

# Wavelets, Ridgelets and Curvelets on the Sphere

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Emmanuel Candes, David Donoho*

## 1) Motivation for a Curvelet Transform on the Sphere

### - The Cosmic Microwave Data Story

====> Data restoration

====> Detection of cosmological non Gaussian signature

## 2) The Curvelet Transform

### - Comparison of several Curvelet implementations for denoising

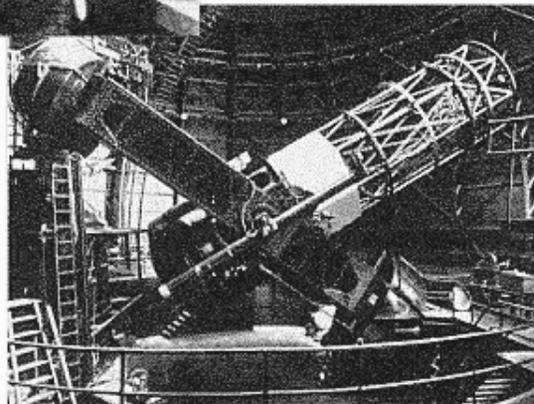
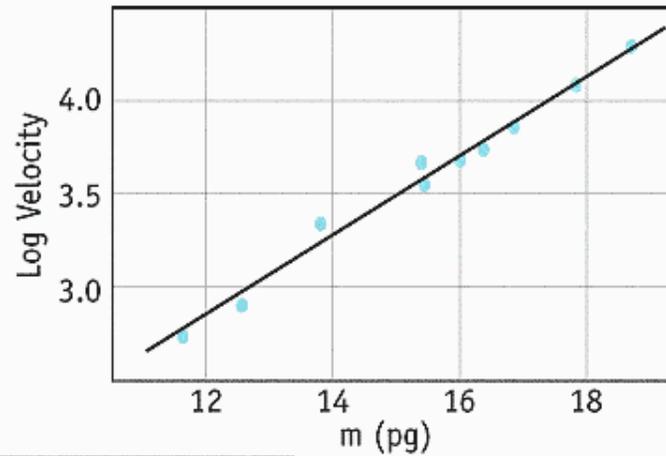
## 3) Wavelet, Ridgelet and Curvelet Transforms on the Sphere

# The Big Bang

## DISCOVERY OF EXPANDING UNIVERSE



Edwin Hubble

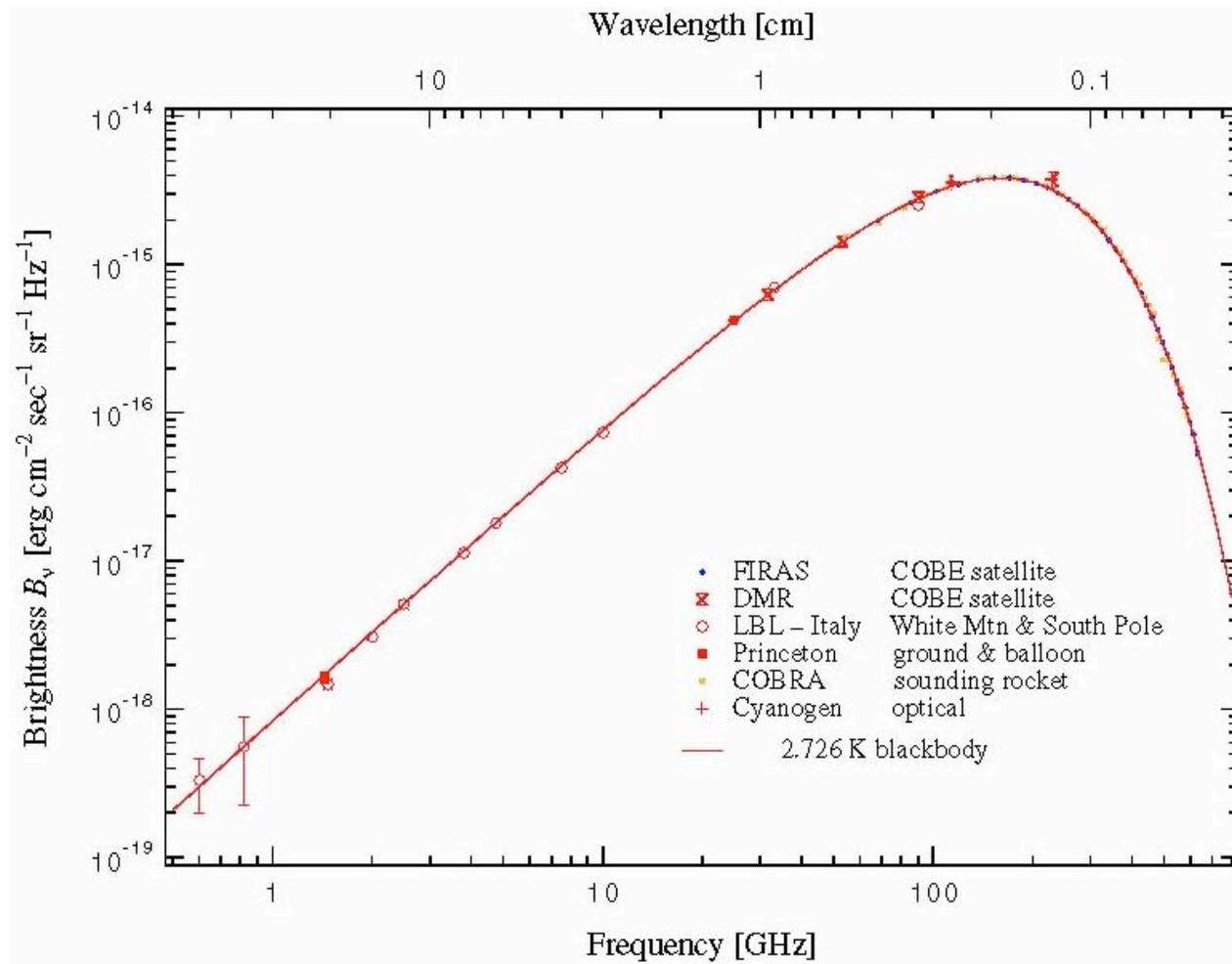


Mt. Wilson  
100 Inch  
Telescope

# The Cosmic Microwave Background

- The Universe is filled with a blackbody radiation field at a temperature of 3K.
- Predicted by G. Gamow in 1948
- Observed for the first time by Penzias and Wilson (1965)
- Confirmed by COBE (1990)

# The Cosmic Microwave Background



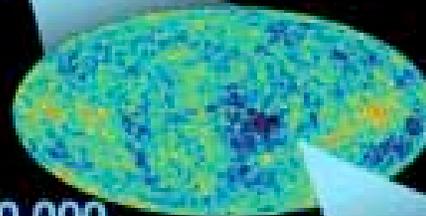
**DAWN  
OF  
TIME**

**tiny fraction  
of a second**



**inflation**

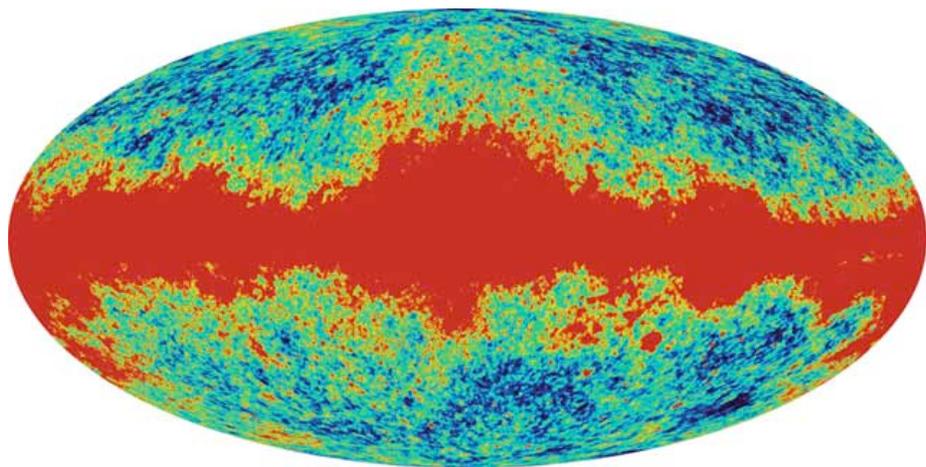
**380,000  
years**



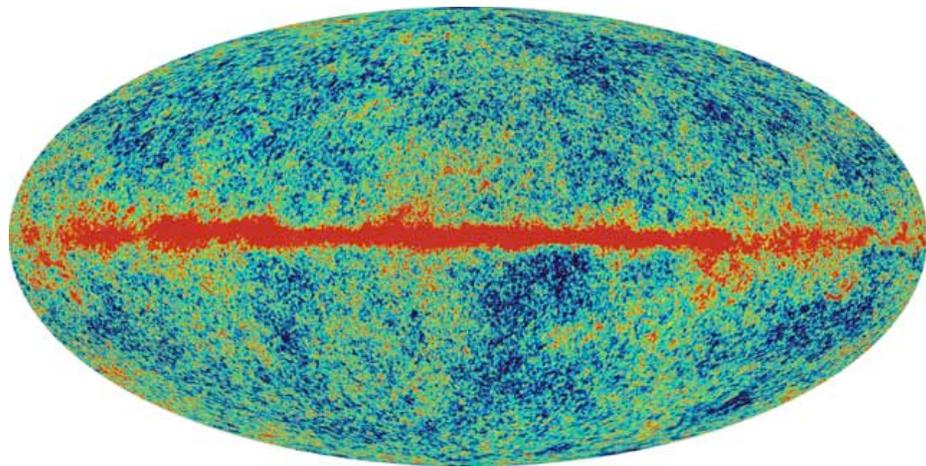
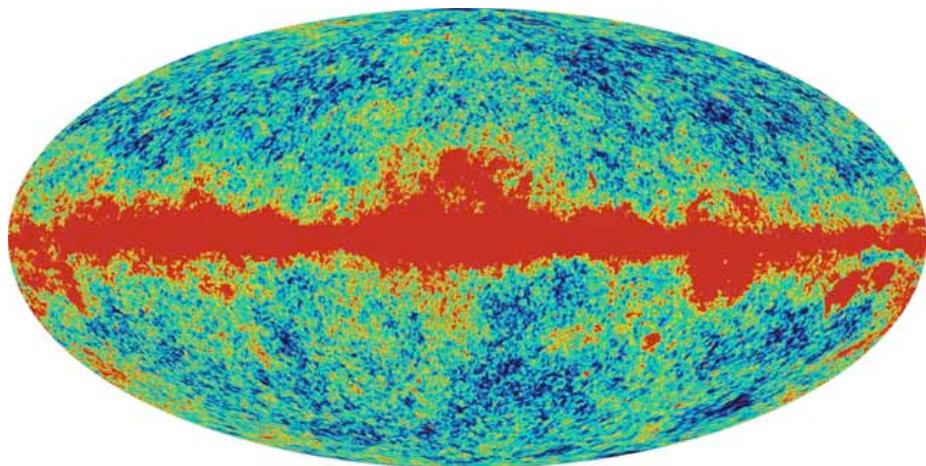
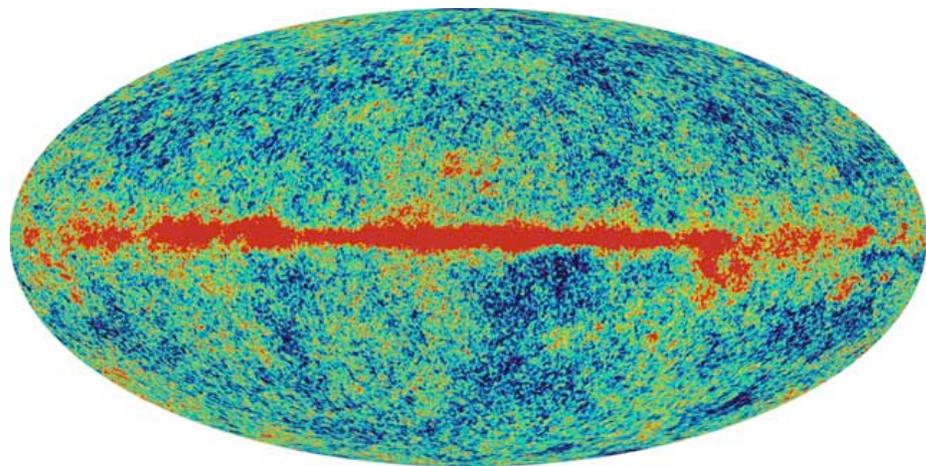
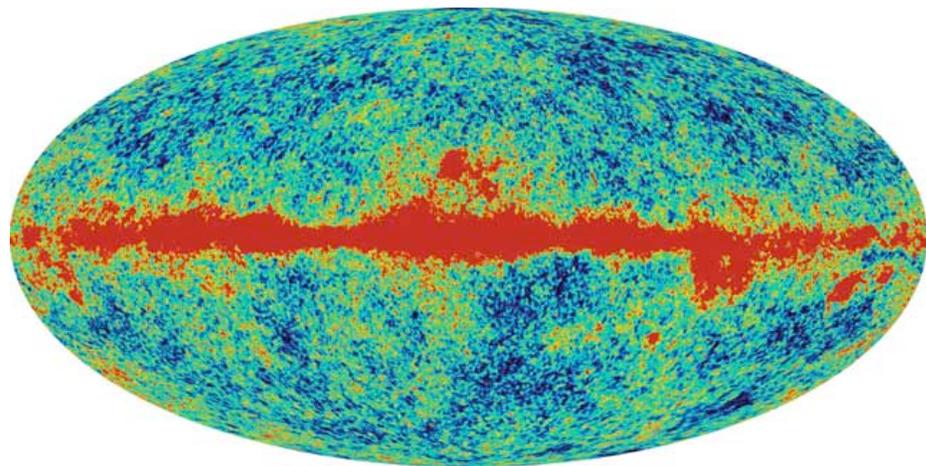
**13.7  
billion  
years**



## WMAP: five frequency maps



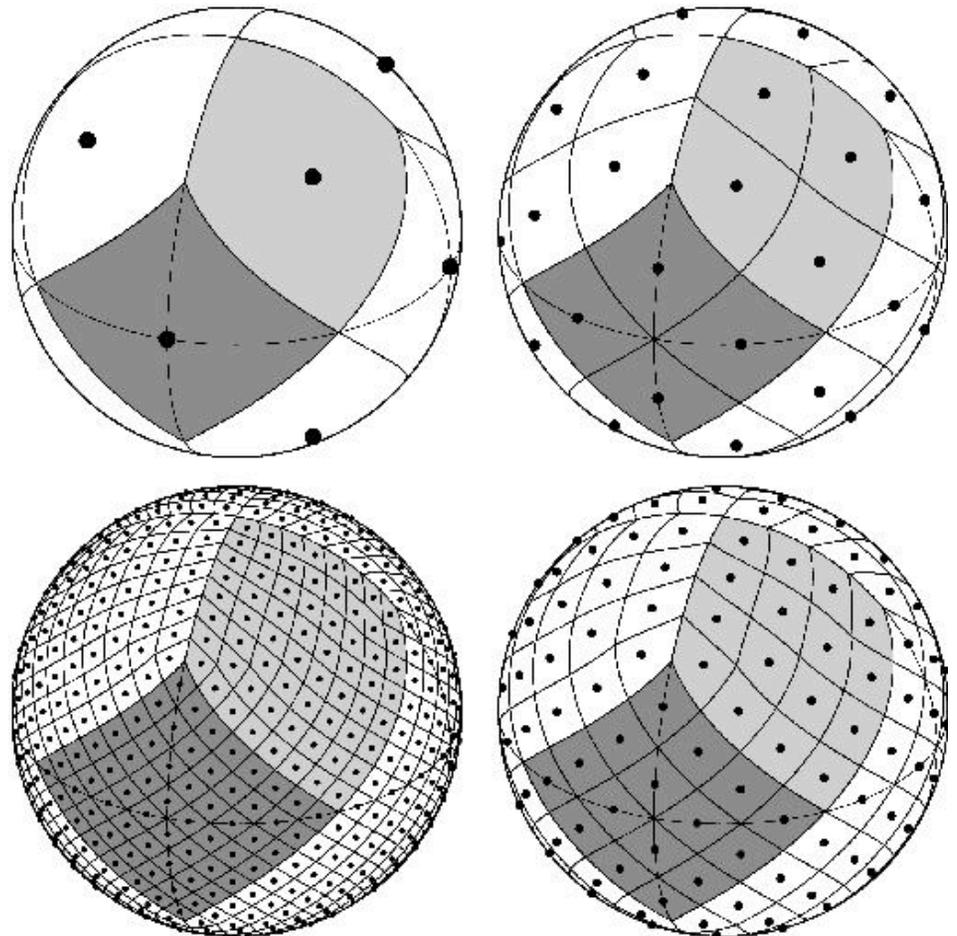
23, 33, 41, 61 and 93 GHz

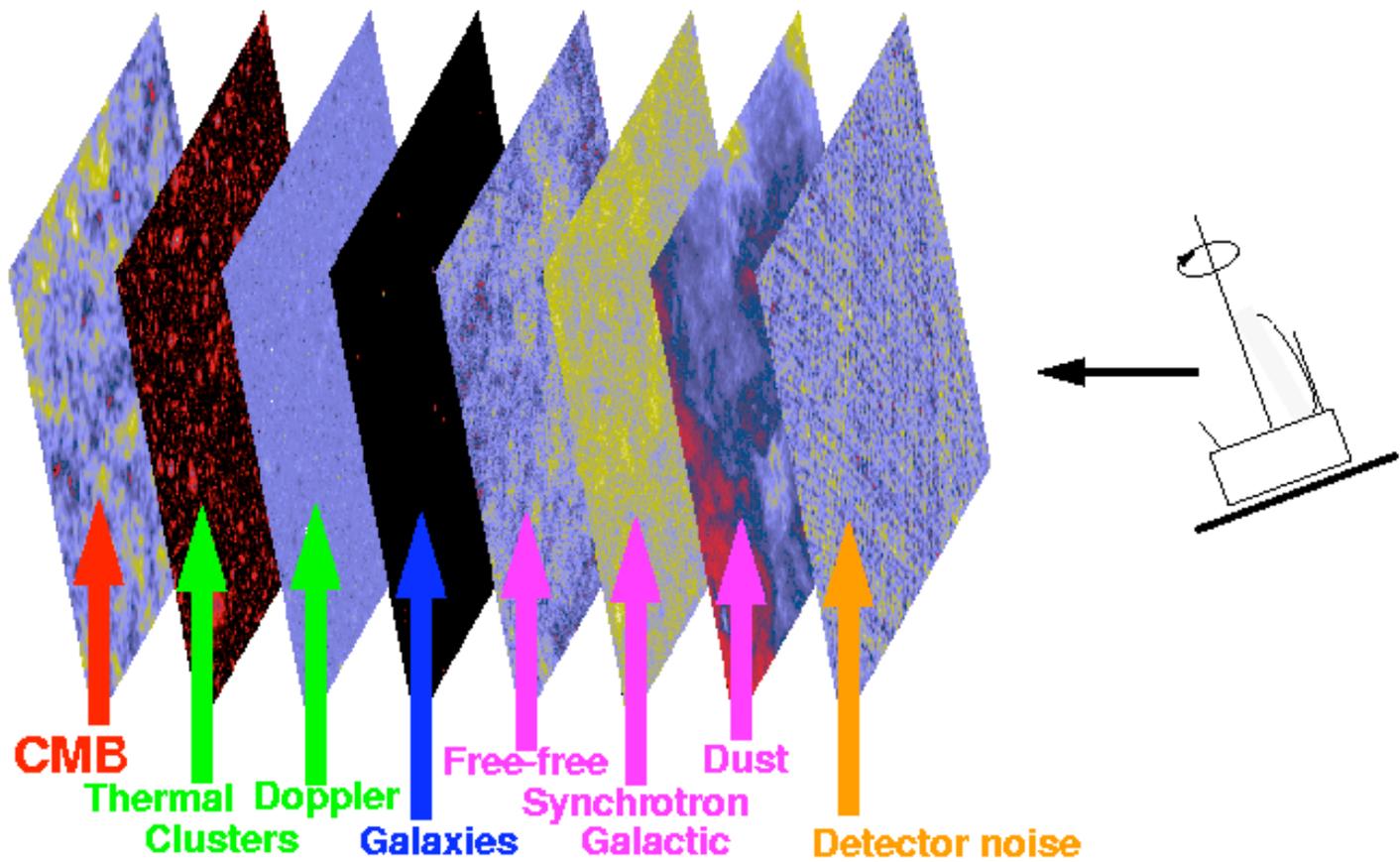


# Healpix

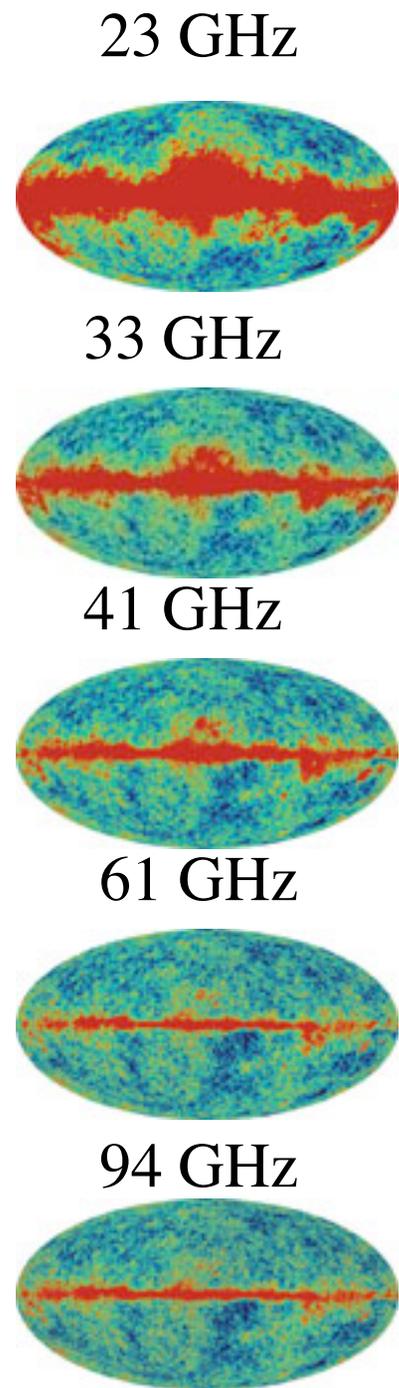
K.M. Gorski et al., 1999, astro-ph/9812350,  
<http://www.eso.org/science/healpix>

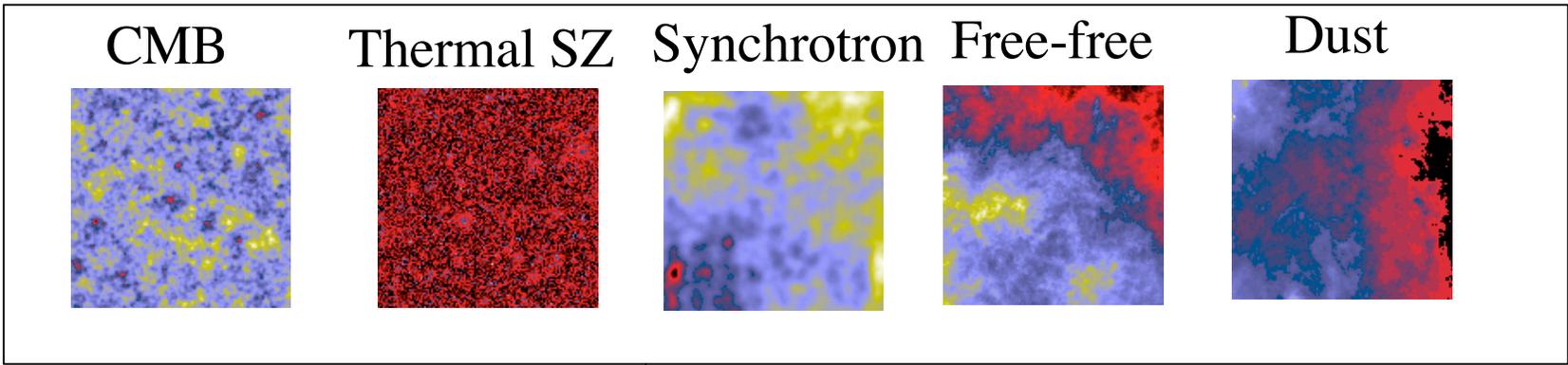
- Pixel = Rhombus
- Same Surfaces
- For a given latitude :  
regularly spaced
- Number of pixels:  
 $12 \times (N_{\text{sides}})^2$
- Included in the software:
  - Anafast
  - Synfast





Synchrotron emission due to cosmic rays electrons accelerated into galactic magnetic fields

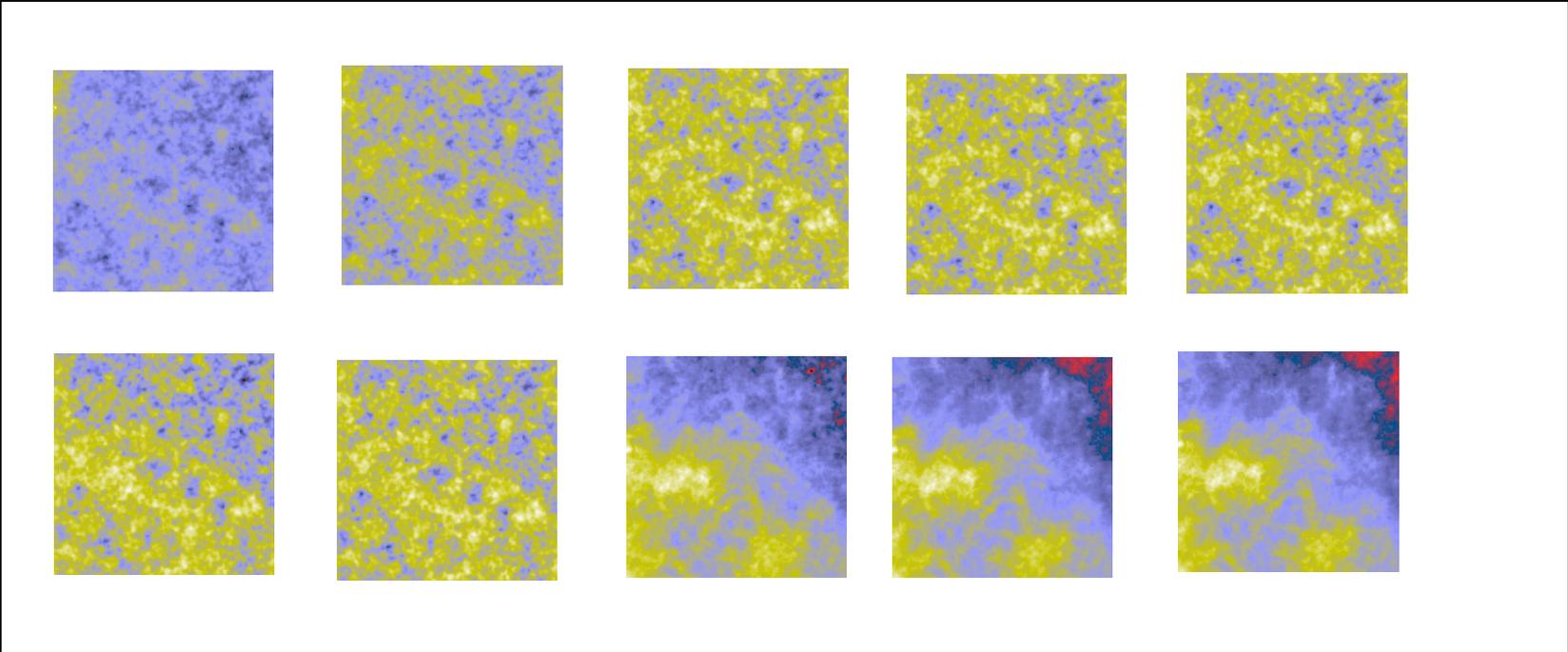




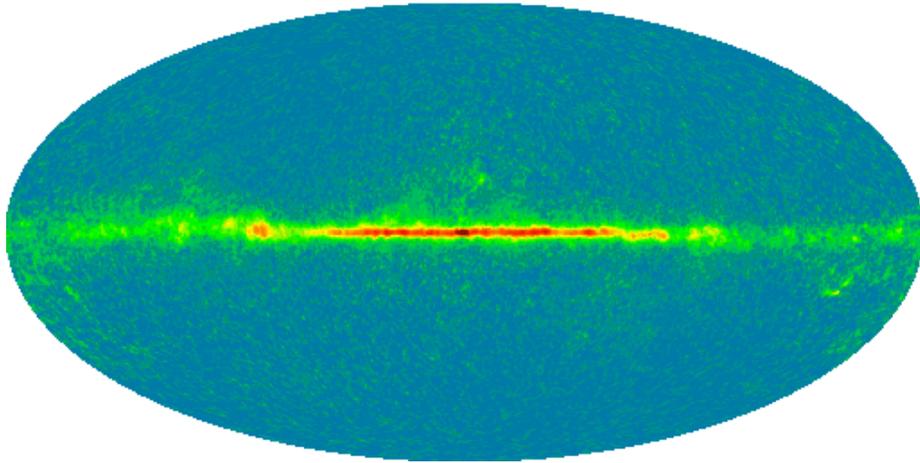
Sky components

Linear combination + PSF + Noise

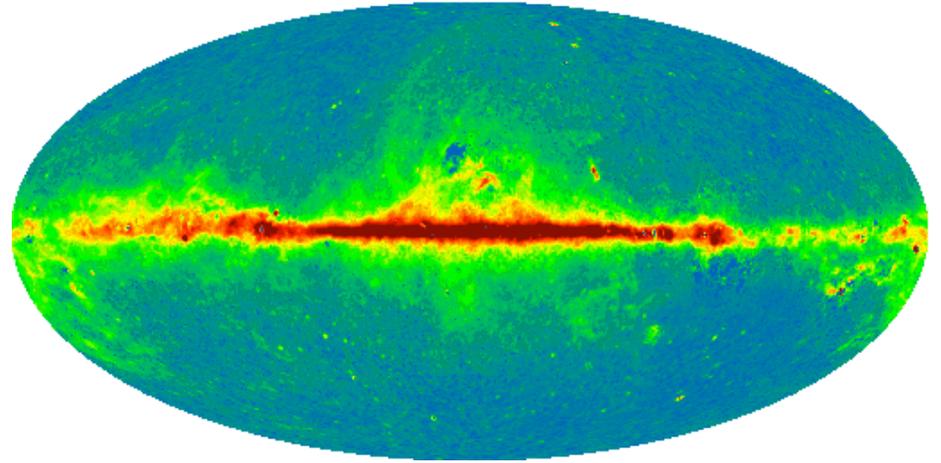
Observations



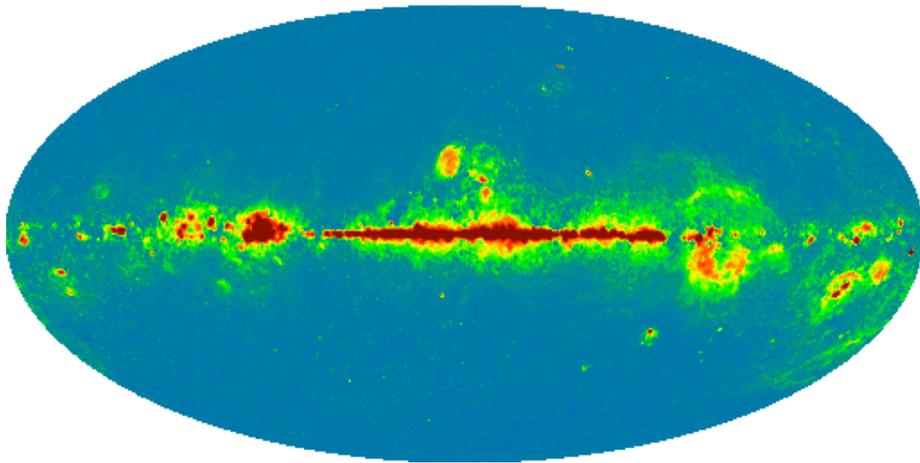
## WMAP derived sky maps



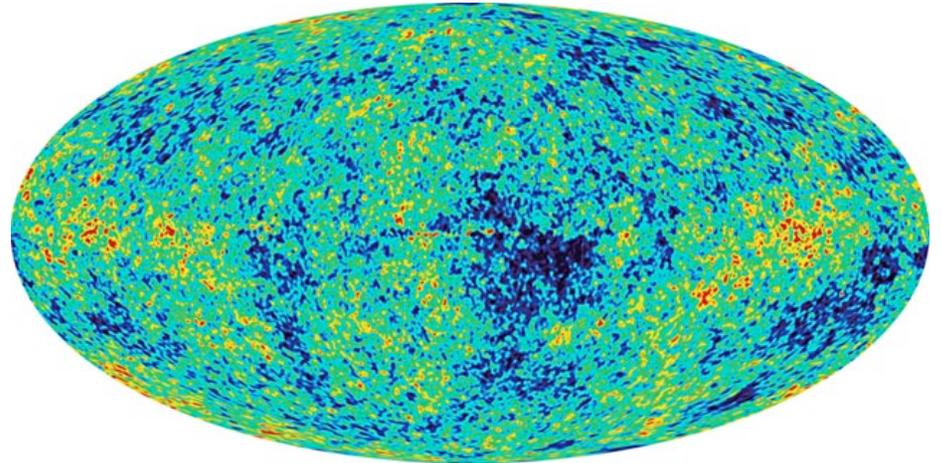
Dust map



Synchrotron map

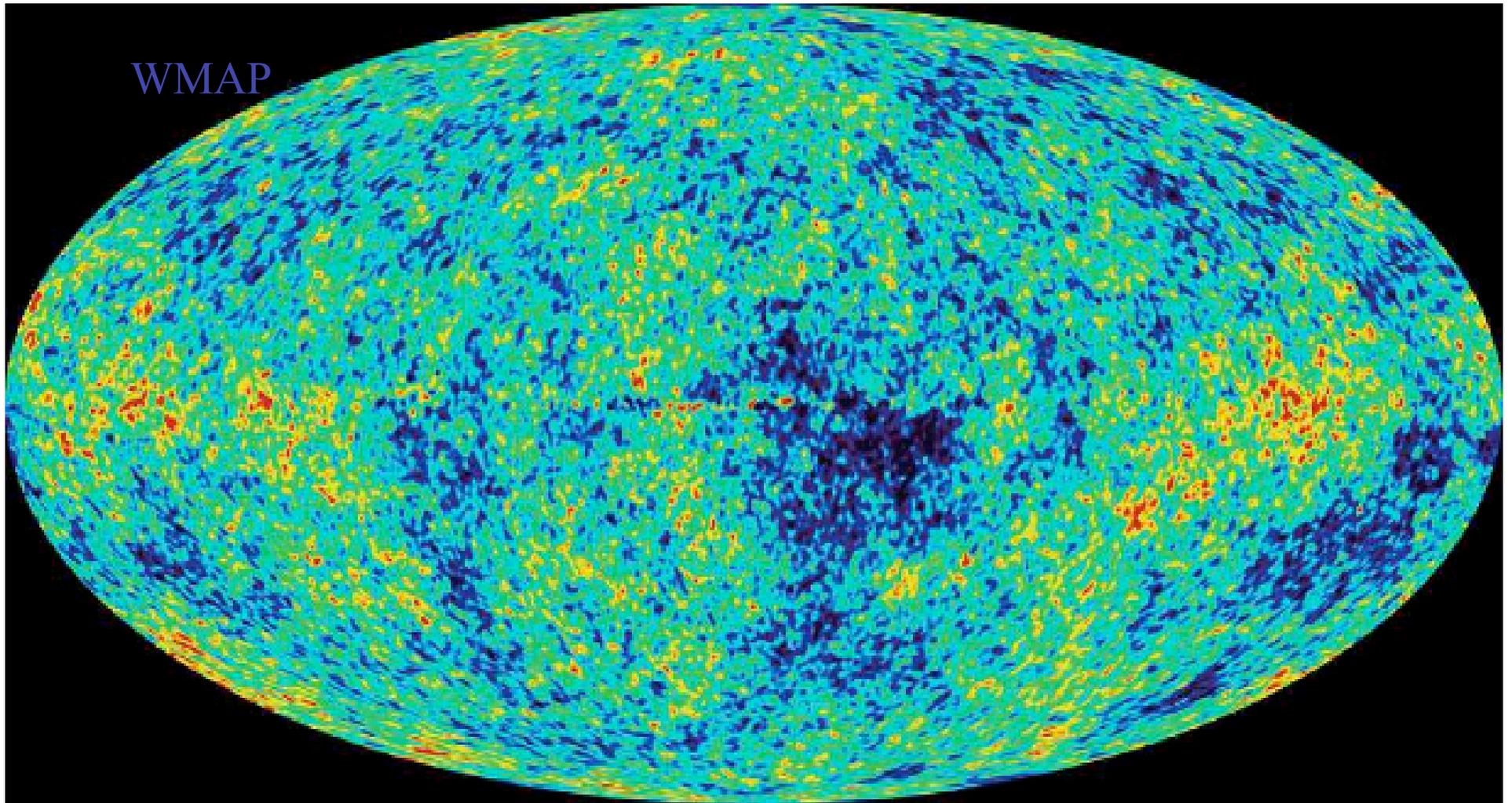


Free-free map

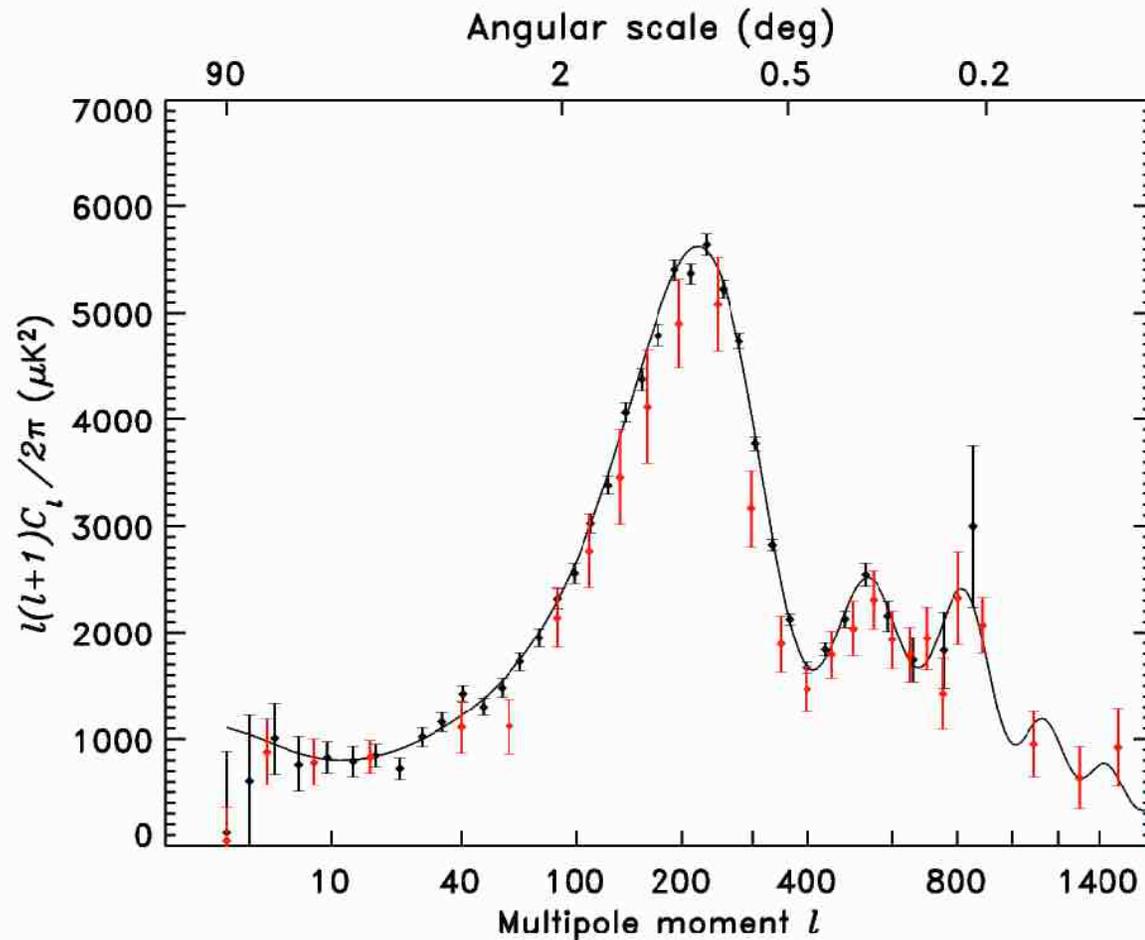


CMB map

# The CMB exhibits Fluctuations



# Power spectrum of WMAP



Remarkably consistent with earlier data

# The Cosmic Microwave Background

- The power spectrum gives constraints on the geometry and the physical state of the Universe.
- It supports the hypothesis of a period of rapid expansion of the early Universe: the Inflationary period.

## The Primary fluctuations

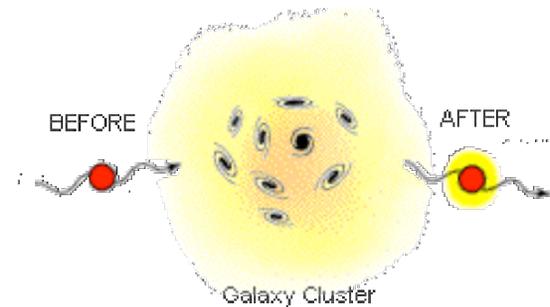
- This process causes randomly distributed seeds and Gaussian distributed fluctuations.
- At the end of this period, topological defects may occur that produce non-Gaussian fluctuations e.g. Cosmic Strings.

## The Secondary fluctuations

The secondary fluctuations arise from the interaction of the CMB photons with the cluster of galaxies. It is the Sunyaev Zel'dovich (SZ) effect.

# The Sunyaev Zel'dovich effect

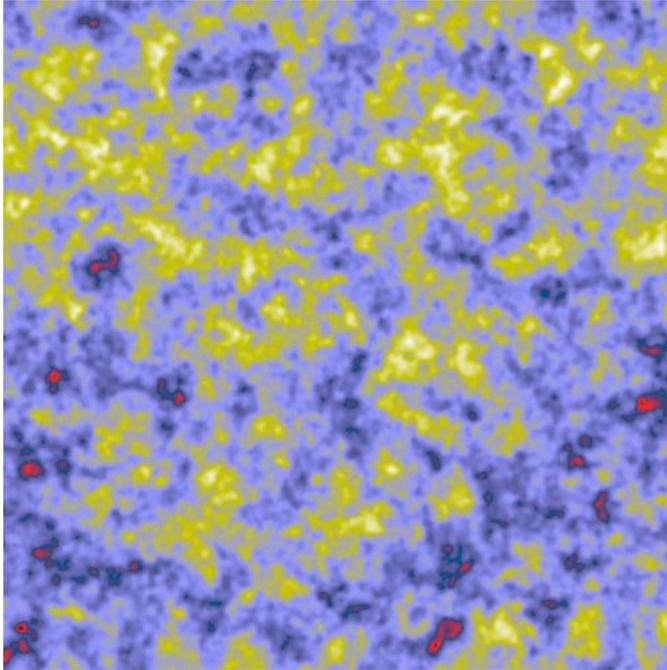
- First predicted by the Russian scientists Sunyaev and Zel'dovich in 1969.
- Galaxy Clusters have hot gas
  - $T_{\text{gas}} \sim 10\text{-}100$  million Kelvin
  - Electron scattering from nuclei produces X-rays, thermal bremsstrahlung.
- Compton scattering occurs between CMB photons and the hot electrons
  - $\sim 1\%$  of CMB photons will interact with the hot electrons
    - Energy will be transferred from the hot electrons to the low energy CMB photons, changing the shape of their intensity vs. frequency plot.
      - Measurements made at low frequencies will have a lower intensity, since photons which originally had these energies were scattered to higher energies. This distorts the spectrum by  $\sim 0.1\%$ .
- Estimates of cosmological parameters (ie.  $H_0$  and  $\Omega_b$ ) can be made by combining these measurements.



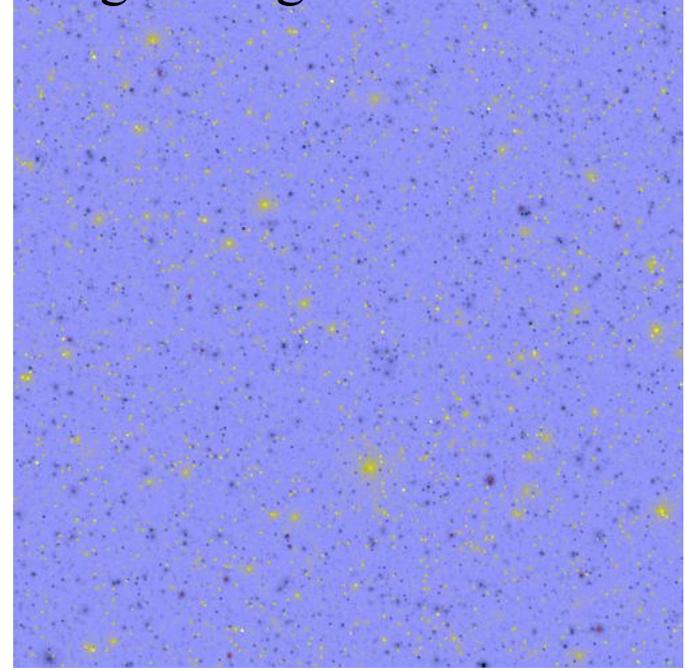
The SZ effect comes from the interaction of the cold CMB photons with the hot electrons (TSZ) of moving (KSZ) galaxy clusters

# Detection of non-Gaussian Cosmological Signatures

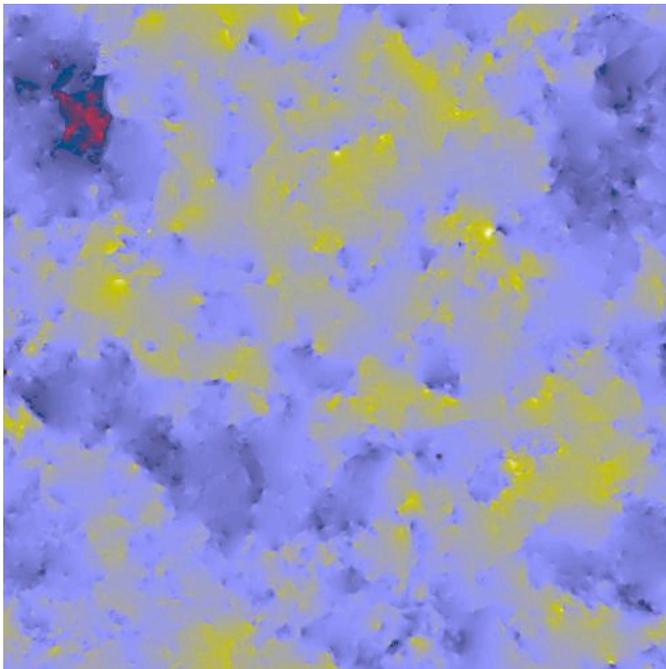
CMB



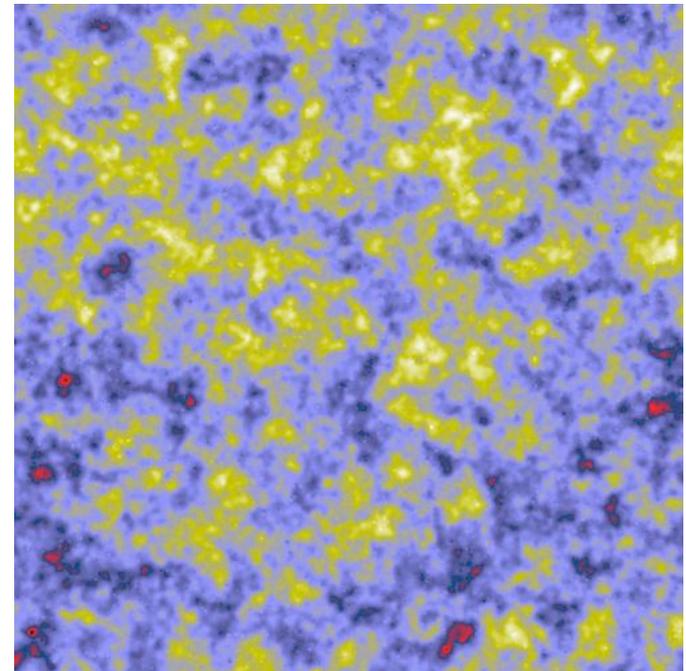
SZ

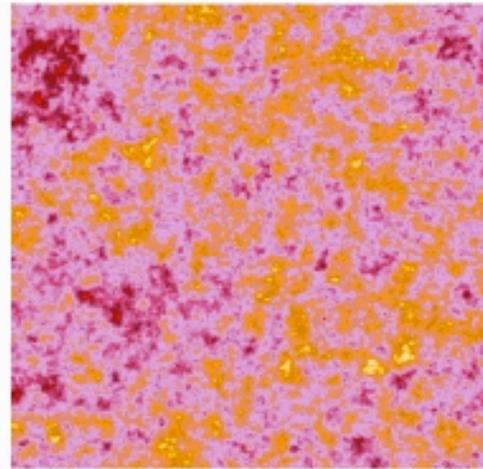
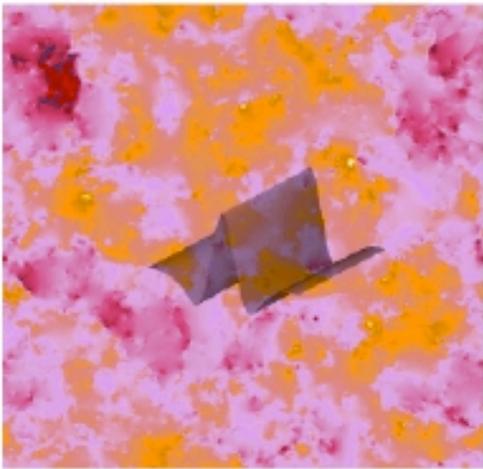
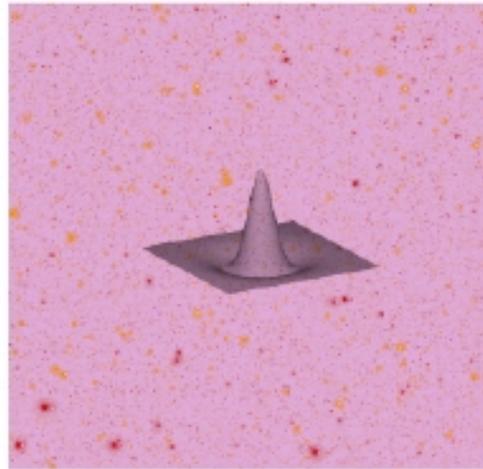
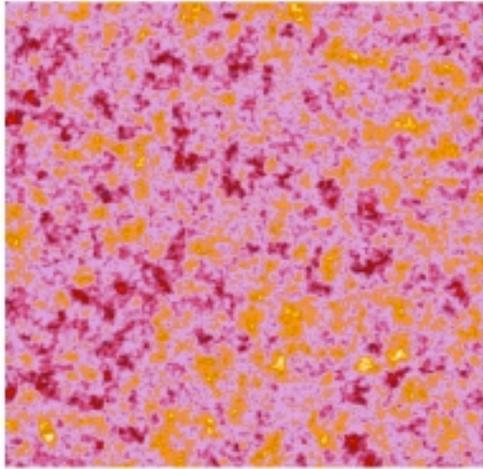


CS



Total

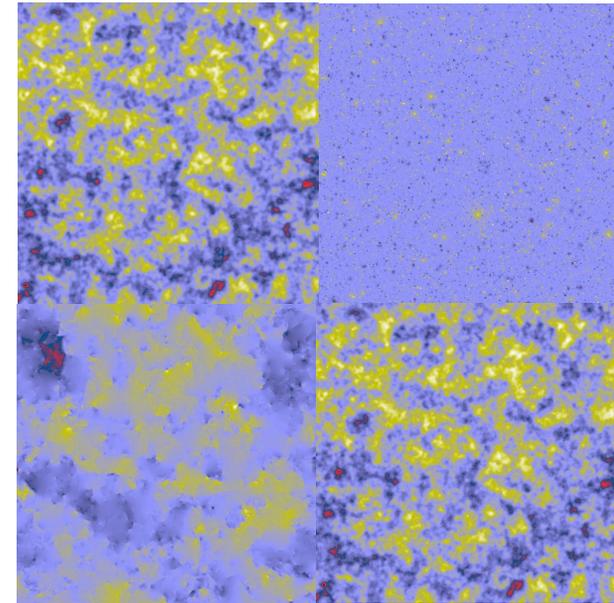




# Multiscale Analysis of the CMB

We have applied the following multiscale transforms

- Isotropic wavelet transform
- Bi-orthogonal wavelet transform
- Ridgelets (block size of 16 pixels)
- Ridgelets (block size of 32 pixels)
- Curvelets



On

1) 100 **CMB + KSZ** + 100 Gaussian realizations with the same power spectrum.

$$K_{CMB-SZ}(i,b) \Rightarrow K_{CMB-SZ}(b) = \text{mean}(K_{CMB-SZ}(1..100,b)), \bar{K}_{CMB-SZ}(b) = \frac{K_{CMB-SZ}(b)}{K_{CMB}(b)}$$

2) 100 **CMB + CS** + 100 Gaussian realizations with the same power spectrum

3) 100 **CMB + KSZ + CS** + 100 Gaussian realizations with the same power spectrum

We compare the normalized kurtosis for the three data set.

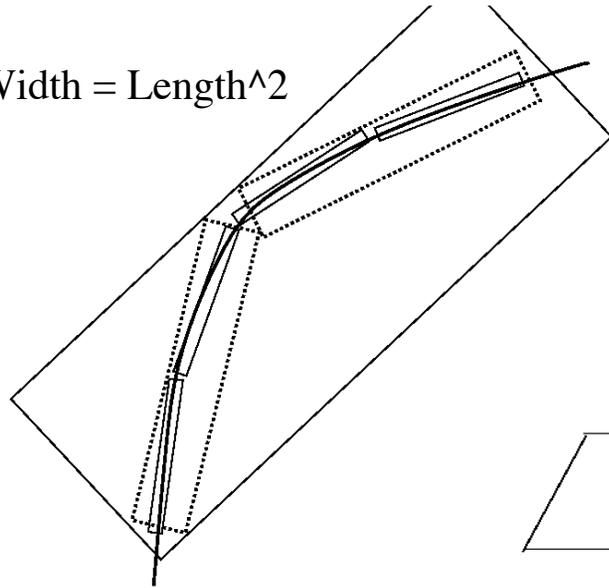
# Results

- **Curvelets are NOT sensitive to KSZ but are sensitive to cosmic strings**

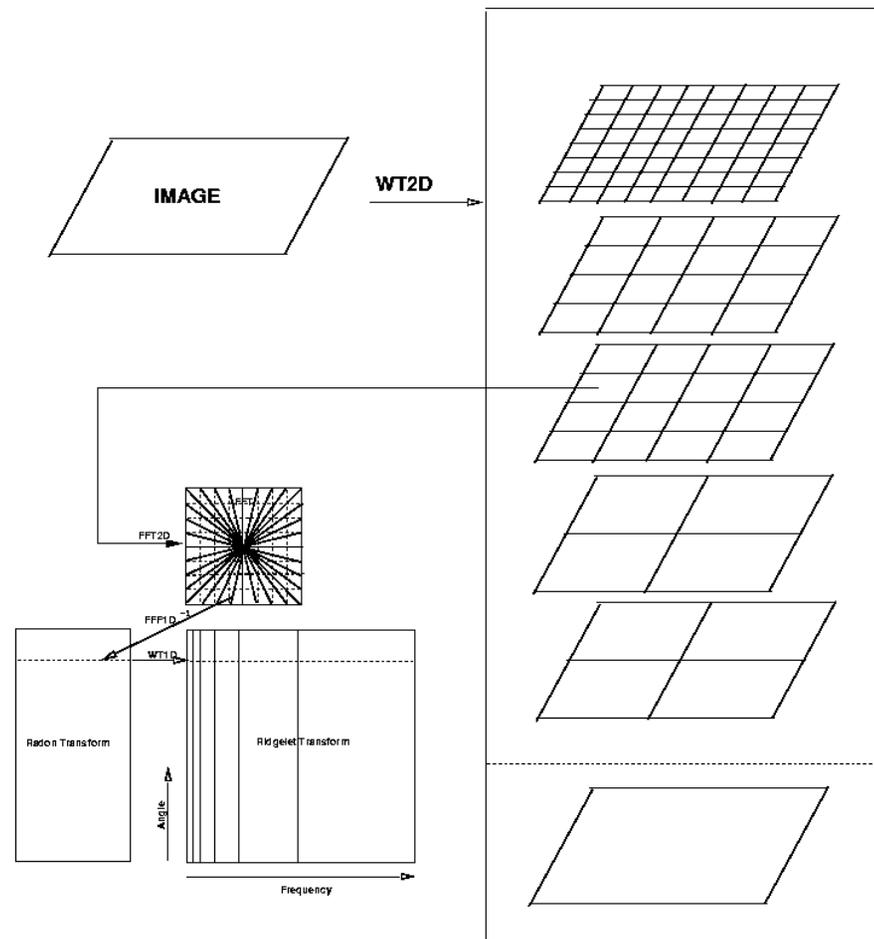
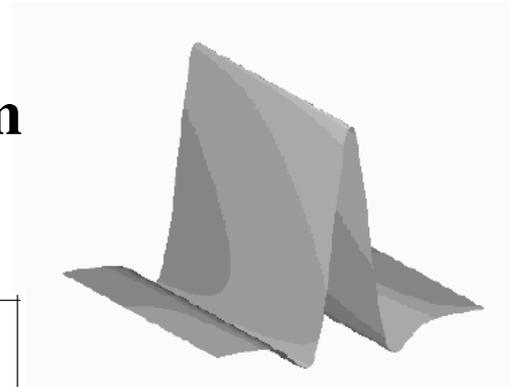
	Bi-orthogonal WT	Ridgelet	Curvelet
CMB+KSZ	1106.	0.1	10.12
CMB+CS	1813.	5.7	198.
CMB+CS+KSZ	1040.	5.9	165.

*Detecting cosmological non-Gaussian signatures by multi-scale methods, Astron. and Astrophys., 416, 9--17, 2004 .*

Width = Length<sup>2</sup>

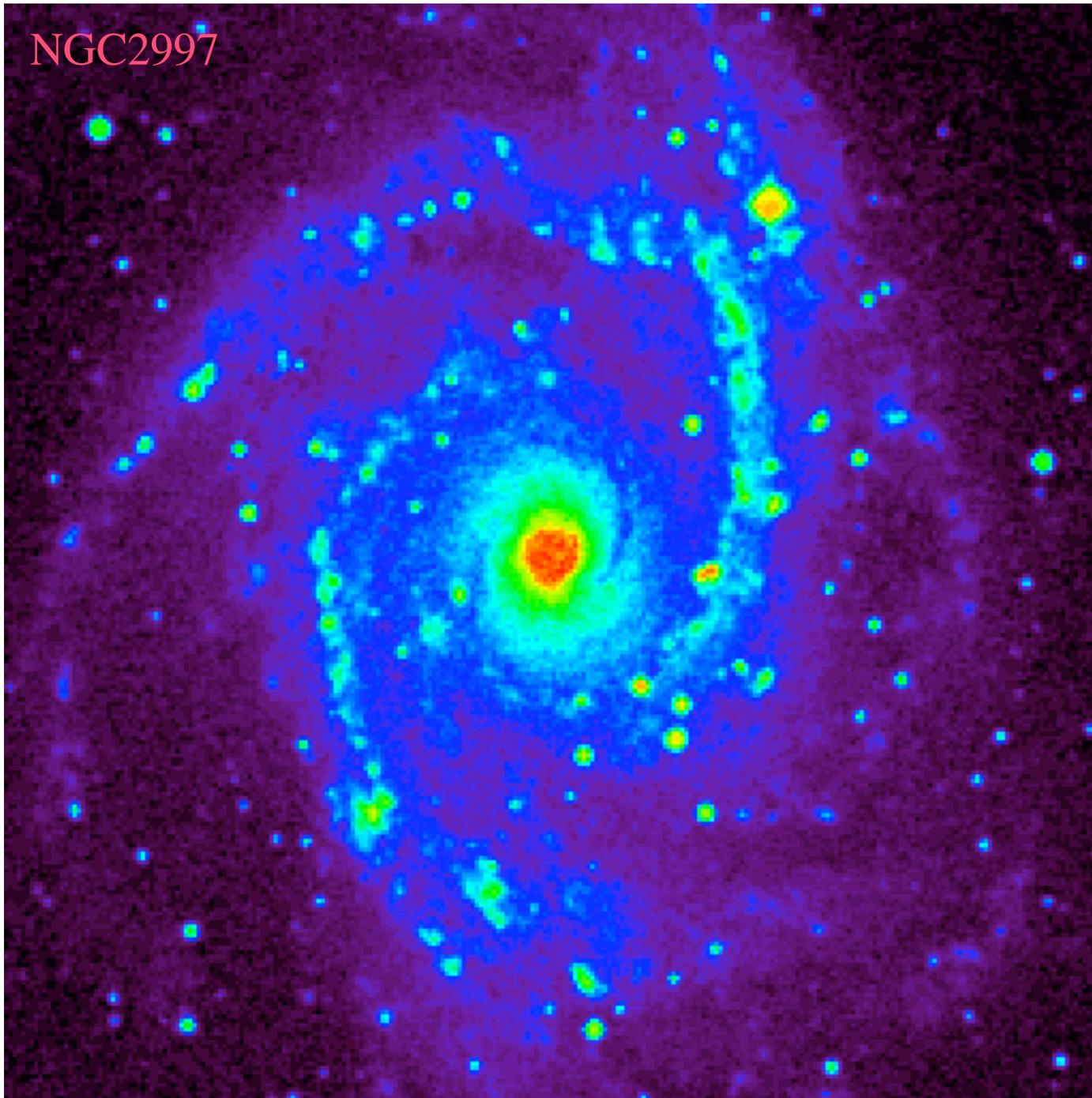


# The Curvelet Transform

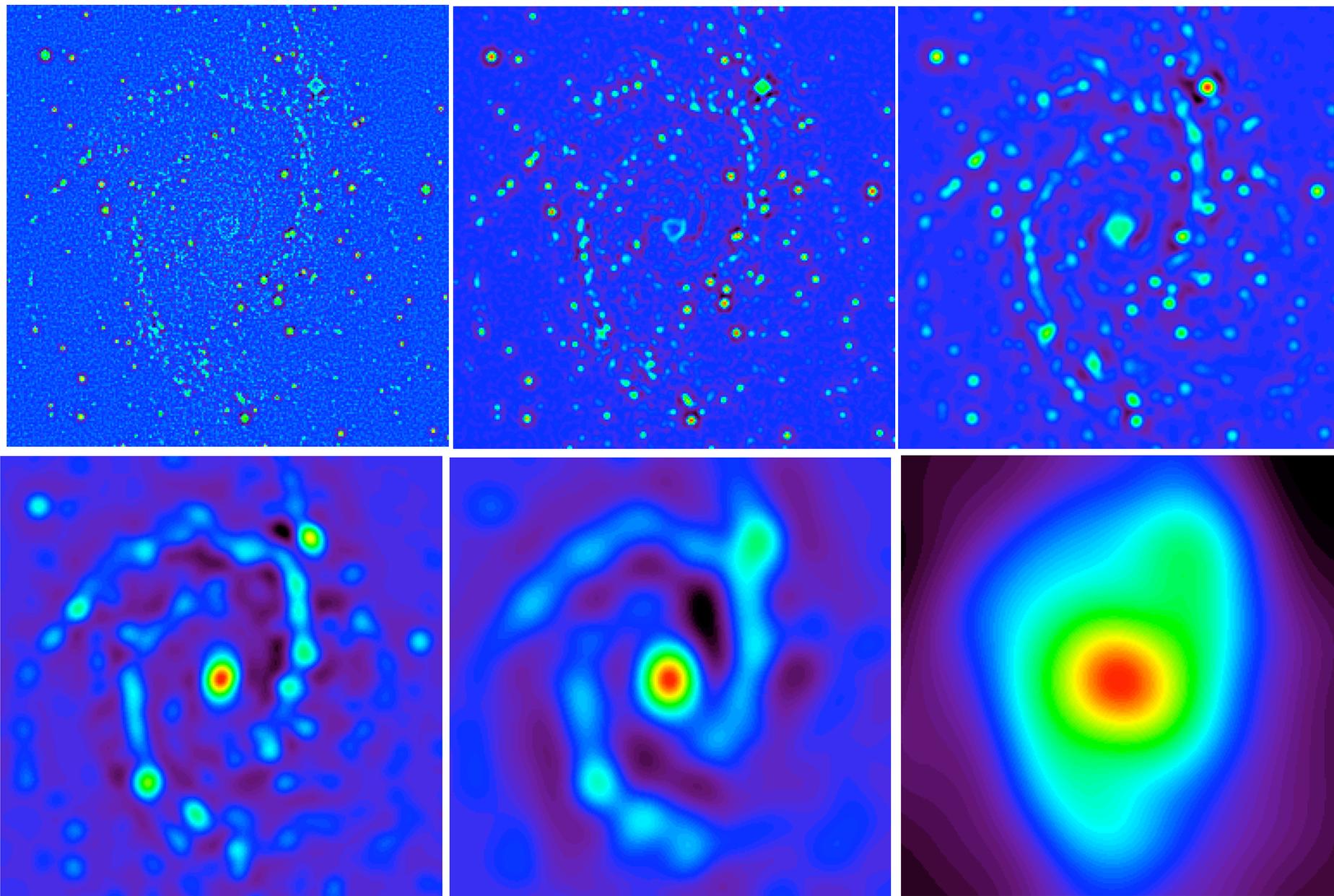


*The Curvelet Transform for Image Denoising, IEEE Transaction on Image Processing, 11, 6, 2002.*

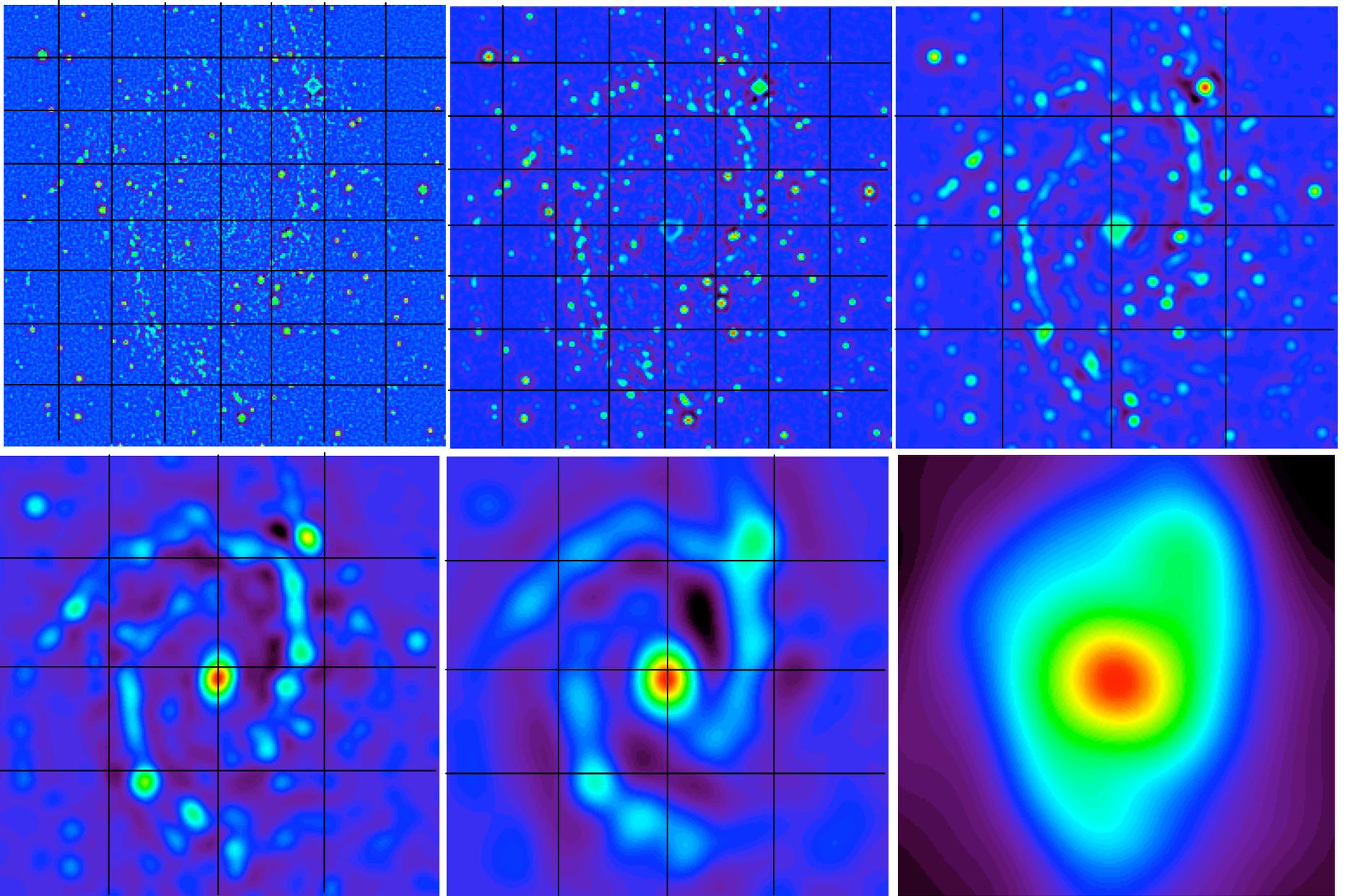
NGC2997



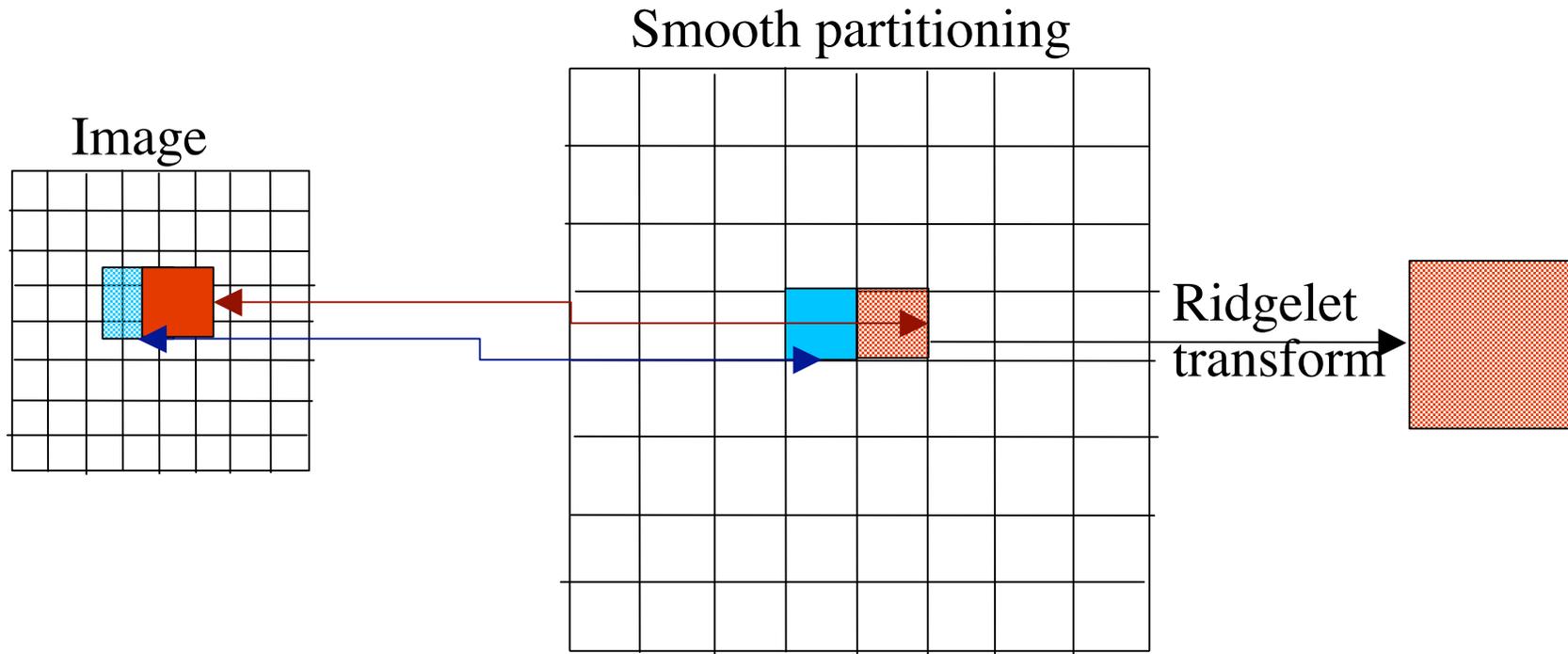
**Undecimated Isotropic WT:**  $I(k, l) = c_{J,k,l} + \sum_{j=1}^J w_{j,k,l}$



# PARTITIONING



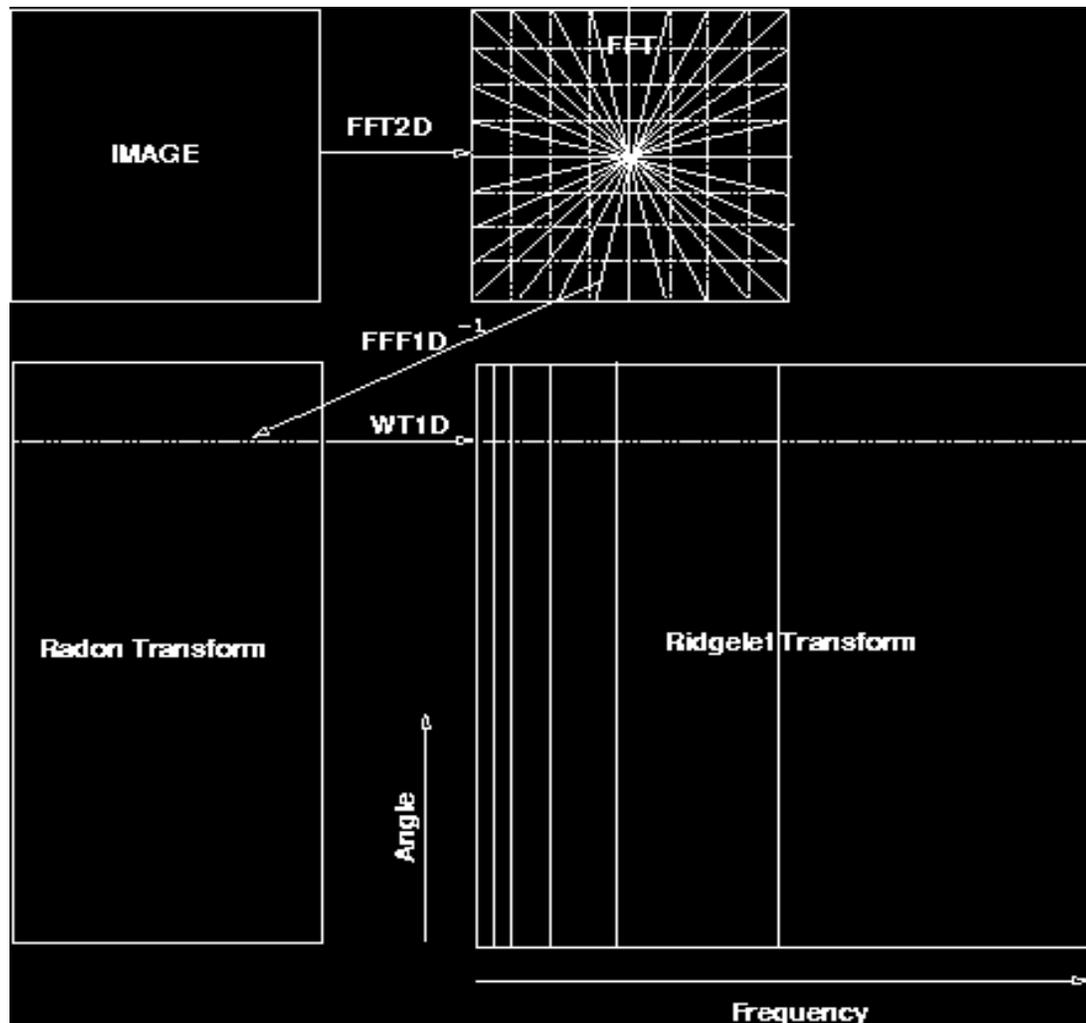
# LOCAL RIDGELET TRANSFORM



The partitioning introduces a redundancy, as a pixel belongs to 4 neighboring blocks.

The ridgelet coefficients of an object  $f$  are given by analysis

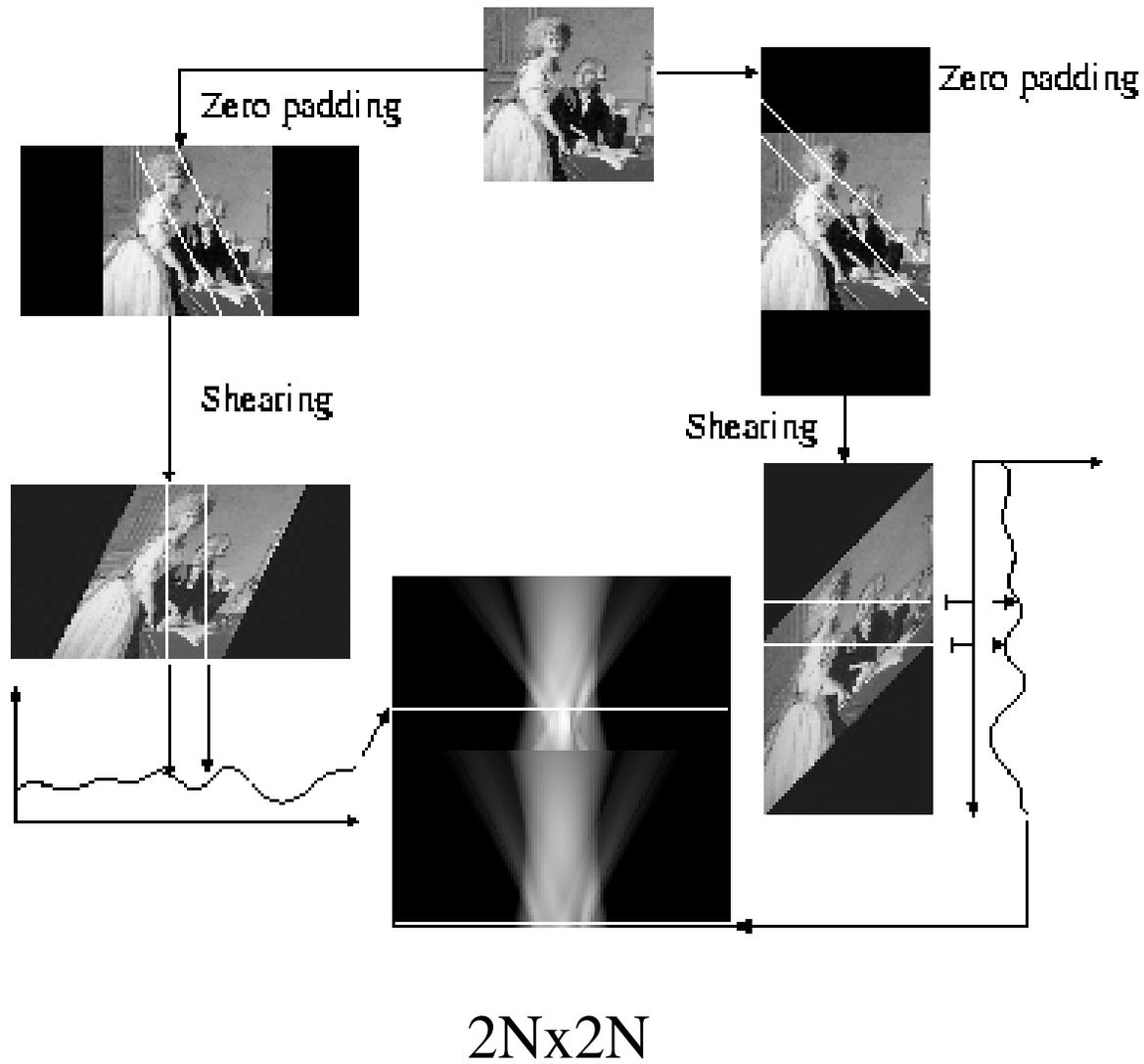
of the Radon transform via: 
$$R_f(a,b,\theta) = \int Rf(\theta,t)\psi\left(\frac{t-b}{a}\right)dt$$

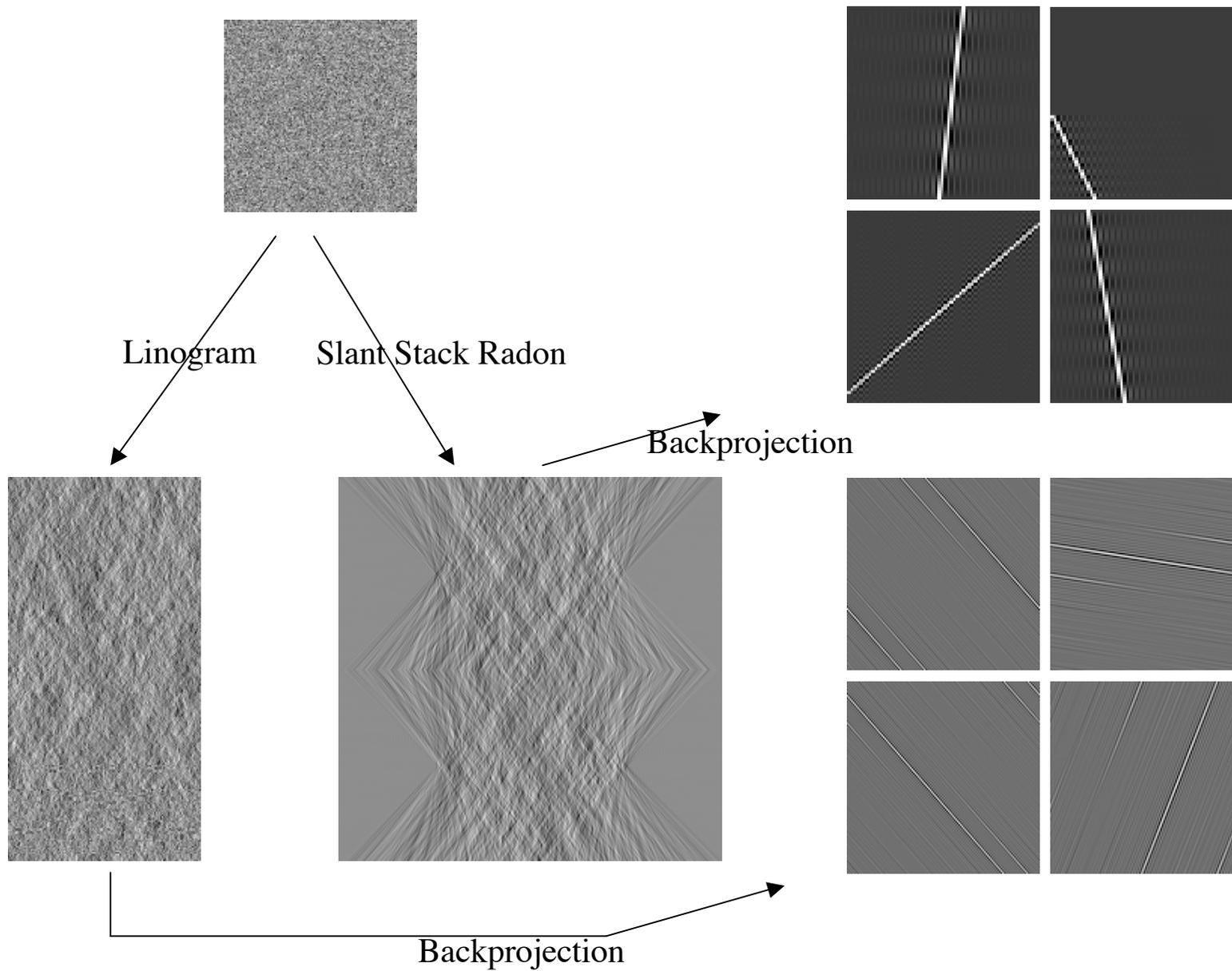


# Curvelet Transform Implementations

1. CUR99 (Donoho, Candes, Duncan, 1999)  
Radon (linogram) + WT1D
2. CUR01 (Starck, Candes, Donoho, 2001)  
Radon (linogram) + WT1D
3. CUR03 (Candes, Donoho, 2003)  
In Fourier space using USFFT
4. CUR04 (Demanet, Candes, 2004)  
In Fourier space using warping

# Slant Stack Radon Transform (Averbuch et al, 2001) or Linogram ?





**CUR01+UWT, PSNR=32.10**



# Combined Filtering

*Very High Quality Image Restoration*, in *Signal and Image Processing IX*, San Diego, 1-4 August, 2001,  
Eds Laine, Andrew F.; Unser, Michael A.; Aldroubi, Akram, Vol. 4478, pp 9-19, 2001.

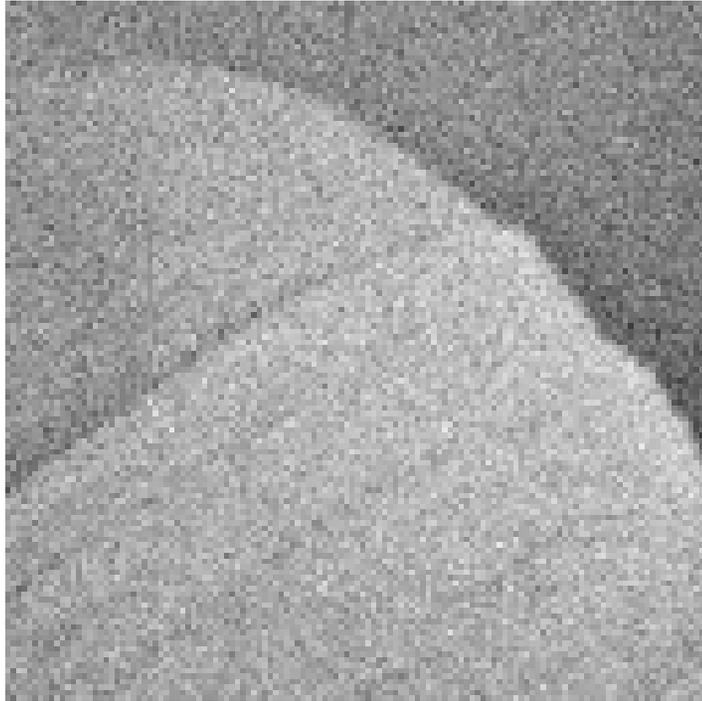
$$\min \quad \text{Complexity\_penalty}(\tilde{s}), \quad \text{subject to } \tilde{s} \in C$$

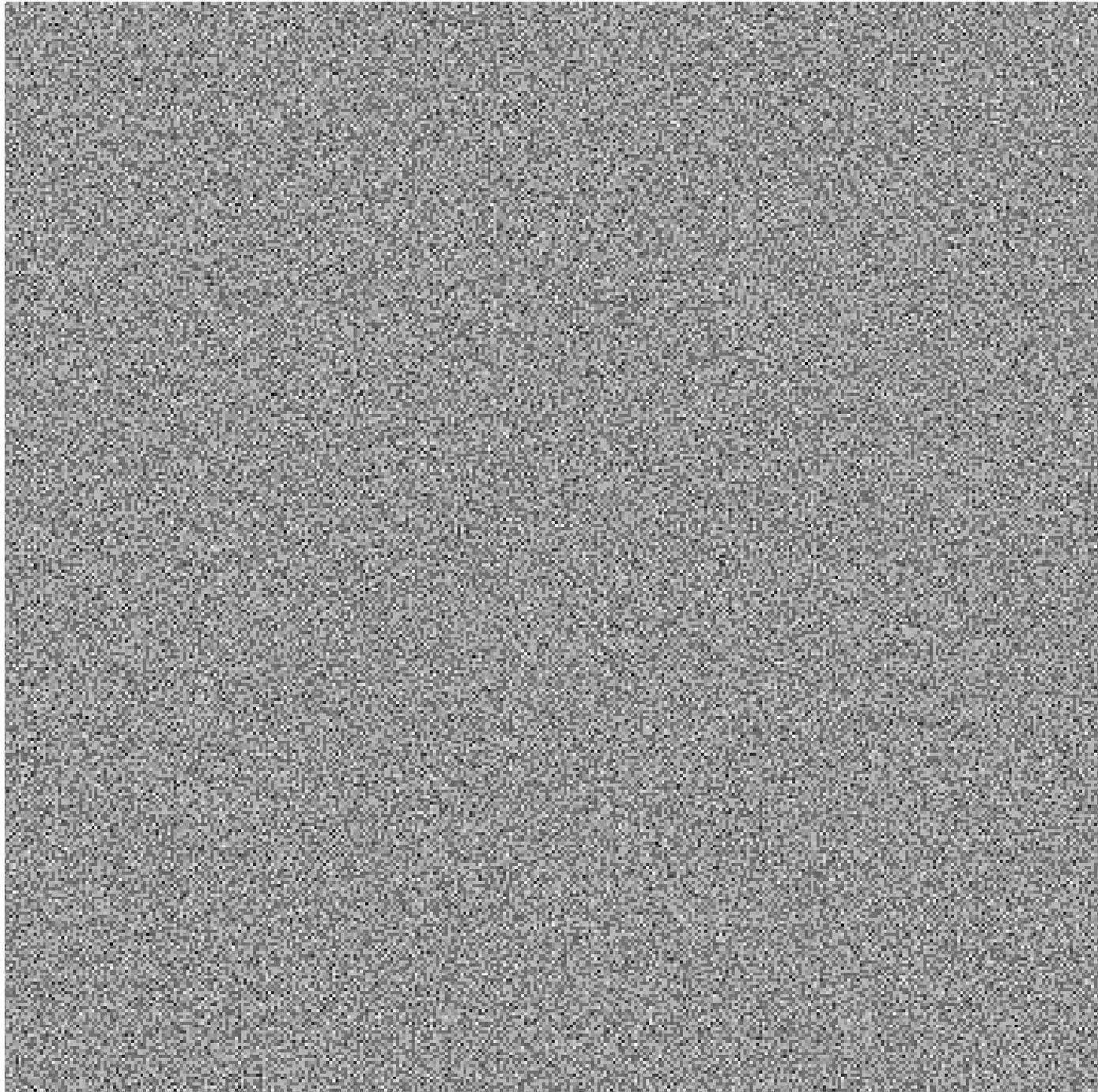
Where  $C$  is the set of vectors which obey the linear constraints:

$$\tilde{s} > 0, \quad \text{positivity constraint}$$

$$\left| (T_k \tilde{s} - T_k s)_l \right| \leq e, \quad \text{if } (T_k s)_l \text{ is significant}$$

The second constraint guarantees that the reconstruction will take into account any pattern which is detected by any of the  $K$  transforms.





**CUR01+UWT, PSNR=32.10**



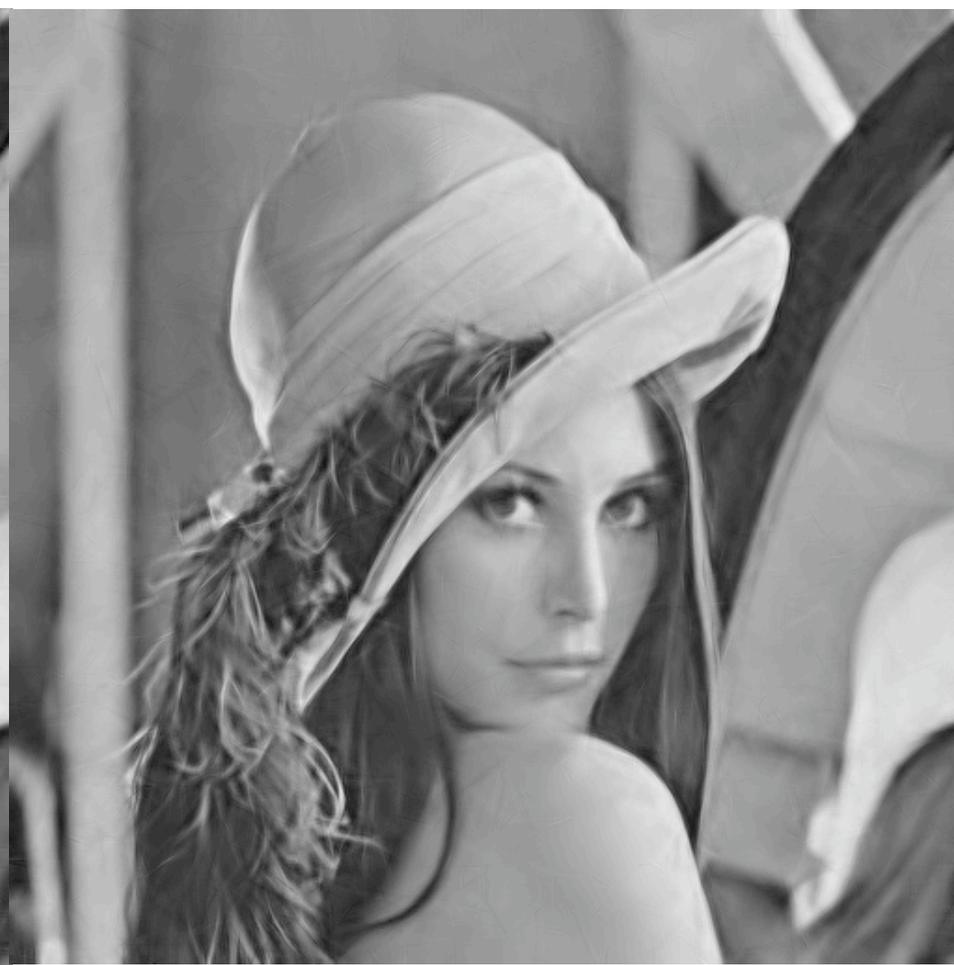
**Simoncelli and Portilla (2003), PSNR=32.70**



CUR01, PSNR=31.52



CUR01+SSR. PSNR=31.65



Notation: SSR: Slant Stack Radon  
NOB: Non Overlapping Blocks

The improvement of the curvelet transform result when using SSR-CUR01 is only 0.1 dB on Lena, almost undistinguishable by eye, and the computation time is multiplied by 20.

If you don't use overlapping blocks, the SSR-CUR01-NOB is much better than CUR01-NOB, but SSR-CUR01-NOB is still 5 times slower than CUR01 (WITH overlapping) and CUR01 is better by 0.5 dB than SSR-CUR01-NOB.

**Conclusion:** there is no need to use SSR in curvelet restoration applications.

# PSNR

$$(-10\log_{10}(\text{Variance}(\text{Error})/255^2))$$

	Sigma=20	Sigma=42
DATA	22.09	15.66
UWT	31.36	28.66
CUR01	31.51	28.74
CUR01+ SSR	31.65	28.83
CUR03	30.90	27.90
CUR04	31.12	28.17
PORTILLA et al.	<b>32.70</b>	<b>28.99</b>
CUR01+UWT	32.11	28.90

# **Need for new representations on the sphere:**

- 1) Isotropic Redundant WT on the Sphere
- 2) Ridgelets on the Sphere
- 3) Curvelet Transform on the Sphere

For restoration, we need an exact reconstruction

# Wavelet on the Sphere

- P. Schroder and W. Sweldens (Orthogonal Haar WT), 1995.
- M. Holschneider, Continuous WT, 1996.
- **W. Freeden and T. Maier, OWT, 1998.**
- J.P. Antoine, Continuous WT, 1999.
- L. Tenerio, A.H. Jaffe, Haar Spherical CWT, (CMB), 1999.
- L. Cayon, J.L Sanz, E. Martinez-Gonzales, Mexican Hat CWT, 2001.
- J.P. Antoine and L. Demanet, Directional CWT, 2002.

## Wavelet transform in the spherical harmonics space

We assume that

$$c_0(\theta, \phi) = c_{-1}(\theta, \phi) * \phi_{l_{max}}(\theta, \phi)$$

where  $\phi_{l_{max}}$  is a low pass filter (scaling function) with a frequency cut-off  $l_{max}$ .

$c_0$  the input map, and  $c_{-1}$  an unknown function. A resolution level  $j$  is related to the HEALPix definition ( $N_{side} = 2^j$ ).

The image at different resolutions is given by:

$$c_1(\theta, \phi) = c_{-1}(\theta, \phi) * \phi_{l_{max}/2}(\theta, \phi)$$

$$c_2(\theta, \phi) = c_{-1}(\theta, \phi) * \phi_{l_{max}/4}(\theta, \phi)$$

⋮

$$c_j(\theta, \phi) = c_{-1}(\theta, \phi) * \phi_{l_{max}/2^j}(\theta, \phi)$$

and the wavelet coefficients of the image are:

$$w_1(\theta, \phi) = c_{-1}(\theta, \phi) * \psi_1(\theta, \phi)$$

$$w_2(\theta, \phi) = c_{-1}(\theta, \phi) * \psi_2(\theta, \phi)$$

⋮

$$w_j(\theta, \phi) = c_{-1}(\theta, \phi) * \psi_j(\theta, \phi)$$

where  $\psi_j$  is the wavelet function at scale  $j$ .

## Choice of Scale Function and Wavelet Function

We define  $\phi_{l_{max}}$  in the Spherical Harmonics space as a B-spline function.  
The B-spline function is

$$B(x) = \frac{1}{12} (|x-2|^3 - 4|x-1|^3 + 6|x|^3 - 4|x+1|^3 + |x+2|^3)$$

and the scaling function  $\phi_{l_{max}}$  is:

$$\hat{\phi}_{l_{max}}(l, m) = \frac{3}{2} B\left(\frac{l}{l_{max}}\right)$$

The wavelet can be chosen as the difference between two resolutions:

$$\psi_j(\theta, \phi) = \phi_{l_{max}/2^{j-1}}(\theta, \phi) - \phi_{l_{max}/2^j}(\theta, \phi)$$

## From one resolution to the next one

The image at the first scale is given by:  $c_0(\theta, \phi) = c_{-1}(\theta, \phi) * \phi_{l_{max}}(\theta, \phi)$ , which gives in space of spherical harmonics (as  $\phi_{l_{max}}$  is azimuthally symmetric):

$$\hat{c}_0(l, m) = \sqrt{\frac{2l+1}{4\pi}} \hat{c}_{-1}(l, m) \hat{\phi}_{l_{max}}(l, m)$$

The coefficients which allow us to go from one resolution to the next one are obtained with the discrete filters  $h$  and  $\hat{h}$ :

$$\hat{h}(l, m) = \begin{cases} \frac{\hat{\phi}_{l_{max}}(2l, m)}{\hat{\phi}_{l_{max}}(l, m)} & \text{if } l < l_{max} \\ 0 & \text{otherwise} \end{cases}$$

$$\hat{g}(l, m) = \begin{cases} \frac{\hat{\psi}_1(2l, m)}{\hat{\phi}_{l_{max}}(l, m)} & \text{if } l < l_{max} \\ 0 & \text{otherwise} \end{cases}$$

We have:

$$\hat{c}_{j+1}(l, m) = \hat{c}_j(l, m) \hat{h}(2^j l, m)$$

$$\hat{w}_{j+1}(l, m) = \hat{c}_j(l, m) \hat{g}(2^j l, m)$$

The cut-off frequency is reduced by a factor 2 at each step.

## Reconstruction

If the wavelet is the difference between two resolutions, an evident reconstruction for a wavelet transform  $\mathcal{W} = \{w_1, \dots, w_J, c_J\}$  is:

$$c_0(\theta, \phi) = c_J(\theta, \phi) + \sum_j w_j(\theta, \phi)$$

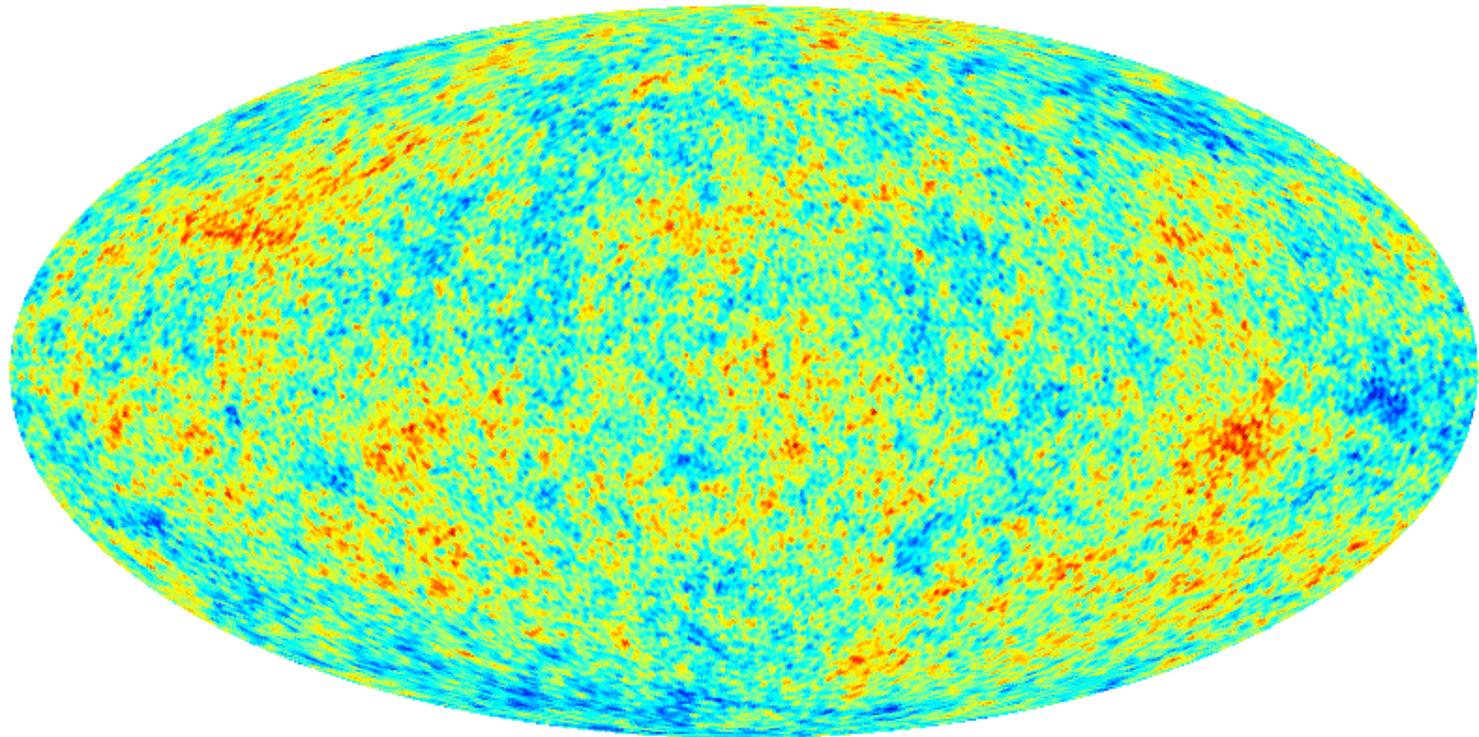
An alternative is to use the conjugate filters defined by

$$\begin{aligned}\hat{\tilde{h}} &= \hat{h}^* / (|\hat{h}|^2 + |\hat{g}|^2) \\ \hat{\tilde{g}} &= \hat{g}^* / (|\hat{h}|^2 + |\hat{g}|^2)\end{aligned}$$

And the reconstruction is obtained by:

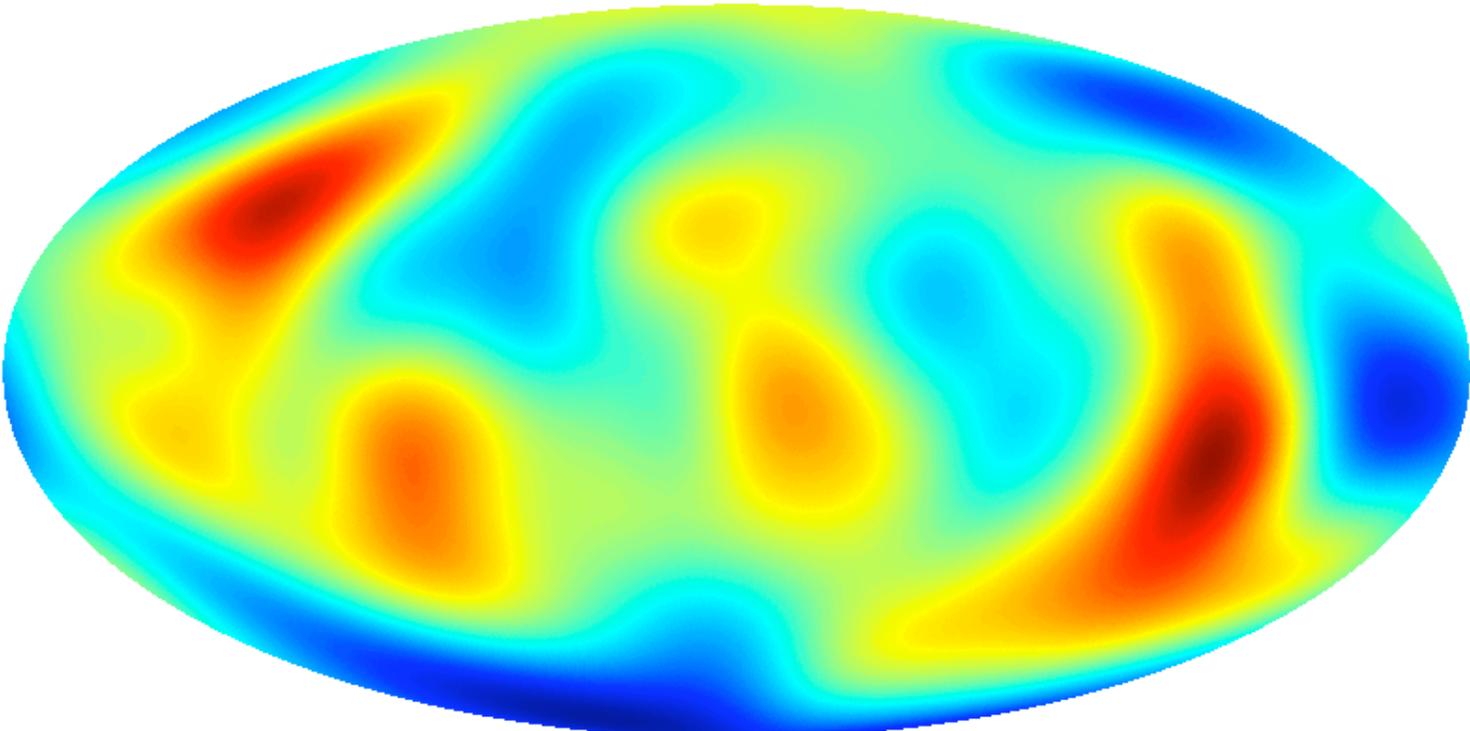
$$\hat{c}_j(l, m) = \hat{c}_{j+1}(l, m) \hat{\tilde{h}}(l_{max}/2^j, m) + \hat{w}_{j+1}(l, m) \hat{\tilde{g}}(l_{max}/2^j, m)$$

ess.fits: TEMPERATURE

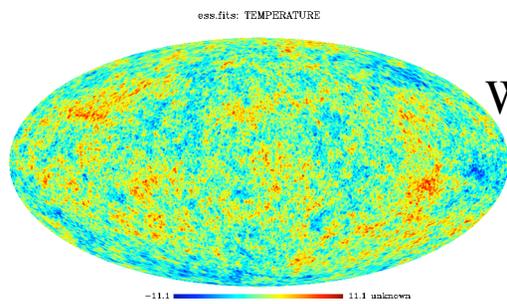


-11.1  11.1 unknown

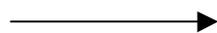
8result\_h.fits: TEMPERATURE



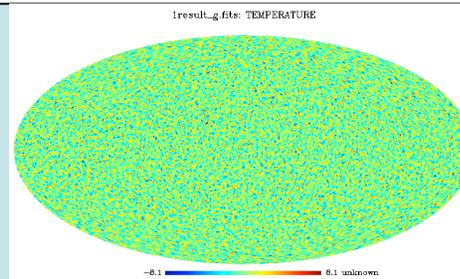
-2.5  2.5 unknown



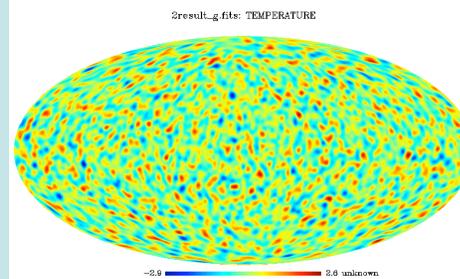
Undecimated  
Wavelet Transform



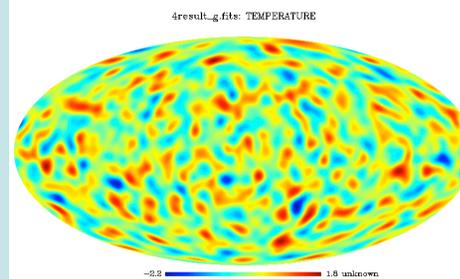
j=1



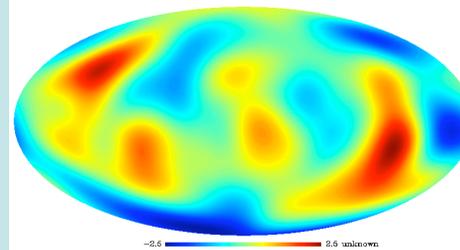
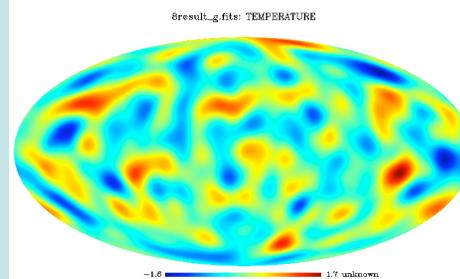
j=2

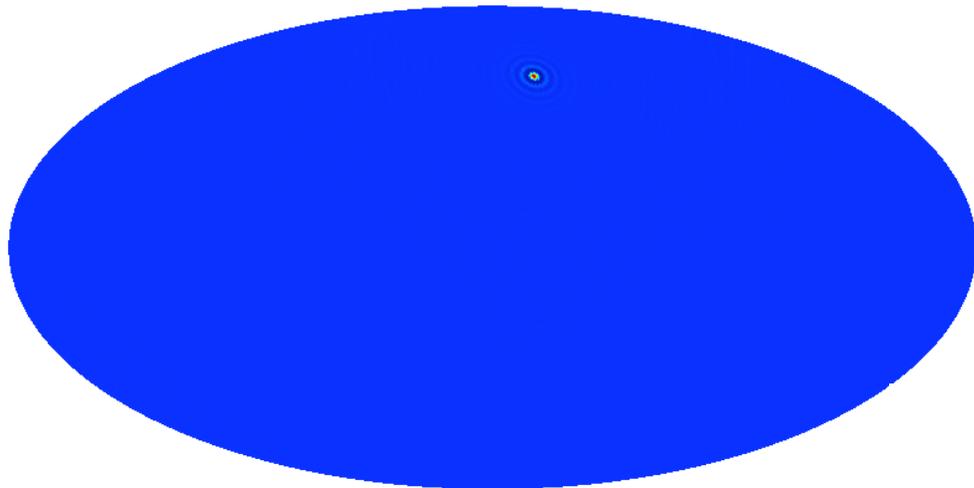


j=3



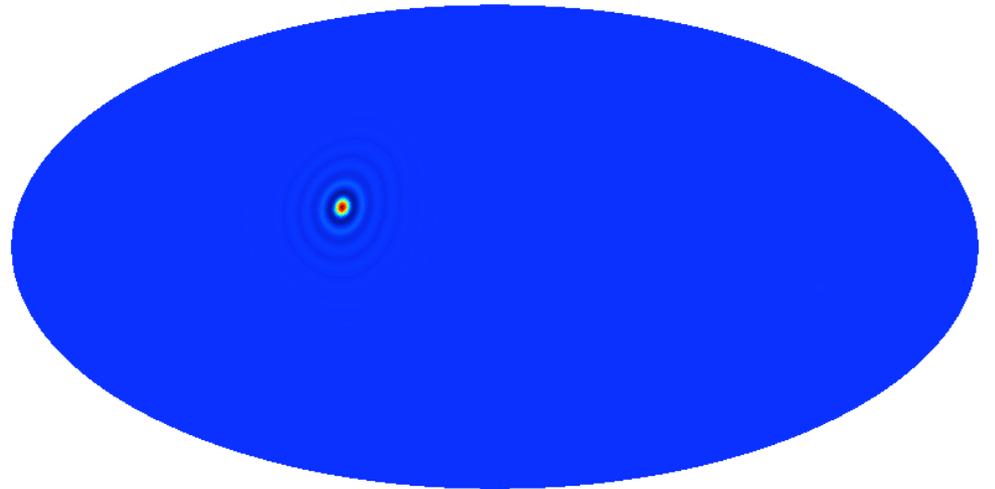
j=4





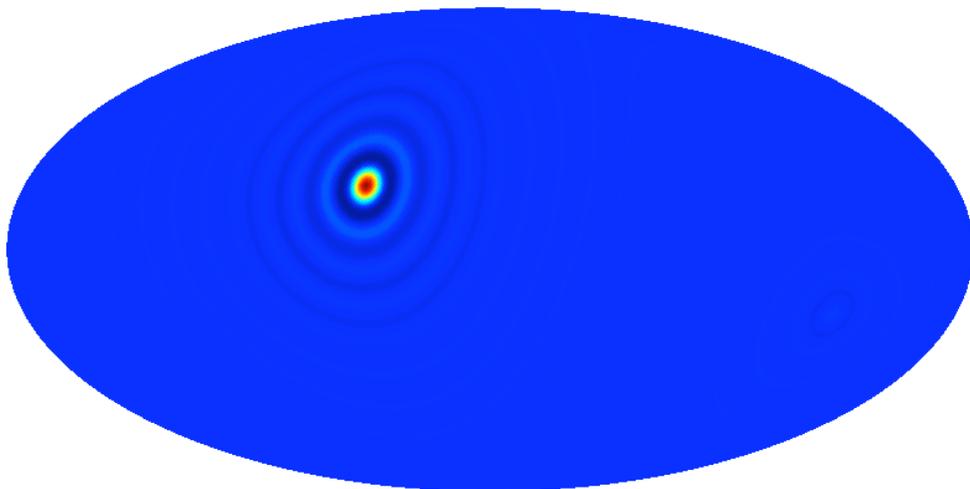
-0.045 0.32

sky\_recons.fits: X0



-0.046 0.34

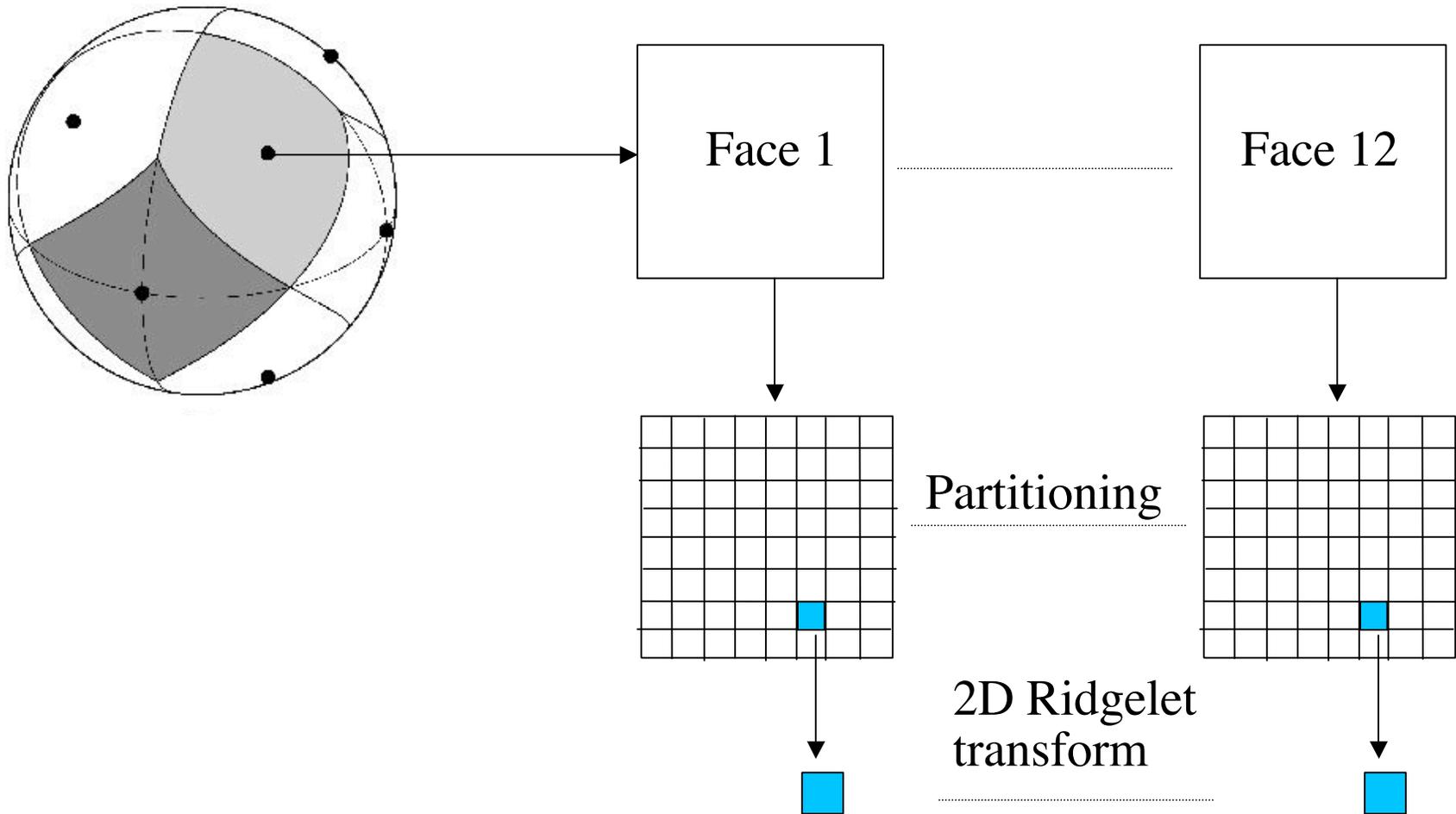
sky\_recons.fits: X0



-0.047 0.35

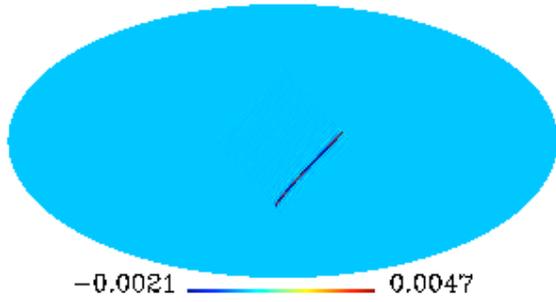


# Ridgelets on the Sphere

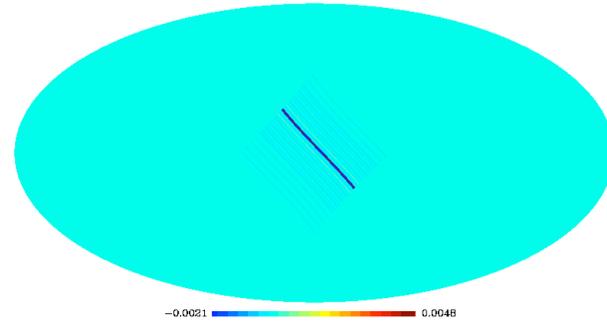


# Example of ridgelet functions on the sphere

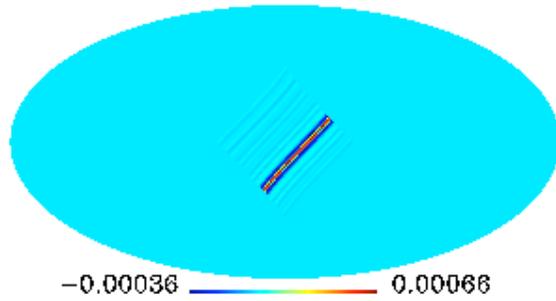
1 : 140 ]: 0



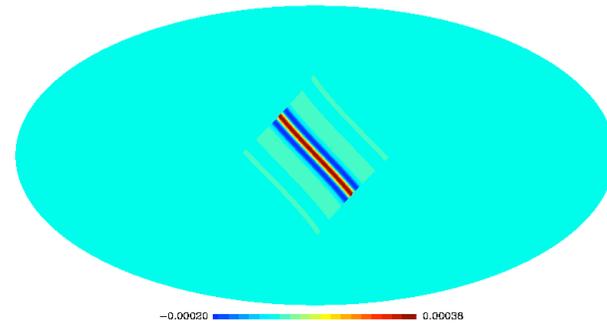
on line processing :



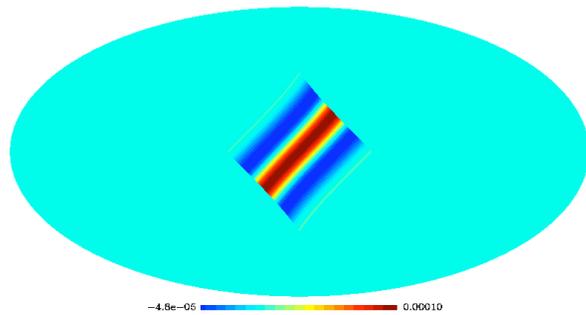
1 : 80 ]: 0



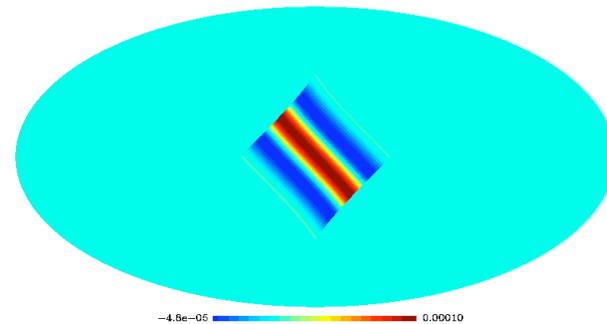
on line processing :



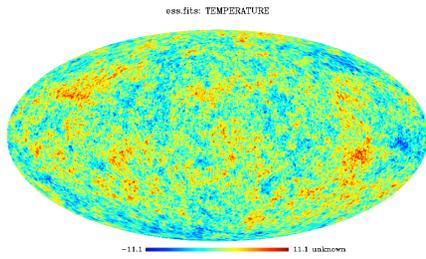
on line processing :



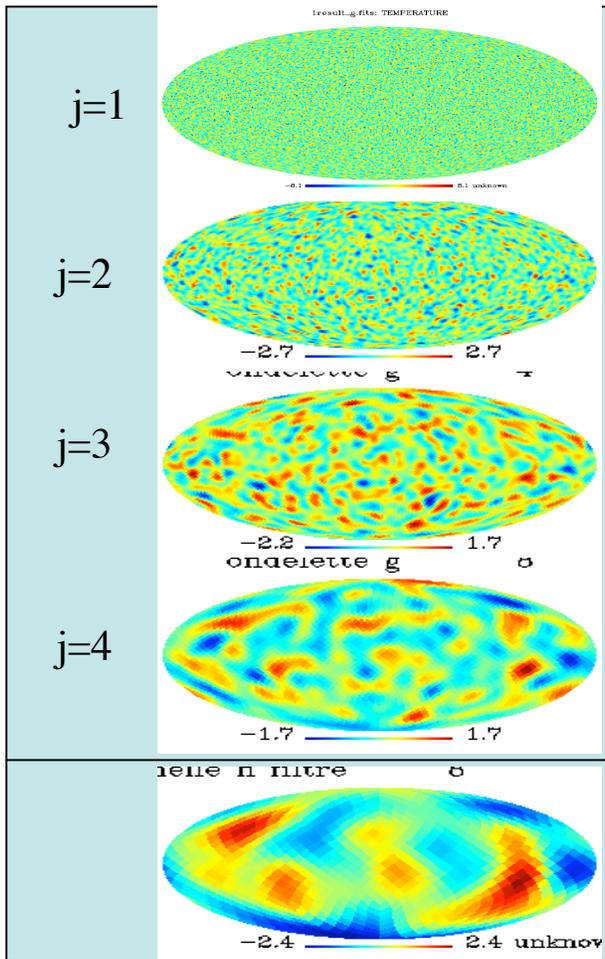
on line processing :



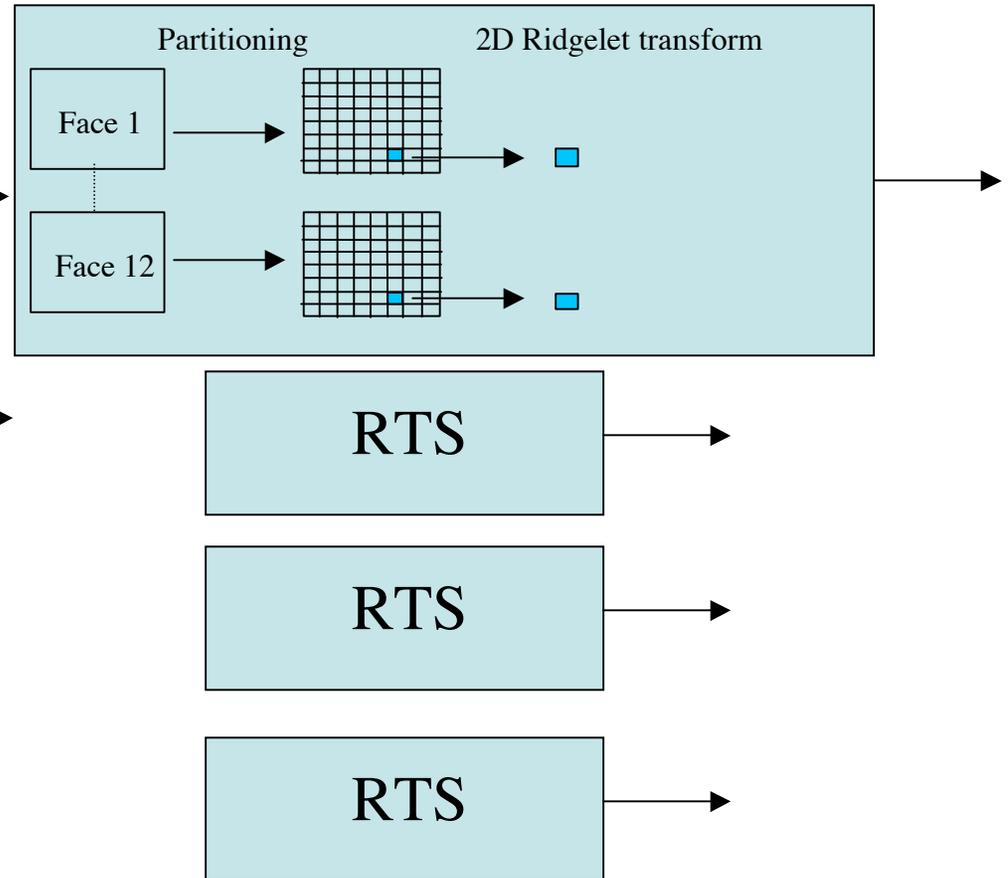
# Curvelets on the Sphere



Pyramidal WT on the Sphere

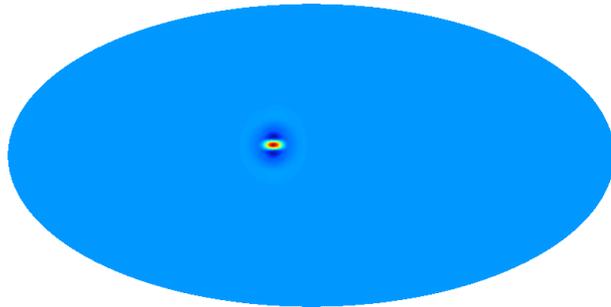


Ridgelet Transform on the Sphere (RTS)



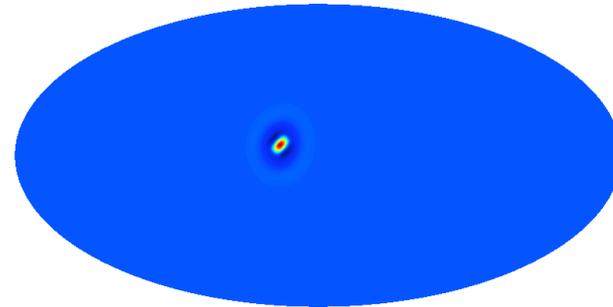
# Example of curvelet functions on the sphere

on line processing :



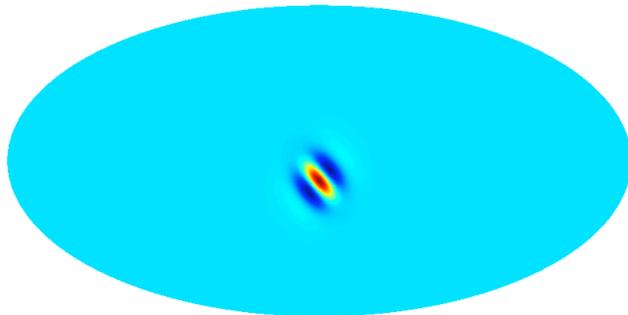
-0.064 0.16

on line processing :



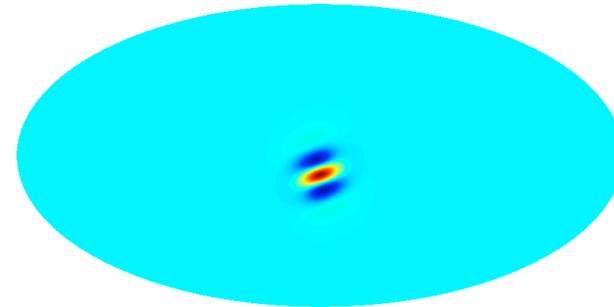
-0.008 0.17

on line processing :



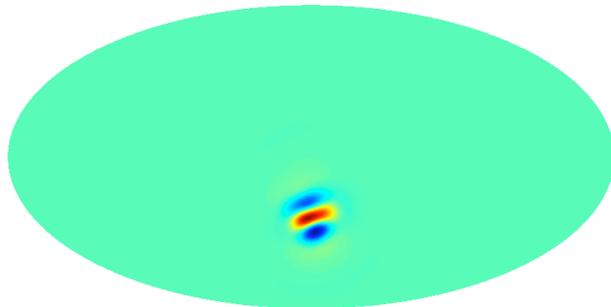
-0.016 0.036

on line processing :



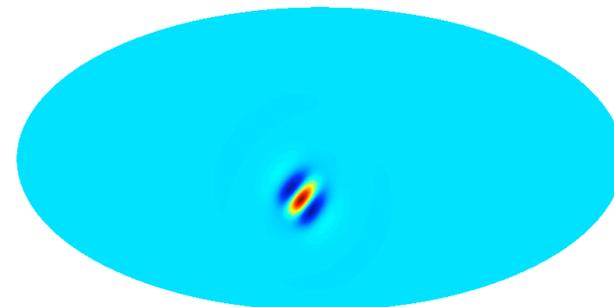
-0.012 0.020

on line processing :



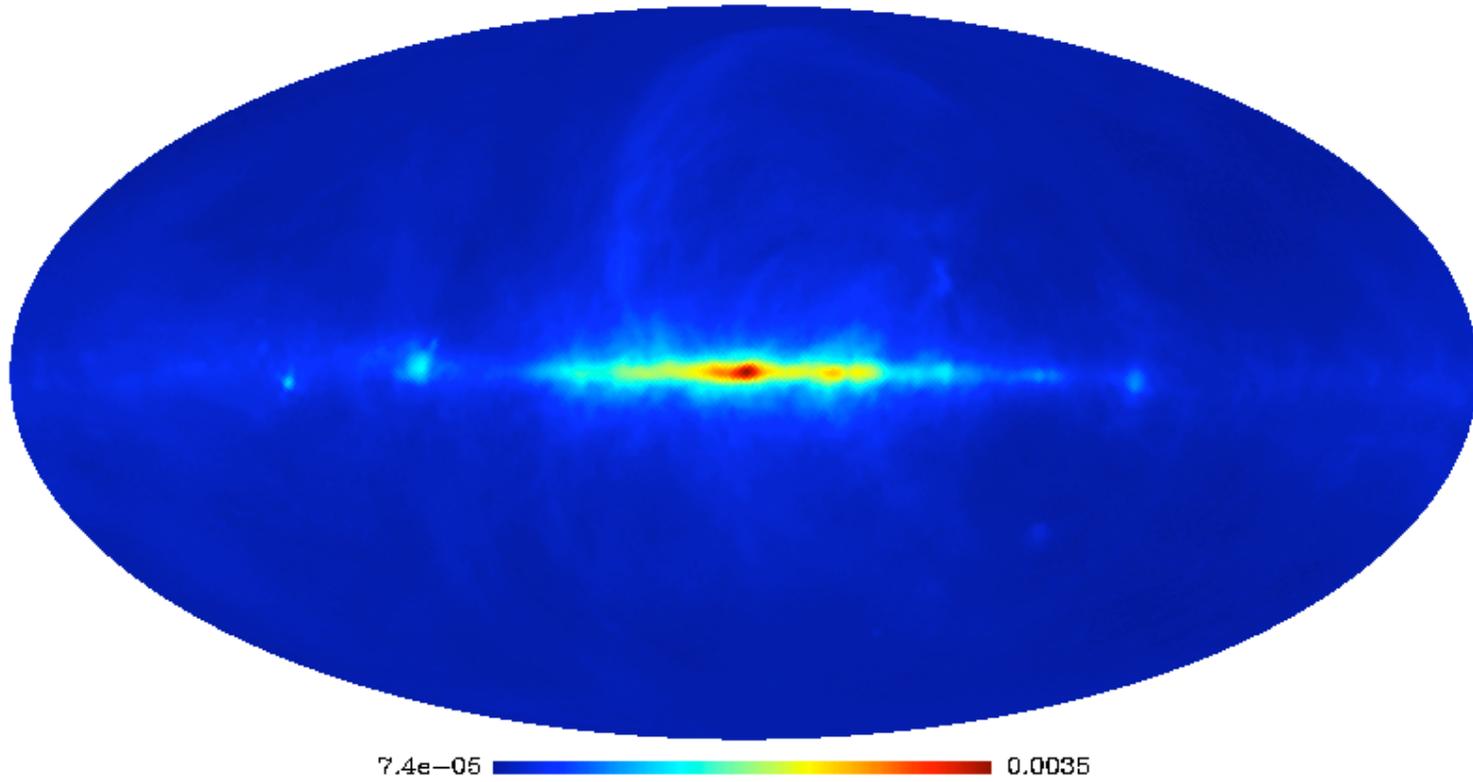
-0.022 0.026

on line processing :

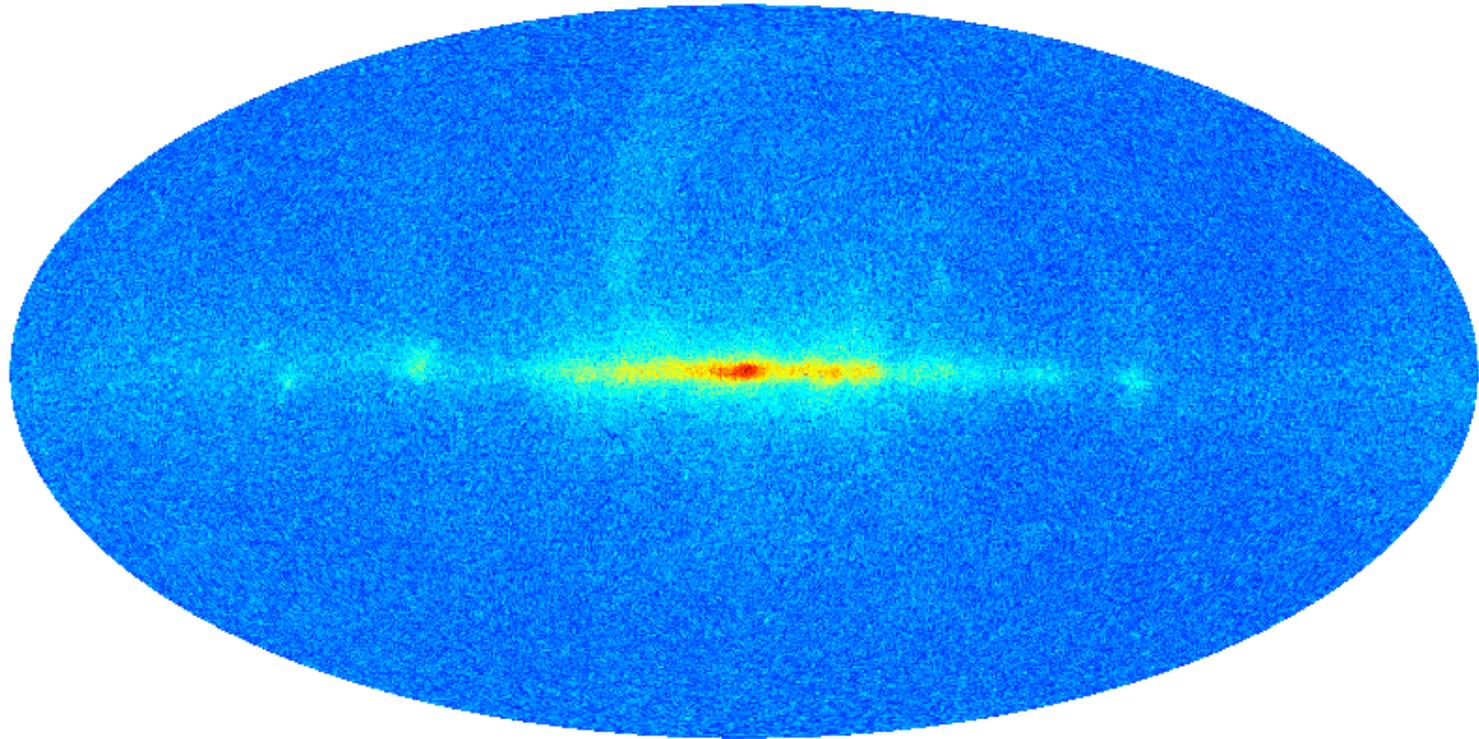


-0.019 0.038

on line processing :

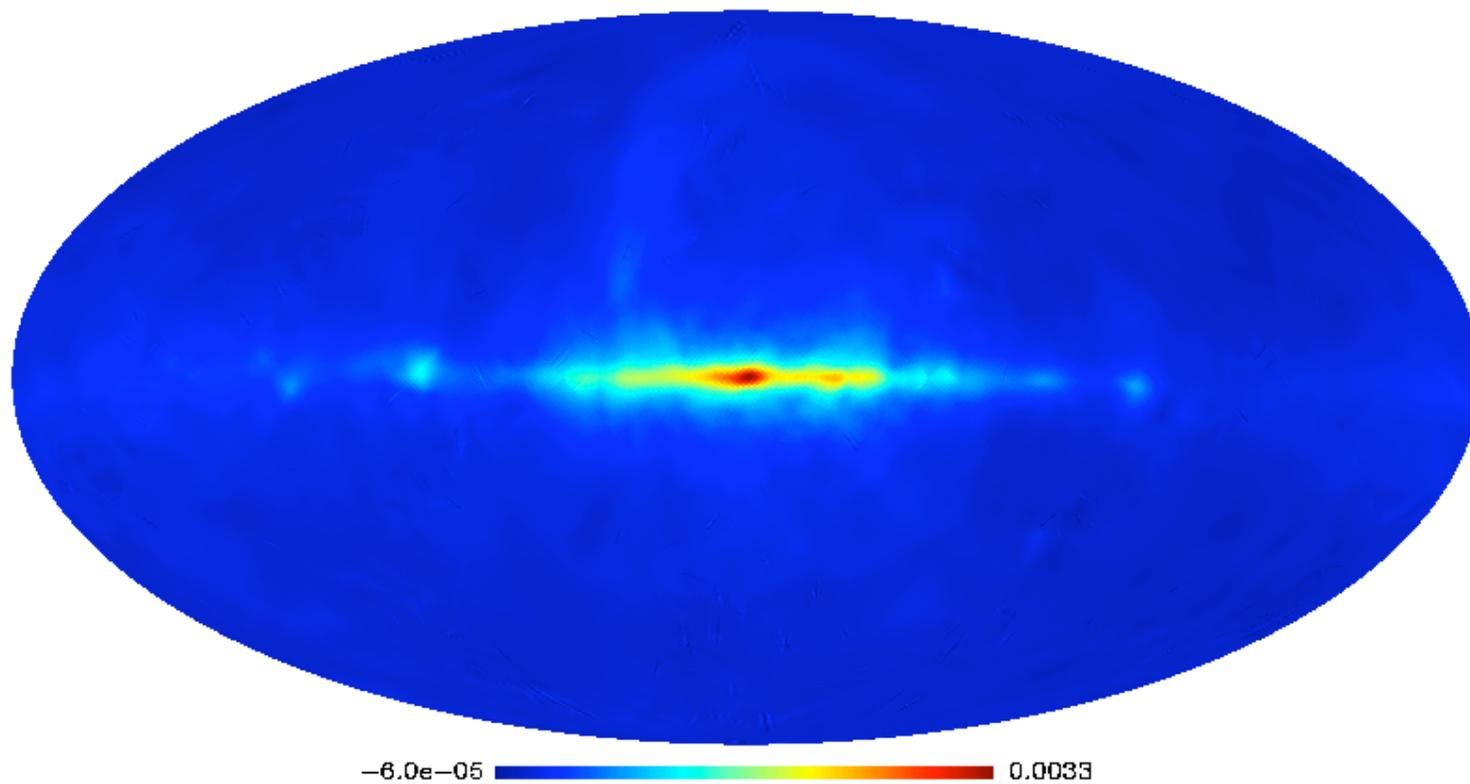


bruit1

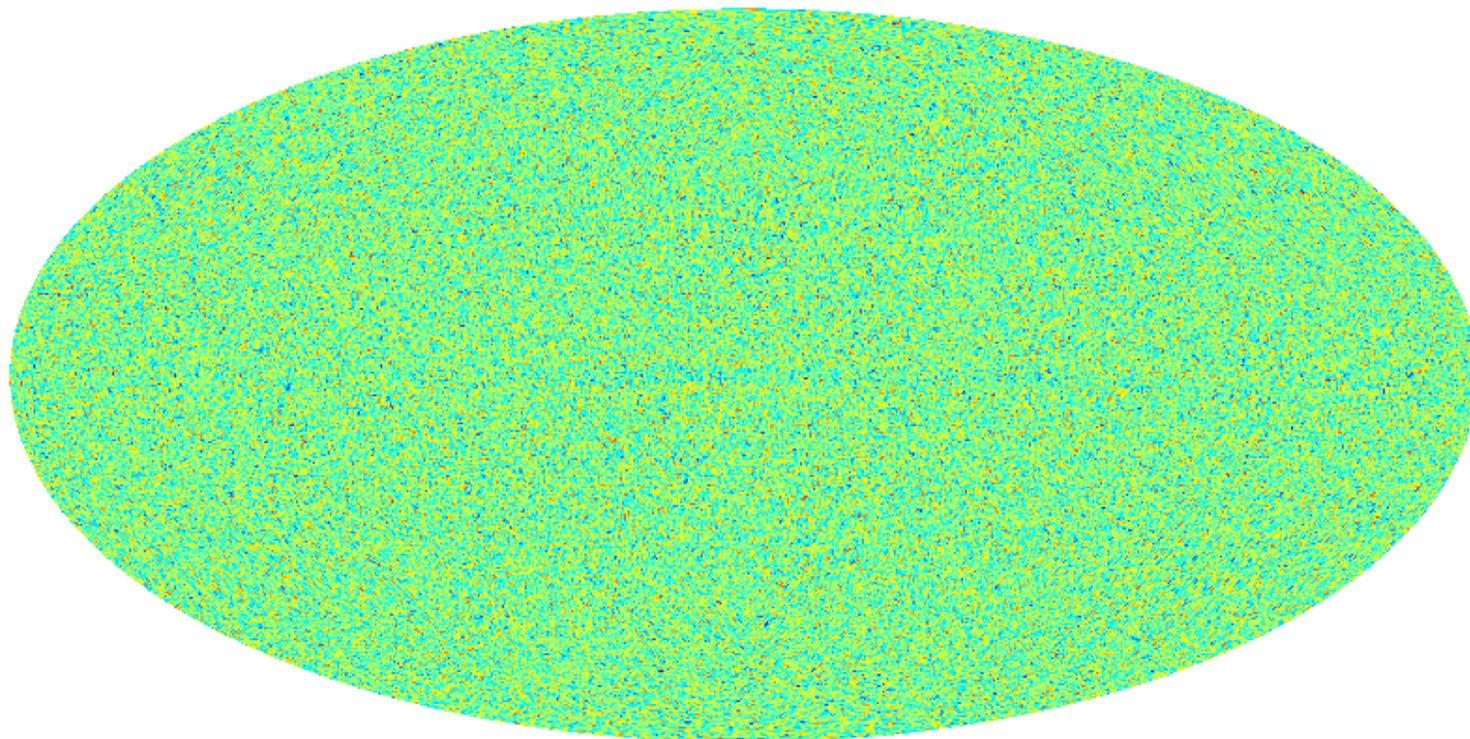


-0.00084 0.0038

denoising cur at 5 sigma

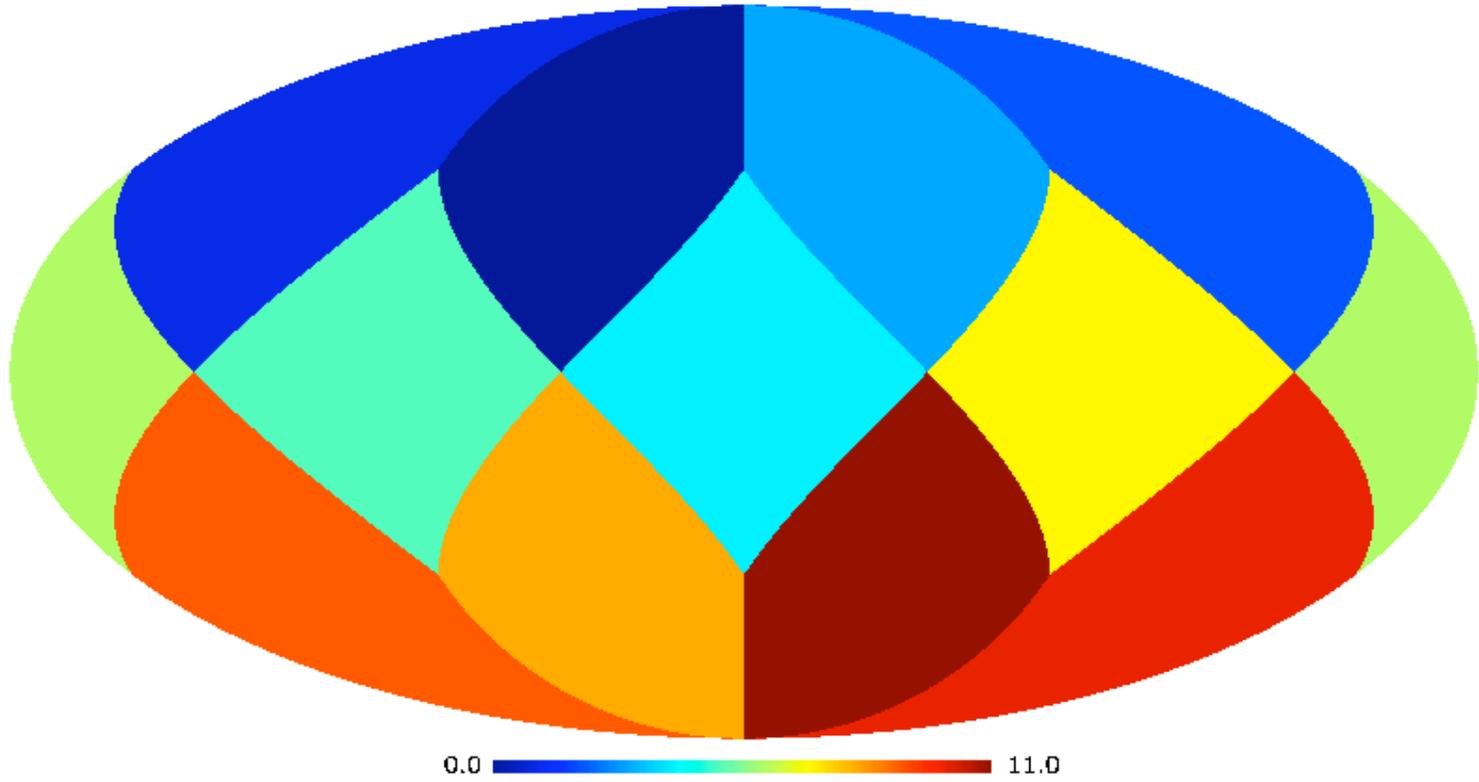


dif 5 sigma

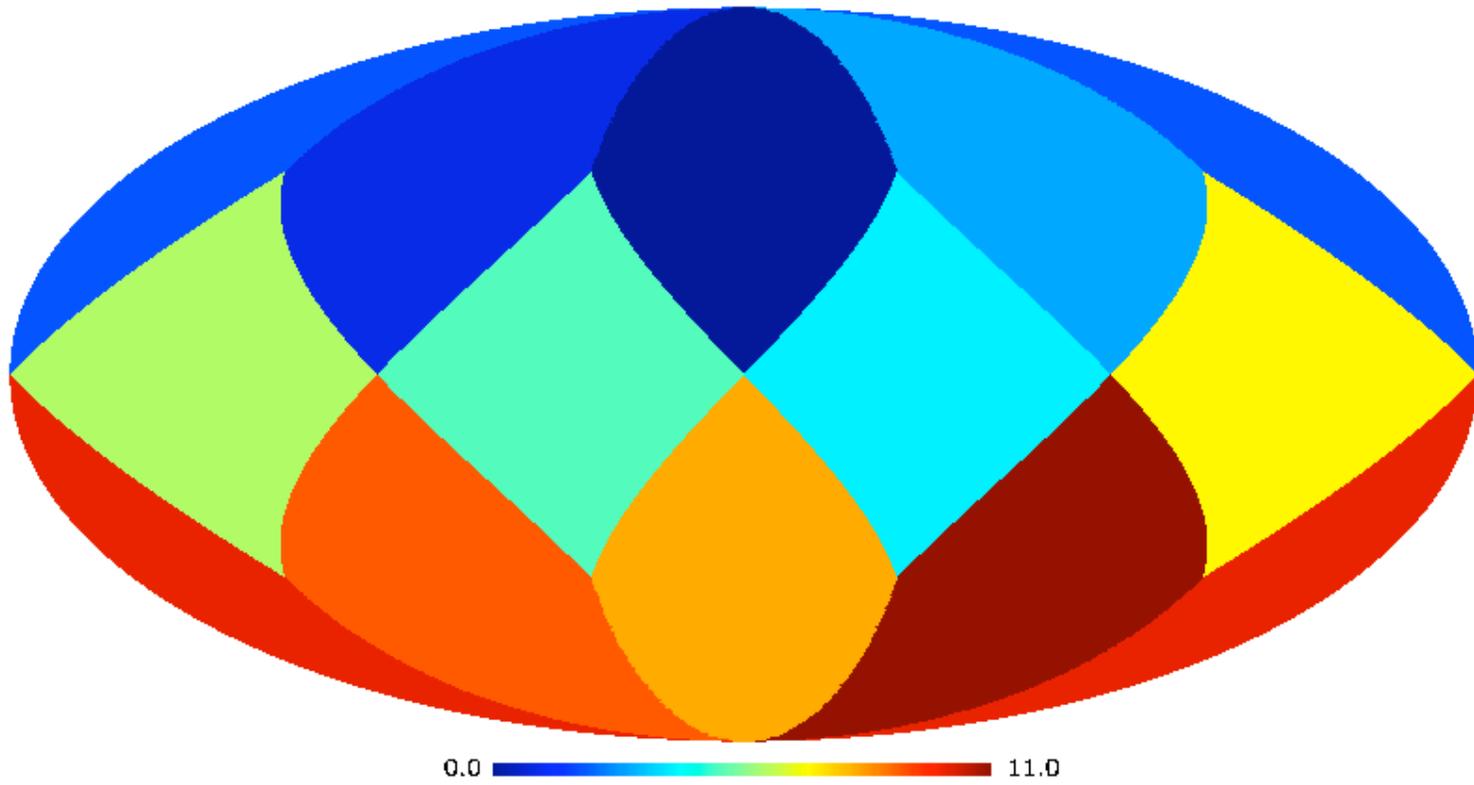


-0.00098 0.00098

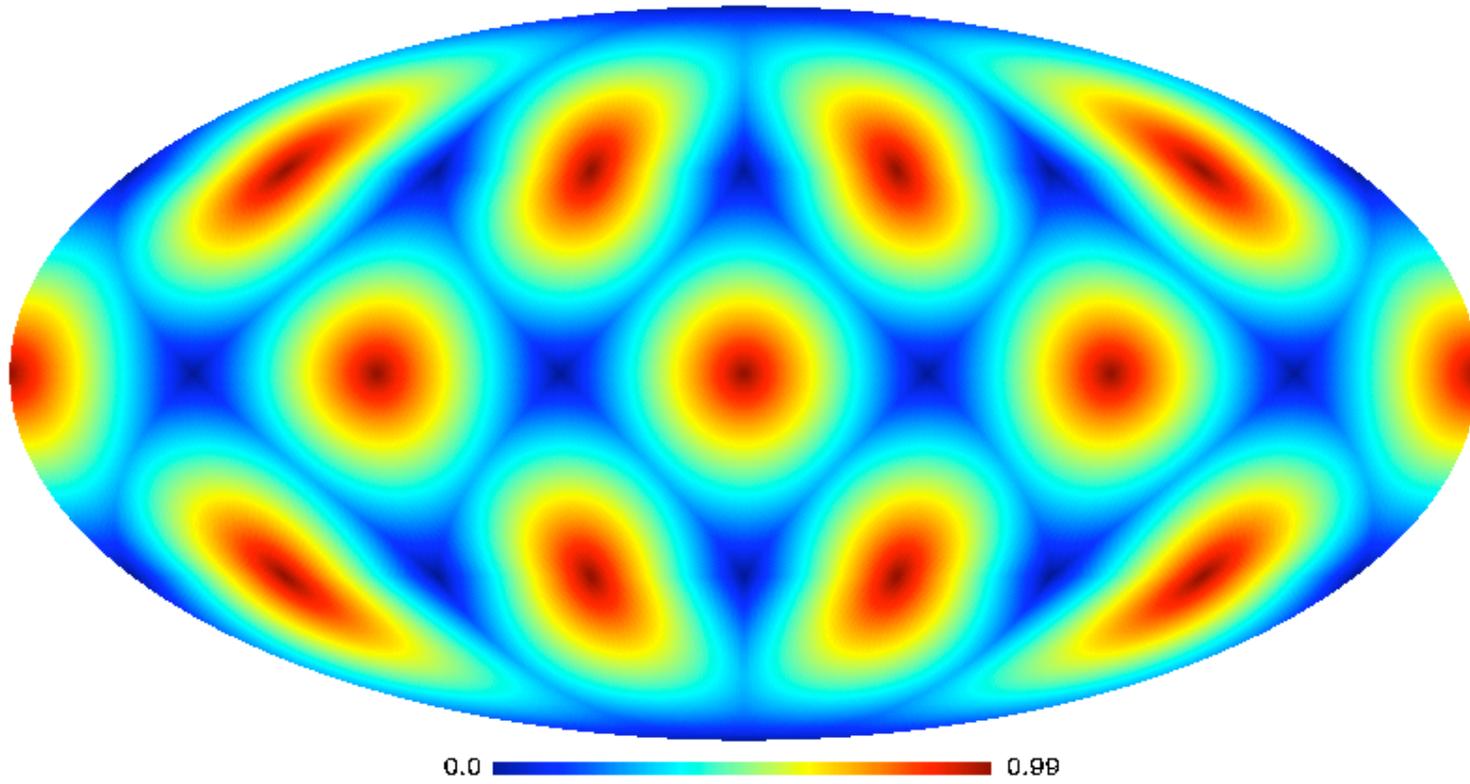
12 faces



12 faces rotation



ponderation



# Conclusions

- 1. Isotropic WT on the Sphere, very similar to the Isotropic a trous WT**
- 2. Extension of CUR01 to the sphere**
- 3. Produce encouraging results in our denoising experiment.**