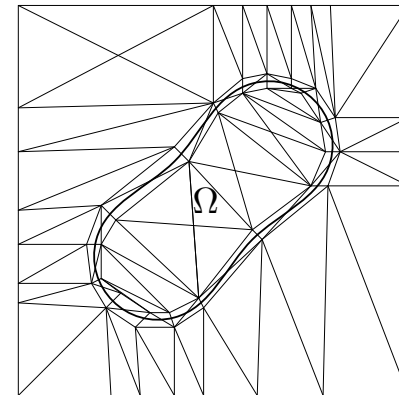
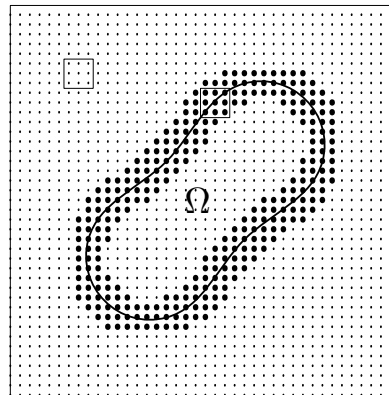
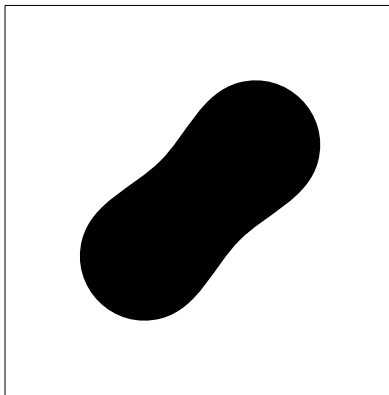


# Geometrical Image Representation with Bandelets

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## Sparse Representations Using Geometric Regularity

- A representation which uses the geometric regularity along contours.
- Adapted for compression, estimation, pattern recognition, inverse problem.
- Let  $f = \mathbb{1}_\Omega$ , where the boundary  $\partial\Omega$  is regular:  $\mathbf{C}^s$  with  $s > 2$ .



- Wavelet Approximation :  $\|f - f_M\| \sim \|f\|_{TV} M^{-1/2}$ .
- With  $M$  geometric elements :  $\|f - f_M\| \sim C M^{-1}$ .
- With higher order elements :  $\|f - f_M\| \sim C M^{-s/2}$ .

## Redundant Dictionary Approximations

- Dictionary of finite elements with arbitrary geometry:

$$\mathcal{D} = \{g_\gamma\}_{\gamma \in \Gamma} .$$

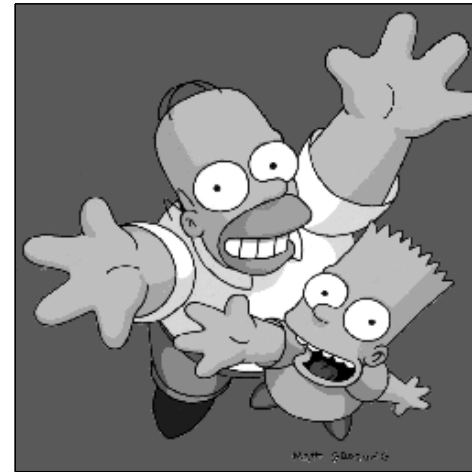
- Non-linear approximation: find  $\{g_{\gamma_m}\}_{1 \leq m \leq M}$  in  $\mathcal{D}$  which minimize

$$\left\| f - \sum_{m=1}^M \alpha_m g_{\gamma_m} \right\| .$$

### Difficulties

- Typically NP hard problem.
- Unstable solutions.
- Greedy algorithms: far from optimal.

## More Structured Approach



- Edge : One dimensional singularity that has regular evolution within image plane.
- An edge detection is needed : difficult.

## Overview

- 1D Foveal Wavelets for singularity detection.
- 2D Bandelet basis with geometry built in.
- Approximation results.

# Foveal Wavelets For Singularities

- Wavelets that zoom on singularities.
- An even wavelet  $\Psi^1$  and an odd wavelet  $\Psi^2$

$$\Psi_{j,u}^k(x) = \frac{1}{2^{j/2}} \Psi^k \left( \frac{x-u}{2^j} \right)$$

## Conditions

- Orthonormal family of foveal wavelets

$$\{ \Psi_{j,u}^1, \Psi_{j,u}^2 \}_{j \in \mathbb{Z}}$$

- Reconstruction of polynomial discontinuities :

$$V_u = \text{span} \left\{ \Psi_{j,u}^k \right\}_{j,k}$$

$$\mathbb{1}_{(-\infty, u]} P(x) \in V_u \quad \text{and} \quad \mathbb{1}_{[u, +\infty)} P(x) \in V_u$$

if  $\deg(P) \leq m$

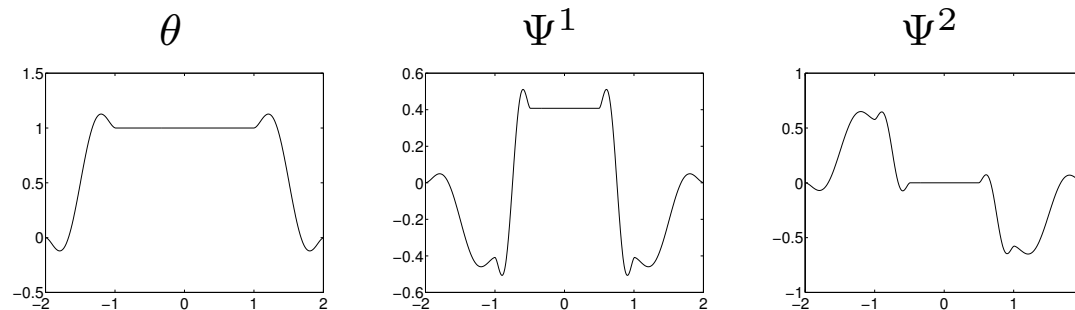
## Construction

- Foveal mother wavelets are defined by

$$\Psi^1(x) = \theta(x) - 2\theta(2x)$$

$$\Psi^2(x) = \text{sgn}(x) [\theta(x) - \theta(2x)]$$

With appropriate conditions on  $\theta$ ,  $\{\Psi_{j,u}^1, \Psi_{j,u}^2\}_j$  is orthonormal.



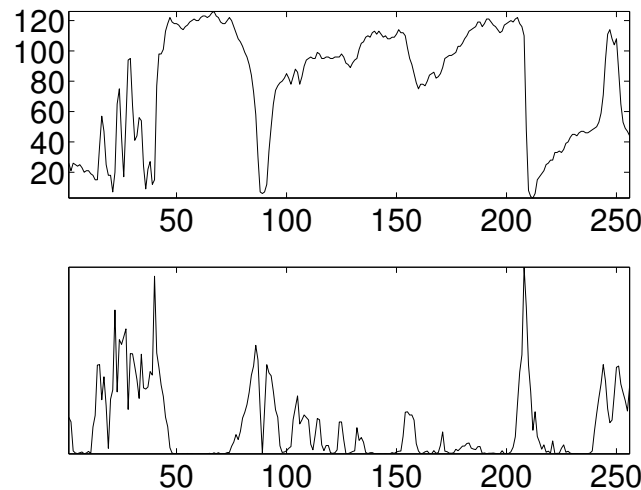
# Detection of Singularities

**Theorem :** A function  $f$  has a non-oscillating singularity Lipschitz of order  $\alpha$  at  $u$  if and only if

$$|\langle f, \Psi_{j,u}^k \rangle| = O(2^{j(1/2+\alpha)})$$

- To detect singularities with  $\alpha < 1$ , compute for each  $u$

$$\epsilon(u) = \sum_{k \in \{1,2\}} \sum_{j \in \mathbb{Z}} 2^{-3j} |\langle f, \Psi_{j,u}^k \rangle|^2$$





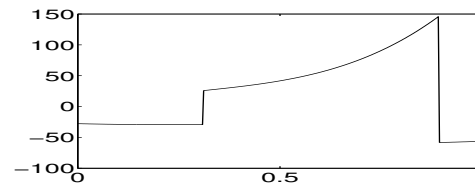
# Singularities Cancellation

**Theorem :** If  $f$  is uniformly Lipschitz  $\alpha < m$  for  $a \leq x < u$  and  $u < x \leq b$  then

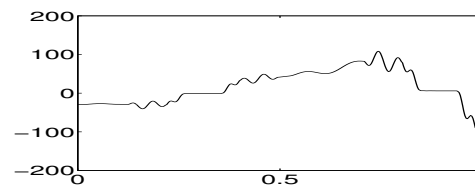
$$f - P_{V_u} f = f - \sum_{k \in \{1,2\}} \sum_{j \in \mathbb{Z}} \langle f, \Psi_{j,u}^k \rangle \Psi_{j,u}^k$$

is uniformly Lipschitz  $\alpha$  on  $[a, b]$ .

Original Signal

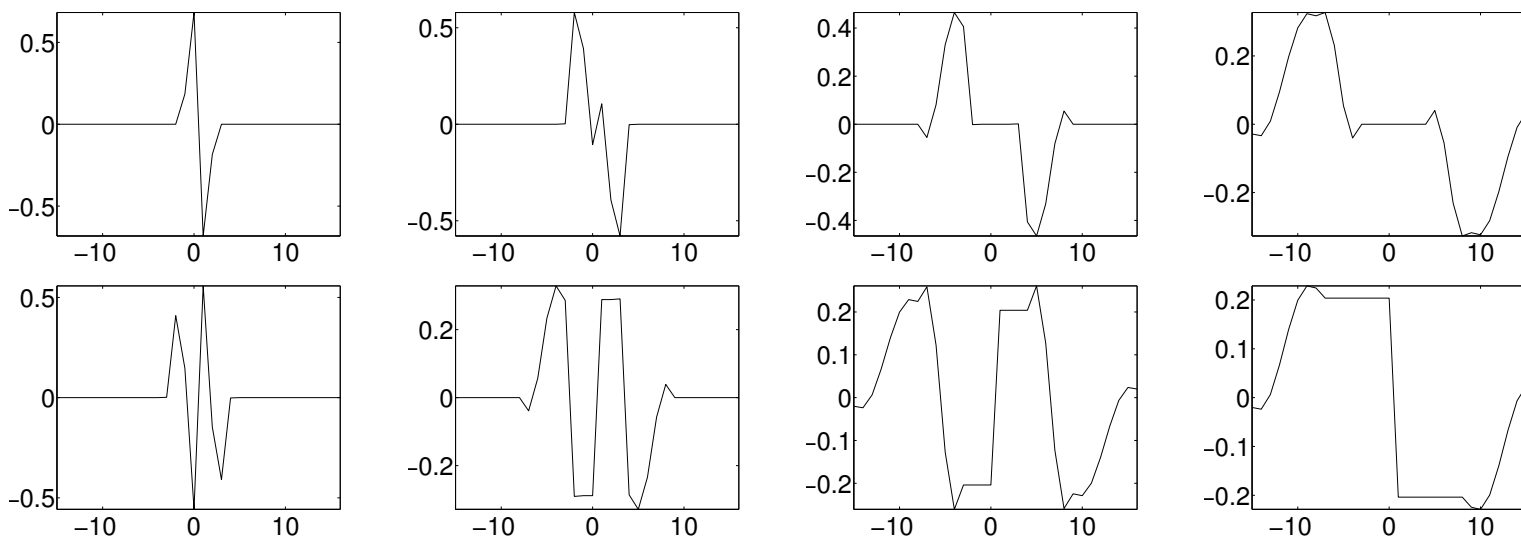


Residual  $Rf = f - P_V f$  : smooth



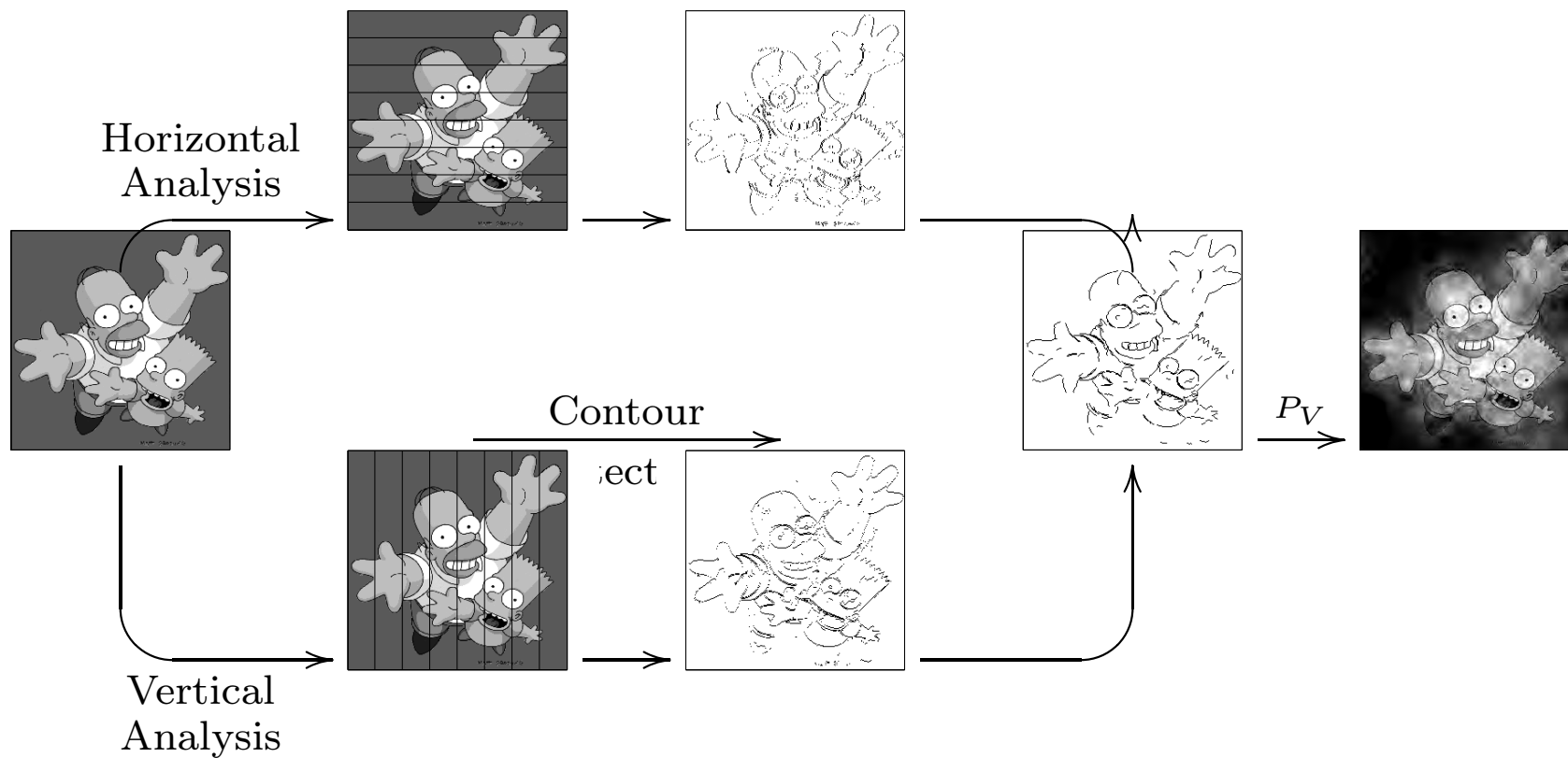
## Foveal Wavelet Packets

- Change of basis among antisymmetric foveal wavelets  $\{\Psi_{j,u}^2\}$ .
- Includes the projection of  $\text{sgn}(x - u)$  in  $V_u$ .

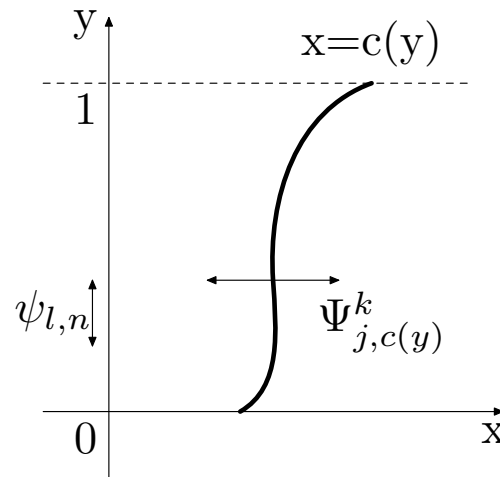


# Images and lines

- Singularities are essentially monodimensionals.
- Use only vertical and horizontal lines for detection and reconstruction.



## 2D Bandelets : geometrical basis



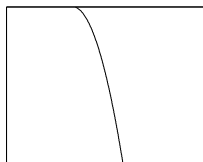
- Use a wavelet orthonormal basis

$$\left\{ \psi_{l,n} = \frac{1}{2^{l/2}} \psi \left( \frac{x - n2^l}{2^l} \right) \right\}_{l \in \mathbb{Z}^+, n \in \mathbb{Z}}$$

- Orthonormal family of bandelets

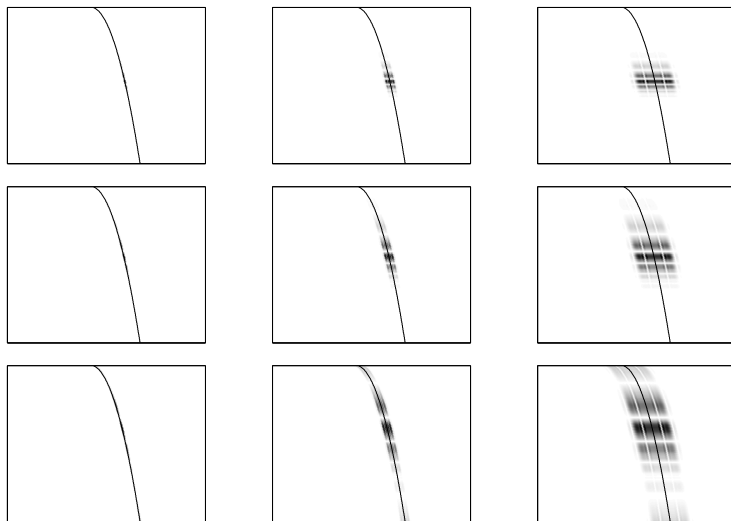
$$\mathcal{B}_c = \left\{ \Psi_{j,c(y)}^1(x) \psi_{l,n}(y), \quad \Psi_{j,c(y)}^2(x) \psi_{l,n}(y) \right\}_{j,l,n} = \{b_p^c\}_{p \in \mathbb{Z}}$$

# Some Bandelets

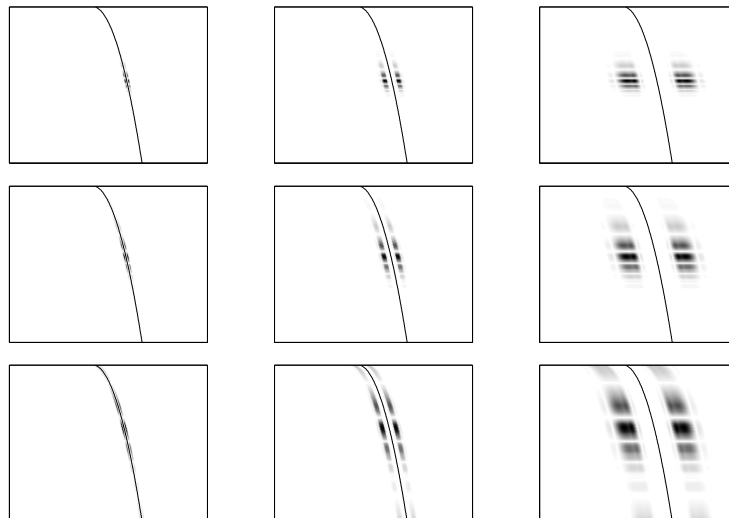


- Bandelet along the curve.

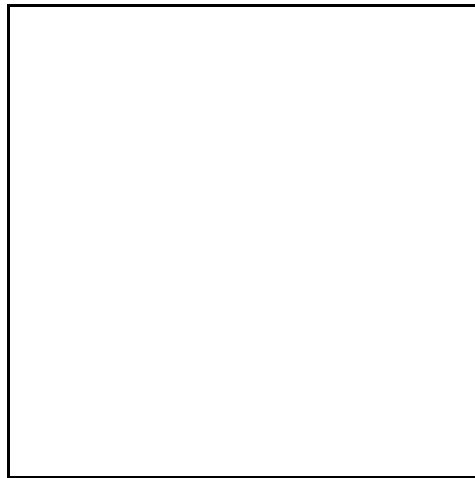
$$\Psi_{j,c(y)}^1(x) \psi_{l,n}(y)$$



$$\Psi_{j,c(y)}^2(x) \psi_{l,n}(y)$$

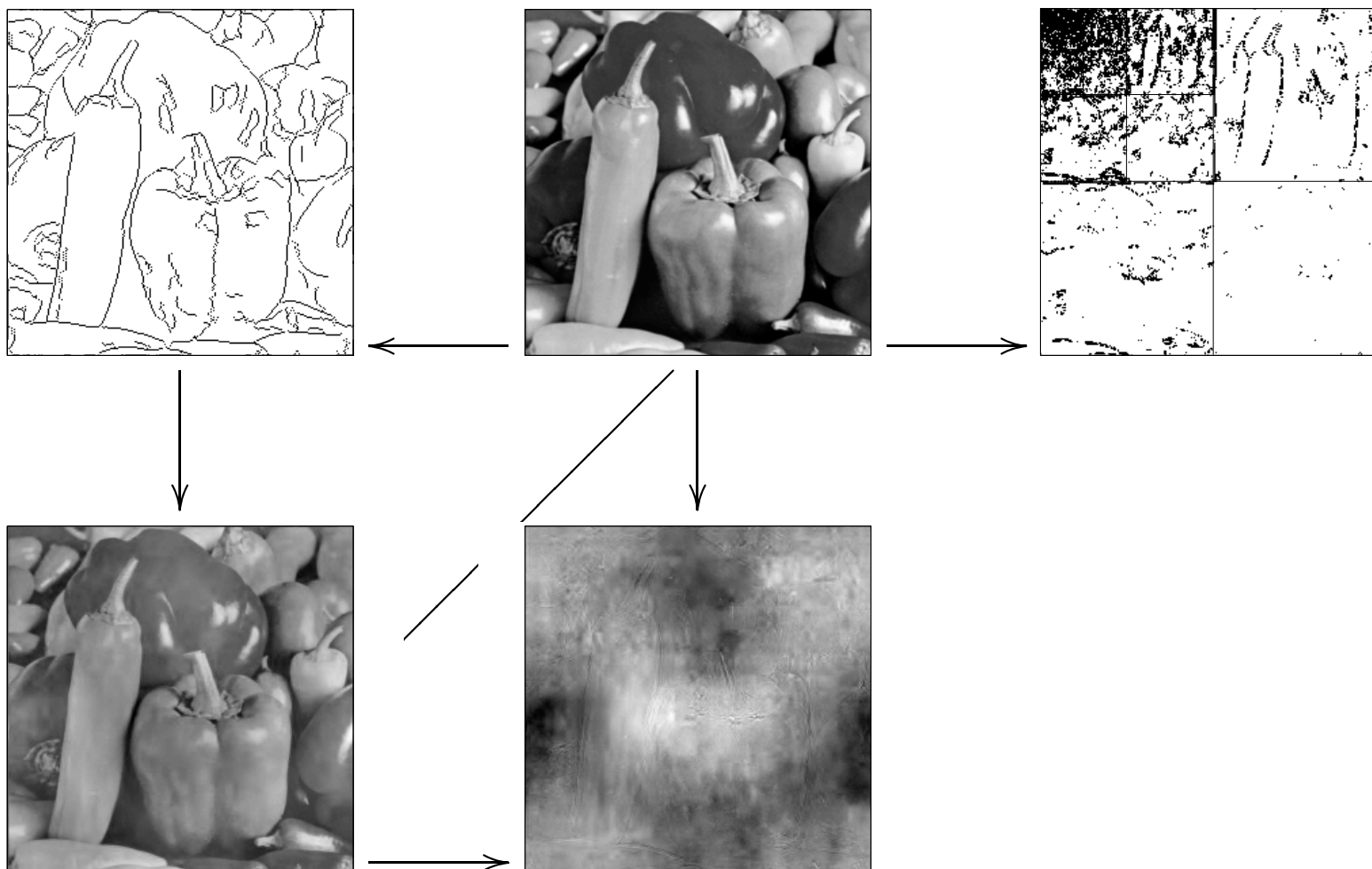


## Bandelet Frame : bandelet families and 2D wavelet basis



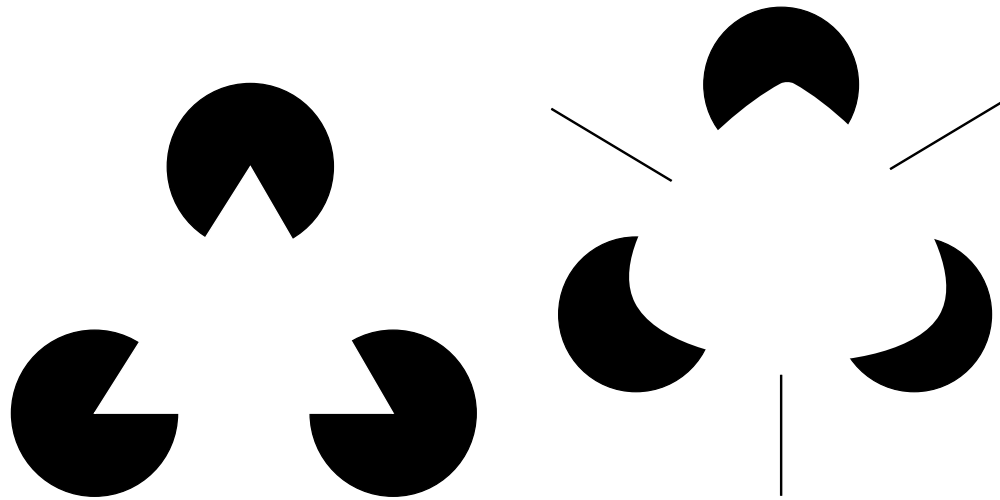
- Depends upon the choice of the  $\{c_i\}$ .

## Bandelet Frame Decomposition





## Where are the Edges?



- **Problem** : Find the edges  $\{c_i\}_i$  which minimize  $\|f - f_M\|$