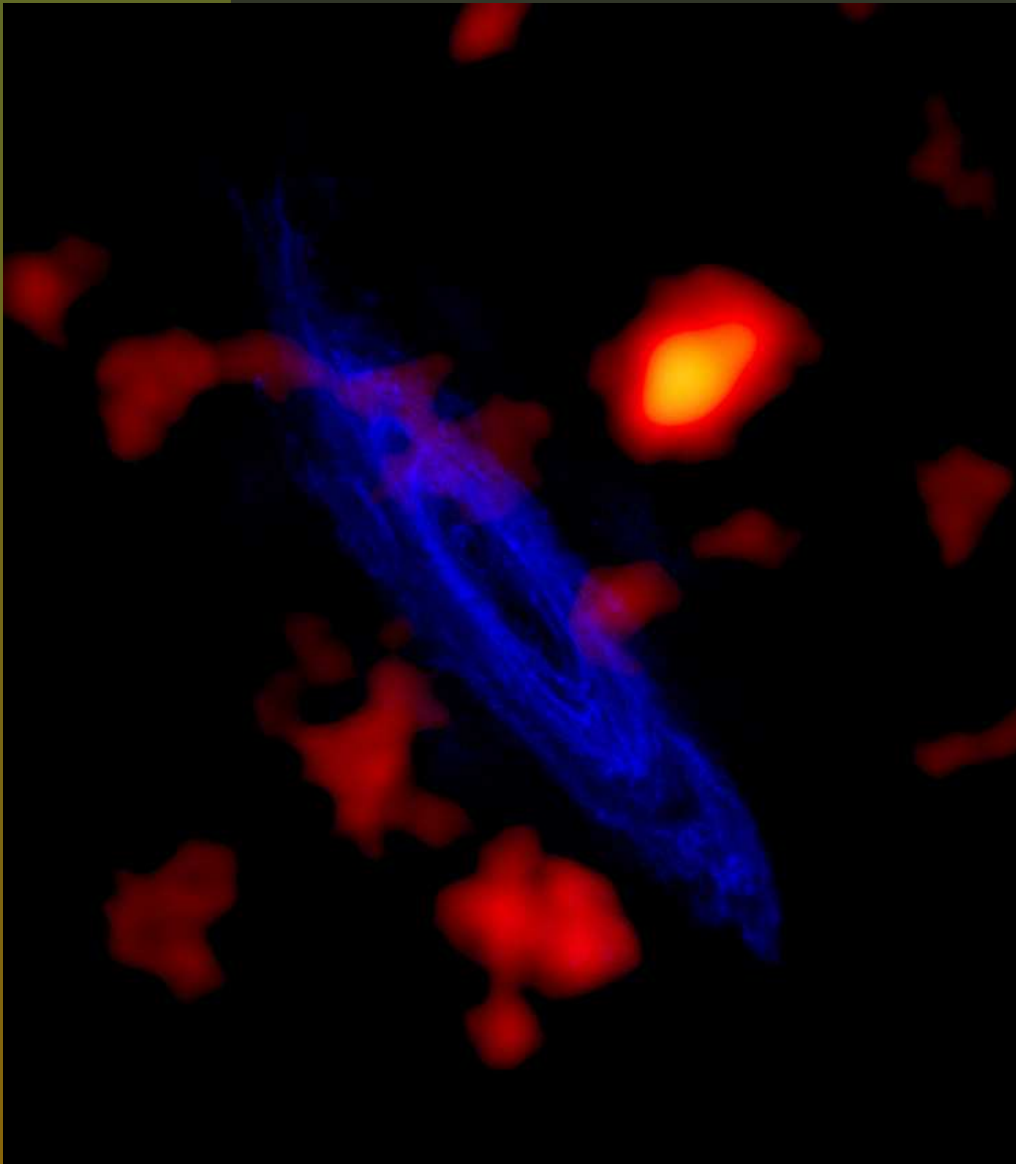
# Example 2: Aperture synthesis



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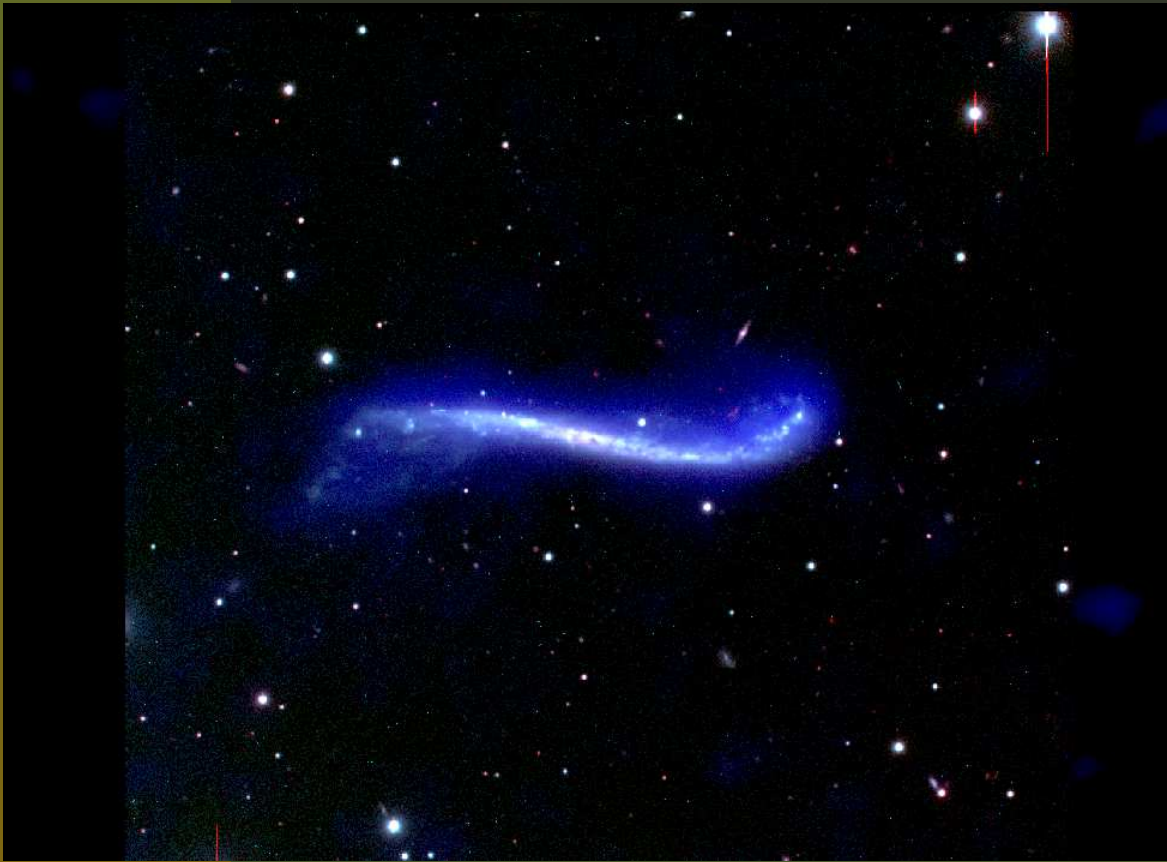
# Andromeda Galaxy



WSRT

*Courtesy of National Radio Astronomy Observatory / Associated Universities, Inc. / National Science Foundation*

# Integral Sign Galaxy ?



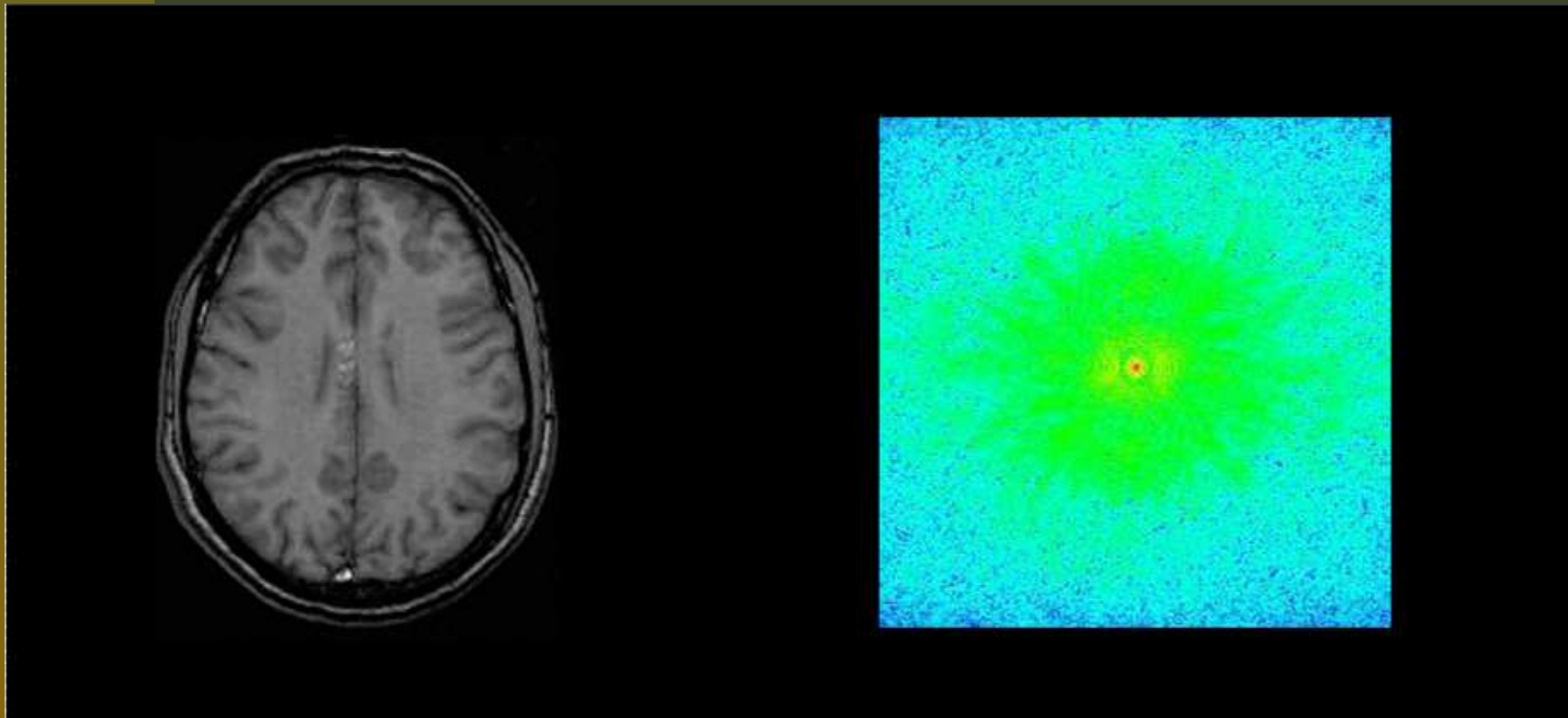
VLA

*Courtesy of National Radio Astronomy Observatory / Associated Universities, Inc. / National Science Foundation*

# Example 3: MRI

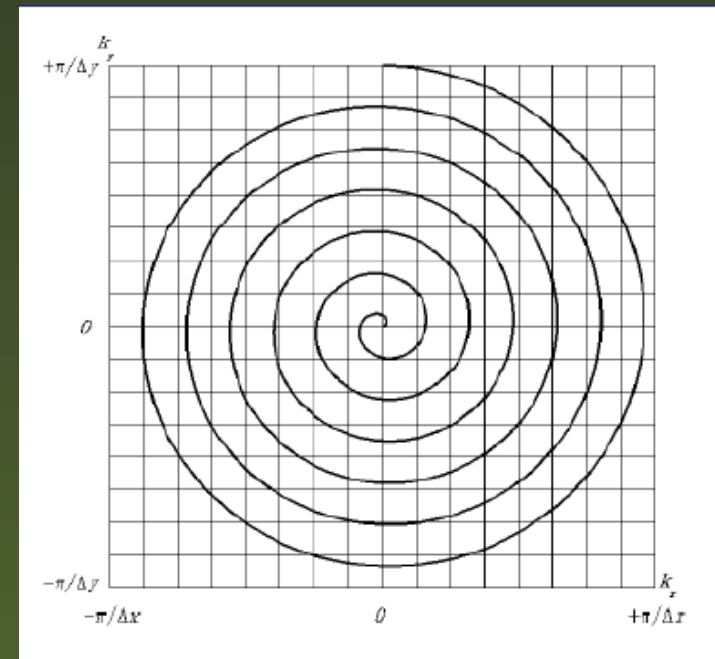
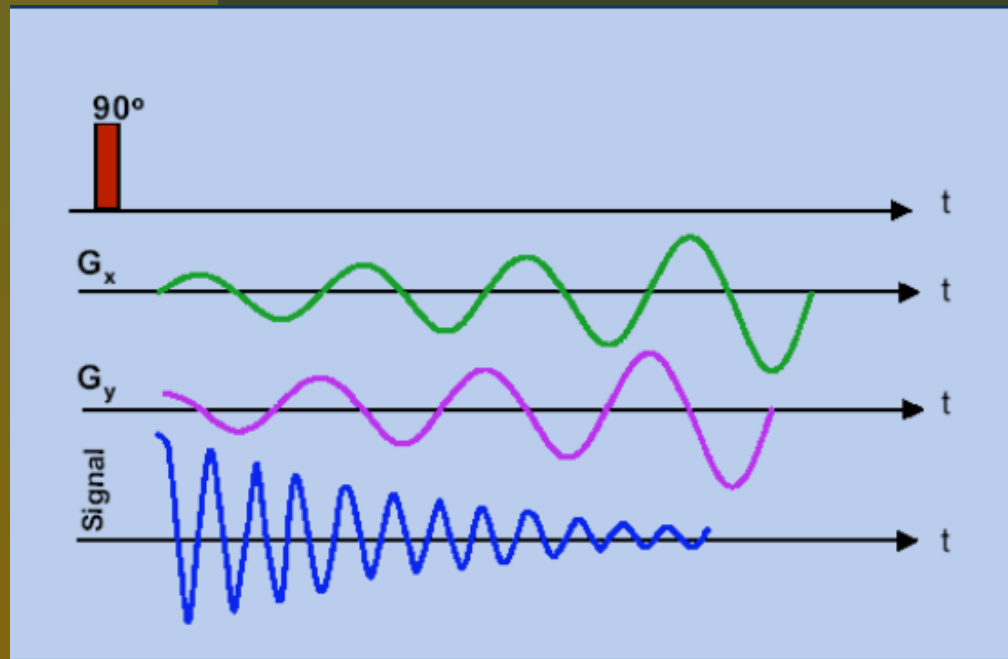
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Standard acquisitions:



# Example 3: MRI

Non-Cartesian and sparse acquisitions:



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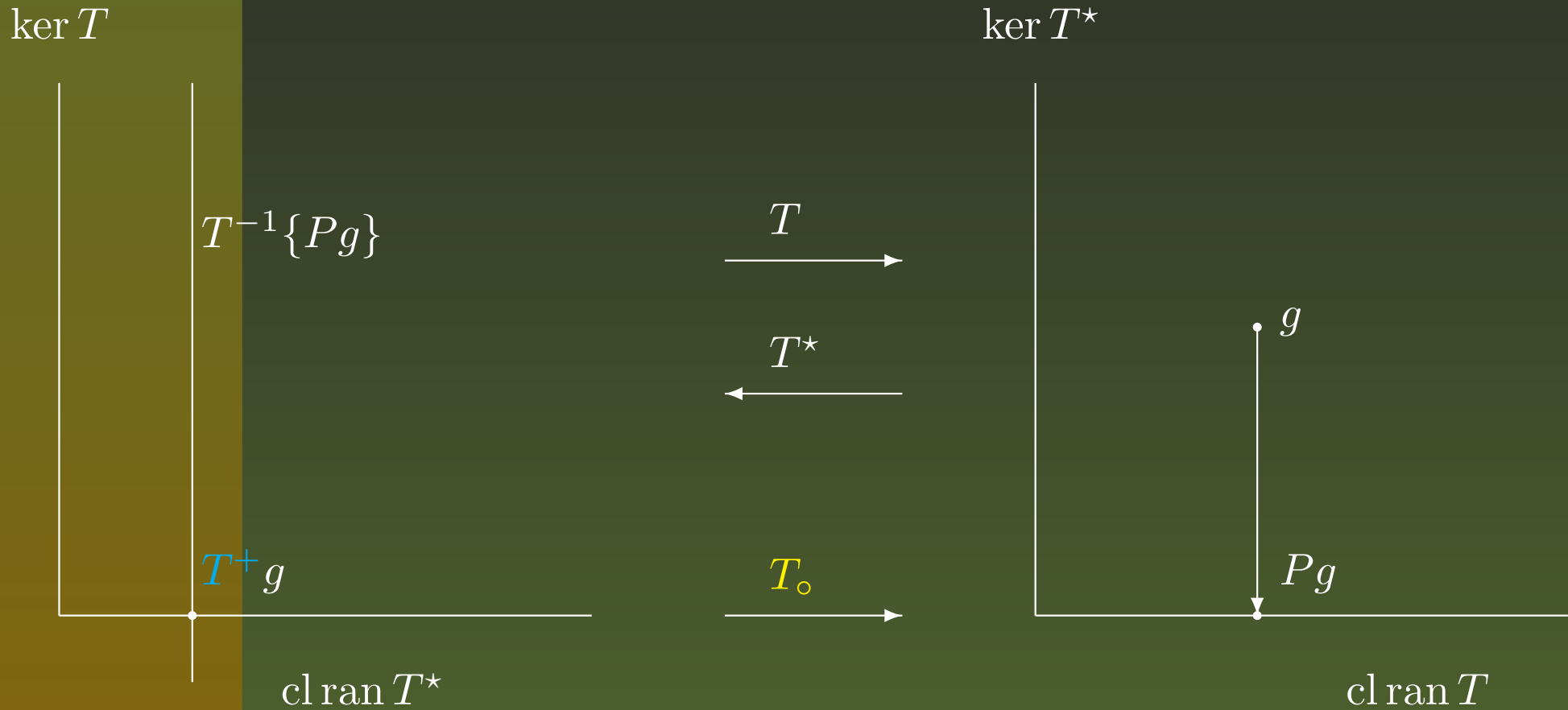
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- Well-posedness:  $\exists$  a unique solution which depends continuously on  $g$
- Well-posedness in the *LS sense*:  $T^+$  is continuous (with domain  $G$ )

# Pseudo-inverse



$T_\circ$ : restriction of  $T$  to  $(\ker T)^\perp = \text{cl ran } T^*$

$$T^+ = T_\circ^{-1} \circ P$$

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- (ii)  $T^{-1} : \text{ran } T \rightarrow F$  is unbounded, and so is  $T^+$
- (iii)  $\text{ran } T$  is not closed, so that

$$\mathcal{D}(T^+) = \text{ran } T + (\text{ran } T)^\perp \subsetneq G$$

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- (i) Clearly,  $T^*T$  is Hermitian, compact, positive and injective. Point (i) is then a particular instance of the spectral theorem for compact hermitian operators.

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(iii) If  $\text{ran } T$  were closed, it would be a Banach space on its own, and the Open Mapping Theorem would imply continuity of  $T^{-1}$ . This would contradict Point (ii). ■

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**Proposition** For all  $g \in \mathcal{D}(T^+)$ ,

$$T^+ g = \sum_{k \in \mathbb{N}^*} \frac{1}{\sqrt{\lambda_k}} \langle g, g_k \rangle f_k.$$

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**Theorem** The Tikhonov's solution  $f_\alpha$  converges strongly to  $f^+ := T^+g$  as  $\alpha \downarrow 0$ .

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*Truncated Fourier operator:*

$$\begin{aligned} T_W : L^2(V) &\longrightarrow L^2(W) \\ f &\longmapsto T_W f := \mathbf{1}_W \hat{f} = \mathbf{1}_W \mathbb{U} f. \end{aligned}$$

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$$(T_W f)(\xi) = \int_{\mathbb{R}^d} \underbrace{e^{-2i\pi\langle x, \xi \rangle} \mathbb{1}_V(x) \mathbb{1}_W(\xi)}_{\alpha(x, \xi) \in L^2(\mathbb{R}^d \times \mathbb{R}^d)} f(x) \, dx.$$

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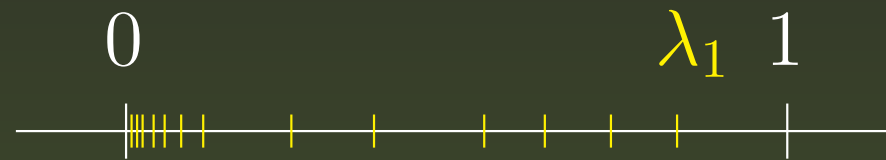
$\hookrightarrow T_W^+$  is unbounded and  $\mathcal{D}(T_W^+) \subsetneq L^2(W)$

$\mathcal{D}(T_W^+)$  is a dense subset of  $L^2(W)$

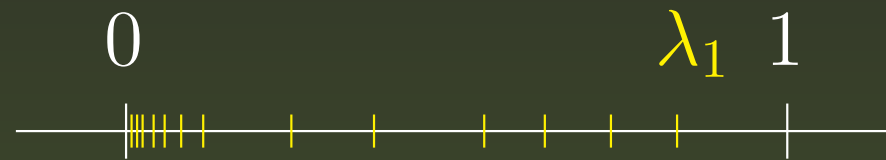
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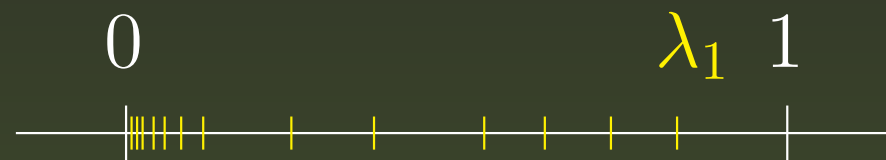


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**Proof**  $T_W^* T_W = \mathbb{1}_V U^{-1} \mathbb{1}_W U$

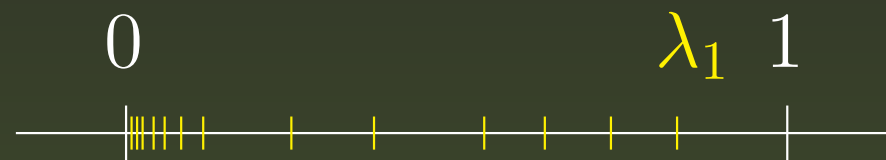
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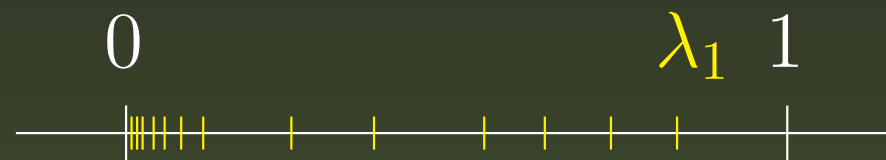
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$$\leq \int_{\mathbb{R}^d} |U^{-1} \mathbb{1}_W U f_1|^2$$

$$= \int_W |U f_1|^2 \quad (\leq 1)$$

$$< 1 \quad (\text{otherwise, } U f_1 \equiv 0 \text{ on } W^c) \quad \blacksquare$$

# Fourier interpolation

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**Proposition** Assume that  $\Omega \subseteq \mathbb{R}^d$  is such that  $\Omega^c$  is bounded. Then,

- (i)  $T_\Omega$  is bounded and injective;
- (ii)  $\text{ran } T_\Omega$  is closed;
- (iii)  $T_\Omega^{-1} : \text{ran } T_\Omega \rightarrow L^2(V)$  is bounded.

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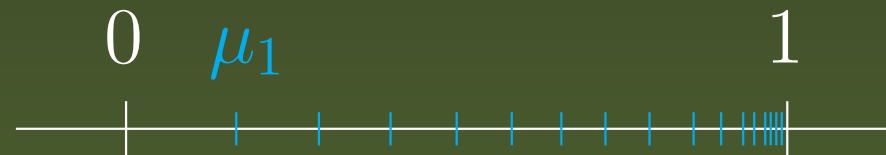


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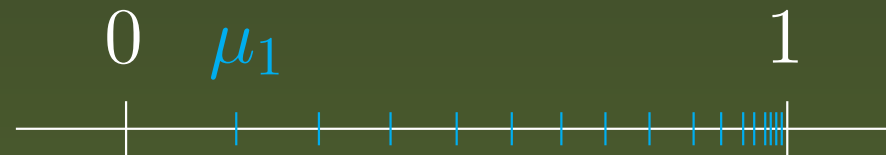
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and  $\text{ran } T_{\Omega}$  is closed ■

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$$\text{Minimize } \frac{1}{2} \|g - T_W f\|_{L^2(W)}^2 + \frac{\alpha}{2} \|\mathbb{1}_{W_\beta} U f\|_{L^2(W_\beta)}^2$$

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*New object to be reconstructed:  $\phi_\beta * f_0$*

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$$\phi_\beta := U^{-1} \mathbb{1}_{B_{1/\beta}}$$

# Regularization

Minimize  $\frac{1}{2} \|g - T_W f\|_{L^2(W)}^2 + \frac{\alpha}{2} \|\mathbb{1}_{W_\beta} U f\|_{L^2(W_\beta)}^2$

$W_\beta$ : complement of  $B_{1/\beta}$

New object to be reconstructed:  $\phi_\beta * f_0$

$$\phi_\beta := U^{-1} \mathbb{1}_{B_{1/\beta}}$$

$$\frac{1}{2} \|\mathbb{1}_{W_\beta} U f\|_{L^2(W_\beta)}^2$$

energy of  $f$  in the high frequency domain

# Apodized version

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$$\begin{array}{l} \text{Minimize} \quad \frac{1}{2} \left\| \hat{\phi}_\beta g - T_W f \right\|_{L^2(W)}^2 + \frac{\alpha}{2} \left\| (1 - \hat{\phi}_\beta) \hat{f} \right\|_{L^2(\mathbb{R}^d)}^2 \\ \text{s.t.} \quad f \in L^2(V) \end{array}$$

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$$\begin{array}{l} \text{Minimize} \\ \text{s.t.} \end{array} \quad \frac{1}{2} \left\| \hat{\phi}_\beta g - T_W f \right\|_{L^2(W)}^2 + \frac{\alpha}{2} \left\| (1 - \hat{\phi}_\beta) \hat{f} \right\|_{L^2(\mathbb{R}^d)}^2$$
$$f \in L^2(V)$$

*Regularized data:  $g_\beta := \hat{\phi}_\beta g$*

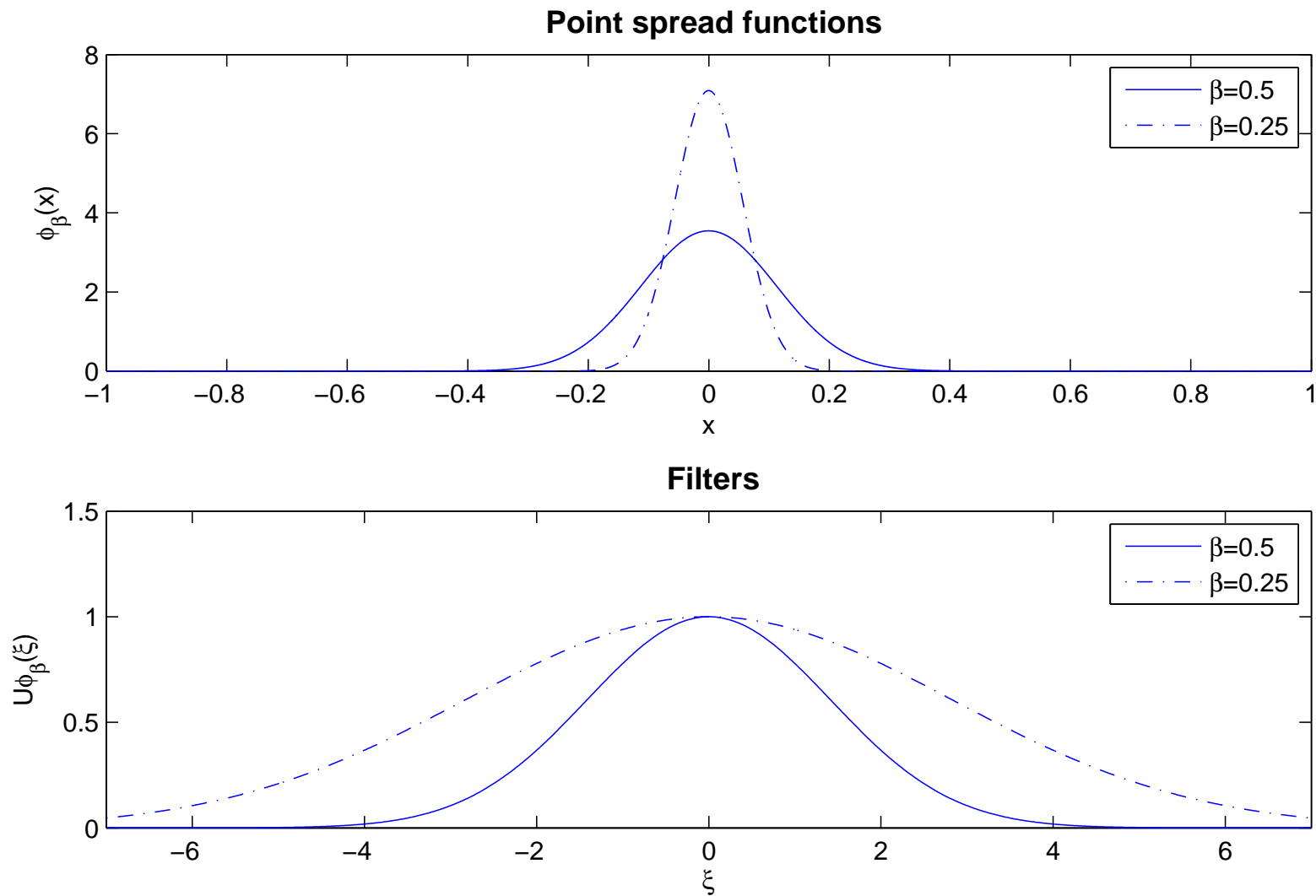
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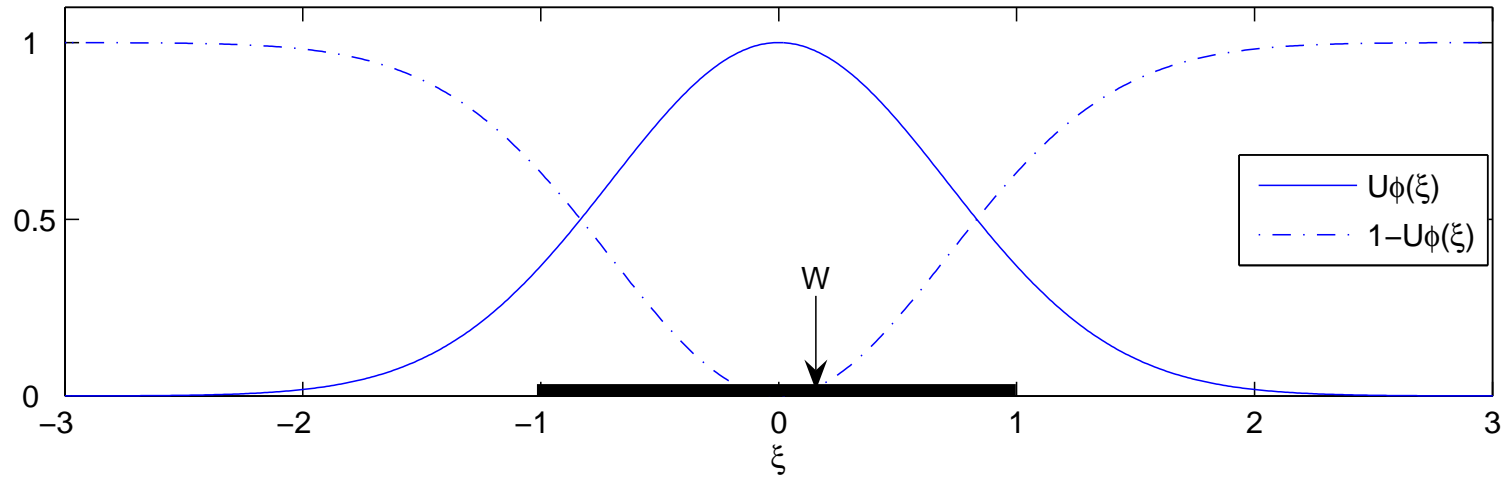
$$\phi_\beta(x) = \frac{1}{\beta^d} \phi\left(\frac{x}{\beta}\right) \quad \hat{\phi}_\beta(\xi) = \hat{\phi}(\beta\xi)$$

# Filters and associated PSF

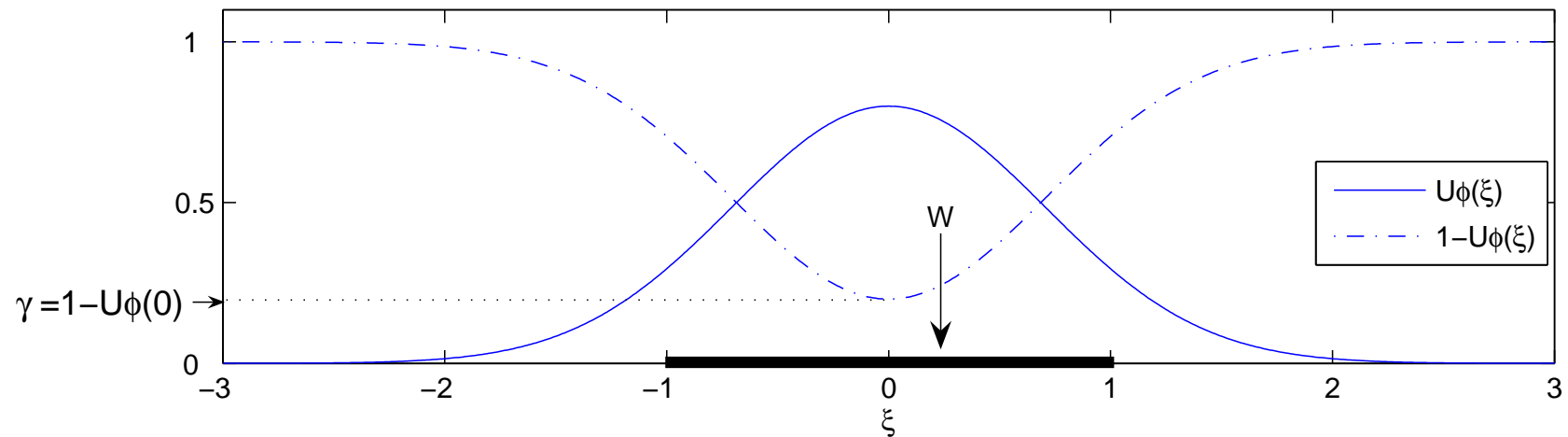


# Filters and their complement

Case  $U\phi(0)=1$



Case  $U\phi(0) \in (0,1)$



# General framework

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$$\left| \begin{array}{l} \text{Minimize } \frac{1}{2} \|g - T_W f\|^2 + \alpha \mathcal{H}(f) \\ \text{s.t. } f \in L^2(V) \end{array} \right.$$

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$$\mathcal{H}_\beta(f) = \frac{1}{2} \left\| (1 - \hat{\phi}_\beta) \hat{f} \right\|_{L^2(\mathbb{R}^d)}^2 = \frac{1}{2} \int_{\mathbb{R}^d} |1 - \hat{\phi}_\beta|^2 |\hat{f}|^2$$

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## Main issues

# General framework

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## Main issues

Well-posedness

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## Main issues

Well-posedness

Asymptotic behavior ( $\alpha$  and/or  $\beta$  tend to zero)

# Well-posedness

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$$\begin{array}{l} (\mathcal{P}_{\alpha,\beta}) \quad \left| \begin{array}{l} \text{Minimize} \quad \frac{1}{2} \left\| \hat{\phi}_\beta g - T_W f \right\|^2 + \frac{\alpha}{2} \left\| (1 - \hat{\phi}_\beta) \hat{f} \right\|^2 \\ \text{s.t.} \quad f \in L^2(V) \end{array} \right. \end{array}$$

# Well-posedness

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**Definition**  $\langle f_1, f_2 \rangle_\beta := \int_{\mathbb{R}^d} |1 - \hat{\phi}_\beta|^2 U f_1 \overline{U f_2}$

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$(\mathcal{P}_{\alpha, \beta})$   $\left\{ \begin{array}{l} \text{Minimize} \quad \frac{1}{2} \left\| \hat{\phi}_\beta g - T_W f \right\|^2 + \frac{\alpha}{2} \left\| (1 - \hat{\phi}_\beta) \hat{f} \right\|^2 \\ \text{s.t.} \quad f \in L^2(V) \end{array} \right.$

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**Lemma**  $\langle \cdot, \cdot \rangle_\beta$  is an inner product which turns  $L^2(V)$  into a Hilbert space. The corresponding norm  $\| \cdot \|_\beta$  is equivalent to  $\| \cdot \|_{L^2(V)}$ .

$$\begin{array}{l} (\mathcal{P}_{\alpha, \beta}) \quad \left| \begin{array}{l} \text{Minimize} \quad \frac{1}{2} \left\| \hat{\phi}_\beta g - T_W f \right\|^2 + \frac{\alpha}{2} \left\| (1 - \hat{\phi}_\beta) \hat{f} \right\|^2 \\ \text{s.t.} \quad f \in L^2(V) \end{array} \right. \end{array}$$

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**Proposition** Let  $\alpha, \beta > 0$  be fixed. Then  $(\mathcal{P}_{\alpha, \beta})$  has a unique solution  $f_{\alpha, \beta}$ , which depends continuously on  $g \in L^2(W)$ .

$$(\mathcal{P}_{\alpha, \beta}) \quad \left| \begin{array}{l} \text{Minimize} \quad \frac{1}{2} \left\| \hat{\phi}_\beta g - T_W f \right\|^2 + \frac{\alpha}{2} \left\| (1 - \hat{\phi}_\beta) \hat{f} \right\|^2 \\ \text{s.t.} \quad f \in L^2(V) \end{array} \right.$$

# Proof

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$(\mathcal{P}_{\alpha,\beta})$  can be rewritten as:

$$\begin{array}{l} \text{Minimize} \quad \frac{1}{2} \left\| \hat{\phi}_\beta g - T_W f \right\|_{L^2(W)}^2 + \frac{\alpha}{2} \|f\|_\beta^2 \\ \text{s.t.} \quad f \in L^2(V) \end{array}$$

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$$f_{\alpha,\beta} = (T_W^\# T_W + \alpha I)^{-1} T_W^\# (\hat{\phi}_\beta g)$$

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The conclusion then follows from Tikhonov's theory and the continuity of the multiplication  $g \mapsto \hat{\phi}_\beta g$ . ■

# Tikhonov-like regularization $(\alpha \downarrow 0)$

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Unique solution:  $T_W^+(\hat{\phi}_\beta g)$

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Unique solution:  $T_W^+(\hat{\phi}_\beta g)$

**Theorem** Let  $\beta > 0$  be fixed and let  $g \in \mathcal{D}(T_W^+)$ .

- (i) If  $\hat{\phi}_\beta g \in \mathcal{D}(T_W^+)$ , then  $f_{\alpha,\beta} \rightarrow T_W^+(\hat{\phi}_\beta g)$  as  $\alpha \downarrow 0$ .
- (ii) If  $\hat{\phi}_\beta g \notin \mathcal{D}(T_W^+)$ , then  $\|f_{\alpha,\beta}\|_{L^2(V)} \rightarrow \infty$  as  $\alpha \downarrow 0$ .

# Tikhonov-like regularization $(\alpha \downarrow 0)$

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Unique solution:  $T_W^+(\hat{\phi}_\beta g)$

**Proposition** Assume that  $\phi \in L^1(\mathbb{R}^d)$  is such that  $\hat{\phi}$  is analytic, and let  $\beta > 0$  be fixed and  $g \in \mathcal{D}(T_W^+)$ . Then, the following are equivalent:

- (i)  $\hat{\phi}_\beta g \in \mathcal{D}(T_W^+)$ ;
- (ii)  $\text{supp}(\phi_\beta * T_W^+ g) \subseteq V$ .

# Outline

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- Introduction
- A reminder on ill-posed problems
- Fourier synthesis
- Asymptotic analysis (joint work with N. Alibaud and Y. Saesor)
- A dual algorithm

# Main Result

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- $\forall \xi \in \mathbb{R}^d \setminus \{0\}, \hat{\phi}(\xi) \neq 1$

If  $g \in T_W(L^2(V) \cap H^s(\mathbb{R}^d))$ , then  $f_{\alpha,\beta} \rightarrow T_W^+ g$  strongly as  $\beta \downarrow 0$ .

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$$\left. \begin{array}{l} (f_n) \text{ bounded} \\ \lim_{R \rightarrow \infty} \sup_n \int_{\|x\| > R} |f_n(x)|^2 dx = 0 \\ \sup_n \|\mathcal{T}_h f_n - f_n\| \rightarrow 0 \text{ as } \|h\| \rightarrow 0 \end{array} \right\} \Rightarrow (f_n) \text{ precompact}$$

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$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

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$\tilde{g} := UT_W^+ g$  (analytic extension of  $g$ )

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$$g \in T_W(L^2(V) \cap H^s(\mathbb{R}^d)) \implies \int_{\mathbb{R}^d} \|\xi\|^{2s} |\tilde{g}(\xi)|^2 d\xi < \infty$$

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$$\begin{aligned} & \left\| \hat{\phi}_\beta g - T_W f_{\alpha,\beta} \right\|^2 + \alpha \left\| (1 - \hat{\phi}_\beta) U f_{\alpha,\beta} \right\|^2 \\ & \leq \left\| \hat{\phi}_\beta g - T_W f \right\|^2 + \alpha \left\| (1 - \hat{\phi}_\beta) U f \right\|^2 \end{aligned}$$

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$$T_W T_W^+ g = g \quad \text{and} \quad U T_W^+ g = \tilde{g}$$

$$\left\| \hat{\phi}_\beta g - T_W f_{\alpha,\beta} \right\|^2 \geq 0$$

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$$\left\| (1 - \hat{\phi}_\beta) U f_{\alpha,\beta} \right\|^2 \leq \frac{1}{\alpha} \left\| \hat{\phi}_\beta g - g \right\|^2 + \left\| (1 - \hat{\phi}_\beta) \tilde{g} \right\|^2$$

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$$\left\| \hat{\phi}_\beta g - g \right\|^2 = \left\| (1 - \hat{\phi}_\beta) g \right\|^2 \leq \left\| (1 - \hat{\phi}_\beta) \tilde{g} \right\|^2$$

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$$\left\| (1 - \hat{\phi}_\beta) \hat{f}_{\alpha,\beta} \right\|^2 \leq \frac{1 + \alpha}{\alpha} \left\| (1 - \hat{\phi}_\beta) \tilde{g} \right\|^2$$

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$$\hat{\phi}_\beta(\xi) = \hat{\phi}(\beta\xi)$$

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$$\left\| (1 - \hat{\phi}_\beta) \hat{f}_{\alpha,\beta} \right\|^2 \leq \frac{1 + \alpha}{\alpha} \left\| (1 - \hat{\phi}_\beta) \tilde{g} \right\|^2$$

$$\int_{\mathbb{R}^d} \left| 1 - \hat{\phi}(\beta\xi) \right|^2 \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \leq \frac{1 + \alpha}{\alpha} \int_{\mathbb{R}^d} \left| 1 - \hat{\phi}(\beta\xi) \right|^2 \left| \tilde{g}(\xi) \right|^2 d\xi$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\left\| (1 - \hat{\phi}_\beta) \hat{f}_{\alpha,\beta} \right\|^2 \leq \frac{1 + \alpha}{\alpha} \left\| (1 - \hat{\phi}_\beta) \tilde{g} \right\|^2$$

$$\int_{\mathbb{R}^d} \left| 1 - \hat{\phi}(\beta\xi) \right|^2 \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \leq \frac{1 + \alpha}{\alpha} \int_{\mathbb{R}^d} \left| 1 - \hat{\phi}(\beta\xi) \right|^2 \left| \tilde{g}(\xi) \right|^2 d\xi$$

$$\begin{aligned} m_\beta \int_{\mathbb{R}^d} \frac{\left| 1 - \hat{\phi}(\beta\xi) \right|^2}{\left| 1 - \hat{\phi}(\beta\xi/\|\xi\|) \right|^2} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \\ \leq \frac{1 + \alpha}{\alpha} M_\beta \int_{\mathbb{R}^d} \frac{\left| 1 - \hat{\phi}(\beta\xi) \right|^2}{\left| 1 - \hat{\phi}(\beta\xi/\|\xi\|) \right|^2} \left| \tilde{g}(\xi) \right|^2 d\xi \end{aligned}$$

$$m_\beta := \min_{\|\xi\|=1} \left| 1 - \hat{\phi}(\beta\xi) \right|^2 > 0 \quad \text{and} \quad M_\beta := \max_{\|\xi\|=1} \left| 1 - \hat{\phi}(\beta\xi) \right|^2 > 0$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\left\| (1 - \hat{\phi}_\beta) \hat{f}_{\alpha,\beta} \right\|^2 \leq \frac{1 + \alpha}{\alpha} \left\| (1 - \hat{\phi}_\beta) \tilde{g} \right\|^2$$

$$\int_{\mathbb{R}^d} \left| 1 - \hat{\phi}(\beta\xi) \right|^2 \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \leq \frac{1 + \alpha}{\alpha} \int_{\mathbb{R}^d} \left| 1 - \hat{\phi}(\beta\xi) \right|^2 \left| \tilde{g}(\xi) \right|^2 d\xi$$

$$\int_{\mathbb{R}^d} \frac{\left| 1 - \hat{\phi}(\beta\xi) \right|^2}{\left| 1 - \hat{\phi}(\beta\xi / \|\xi\|) \right|^2} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi$$

$$\leq \frac{1 + \alpha}{\alpha} \frac{M_\beta}{m_\beta} \int_{\mathbb{R}^d} \frac{\left| 1 - \hat{\phi}(\beta\xi) \right|^2}{\left| 1 - \hat{\phi}(\beta\xi / \|\xi\|) \right|^2} \left| \tilde{g}(\xi) \right|^2 d\xi$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\begin{aligned} \int_{\mathbb{R}^d} \frac{|1 - \hat{\phi}(\beta\xi)|^2}{|1 - \hat{\phi}(\beta\xi/\|\xi\|)|^2} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \\ \leq \frac{1 + \alpha M_\beta}{\alpha m_\beta} \int_{\mathbb{R}^d} \frac{|1 - \hat{\phi}(\beta\xi)|^2}{|1 - \hat{\phi}(\beta\xi/\|\xi\|)|^2} \left| \tilde{g}(\xi) \right|^2 d\xi \end{aligned}$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\int_{\mathbb{R}^d} \frac{|1 - \hat{\phi}(\beta\xi)|^2}{|1 - \hat{\phi}(\beta\xi/\|\xi\|)|^2} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi$$
$$\leq \frac{1 + \alpha M_\beta}{\alpha m_\beta} \int_{\mathbb{R}^d} \frac{|1 - \hat{\phi}(\beta\xi)|^2}{|1 - \hat{\phi}(\beta\xi/\|\xi\|)|^2} |\tilde{g}(\xi)|^2 d\xi$$

**Lemma**  $\exists \nu_0, C_0 > 0$  such that  $\forall \beta \in (0, 1], \forall \xi \in \mathbb{R}^d \setminus \{0\}$ ,

$$\nu_0 \left( \|\xi\|^{2s} \mathbf{1}_{B_{1/\beta}}(\xi) + \frac{1}{M_\beta} \mathbf{1}_{B_{1/\beta}^c}(\xi) \right) \leq \frac{|1 - \hat{\phi}(\beta\xi)|^2}{|1 - \hat{\phi}(\beta\xi/|\xi|)|^2} \leq C_0 \|\xi\|^{2s}$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\int_{\mathbb{R}^d} \frac{|1 - \hat{\phi}(\beta\xi)|^2}{|1 - \hat{\phi}(\beta\xi/\|\xi\|)|^2} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi$$
$$\leq \frac{1 + \alpha M_\beta}{\alpha m_\beta} \int_{\mathbb{R}^d} \frac{|1 - \hat{\phi}(\beta\xi)|^2}{|1 - \hat{\phi}(\beta\xi/\|\xi\|)|^2} |\tilde{g}(\xi)|^2 d\xi$$

**Lemma**  $\exists \nu_0, C_0 > 0$  such that  $\forall \beta \in (0, 1], \forall \xi \in \mathbb{R}^d \setminus \{0\}$ ,

$$\nu_0 \left( \|\xi\|^{2s} \mathbf{1}_{B_{1/\beta}}(\xi) + \frac{1}{M_\beta} \mathbf{1}_{B_{1/\beta}^c}(\xi) \right) \leq \frac{|1 - \hat{\phi}(\beta\xi)|^2}{|1 - \hat{\phi}(\beta\xi/|\xi|)|^2} \leq C_0 \|\xi\|^{2s}$$

$$\nu_0 \int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi + \frac{\nu_0}{M_\beta} \int_{\|\xi\| > 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi$$
$$\leq \frac{1 + \alpha M_\beta}{\alpha m_\beta} C_0 \int_{\mathbb{R}^d} \|\xi\|^{2s} |\tilde{g}(\xi)|^2 d\xi$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\begin{aligned} \nu_0 \int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi + \frac{\nu_0}{M_\beta} \int_{\|\xi\| > 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \\ \leq \frac{1 + \alpha M_\beta}{\alpha m_\beta} C_0 \int_{\mathbb{R}^d} \|\xi\|^{2s} \left| \tilde{g}(\xi) \right|^2 d\xi \end{aligned}$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\begin{aligned} & \nu_0 \int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi + \frac{\nu_0}{M_\beta} \int_{\|\xi\| > 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \\ & \leq \frac{1 + \alpha}{\alpha} \frac{M_\beta}{m_\beta} C_0 \underbrace{\int_{\mathbb{R}^d} \|\xi\|^{2s} \left| \tilde{g}(\xi) \right|^2 d\xi}_{\leq C_1} \end{aligned}$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\begin{aligned} \nu_0 \int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi + \frac{\nu_0}{M_\beta} \int_{\|\xi\| > 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \\ \leq \frac{1 + \alpha M_\beta}{\alpha m_\beta} C_0 \underbrace{\int_{\mathbb{R}^d} \|\xi\|^{2s} \left| \tilde{g}(\xi) \right|^2 d\xi}_{\leq C_1} \end{aligned}$$

$$\left. \begin{array}{l} \frac{M_\beta}{m_\beta} \rightarrow 1 \quad \text{as } \beta \rightarrow 0 \text{ and } \beta \rightarrow \infty \\ \frac{M_\beta}{m_\beta} \text{ is continuous} \end{array} \right\} \implies \frac{M_\beta}{m_\beta} \text{ is bounded}$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\begin{aligned} \nu_0 \int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi + \frac{\nu_0}{M_\beta} \int_{\|\xi\| > 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \\ \leq \frac{1 + \alpha}{\alpha} \frac{M_\beta}{m_\beta} C_0 \underbrace{\int_{\mathbb{R}^d} \|\xi\|^{2s} \left| \tilde{g}(\xi) \right|^2 d\xi}_{\leq C_1} \end{aligned}$$

$$\left. \begin{array}{l} \frac{M_\beta}{m_\beta} \rightarrow 1 \quad \text{as } \beta \rightarrow 0 \text{ and } \beta \rightarrow \infty \\ \frac{M_\beta}{m_\beta} \text{ is continuous} \end{array} \right\} \implies \frac{M_\beta}{m_\beta} \text{ is bounded}$$

$$\begin{aligned} \nu_0 \int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi + \frac{\nu_0}{M_\beta} \int_{\|\xi\| > 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \\ \leq \frac{1 + \alpha}{\alpha} C_2 C_0 C_1 = K \end{aligned}$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi + \frac{1}{M_\beta} \int_{\|\xi\| > 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \leq \frac{K}{\nu_0}$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\underbrace{\int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi + \frac{1}{M_\beta} \int_{\|\xi\| > 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi}_{\geq} \leq \frac{K}{\nu_0}$$
$$\geq \int_{1 \leq \|\xi\| \leq 1/\beta} \|\xi\|^{2s} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\underbrace{\int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi + \frac{1}{M_\beta} \int_{\|\xi\| > 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi}_{\geq \int_{1 \leq \|\xi\| \leq 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi} \leq \frac{K}{\nu_0}$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\underbrace{\int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi}_{\geq \int_{1 \leq \|\xi\| \leq 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi} + \frac{1}{M_\beta} \int_{\|\xi\| > 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \leq \frac{K}{\nu_0}$$
$$M_\beta := \max_{\|\xi\|=1} \left| 1 - \hat{\phi}(\beta\xi) \right|^2$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\underbrace{\int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi + \frac{1}{M_\beta} \int_{\|\xi\| > 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi}_{\geq \int_{1 \leq \|\xi\| \leq 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi} \leq \frac{K}{\nu_0}$$
$$M_\beta := \max_{\|\xi\|=1} \left| 1 - \hat{\phi}(\beta\xi) \right|^2 \leq \left( 1 + \max_{\|\xi\|=1} \left| \hat{\phi}(\beta\xi) \right| \right)^2$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\underbrace{\int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi + \frac{1}{M_\beta} \int_{\|\xi\| > 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi}_{\geq \int_{1 \leq \|\xi\| \leq 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi} \leq \frac{K}{\nu_0}$$
$$M_\beta := \max_{\|\xi\|=1} \left| 1 - \hat{\phi}(\beta\xi) \right|^2$$
$$\leq \left( 1 + \max_{\|\xi\|=1} \left| \hat{\phi}(\beta\xi) \right| \right)^2$$
$$\leq (1 + \|\phi\|_1)^2$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\begin{aligned} & \underbrace{\int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi + \frac{1}{M_\beta} \int_{\|\xi\| > 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi}_{\geq} \leq \frac{K}{\nu_0} \\ & \geq \int_{1 \leq \|\xi\| \leq 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \quad M_\beta := \max_{\|\xi\|=1} \left| 1 - \hat{\phi}(\beta\xi) \right|^2 \\ & \leq \left( 1 + \max_{\|\xi\|=1} \left| \hat{\phi}(\beta\xi) \right| \right)^2 \\ & \leq (1 + \|\phi\|_1)^2 \\ & \int_{1 \leq \|\xi\| \leq 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \\ & \quad + \frac{1}{(1 + \|\phi\|_1)^2} \int_{\|\xi\| > 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \leq \frac{K}{\nu_0} \end{aligned}$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\underbrace{\int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi + \frac{1}{M_\beta} \int_{\|\xi\| > 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi} \leq \frac{K}{\nu_0}$$

$$\begin{aligned} &\geq \int_{1 \leq \|\xi\| \leq 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi & M_\beta &:= \max_{\|\xi\|=1} \left| 1 - \hat{\phi}(\beta\xi) \right|^2 \\ & & &\leq \left( 1 + \max_{\|\xi\|=1} \left| \hat{\phi}(\beta\xi) \right| \right)^2 \\ & & &\leq (1 + \|\phi\|_1)^2 \end{aligned}$$

$$\frac{1}{(1 + \|\phi\|_1)^2} \int_{1 \leq \|\xi\| \leq 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi$$

$$+ \frac{1}{(1 + \|\phi\|_1)^2} \int_{\|\xi\| > 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \leq \frac{K}{\nu_0}$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\underbrace{\int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi + \frac{1}{M_\beta} \int_{\|\xi\| > 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi}_{\geq \int_{1 \leq \|\xi\| \leq 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi} \leq \frac{K}{\nu_0}$$
$$M_\beta := \max_{\|\xi\|=1} \left| 1 - \hat{\phi}(\beta\xi) \right|^2$$
$$\leq \left( 1 + \max_{\|\xi\|=1} \left| \hat{\phi}(\beta\xi) \right| \right)^2$$
$$\leq (1 + \|\phi\|_1)^2$$

$$\int_{\|\xi\| \geq 1} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \leq \frac{K}{\nu_0} (1 + \|\phi\|_1)^2 =: K_0$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\int_{\|\xi\| \geq 1} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \leq K_0$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\int_{\|\xi\| \geq 1} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \leq K_0$$

$$\int_{\mathbb{R}^d} \left| \mathbf{1}_{B_1^c} \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \leq K_0$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\int_{\|\xi\| \geq 1} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \leq K_0$$

$$\int_{\mathbb{R}^d} \left| \mathbb{1}_{B_1^c} \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \leq K_0$$

$$\int_{\mathbb{R}^d} \left| T_{B_1^c} f_{\alpha,\beta}(\xi) \right|^2 d\xi \leq K_0$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\int_{\|\xi\| \geq 1} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \leq K_0$$

$$\int_{\mathbb{R}^d} \left| \mathbf{1}_{B_1^c} \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \leq K_0$$

$$\int_{\mathbb{R}^d} \left| T_{B_1^c} f_{\alpha,\beta}(\xi) \right|^2 d\xi \leq K_0$$

$$\|f_{\alpha,\beta}\|^2 = \left\| T_{B_1^c}^{-1} T_{B_1^c} f_{\alpha,\beta} \right\|^2 \leq \left\| T_{B_1^c}^{-1} \right\|^2 \left\| T_{B_1^c} f_{\alpha,\beta} \right\|^2$$

# Step 1

$(f_{\alpha,\beta})_{\beta \in (0,1]}$  is bounded

$$\int_{\|\xi\| \geq 1} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \leq K_0$$

$$\int_{\mathbb{R}^d} \left| \mathbf{1}_{B_1^c} \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \leq K_0$$

$$\int_{\mathbb{R}^d} \left| T_{B_1^c} f_{\alpha,\beta}(\xi) \right|^2 d\xi \leq K_0$$

$$\|f_{\alpha,\beta}\|^2 = \left\| T_{B_1^c}^{-1} T_{B_1^c} f_{\alpha,\beta} \right\|^2 \leq \left\| T_{B_1^c}^{-1} \right\|^2 \left\| T_{B_1^c} f_{\alpha,\beta} \right\|^2$$

$$\|f_{\alpha,\beta}\|^2 \leq \left\| T_{B_1^c}^{-1} \right\|^2 K_0$$

# Overview of the proof

Step 1:  $(f_{\alpha,\beta})_{\beta \in (0,1]}$  is **bounded**

Step 2:  $(f_{\alpha,\beta})_{\beta \in (0,1]}$  converges weakly to  $T_W^+ g$

$$\beta_n \downarrow 0, f_n := f_{\alpha,\beta_n}$$

■  $\exists (f_{n_k}) \rightharpoonup T_W^+ g$

Step 3: The convergence is in fact strong

■  $(f_n)$  is bounded (Step 1)

■  $V$  bounded  $\iff \limsup_{R \rightarrow \infty} \sup_n \int_{\|x\| > R} |f_n(x)|^2 dx = 0$

■  $\sup_n \|\mathcal{T}_h f_n - f_n\| \rightarrow 0$  as  $\|h\| \rightarrow 0$

# Step 2

$$\exists(f_{n_k}) \rightharpoonup T_W^+ g$$

$$\beta_n \downarrow 0, \quad f_n := f_{\alpha, \beta_n}, \quad \phi_{n_k} := \phi_{\beta_{n_k}}$$

# Step 2

$$\exists(f_{n_k}) \rightharpoonup T_W^+ g$$

$$\beta_n \downarrow 0, \quad f_n := f_{\alpha, \beta_n}, \quad \phi_{n_k} := \phi_{\beta_{n_k}}$$

$$\left. \begin{array}{l} \text{Step 1} \implies (f_n) \text{ bounded} \\ \text{Weak Compactness Theorem} \end{array} \right\} \implies \exists(f_{n_k}) \rightharpoonup f'$$

# Step 2

$$\exists(f_{n_k}) \rightharpoonup T_W^+ g$$

$$\beta_n \downarrow 0, \quad f_n := f_{\alpha, \beta_n}, \quad \phi_{n_k} := \phi_{\beta_{n_k}}$$

$$\left. \begin{array}{l} \text{Step 1} \implies (f_n) \text{ bounded} \\ \text{Weak Compactness Theorem} \end{array} \right\} \implies \exists(f_{n_k}) \rightharpoonup f'$$

To be shown:  $f' = T_W^+ g$

# Step 2

$$\exists(f_{n_k}) \rightharpoonup T_W^+ g$$

$$\beta_n \downarrow 0, \quad f_n := f_{\alpha, \beta_n}, \quad \phi_{n_k} := \phi_{\beta_{n_k}}$$

$$\left. \begin{array}{l} \text{Step 1} \implies (f_n) \text{ bounded} \\ \text{Weak Compactness Theorem} \end{array} \right\} \implies \exists(f_{n_k}) \rightharpoonup f'$$

To be shown:  $f' = T_W^+ g$

$$\begin{aligned} \left\| \hat{\phi}_{n_k} g - T_W f_{n_k} \right\|^2 + \alpha \left\| (1 - \hat{\phi}_{n_k}) \hat{f}_{n_k} \right\|^2 \\ \leq \left\| \hat{\phi}_{n_k} g - T_W f \right\|^2 + \alpha \left\| (1 - \hat{\phi}_{n_k}) \hat{f} \right\|^2 \end{aligned}$$

# Step 2

$$\exists(f_{n_k}) \rightharpoonup T_W^+ g$$

$$\beta_n \downarrow 0, \quad f_n := f_{\alpha, \beta_n}, \quad \phi_{n_k} := \phi_{\beta_{n_k}}$$

$$\left. \begin{array}{l} \text{Step 1} \implies (f_n) \text{ bounded} \\ \text{Weak Compactness Theorem} \end{array} \right\} \implies \exists(f_{n_k}) \rightharpoonup f'$$

To be shown:  $f' = T_W^+ g$

$$\begin{aligned} \left\| \hat{\phi}_{n_k} g - T_W f_{n_k} \right\|^2 + \alpha \left\| (1 - \hat{\phi}_{n_k}) \hat{f}_{n_k} \right\|^2 \\ \leq \left\| \hat{\phi}_{n_k} g - T_W f \right\|^2 + \alpha \left\| (1 - \hat{\phi}_{n_k}) \hat{f} \right\|^2 \end{aligned}$$

# Step 2

$$\exists(f_{n_k}) \rightharpoonup T_W^+ g$$

$$\beta_n \downarrow 0, \quad f_n := f_{\alpha, \beta_n}, \quad \phi_{n_k} := \phi_{\beta_{n_k}}$$

$$\left. \begin{array}{l} \text{Step 1} \implies (f_n) \text{ bounded} \\ \text{Weak Compactness Theorem} \end{array} \right\} \implies \exists(f_{n_k}) \rightharpoonup f'$$

To be shown:  $f' = T_W^+ g$

$$\left\| \hat{\phi}_{n_k} g - T_W f_{n_k} \right\|^2 \leq \left\| \hat{\phi}_{n_k} g - T_W f \right\|^2 + \alpha \left\| (1 - \hat{\phi}_{n_k}) \hat{f} \right\|^2$$

# Step 2

$$\exists(f_{n_k}) \rightharpoonup T_W^+ g$$

$$\beta_n \downarrow 0, \quad f_n := f_{\alpha, \beta_n}, \quad \phi_{n_k} := \phi_{\beta_{n_k}}$$

$$\left. \begin{array}{l} \text{Step 1} \implies (f_n) \text{ bounded} \\ \text{Weak Compactness Theorem} \end{array} \right\} \implies \exists(f_{n_k}) \rightharpoonup f'$$

To be shown:  $f' = T_W^+ g$

$$\left\| \hat{\phi}_{n_k} g - T_W f_{n_k} \right\|^2 \leq \left\| \hat{\phi}_{n_k} g - T_W f \right\|^2 + \alpha \underbrace{\left\| (1 - \hat{\phi}_{n_k}) \hat{f} \right\|^2}_{\text{term to be shown}}$$

$$(1 - \hat{\phi}_{n_k}) \hat{f} \rightarrow \hat{f} - \hat{\phi}(0) \hat{f} = 0$$

# Step 2

$$\exists(f_{n_k}) \rightharpoonup T_W^+ g$$

$$\beta_n \downarrow 0, \quad f_n := f_{\alpha, \beta_n}, \quad \phi_{n_k} := \phi_{\beta_{n_k}}$$

$$\left. \begin{array}{l} \text{Step 1} \implies (f_n) \text{ bounded} \\ \text{Weak Compactness Theorem} \end{array} \right\} \implies \exists(f_{n_k}) \rightharpoonup f'$$

To be shown:  $f' = T_W^+ g$

$$\left\| \hat{\phi}_{n_k} g - T_W f_{n_k} \right\|^2 \leq \underbrace{\left\| \hat{\phi}_{n_k} g - T_W f \right\|^2}_{\alpha} + \alpha \left\| (1 - \hat{\phi}_{n_k}) \hat{f} \right\|^2$$

$$\hat{\phi}_{n_k} g - T_W f \rightarrow g - T_W f$$

# Step 2

$$\exists(f_{n_k}) \rightharpoonup T_W^+ g$$

$$\beta_n \downarrow 0, \quad f_n := f_{\alpha, \beta_n}, \quad \phi_{n_k} := \phi_{\beta_{n_k}}$$

$$\left. \begin{array}{l} \text{Step 1} \implies (f_n) \text{ bounded} \\ \text{Weak Compactness Theorem} \end{array} \right\} \implies \exists(f_{n_k}) \rightharpoonup f'$$

To be shown:  $f' = T_W^+ g$

$$\underbrace{\left\| \hat{\phi}_{n_k} g - T_W f_{n_k} \right\|^2}_{\text{}} \leq \left\| \hat{\phi}_{n_k} g - T_W f \right\|^2 + \alpha \left\| (1 - \hat{\phi}_{n_k}) \hat{f} \right\|^2$$

$$\left. \begin{array}{l} \hat{\phi}_{n_k} g \rightarrow \hat{\phi}(0)g = g \\ T_W f_{n_k} \rightharpoonup T_W f' \end{array} \right\} \implies \hat{\phi}_{n_k} g - T_W f_{n_k} \rightharpoonup g - T_W f'$$

# Step 2

$$\exists(f_{n_k}) \rightharpoonup T_W^+ g$$

$$\beta_n \downarrow 0, \quad f_n := f_{\alpha, \beta_n}, \quad \phi_{n_k} := \phi_{\beta_{n_k}}$$

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$$\left\| g - T_W f' \right\|^2 \leq \liminf_{k \rightarrow \infty} \left\| \hat{\phi}_{n_k} g - T_W f_{n_k} \right\|^2.$$

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$$\hookrightarrow f' \text{ minimizes } f \mapsto \left\| g - T_W f \right\|^2$$

$$\hookrightarrow f' = T_W^+ g$$

# Overview of the proof

Step 1:  $(f_{\alpha,\beta})_{\beta \in (0,1]}$  is **bounded**

Step 2:  $(f_{\alpha,\beta})_{\beta \in (0,1]}$  converges weakly to  $T_W^+ g$

$$\beta_n \downarrow 0, f_n := f_{\alpha,\beta_n}$$

■  $\exists (f_{n_k}) \rightharpoonup T_W^+ g$

Step 3: The convergence is in fact strong

■  $(f_n)$  is bounded (Step 1)

■  $V$  bounded  $\iff \lim_{R \rightarrow \infty} \sup_n \int_{\|x\| > R} |f_n(x)|^2 dx = 0$

■  $\sup_n \|\mathcal{T}_h f_n - f_n\| \rightarrow 0$  as  $\|h\| \rightarrow 0$

# Step 3

$$\sup_n \|\mathcal{T}_h f_n - f_n\| \rightarrow 0 \text{ as } \|h\| \rightarrow 0$$

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$$\|\mathcal{T}_h f_{\alpha,\beta} - f_{\alpha,\beta}\|^2 = \|U(\mathcal{T}_h f_{\alpha,\beta} - f_{\alpha,\beta})\|^2$$

# Step 3

$$\sup_n \|\mathcal{T}_h f_n - f_n\| \rightarrow 0 \text{ as } \|h\| \rightarrow 0$$

$$\begin{aligned} \|\mathcal{T}_h f_{\alpha,\beta} - f_{\alpha,\beta}\|^2 &= \|U(\mathcal{T}_h f_{\alpha,\beta} - f_{\alpha,\beta})\|^2 \\ &= \int_{\mathbb{R}^d} |e^{-2i\pi\langle h,\xi\rangle} - 1|^2 |\hat{f}_{\alpha,\beta}(\xi)|^2 d\xi \end{aligned}$$

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$$I_1 := \int_{\|\xi\| \leq 1/\beta} |e^{-2i\pi\langle h,\xi\rangle} - 1|^2 |\hat{f}_{\alpha,\beta}(\xi)|^2 d\xi$$

$$I_2 := \int_{\|\xi\| > 1/\beta} |e^{-2i\pi\langle h,\xi\rangle} - 1|^2 |\hat{f}_{\alpha,\beta}(\xi)|^2 d\xi$$

# Step 3

$$\sup_n \|\mathcal{T}_h f_n - f_n\| \rightarrow 0 \text{ as } \|h\| \rightarrow 0$$

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$$I_1 = \int_{0 < \|\xi\| \leq 1/\beta} \frac{|e^{-2i\pi\langle h, \xi \rangle} - 1|^2}{\|\xi\|^{2s'}} \|\xi\|^{2s'} \left| \hat{f}_{\alpha, \beta}(\xi) \right|^2 d\xi$$

# Step 3

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$$\begin{aligned} I_1 &= \int_{0 < \|\xi\| \leq 1/\beta} \frac{|e^{-2i\pi\langle h, \xi \rangle} - 1|^2}{\|\xi\|^{2s'}} \|\xi\|^{2s'} \left| \hat{f}_{\alpha, \beta}(\xi) \right|^2 d\xi \\ &\leq \sup_{\xi \neq 0} \frac{|e^{-2i\pi\langle h, \xi \rangle} - 1|^2}{\|\xi\|^{2s'}} \int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s'} \left| \hat{f}_{\alpha, \beta}(\xi) \right|^2 d\xi \end{aligned}$$

# Step 3

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$$\begin{aligned} I_1 &= \int_{0 < \|\xi\| \leq 1/\beta} \frac{|e^{-2i\pi\langle h, \xi \rangle} - 1|^2}{\|\xi\|^{2s'}} \|\xi\|^{2s'} \left| \hat{f}_{\alpha, \beta}(\xi) \right|^2 d\xi \\ &\leq \sup_{\xi \neq 0} \frac{|e^{-2i\pi\langle h, \xi \rangle} - 1|^2}{\|\xi\|^{2s'}} \int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s'} \left| \hat{f}_{\alpha, \beta}(\xi) \right|^2 d\xi \\ &= \|h\|^{2s'} \sup_{\xi' \neq 0} \frac{|e^{-2i\pi\langle \|h\|^{-1}h, \xi' \rangle} - 1|^2}{\|\xi'\|^{2s'}} \int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s'} \left| \hat{f}_{\alpha, \beta}(\xi) \right|^2 d\xi \end{aligned}$$

$$\xi' = \|h\| \xi$$

# Step 3

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$$\begin{aligned} I_1 &= \int_{0 < \|\xi\| \leq 1/\beta} \frac{|e^{-2i\pi\langle h, \xi \rangle} - 1|^2}{\|\xi\|^{2s'}} \|\xi\|^{2s'} \left| \hat{f}_{\alpha, \beta}(\xi) \right|^2 d\xi \\ &\leq \sup_{\xi \neq 0} \frac{|e^{-2i\pi\langle h, \xi \rangle} - 1|^2}{\|\xi\|^{2s'}} \int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s'} \left| \hat{f}_{\alpha, \beta}(\xi) \right|^2 d\xi \\ &= \underbrace{\|h\|^{2s'} \sup_{\xi' \neq 0} \frac{|e^{-2i\pi\langle \|h\|^{-1}h, \xi' \rangle} - 1|^2}{\|\xi'\|^{2s'}}}_{\leq K_1} \int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s'} \left| \hat{f}_{\alpha, \beta}(\xi) \right|^2 d\xi \end{aligned}$$

since  $|e^{-2i\pi\langle \|h\|^{-1}h, \xi' \rangle} - 1| = \mathcal{O}(\|\xi'\|)$  near the origin

# Step 3

$$\sup_n \|\mathcal{T}_h f_n - f_n\| \rightarrow 0 \text{ as } \|h\| \rightarrow 0$$

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$$I_1 \leq K_1 \|h\|^{2s'} \int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s'} \left| \hat{f}_{\alpha, \beta}(\xi) \right|^2 d\xi$$

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$$s' := \min\{1, s\} \implies \|\xi\|^{2s'} \leq 1 + \|\xi\|^{2s}$$

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$$I_1 \leq K_1 \|h\|^{2s'} \left( \int_{\|\xi\| \leq 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi + \int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \right)$$

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$$\begin{aligned} I_1 &\leq K_1 \|h\|^{2s'} \left( \underbrace{\int_{\|\xi\| \leq 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi}_{\leq \|\hat{f}_{\alpha,\beta}\|^2} + \int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \right) \\ &\leq \|\hat{f}_{\alpha,\beta}\|^2 = \|f_{\alpha,\beta}\|^2 \leq K_2 \end{aligned}$$

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$$\leq K_1 \|h\|^{2s'} \left( K_2 + \int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \right)$$

# Step 3

$$\sup_n \|\mathcal{T}_h f_n - f_n\| \rightarrow 0 \text{ as } \|h\| \rightarrow 0$$

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Reminder from Step 1:

$$\int_{\|\xi\| \leq 1/\beta} \|\xi\|^{2s} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi + \frac{1}{M_\beta} \int_{\|\xi\| > 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \leq \frac{K}{\nu_0}$$

# Step 3

$$\sup_n \|\mathcal{T}_h f_n - f_n\| \rightarrow 0 \text{ as } \|h\| \rightarrow 0$$

$$I_1 \leq K_1 \|h\|^{2s'} (K_2 + K_3) = K_4 \|h\|^{2s'}$$

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$$\begin{aligned} I_2 &:= \int_{\|\xi\| > 1/\beta} \left| e^{-2i\pi\langle h, \xi \rangle} - 1 \right|^2 \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \\ &\leq 4 \int_{\|\xi\| > 1/\beta} \left| \hat{f}_{\alpha,\beta}(\xi) \right|^2 d\xi \\ &\leq \frac{4K}{\nu_0} M_\beta \end{aligned}$$

# Step 3

$$\sup_n \|\mathcal{T}_h f_n - f_n\| \rightarrow 0 \text{ as } \|h\| \rightarrow 0$$

$$\left. \begin{array}{l} I_1 \leq K_4 \|h\|^{2s'} \\ I_2 \leq \frac{4K}{\nu_0} M_\beta \end{array} \right\} \implies I_1 + I_2 \leq \max \left\{ K_4, \frac{4K}{\nu_0} \right\} \left( \|h\|^{2s'} + M_\beta \right)$$

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$$\|\mathcal{T}_h f_{\alpha,\beta} - f_{\alpha,\beta}\|^2 \leq K_5 (\|h\|^{2s'} + M_\beta)$$

# Step 3

$$\sup_n \|\mathcal{T}_h f_n - f_n\| \rightarrow 0 \text{ as } \|h\| \rightarrow 0$$

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$$\|\mathcal{T}_h f_{\alpha,\beta} - f_{\alpha,\beta}\|^2 \leq K_5 (\|h\|^{2s'} + M_\beta)$$

$$M_\beta := \max_{\|\xi\|=1} \left| 1 - \hat{\phi}(\beta\xi) \right|^2 \rightarrow 0 \text{ as } \beta \rightarrow 0$$

# Step 3

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$$\hookrightarrow \forall \varepsilon > 0, \exists n_0 \in \mathbb{N}^*: \forall n \geq n_0, M_{\beta_n} \leq \varepsilon$$

# Step 3

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$\rightarrow 0 \text{ as } h \rightarrow 0$

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$$\rightarrow 0 \text{ as } h \rightarrow 0$$

$$\limsup_{h \rightarrow 0} \sup_{n \in \mathbb{N}^*} \|\mathcal{T}_h f_n - f_n\|^2 \leq K_5 \varepsilon$$

# Examples: Lévy kernels

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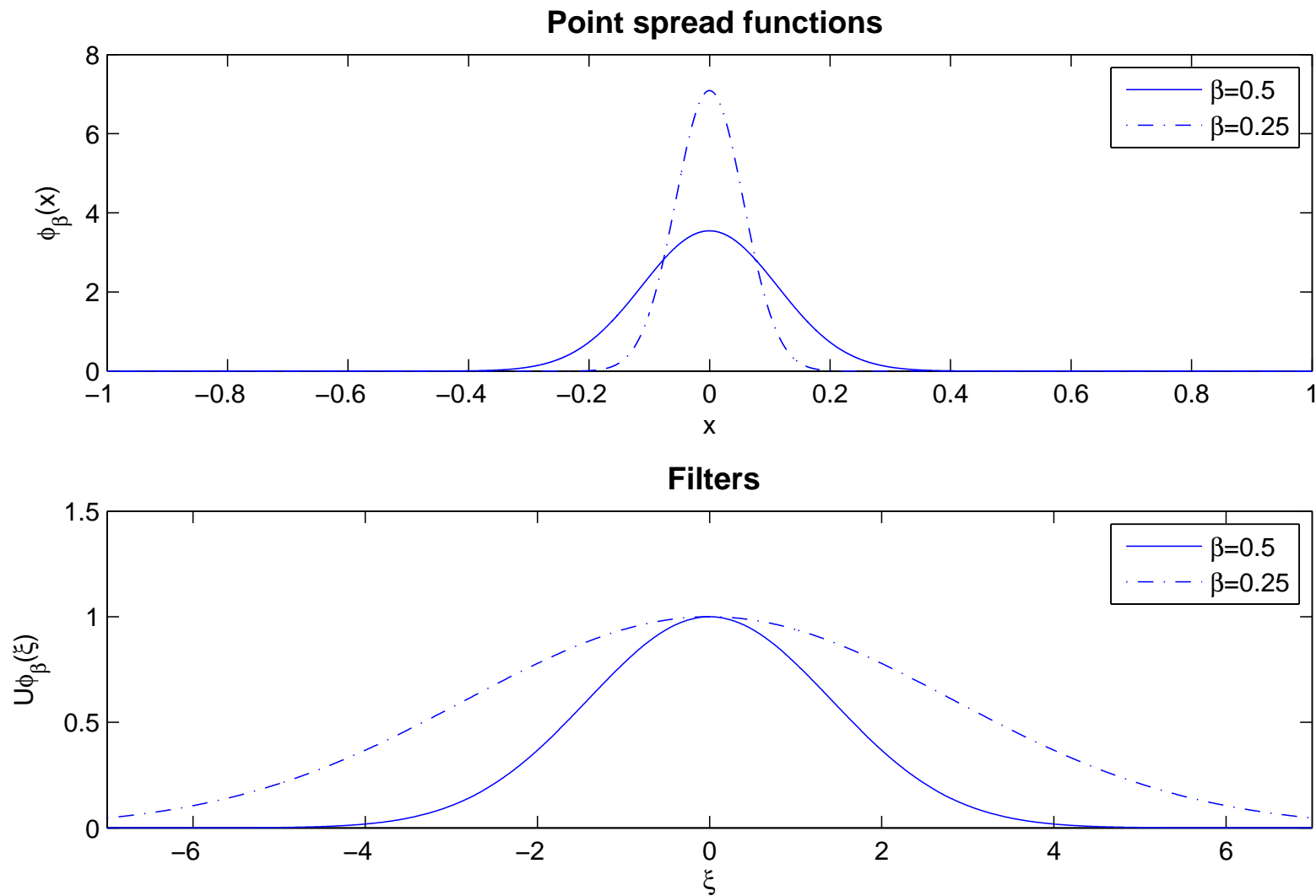
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$$\hat{\phi}: \xi \mapsto \exp(-\|\xi\|^s), \quad s \in [0, 2]$$

$$\phi: x \mapsto U^{-1} \exp(-\|\cdot\|^s)(x)$$

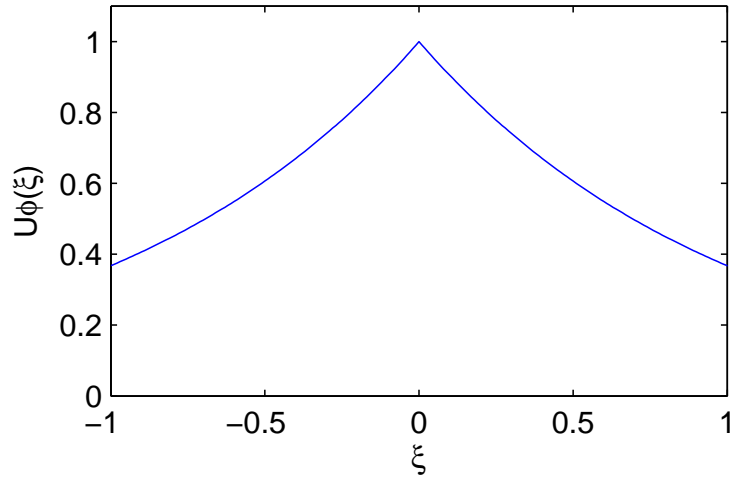
$\hookrightarrow \phi$  is positive, isotropic, radially decreasing,  $C^\infty$

# Examples: Lévy kernels

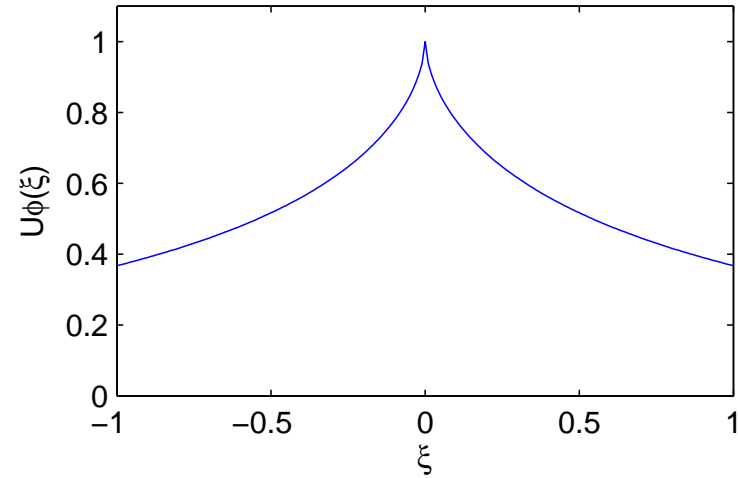


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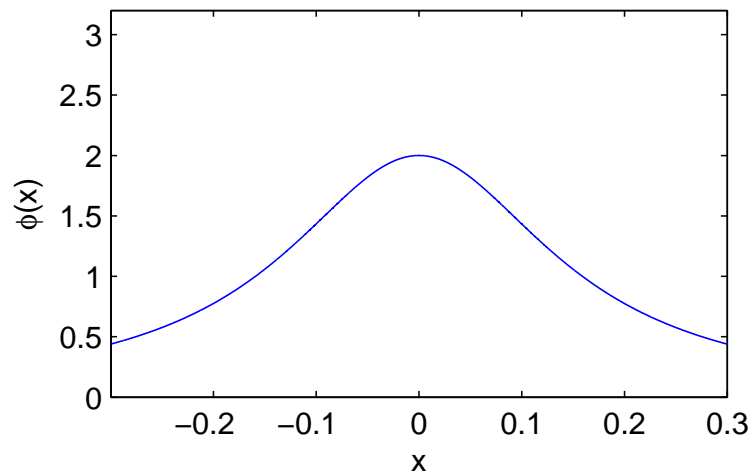
Cauchy filter (s=1)



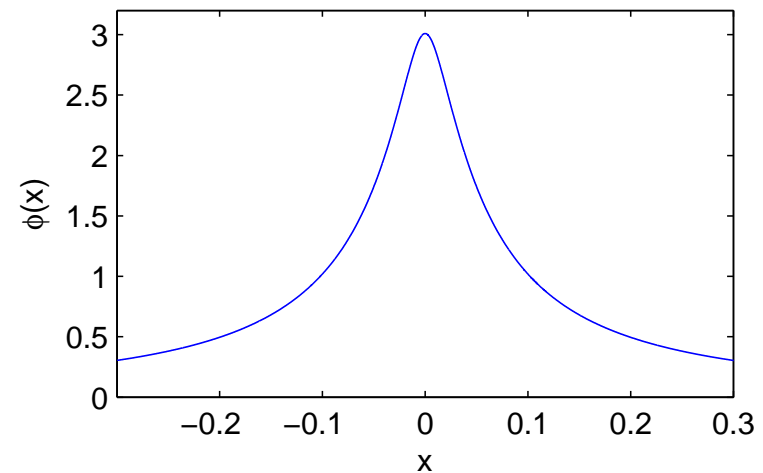
Filter for s=0.6



Cauchy kernel (s=1)



Kernel for s=0.6



# Hybrid approach

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---

**Theorem** Assume that

- $\phi \in L^1(\mathbb{R}^d)$  and  $\hat{\phi}(0) = \int \phi(x) dx \in (0, 1)$

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# Hybrid approach

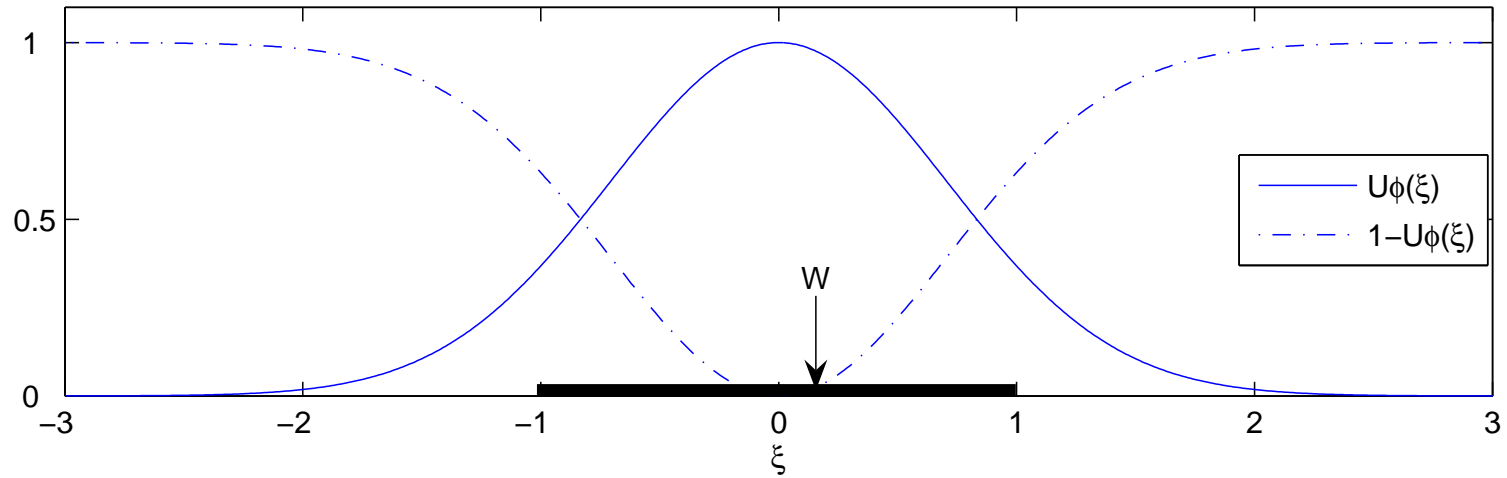
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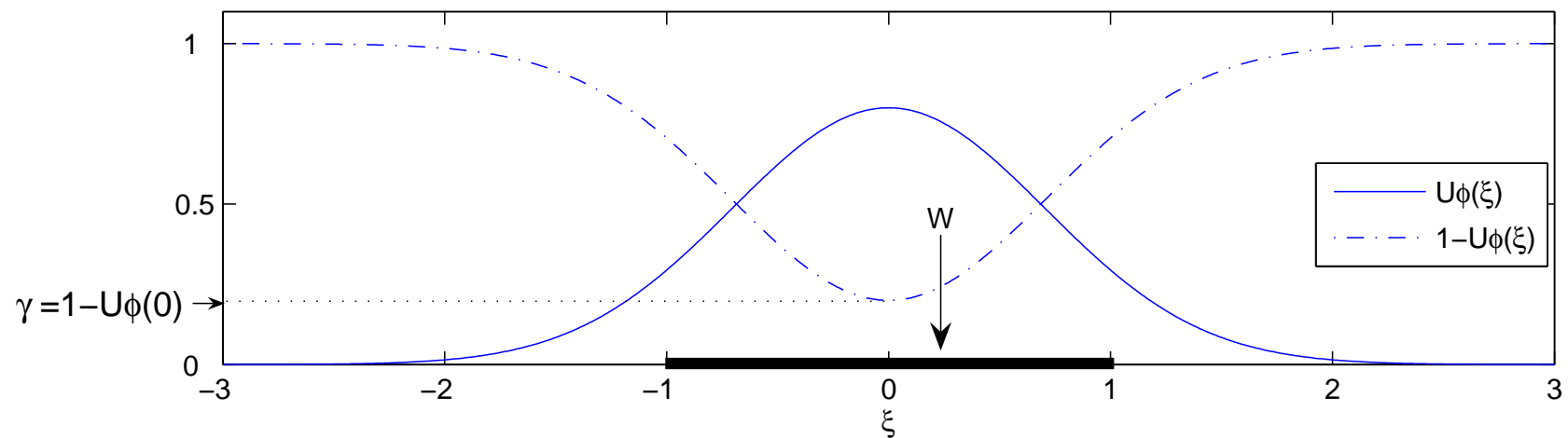
Let  $f_n := f_{\alpha_n, \beta_n}$ . If  $g \in \mathcal{D}(T_W^+)$ , then  $\hat{\phi}(0)^{-1} f_n \rightarrow T_W^+ g$ ,  
as  $n \rightarrow \infty$ .

# Hybrid approach

Case  $U\phi(0)=1$



Case  $U\phi(0) \in (0,1)$



# Outline

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- Introduction
- A reminder on ill-posed problems
- Fourier synthesis
- Asymptotic analysis
- A dual algorithm (joint work with D. Wallach)

# Finite dimensional setting

---

$$(P) \quad \left| \begin{array}{l} \text{Minimize} \quad \frac{1}{2} \|\mathbf{g} - T\mathbf{f}\|_{\mathbb{C}^m}^2 + \frac{\alpha}{2} \|H\mathbf{f}\|_{\mathbb{C}^{m_r}}^2 \\ \text{s.t.} \quad \mathbf{f} \in \mathbb{R}^n \end{array} \right.$$

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$$g_k \simeq \hat{f}_0(\xi_k) = \int_V e^{-2i\pi\langle x, \xi_k \rangle} f_0(x) \, dx, \quad k = 1, \dots, m$$

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$$(P) \quad \left| \begin{array}{l} \text{Minimize} \quad \frac{1}{2} \| \mathbf{g} - T\mathbf{f} \|^2_{\mathbb{C}^m} + \frac{\alpha}{2} \| H\mathbf{f} \|^2_{\mathbb{C}^{m_r}} \\ \text{s.t.} \quad \mathbf{f} \in \mathbb{R}^n \end{array} \right.$$

The components of  $\mathbf{f} \in \mathbb{R}^n$  (or  $\mathbb{C}^n$ ) are the coordinates of the approximation of  $f$  in the finite dimensional subspace generated by some *interpolation family*  $\{e_1, \dots, e_n\}$ .

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$T \in M_{m \times n}(\mathbb{C})$  is the matrix whose entry  $(j, k)$  is the Fourier transform of  $e_k$  at  $\xi_j$

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$H \in M_{m_r \times n}(\mathbb{C})$  is formed likewise, with  $\xi_j$  replaced by *regularizing sampling points*.

# Finite dimensional setting

---

Framework:

Minimize  $\Phi(\mathbf{f}) - \Psi(T\mathbf{f})$  with  $\left\{ \begin{array}{l} \Phi \text{ convex} \\ \Psi \text{ concave} \end{array} \right.$

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Fenchel duality:

$$\inf_{\mathbf{f} \in \mathbb{R}^n} \{ \Phi(\mathbf{f}) - \Psi(T\mathbf{f}) \} = \max_{\boldsymbol{\lambda} \in \mathbb{R}^m} \{ \Psi_{\star}(\boldsymbol{\lambda}) - \Phi^*(T^*\boldsymbol{\lambda}) \}$$

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$$\bar{\mathbf{f}} = \nabla \Phi^*(T^*\bar{\boldsymbol{\lambda}})$$

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# Conjugacy: real/complex

---

$$\Psi_{\star}(\xi) := \inf_x \{ \langle \xi, x \rangle - \Psi(x) \} \quad (\text{convex})$$

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The complex case:

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Then,

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---

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**Remark** the Constraint Qualification (CQ) is always satisfied by Problem  $(\mathcal{P})$

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Together with the assumptions of Fenchel's Theorem, suppose that

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---

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$$\begin{aligned}\Phi^*(\xi) &= \sup \{ \langle \xi, x \rangle - F(Tx) \mid x \in X \} \\ &= \sup \{ \langle \xi, T^{-1}y \rangle - F(y) \mid y \in Y \} \\ &= \sup \{ \langle (T^{-1})^*\xi, y \rangle - F(y) \mid y \in Y \} \\ &= F^*((T^*)^{-1}\xi) \quad \blacksquare\end{aligned}$$

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## Lemma

- $Y$  be a (real or complex) Hilbert space with inner product  $\langle \cdot, \cdot \rangle$  and norm  $\| \cdot \|$

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Then  $\Phi$  is self-conjugate, that is,  $\Phi^* = \Phi$ .

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# Computing $D(\lambda)$ and $\nabla D(\lambda)$

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thus it is a Hilbert space on its own

# Computing $D(\lambda)$ and $\nabla D(\lambda)$

$\text{ran } T_{W_\beta}$  is closed

$$\begin{aligned}\Phi^*(\varphi) &= \frac{1}{2} \left\| (T_{W_\beta}^*)^{-1} \varphi \right\|_{L^2(W_\beta)}^2 \\ &= \frac{1}{2} \left\langle (T_{W_\beta}^*)^{-1} \varphi, (T_{W_\beta}^{-1})^* \varphi \right\rangle_{L^2(W_\beta)} \\ &= \frac{1}{2} \left\langle (T_{W_\beta}^* T_{W_\beta})^{-1} \varphi, \varphi \right\rangle_{L^2(V)}\end{aligned}$$

# Computing $D(\lambda)$ and $\nabla D(\lambda)$

---

$$T_{W_\beta}^* T_{W_\beta} = I - T_{B_{1/\beta}}^* T_{B_{1/\beta}}$$

$$\|T_{B_{1/\beta}}^* T_{B_{1/\beta}}\| < 1$$

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$$\Phi^*(\varphi) = \frac{1}{2} \sum_{k=0}^{\infty} \left\langle (T_{B_{1/\beta}}^* T_{B_{1/\beta}})^k \varphi, \varphi \right\rangle_{L^2(V)}$$

# Illustration

