

Equivalence of smoothness spaces by means of frames of discrete shearlets on the cone and curvelets

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Outline

Motivation

- Art Authentication

Background

- Frames

 - Curvelets

 - Shearlets on the cone

- Approximation Spaces

- Decomposition Spaces

Function Spaces via Frame Decompositions

- Equivalence of Decomposition Spaces

- Sketch of the Proof

Future Work

- From Decomposition Spaces to Approximation Spaces



Art Authentication

Goals

The goal of research in [Art Authentication](#) is to generate a mathematical and computational characterization of a painter's technique (including brushwork, composition and color choice)

- ▶ Dating
- ▶ Attribution
- ▶ Distinguishing feature extraction



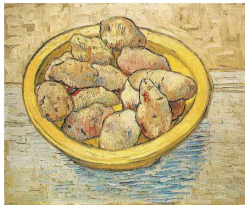
Art Authentication

The Van Gogh Project—Johnson, Daubechies, Wang, Postma

Which one is a fake?



Crab on its back



Still life: Potatoes
in a yellow dish



Willows at sunset

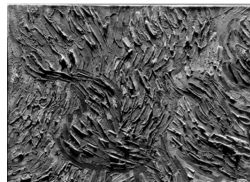


Art Authentication

The Van Gogh Project—Johnson, Daubechies, Wang, Postma

Historical approach to visible light examination of Van Gogh's brushstrokes

- ▶ brushstroke width and shape
- ▶ impact of paint consistency and substrate texture
- ▶ artist-characteristic brushstroke



Raking light detail
to accentuate
brushwork texture in
*The Garden of Saint
Paul's Hospital*



How we choose to represent the data matters

Piecewise constant representation vs. Wavelet representation



$1,024 \times 1,024 = 1,048,576$ pixels

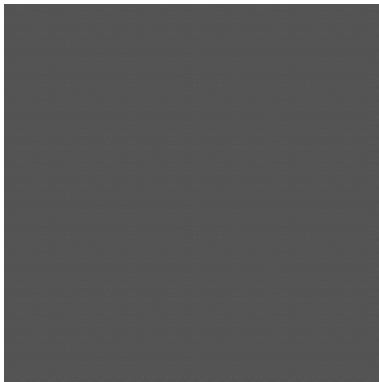


How we choose to represent the data matters

Piecewise constant representation vs. Wavelet representation



$1,024 \times 1,024 = 1,048,576$ pixels



1 wavelet coefficient

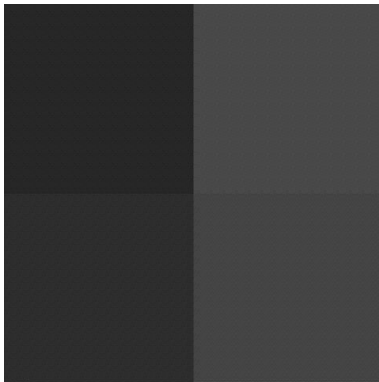


How we choose to represent the data matters

Piecewise constant representation vs. Wavelet representation



$1,024 \times 1,024 = 1,048,576$ pixels



$1 + 4 = 5$ wavelet coefficients

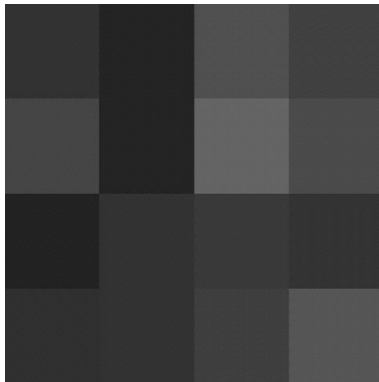


How we choose to represent the data matters

Piecewise constant representation vs. Wavelet representation



$1,024 \times 1,024 = 1,048,576$ pixels



$1 + 4 + 16 = 21$ wavelet coefficients

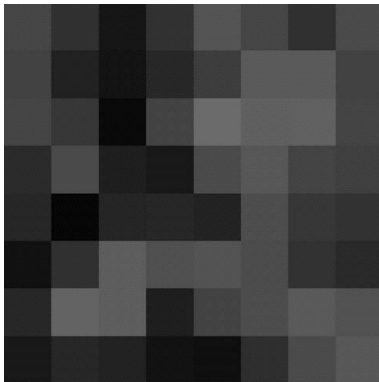


How we choose to represent the data matters

Piecewise constant representation vs. Wavelet representation



$1,024 \times 1,024 = 1,048,576$ pixels



$1 + 4 + 16 + 64 = 85$ wavelet coefficients

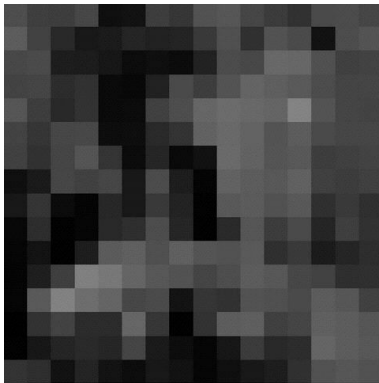


How we choose to represent the data matters

Piecewise constant representation vs. Wavelet representation



$1,024 \times 1,024 = 1,048,576$ pixels



$1 + 4 + 16 + 64 + 256 = 341$ wavelet coefficients

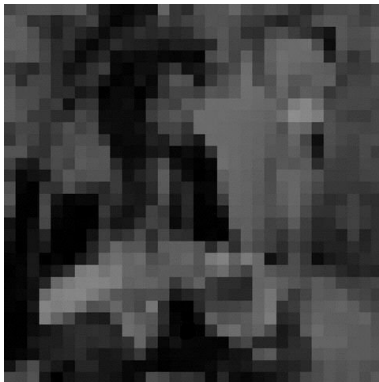


How we choose to represent the data matters

Piecewise constant representation vs. Wavelet representation



$1,024 \times 1,024 = 1,048,576$ pixels



$1 + 4 + 16 + 64 + 256 + 1024 = 1365$ wavelet coefficients



How we choose to represent the data matters

Piecewise constant representation vs. Wavelet representation



$1,024 \times 1,024 = 1,048,576$ pixels



$1 + 4 + 16 + 64 + 256 + 1024 + 4096 = 5461$ wavelet coefficients



A sequence $\{\phi_\lambda\}_{\lambda \in \Lambda}$ is a **frame** for a Hilbert space \mathcal{H} if

$$A\|f\|^2 \leq \sum_{\lambda \in \Lambda} |\langle f, \phi_\lambda \rangle|^2 \leq B\|f\|^2$$

If $A = B$, the frame is **tight**.



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Trivial description of \mathcal{H} by means of frame coefficients:

$$f \in \mathcal{H} \text{ if and only if } \sum_{\lambda \in \Lambda} |\langle f, \phi_\lambda \rangle|^2 < \infty$$



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Trivial description of \mathcal{H} by means of frame coefficients:

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Frames are not bases (in general) but still,

$$f = \sum_{\lambda \in \Lambda} \langle f, \phi_\lambda \rangle \phi_\lambda$$



Frames of Curvelets and Shearlets

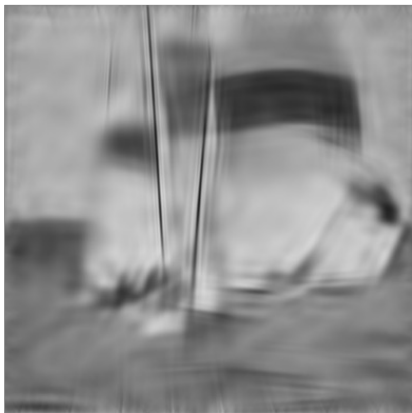


$$f = \sum_{n=0}^{\infty} \sum_k \sum_{z \in \mathbb{Z}^2} \langle f, \phi_{nzk} \rangle \phi_{nzk} \\ + \sum_{z \in \mathbb{Z}^2} \langle f, \Phi_z \rangle \Phi_z$$

$$\|f\|_{L_2(\mathbb{R}^2)}^2 = \sum_{n=0}^{\infty} \sum_k \sum_{z \in \mathbb{Z}^2} |\langle f, \phi_{nzk} \rangle|^2 \\ + \sum_{z \in \mathbb{Z}^2} |\langle f, \Phi_z \rangle|^2$$



Frames of Curvelets and Shearlets

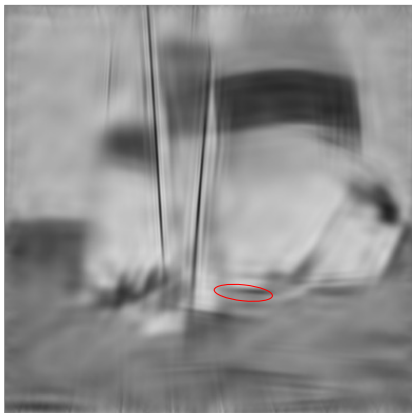


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Frames of Curvelets and Shearlets

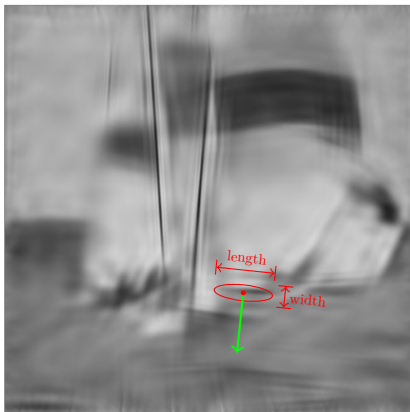


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Frames of Curvelets and Shearlets



$$f = \sum_{n=0}^{\infty} \sum_k \sum_{z \in \mathbb{Z}^2} \langle f, \phi_{nzk} \rangle \phi_{nzk} + \sum_{z \in \mathbb{Z}^2} \langle f, \Phi_z \rangle \Phi_z$$

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$$\text{width}_n / \text{length}_n \asymp 2^n.$$



Definition (Candès, Donoho: Curvelets '03)

$\phi_{nzk}: \mathbb{R}^2 \rightarrow \mathbb{C}$ with parameters $n \in \mathbb{Z}$ (shape AND scaling), $\mathbf{z} \in \mathbb{Z}^2$ (location), and $1 \leq k \leq 2^{\lceil n/2 \rceil + 2}$ (direction).

$$\widehat{(\phi_{nzk})}(\xi) = W_n(|\xi|) V_{n,k}(\xi/|\xi|) e^{2\pi i(\beta_{nzk} \cdot \xi)}$$

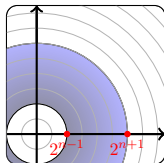


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- $W_n(\xi)$ amplitude window.

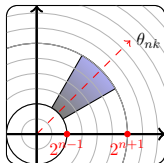


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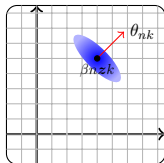


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- ▶ $W_n(\xi)$ amplitude window.
- ▶ $V_{n,k}(\xi/|\xi|)$ angular window.
- ▶ $\phi_{nzk}(x) = \phi_{n0k}(x - \beta_{nzk})$



Definition (Kutyniok, Labate: Shearlets on the cone '05)

$\phi_{nzk}: \mathbb{R}^2 \rightarrow \mathbb{R}$ with parameters $n \in \mathbb{Z}$ (shape AND scaling), $z \in \mathbb{Z}^2$ (location), and $-2^n \leq k < 2^n$ (direction/shear).

$$\widehat{(\phi_{nzk})}(\xi) = W_n(|\xi_1|) V_{n,k}(\xi_2/\xi_1) e^{2\pi i(\beta_{nzk} \cdot \xi)}$$

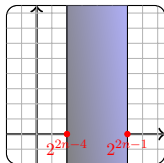


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► $W_n(|\xi_1|)$ “first coordinate” window.

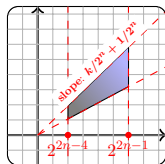


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- ▶ $W_n(|\xi_1|)$ “first coordinate” window.
- ▶ $V_{n,k}(\xi_2/\xi_1)$ slope window.

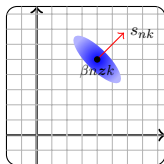


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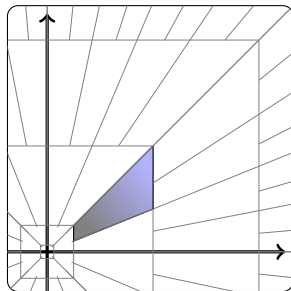
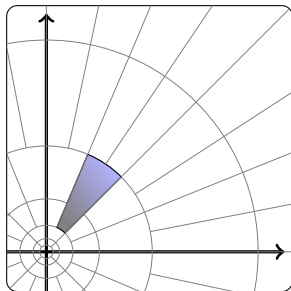
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- ▶ $\phi_{nzk}(x) = \phi_{n0k}(x - \beta_{nzk})$



Different frames?

- ▶ Same
spatial localization, scaling, directional sensitivity, sparsity.
- ▶ Different
frequency localization, generation at different levels.



How to prove equivalence of these two frames?

- ▶ Same description of Smoothness Spaces



How to prove equivalence of these two frames?

- ▶ Same description of Smoothness Spaces
- ▶ Same Approximation Spaces

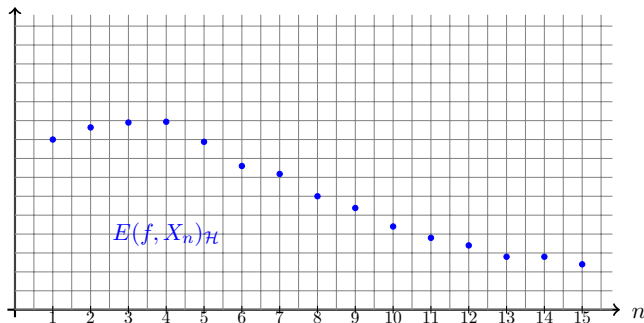
$$X_N = \left\{ \sum_{\ell \in \Lambda_N} \mathbf{c}_\ell \phi_\ell : \#\Lambda_N = N \right\}$$
$$\mathcal{A}_q^s(\mathcal{H}, \{X_N^{\text{curv}}\}_{N \in \mathbb{N}}) = \mathcal{A}_q^s(\mathcal{H}, \{X_N^{\text{shear}}\}_{N \in \mathbb{N}})$$



Approximation Spaces

$$\mathcal{A}_q^s(\mathcal{H}, \{X_n\}_{n \in \mathbb{N}}) = \left\{ f \in \mathcal{H} : \left(\sum_{n=1}^{\infty} \frac{1}{n} (n^s E(f, X_n)_{\mathcal{H}})^q \right)^{1/q} < \infty \right\},$$

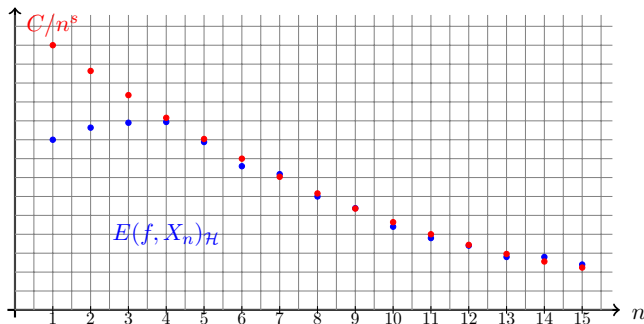
$$\mathcal{A}_{\infty}^s(\mathcal{H}, \{X_n\}_{n \in \mathbb{N}}) = \left\{ f \in \mathcal{H} : \sup_{n \in \mathbb{N}} n^s E(f, X_n)_{\mathcal{H}} < \infty \right\}.$$



Approximation Spaces

$$\mathcal{A}_q^s(\mathcal{H}, \{X_n\}_{n \in \mathbb{N}}) = \left\{ f \in \mathcal{H} : \left(\sum_{n=1}^{\infty} \frac{1}{n} (n^s E(f, X_n)_{\mathcal{H}})^q \right)^{1/q} < \infty \right\},$$

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Decomposition Spaces (Feichtinger, Gröbner) '80s

- ▶ $\{Q_\lambda\}_{\lambda \in \Lambda}$ a covering of \mathbb{R}^2 satisfying:

$$\exists N > 0, \forall \lambda_0, \#\{\lambda : Q_\lambda \cap Q_{\lambda_0} \neq \emptyset\} \leq N.$$

- ▶ $\{\psi_\lambda\}_{\lambda \in \Lambda}$ a partition of unity satisfying:

- ▶ $\text{supp } \psi_\lambda \subset Q_\lambda$
- ▶ $\sup_{\lambda \in \Lambda} |Q_\lambda|^{1/p-1} \|\mathcal{F}^{-1} \psi_\lambda\|_{L_p(\mathbb{R}^2)} < \infty$ for $0 < p < 1$.

- ▶ A moderate weight $\omega = \{\omega_\lambda = \omega(x_\lambda)\}_{\lambda \in \Lambda}$:

- ▶ $\omega: \mathbb{R}^2 \rightarrow \mathbb{R}^+$ satisfying $\omega(x) \leq C\omega(y)$, $x, y \in Q_\lambda$.
- ▶ $x_\lambda \in Q_\lambda$.



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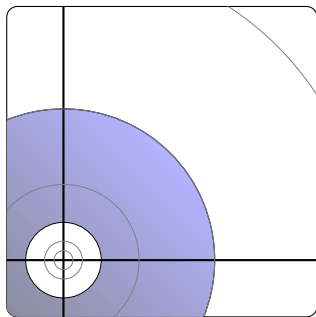
$$\mathfrak{D}(\{Q_\lambda\}_\Lambda, \{\psi_\lambda\}_\Lambda)^{\ell_q(\Lambda, \omega)}_{L_p(\mathbb{R}^2)} = \left\{ f \in L_p(\mathbb{R}^2) : \{\|\mathcal{F}^{-1}(\psi_\lambda \hat{f})\|_{L_p(\mathbb{R}^2)}\}_{\lambda \in \Lambda} \in \ell_q(\Lambda, \omega) \right\}$$



Besov Spaces (Frazier, Jawerth, Weiss) '91

$$f \in B_q^s(L_p(\mathbb{R}^2)) \text{ iff } \sum_{n \in \mathbb{Z}} \left(\int_{\mathbb{R}^2} (1 + 2^{sn})^p |\mathcal{D}_n f(x)|^p dx \right)^{q/p} < \infty$$

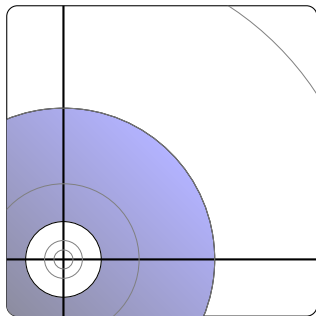
- Atoms $(\mathcal{D}_n f)^\wedge(\xi) = \psi_n(\xi) \hat{f}(\xi)$,
where
 $\text{supp } \psi_n = \{2^{n-1} \leq |\xi| \leq 2^{n+1}\}$,
 $\sum_{n \in \mathbb{Z}} \psi_n(\xi) = 1$.
- Weights $\omega_n = 1 + 2^{sn}$



Besov Spaces (Frazier, Jawerth, Weiss) '91

- ▶ $\Lambda = \mathbb{Z}$.
- ▶ **Covering:**
 $Q_n = \{2^{n-1} < |\xi| < 2^{n+1}\}.$
- ▶ **Partition of Unity:** ψ_n radially symmetric, $\sum_{n \in \mathbb{Z}} \psi_n(\xi) = 1$ for all $\xi \in \mathbb{R}^2 \setminus \{0\}$.
- ▶ **Moderate weight:** $\omega_n = 1 + 2^{ns}$.

$$B_q^s(L_p(\mathbb{R}^2)) = \mathfrak{D}(\{Q_n\}, \{\psi_n\})_{L_p(\mathbb{R}^2)}^{\ell_q(\Lambda, \omega)}$$



Curvelet Decomposition Spaces

► $\Lambda = \{n \in \mathbb{N}; k = 1, \dots, 2^{\lceil n/2 \rceil + 2}\}.$

► **Covering:**

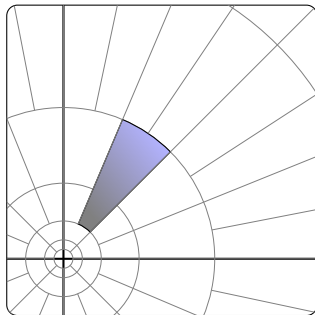
$$Q_{nk} = \text{supp } \phi_{n0k}.$$

► **Partition of Unity:**

$$\psi_{nk} = |\phi_{n0k}|^2.$$

► **Moderate weight:** $\omega_{nk} = 2^{ns}.$

$$\mathfrak{D}(\{Q_{nk}\}, \{\psi_{nk}\})_{L_p(\mathbb{R}^2)}^{\ell_q(\Lambda, \omega)}$$



Theorem (B-S '09)

Define $\psi_{nk} = |\phi_{n\mathbf{0}k}|^2$, $Q_{nk} = \text{supp } \psi_{nk}$, $\Lambda = \{(n, k)\}$. Then, for the same moderate weight ω ,

$$\mathfrak{D}(\{Q_{nk}^{\text{curv}}\}_{\Lambda}, \{\psi_{nk}^{\text{curv}}\}_{\Lambda})_{L_p(\mathbb{R}^2)}^{\ell_q(\Lambda, \omega)} = \mathfrak{D}(\{Q_{nk}^{\text{shear}}\}_{\Lambda}, \{\psi_{nk}^{\text{shear}}\}_{\Lambda})_{L_p(\mathbb{R}^2)}^{\ell_q(\Lambda, \omega)}$$

with equivalent norms.



Sketch of the proof

- The coverings $\mathfrak{Q}^{\text{curv}} = \{Q_{nk}^{\text{curv}}\}$ and $\mathfrak{Q}^{\text{shear}} = \{Q_{nk}^{\text{shear}}\}$ are “equivalent”:

$\mathfrak{Q}^{\text{curv}}$ is subordinate to $\{[Q]_7^{\text{shear}} : Q \in \mathfrak{Q}^{\text{shear}}\}$

$\mathfrak{Q}^{\text{shear}}$ is subordinate to $\{[Q]_2^{\text{curv}} : Q \in \mathfrak{Q}^{\text{curv}}\}$



Sketch of the proof

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$\mathfrak{Q}^{\text{shear}}$ is subordinate to $\{[Q]_2^{\text{curv}} : Q \in \mathfrak{Q}^{\text{curv}}\}$

- The families $\{\psi_{nk}^{\text{curv}}\}$ and $\{\psi_{nk}^{\text{shear}}\}$ are partitions of unity satisfying

$$\sup_{(n,k) \in \Lambda} |Q_{nk}|^{1/p-1} \|\mathcal{F}^{-1} \psi_{nk}\|_p < \infty \text{ for all } 0 < p \leq 1$$



Sketch of the proof

- The coverings $\Omega^{\text{curv}} = \{Q_{nk}^{\text{curv}}\}$ and $\Omega^{\text{shear}} = \{Q_{nk}^{\text{shear}}\}$ are “equivalent”:

Ω^{curv} is subordinate to $\{[Q]_7^{\text{shear}} : Q \in \Omega^{\text{shear}}\}$

Ω^{shear} is subordinate to $\{[Q]_2^{\text{curv}} : Q \in \Omega^{\text{curv}}\}$

- The families $\{\psi_{nk}^{\text{curv}}\}$ and $\{\psi_{nk}^{\text{shear}}\}$ are partitions of unity satisfying

$$\sup_{(n,k) \in \Lambda} |Q_{nk}|^{1/p-1} \|\mathcal{F}^{-1} \psi_{nk}\|_p < \infty \text{ for all } 0 < p \leq 1$$

- Under these conditions, a Theorem by Feichtinger and Groebner (1985) states that [the corresponding decomposition spaces must be equal, with equivalent norms.](#)



Next steps?

► Embedding Theorems

Theorem (Borup, Nielsen '06)

For $0 < p \leq \infty$, $0 < s, q < \infty$ and $s' = \frac{\max(1, 1/p) - \min(1, 1/q)}{2}$,

$$B_q^{s+1/(2q)}(L_p(\mathbb{R}^2)) \hookrightarrow \mathfrak{D}_q^s(L_p(\mathbb{R}^2)) \hookrightarrow B_q^{s-s'}(L_p(\mathbb{R}^2))$$



Next steps?

- Embedding Theorems

Theorem (Borup, Nielsen '06)

For $0 < p \leq \infty$, $0 < s, q < \infty$ and $s' = \frac{\max(1, 1/p) - \min(1, 1/q)}{2}$,

$$B_q^{s+1/(2q)}(L_p(\mathbb{R}^2)) \hookrightarrow \mathfrak{D}_q^s(L_p(\mathbb{R}^2)) \hookrightarrow B_q^{s-s'}(L_p(\mathbb{R}^2))$$

- Equivalence of Approximation Spaces

