



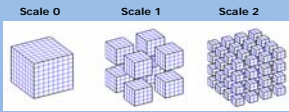
Real Time Multi-Scale Geometric Segmentation of 3D Images

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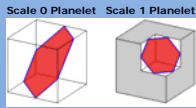
3D Planelets (Donoho and Levi)

The Planelet Transform

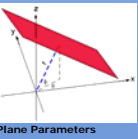
The Planelets Transform is a Discrete Multi-Scale 3D Radon Transform applied on a dyadic cubes structure



Planelets are a set of $O(n^3 \log(n))$ plane patches in a variety of locations, orientations and scales, which can be used to effectively approximate smooth surfaces in a 3D image



The 3D Radon Transform



Plane Parameters

$$P = \{(\cos(\phi)\cos(\theta)x + \cos(\phi)\sin(\theta)y + \sin(\phi)z = r) : \phi \in [0, \Pi], \theta \in [0, 2\Pi]\}$$

$$Rf(r, \phi, \theta) = \iiint f(x, y, z) \delta(r - \sin(\phi)\cos(\theta)x - \sin(\phi)\sin(\theta)y - \cos(\phi)z) dx dy dz$$

$$f(x, y, z), (x, y, z) \in \mathbb{R}^3$$

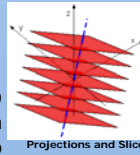
$$Rf(P) = \int_P f(s) ds$$

$$R: \mathbb{R}^3 \rightarrow \mathbb{R}^3$$

Projection-Slice Theorem

$$R(r, \phi, \theta) = F_1^{-1} F_3 f(x, y, z)$$

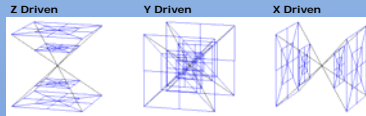
$$f(x, y, z) = F_3^{-1} F_1 R(r, \phi, \theta)$$



This theorem is the key to the 3D Fast Slant-Stack Transform, which uses a Pseudo-Spherical grid to compute a Discrete 3D Radon transform

Pseudo-Spherical (PS) Grid

- Concentric boxes instead of Spheres.
- Uniformly spaced slopes instead of angles.
- Allows to compute the PSFT in $O(n^3 \log(n))$ instead of $O(n^6)$ in direct computation.



PSFFT Algorithm (Averbuch et.al.)

Apply 1D FFT's along x.
Apply Fractional FFT's along y with varying Alpha.
Apply Fractional FFT's along z with varying Alpha.

$$\tilde{f}_\alpha(k) = \sum_{m=-n/2}^{n/2-1} f(m) \exp(-i2\pi\alpha \frac{km}{n}), k=1, \dots, n$$

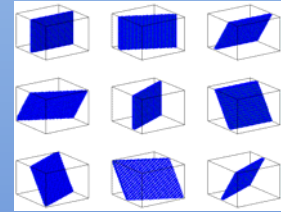
Coefficient set of one derivation



The 3D Fast Slant-Stack Algorithm



3D FSS Back-Projections

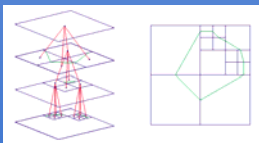


The Planelet Transform is a multi-scale 3D FSS. It also has an overall complexity of $O(n^3 \log(n))$.

3D Geometric Segmentation using Planelets

Planelet Decorated Quad-Tree

- A natural generalization of the "Beamlet Decorated Quad-Tree" (Donoho and Huo):



- We apply a similar idea in 3D using planelets transform as means of computation of SSEs of sections on different sides of each dividing planelet.
- The Decorated QT in our case is made of dyadic cubes in different scales and their divisions by one significant planelet.

Division Criteria

- The division into two parts of a dyadic cube in certain scale is done by an F-test (assuming white noise), which uses the planelets transform to compute the test statistic:

$$SSE(V) = \sum_{v_i \in V_{axes}} v_i^2 - N_v \cdot \bar{v}^2$$

$$N_v = Volume(V)$$

H0: Good Split

H1: No Split

$$F_{st}(V_1, V_2) = \frac{(SSE(V) - SSE(V_1) - SSE(V_2))/1}{(SSE(V_1) + SSE(V_2))/(N_{V_1} + N_{V_2} - 2)}$$

$$F_{st}(V_1, V_2) \sim F_{1, N_v - 2}$$

- Computing the p-value for the given test it can easily be decided if division is needed.

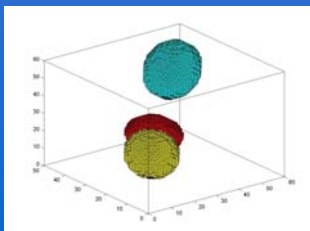
The Algorithm

0. Begin with the entire cube, Determine α
1. Find best Planelet Split
2. Check division Criteria (P-value < α)
 - a) if holds split and stop
 - b) if does test cube homogeneity
 - if the cube is homogenous – stop
 - else apply a quad split and repeat recursively from 1 for each sub cube.

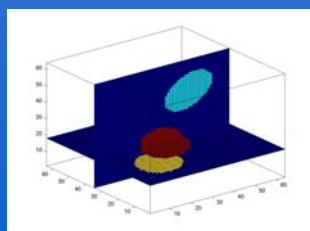
Remarks

- The algorithm uses a single sensitivity parameter, α , which is statistically meaningful (level of significance) -
- The overall complexity of the algorithm is $O(n^3 \log(n))$

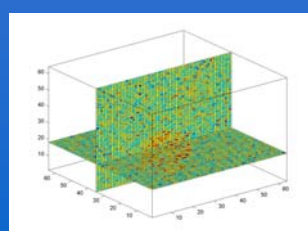
Noiseless 3D Image



Slices View



Slices View of the Noisy Image



Reconstruction from the Noisy Image

