Released to public domain under Creative Commons license type BY (https://creativecommons.org/licenses/by/4.0). Copyright (c) 2018 SINBAD consortium - SLIM group @ The University of British Columbia.



## Context

- "Fast Discrete Curvelet Transforms (FDCT)" by Candes, E., Demanet, L., Donoho, D., and Ying, L., explains the curvelet transform in detail
- "Wave Atoms and Sparsity of Oscillatory Patterns" by Demanet, and D., and Ying, L., explains the wave atom transform in detail
- 3-D curvelet extension by Ying.
- Courtesy Demanet (figures from his SIAM & IPAM talks)
- Check <u>http://www.curvelet.org</u>/ for curvelets
- Check for wave atoms <u>http://</u> <u>www.acm.caltech.edu/~demanet/software/</u> <u>WaveAtom.tar.gz</u>

#### OUR MOTIVATION

Devise a data representation scheme that

- is non-parametric/non-adaptive
- truly exploits the 3-D continuity along wavefronts in  $\mathbf{d}(r, s, t)$
- is noise resilient
- exploits redundant frames
- is  $n \log n$
- deals with *intermittent* regularity!



# Phase space tilings

Need a representation that is

- Iocal in phase space (space-wave number)
- directional

Phase space is too big ... need a tiling & sampling

#### Many possibilities

Need to preserve directivity of wavefronts



SLIM



## Phase space tilings

Phase-space:  $(x, \xi) = (\text{position, frequency})$ . Here  $x \in \mathbb{R}^2$ . Covered by a family of wave packets, indexed by  $\mu = (j, k_1, k_2, \ell, m)$ :

$$\begin{split} \varphi_{\mu}(x) &\simeq 2^{(\alpha+\beta)j/2} \cos((pR_{\theta_{j\ell}}x-k)\cdot(1,0))\varphi(D_jR_{\theta_{j\ell}}x-k)\\ \text{some profile } \phi(x) \text{ like } e^{-x^2/100}, \text{ wave vector } p = 2^j + m2^{\alpha j}, \end{split}$$

 $D_j = \begin{pmatrix} 2^{\alpha j} & 0\\ 0 & 2^{\beta j} \end{pmatrix}$ 

$$\theta_{i\ell} = \ell 2^{(\beta-1)j}$$

SLIM

2 parameters:  $\alpha$  (multiscaleness) and  $\beta$  (aspect ratio).

[From Demanet '05]

for





[From Demanet '05]





















## Other choices

• Directional wavelets:  $\alpha = 1, \beta = 1$ 

 $\varphi_{\mu}(x) = 2^{j} \psi(2^{j} R_{\theta_{j\ell}} x - k)$ 

- with  $\psi(x) = \cos(x_1)\phi(x)$  some profile, oscillatory in one direction.
- Curvelets/contourlets:  $\alpha = 1, \beta = 1/2$

$$\varphi_{\mu}(x) = 2^{3j/4} \psi(\text{diag}(2^{j}, 2^{j/2}) R_{\theta_{j\ell}} x - k)$$

with  $\psi(x) = \cos(x_1)\phi(x)$ .

[From Demanet '05]







~











































# Observations

Different tilings of phase space Given the multiscale and multidirection behavior of seismic data and images

- multiscale partitioning
- multidirectional partitioning

Invariance under FIO's (high-freq. solution operators of the wave equation)

- parabolic scaling principle
- two candidates
  - curvelets
  - wave atoms

How does this work?



































### **Observations**

Curvelets are sparse on distinct **high-frequency** wavefronts

Wave atoms are sparse on **oscillatory** wavefronts Both redundant with fast relative decay

Wave atom simpler to implement & less redundant. Both can unstructured.

Which one will perform better?

seismic data holds middle between high-freq. and "single" frequency

subsurface contains distinct singularities

Both invariant under FIO's/migration ...











