

# New tools for surface analysis

Julie Digne

Joint work with Sébastien Valette, Raphaëlle Chaine and Yohann Béarzi

LIRIS - CNRS / Univ. Lyon



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# Sampled surfaces



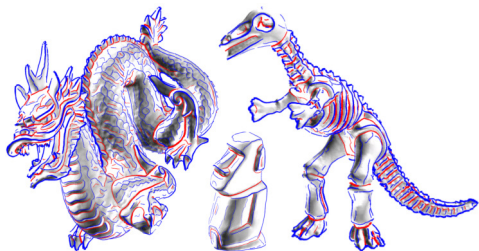
Musée de Lyon Fourvière, LIRIS, projet PAPS

# Sampled surfaces



# Local Shape Analysis

- ▶ Surface normals
- ▶ Surface Curvatures
- ▶ Curvature lines



[Ohtake et al. 2004]

## Estimation

Need to estimate differential quantities on sampled surfaces.

⇒ Can be irregularly sampled, noisy, missing data.

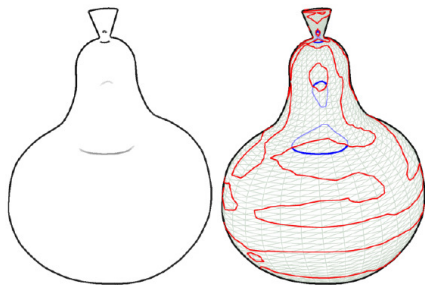
# Curvature Estimation

On point sets:

- ▶ Osculating Jets [Cazals 03], Wavejets [Béarzi 2018]
- ▶ Voronoi Curvature Measure [Mérigot 10]
- ▶ Curvature tensor estimation [Kalogerakis 07,09]

On meshes

- ▶ Curvature and Curvature derivatives estimation [Rusinkiewicz03]
- ▶ Normal Cycles [Morvan, Cohen-Steiner 03]
- ▶ Laplace Beltrami discretization [Meyer02, Wardetzky07, Vallet08]



Per point/vertex computation

# Tangent Vector Fields

## Goal

Compute a smooth tangent vector field with user-prescribed constraints optimizing some regularity criterion.

- ▶ N-symmetry direction fields [Ray 08]
- ▶ Equivalent to a Riemannian metric design problem [Lai 10]
- ▶ Smoothness constraints [Crane10,Knoppel13], symmetry constraints [Panozzo14]

More global methods: permit to constrain directions from a global point of view.

# Higher order Information?

- ▶ Curvature derivatives: helps finding suggestive contours [Rusinkiewicz 03]

## **In this talk**

Can we define principal directions of higher order, and would they reveal something on the surface?

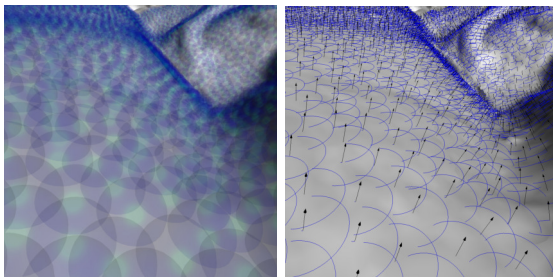
# Assumptions

## Underlying surface $S$ :

- ▶  $S$  can locally be expressed as a height field over a planar parameterization in neighborhoods of *fixed* radius  $r$
- ▶  $S$  is smooth,  $C^\infty$

## Discretization

- ▶ *Sampling condition*:  $r$ -neighborhood of a seed containing enough points.
- ▶ *Noise level*: Noise magnitude strictly below radius  $r$ .





# Local surface representation

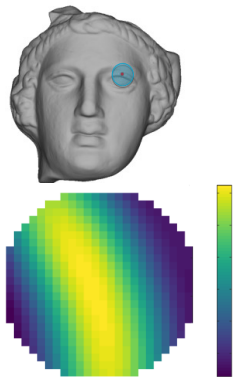
## Height-fields

- ▶ Height-field over a plane:

$$p(x, y, h = f(x, y))$$

- ▶ Taylor expansion at  $(0, 0)$

$$f(x, y) = \sum_{k=0}^{\infty} \sum_{i=0}^k \frac{1}{(k-i)!i!} \frac{\partial^k f}{\partial x^i \partial y^{k-i}}(0, 0) x^i y^{k-i}$$



# A small detour by symmetric tensors

## Def. symmetric tensor

A  $m$ -dimensional symmetric tensor  $T$  of order  $k$  is a  $m$ -dimensional array such that given index  $I = (i_j)_{j \in \llbracket 0, m \rrbracket}$ , for any permutation  $p$  on  $I$ ,  $T_I = T_{p(I)}$

- ▶  $m = 2$  let  $\mathbf{v} = (x, y)$ ,  $T = (T_x, T_y)$  symmetric tensor of order  $k$ , then  $T\mathbf{v} = xT_x + yT_y$ .
- ▶  $T\mathbf{v}$  is a symmetric tensor of order  $k - 1$
- ▶  $T\mathbf{v}^j$  is the result of contracting  $T$  by  $\mathbf{v}$   $j$  times.

# $E$ -eigenvalues of symmetric tensors

## Eigenvalues [Qi 2005,2006,2007]

Given  $T$  a symmetric tensor of order  $k$ , if there exists  $\lambda \in \mathbb{C}$  and a vector  $\mathbf{v} \in \mathbb{R}^2$  such that:

$$\begin{cases} T\mathbf{v}^{k-1} & = & \lambda\mathbf{v} \\ \mathbf{v}^T\mathbf{v} & = & 1 \end{cases} \quad (1)$$

Then  $\lambda$  is called an  $E$ -eigenvalue of  $T$  and  $\mathbf{v}$  is called an  $E$ -eigenvector of  $T$ . The set of  $\lambda$  satisfying (??) are the roots of a polynomial called the  $E$ -characteristic polynomial.

## Disclaimer

Nomenclatura: Supermatrix [Qi] or Tensor.

# Arbitrary order differential tensor

## Differential tensor

$f$  defined on  $\mathbb{R}^2$  with values in  $\mathbb{R}$ .  $T_k$  is a symmetric tensor of order  $k$ , where coefficients are as follows: let  $x_0 = x, x_1 = y$ ,

$$(T_k)_{(i_0, \dots, i_k)} = \frac{\partial^k f}{\partial x_{i_0} \dots \partial x_{i_k}}(0, 0) \quad (2)$$

$$T_k \mathbf{v}^k = \sum_{i=0}^k \binom{k}{i} \frac{\partial^k f}{\partial x^i \partial y^{k-i}}(0, 0) x^i y^{k-i} \quad (3)$$

## Writing $f$ with $T_k$

$$f(\mathbf{v}) = \sum_{k=0}^{\infty} \frac{1}{k!} T_k \mathbf{v}^k + o(\|\mathbf{v}\|^K)$$

# Tensor differentiation

## Expansion

Differentiating a symmetric tensor of order  $k$ ,  $T(\mathbf{v})$  wrt a vector  $\mathbf{v}$  yields a symmetric tensor of order  $k + 1$

## Lemma

Let  $T$  be a symmetric tensor. Let  $\mathbf{v} = (x, y)^T \in \mathbb{R}^2$  be a vector.

$$\frac{\partial T \mathbf{v}^k}{\partial \mathbf{v}} = k T \mathbf{v}^{k-1} \quad (4)$$

# Eigenvectors

## Theorem

Given  $\mathbf{v} = (x, y)$ ,  $T_k$  a real symmetric tensor of order  $k > 1$  representing the derivatives of order  $k$  of a smooth function  $f$  in  $C^k$ , the set of vectors  $\mathbf{v} = (x, y) = (r \cos \theta, r \sin \theta)$  such that  $\frac{\partial}{\partial \theta} T_k \mathbf{v}^k = 0$  and  $\|\mathbf{v}\| = 1$  are  $E$ -eigenvectors of  $T_k$ :

$$\begin{cases} T_k \mathbf{v}^{k-1} = \mathbf{v} T_k \mathbf{v}^k \\ \|\mathbf{v}\| = 1 \end{cases} \quad (5)$$

# Sketch of the proof

- ▶ Show that  $\frac{\partial}{\partial r} T_k \mathbf{v}^k = \frac{k}{r} T_k \mathbf{v}$
- ▶ Show that  $T_k \mathbf{v}^{k-1} = \frac{T_k \mathbf{v}^k}{\|\mathbf{v}\|^2} \mathbf{v}$
- ▶ Since  $\|\mathbf{v}\| = 1$  and by setting  $\lambda = T_k \mathbf{v}^k$ , we get  $T_k \mathbf{v}^{k-1} = \lambda \mathbf{v}$ .

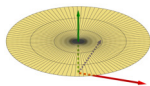
# Expressing $T$ in the *Wavejets* basis

- ▶ Switching to polar coordinates  $\mathbf{v} = (x, y) = (r \cos \theta, r \sin \theta)$
- ▶ Wavejets Basis definition:

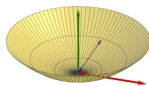
$$f(r, \theta) = \sum_{k=0}^{\infty} \sum_{n=-k}^k \phi_{k,n} B_{k,n}(r, \theta) = \sum_{k=0}^{\infty} \sum_{n=-k}^k \phi_{k,n} r^k e^{in\theta}$$



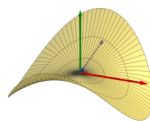
# Wavejets Basis [Béarzi et al. 2018]



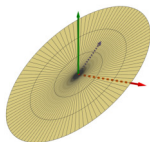
$B_{0,0}$



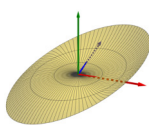
$B_{2,0}$



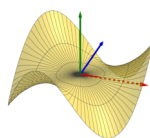
$B_{2,2} + B_{2,-2}$



$B_{1,1} + B_{1,-1}$



$B_{3,1} + B_{3,-1}$



$B_{3,3} + B_{3,-3}$

# Consequence

## Corollary

Given  $\mathbf{v} = (x, y)$ , the directions of the  $E$ -eigenvectors of a tensor  $T_k$  of order  $k$  can be retrieved out of the Wavejet decomposition of  $T_k \mathbf{v}^k$  by looking at the zeros of:

$$\frac{\partial}{\partial \theta} \sum_{n=-k}^k \phi_{k,n} e^{in\theta} = \sum_{n=-k}^k i n \phi_{k,n} e^{in\theta} \quad (6)$$

## Principal directions

For any order  $k$  we can extract eigenvectors of the  $k^{\text{th}}$  order symmetric tensor corresponding to the  $k^{\text{th}}$  order differential tensor

# Maximum and Minimum Principal Curvatures

## Definition

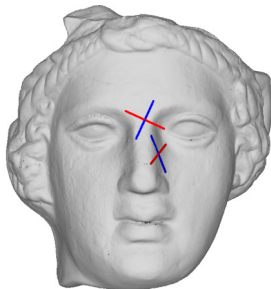
*Maximum* principal directions (resp. *minimum principal directions*) correspond to local maxima (resp. local minima) of  $g_k(\theta) = \sum_{n=-k}^k \phi_{k,n} e^{in\theta} = \frac{T_k(r \cos \theta, r \sin \theta)^k}{k! r^k}$ .

# Principal directions

- ▶ Principal curvatures:

$$\kappa_1 = 2(\phi_{2,0} + \phi_{2,2} + \phi_{2,-2}) \quad \text{and} \quad \kappa_2 = 2(\phi_{2,0} - \phi_{2,2} - \phi_{2,-2}) \quad (7)$$

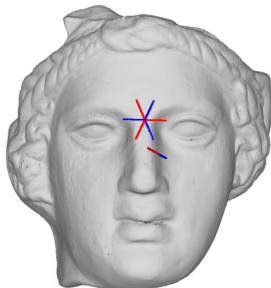
- ▶  $\sum_{\substack{-2 \leq n \leq 2 \\ n \text{ even}}} \phi_{2,n} e^{in\theta} + \phi_{2,-n} e^{-in\theta}$  has 2 maxima aligned with the principal directions



# Higher order principal directions

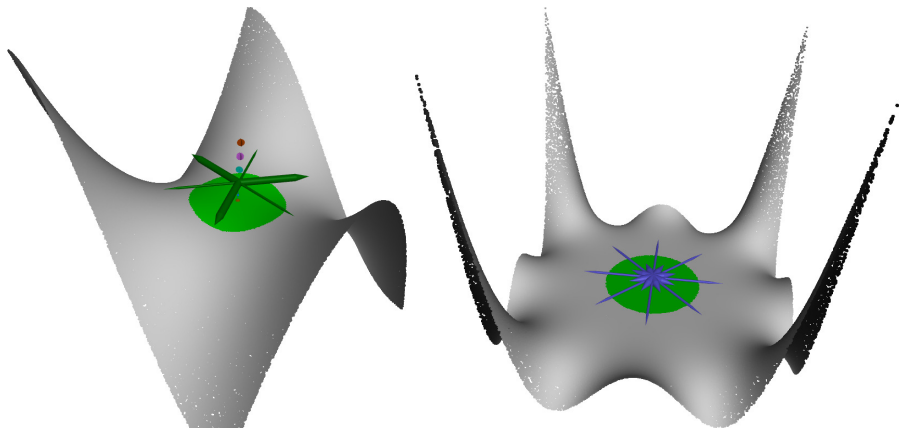
Order 3

- ▶  $\sum_{\substack{-n \leq 3 \\ n \text{ odd}}} \phi_{3,n} e^{in\theta} + \phi_{3,-n} e^{-in\theta}$  has at most 3 maxima (either 1 or 3)



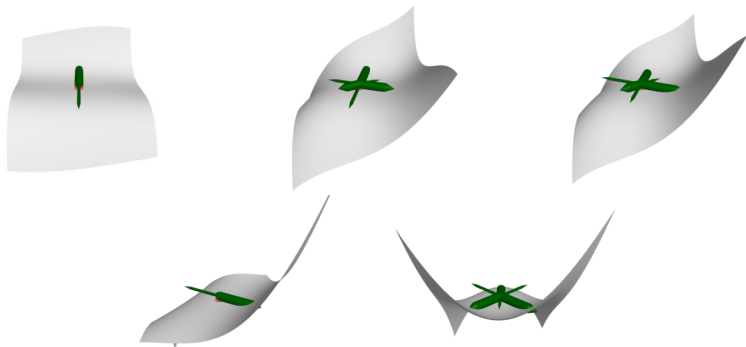
Order 3 maxima directions

# Synthetic Examples



Two synthetic surfaces with relevant principal directions of order 3 and order 8.  
Other orders vanish and exhibit no principal directions.

# Synthetic Examples



Order 3 principal directions on a synthetic surface controlled by its Wavejets coefficients.

# Properties of order $k$ directions

- ▶ If  $k$  is even: if  $\theta_0$  corresponds to a maximum principal direction,  $\theta_0 + \pi$  also corresponds to a maximum principal direction.
- ▶ If  $k$  is odd: if  $\theta_0$  corresponds to a maximum principal direction,  $\theta_0 + \pi$  corresponds to a minimum principal direction.
- ▶ At most  $2k$  principal directions of order  $k$  (roots of a real polynomial of order  $2k$ )
- ▶ Regularity: Order  $k$  principal directions are regular iff  $\phi_{k,n} = 0$  for  $n \neq \pm k$ .



# Practical computation: Truncating the Taylor Expansion

Osculating Jets [Cazals03]

- ▶ Surface parameterized w.r.t.  $\mathcal{P}(p)$  **Not necessarily equal to  $\mathcal{T}(p)$  (tangent plane)**

## Truncated Taylor expansion

$\mathcal{S}$  surface locally homeomorphic to a disk in a small neighborhood around a point  $p$ , expressed as  $f(x, y)$  over a plane  $\mathcal{P}(p)$  passing through  $p$ . The neighborhood of  $p$  can be expressed as a *truncated* Taylor Expansion at order  $K$ :

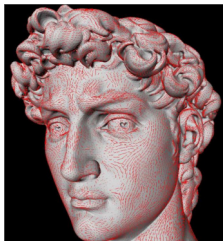
$$f(x, y) = \sum_{k=0}^{\infty} \sum_{j=0}^K \frac{f_{x^{k-j}y^j}(0, 0)}{(k-j)!j!} x^{k-j} y^j \quad (8)$$

where  $f_{x^{k-j}y^j} = \frac{\partial^k f}{\partial x^{k-j} \partial y^j}$ .

# Practical computation: Truncation order

## Accuracy theorem [Cazals03]

Given a Taylor expansion of order  $K$  in a neighborhood of radius  $r$ , the precision of all  $k$  order derivatives is  $o(r^{K-k})$ .



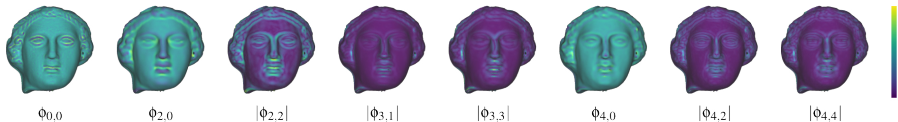
[Cazals 2003]

- ▶ In practice: Computation of the coefficients at each vertex or point by linear system solve.

# Practical computation

## Wavejets

The Wavejets expansion can be truncated similarly to the Osculating Jets expansion.



- ▶  $(r_\ell, \theta_\ell, h_\ell)_{\ell \in \llbracket 1, N \rrbracket}$  : local coordinates around  $\mathbf{p}(0, 0)$

$$\underbrace{\begin{pmatrix} B_{0,0}(r_1, \theta_1) & B_{1,-1}(r_1, \theta_1) & B_{1,1}(r_1, \theta_1) & \cdots & B_{K,K}(r_1, \theta_1) \\ B_{0,0}(r_2, \theta_2) & B_{1,-1}(r_2, \theta_2) & B_{1,1}(r_2, \theta_2) & \cdots & B_{K,K}(r_2, \theta_2) \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ B_{0,0}(r_N, \theta_N) & B_{1,-1}(r_N, \theta_N) & B_{1,1}(r_N, \theta_N) & \cdots & B_{K,K}(r_N, \theta_N) \end{pmatrix}}_M \times \underbrace{\begin{pmatrix} \phi_{0,0} \\ \phi_{1,-1} \\ \phi_{1,1} \\ \vdots \\ \phi_{K,K} \end{pmatrix}}_\Phi = \underbrace{\begin{pmatrix} h_1 \\ h_2 \\ \vdots \\ h_N \end{pmatrix}}_H$$

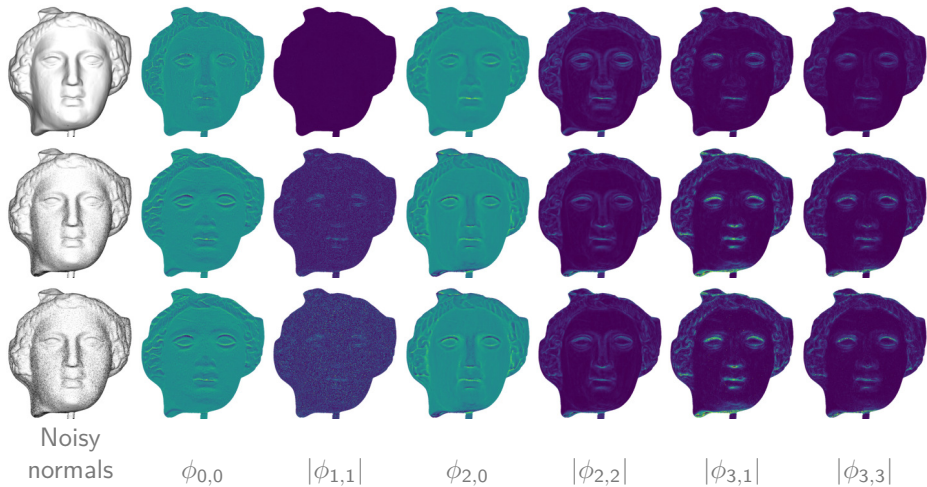
- ▶ Solve using QR decomposition

$$\operatorname{argmin}_{\Phi} \|M\Phi - H\|^2$$

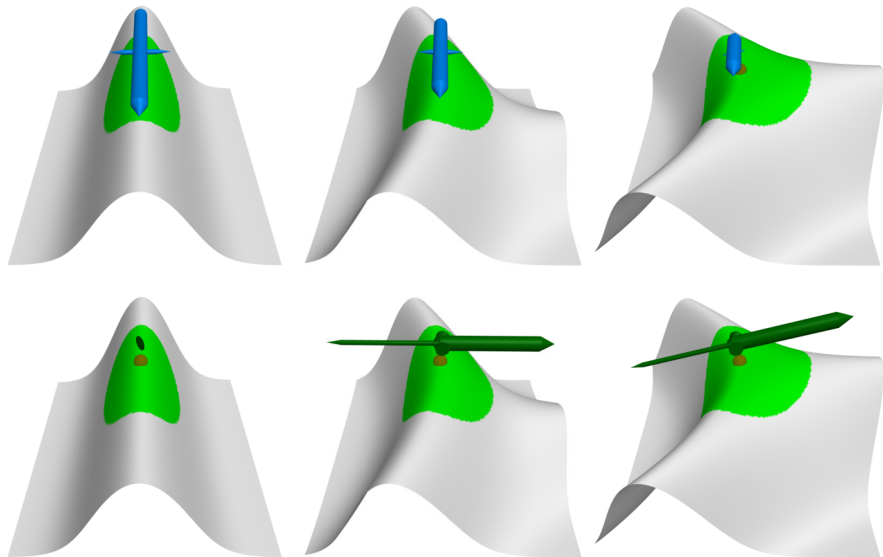
# Properties

- ▶ Adding a weight depending on the distance of the neighbor to  $p$
- ▶ If the weight is smooth and radially decreasing:
  - ▶  $\ell^2$  regression yields smooth coefficients [Levin15]....
  - ▶  $\ell^1$  no such guarantee.

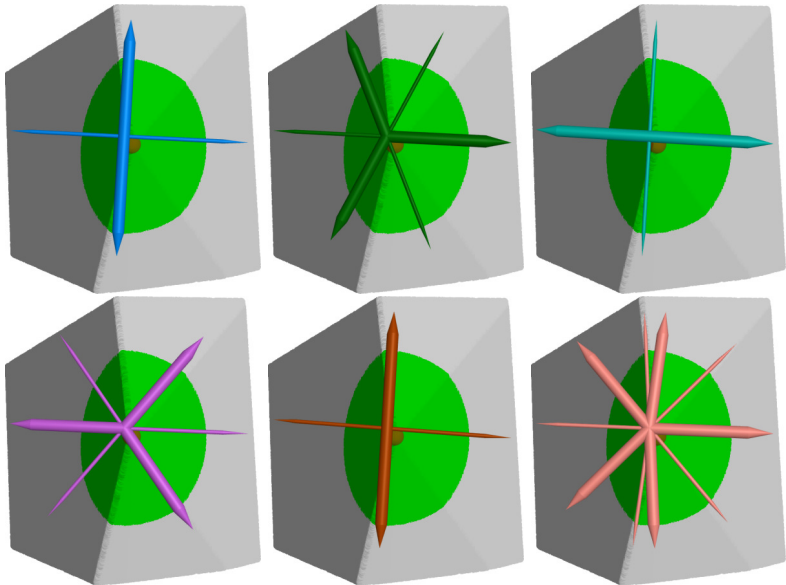
# Results



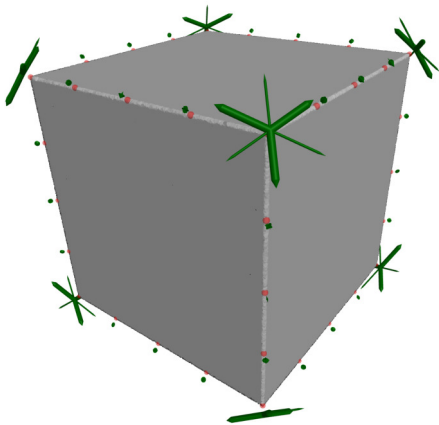
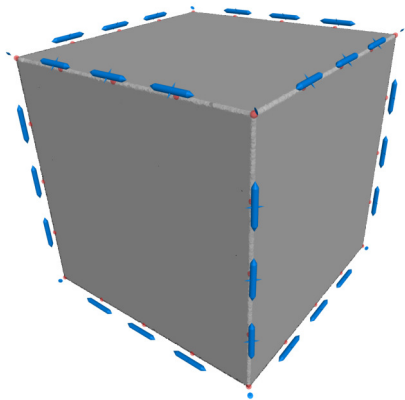
# Experiments



Order 2 (top) and 3 (bottom) principal directions on a surface evolving from a ridge (left) to a smooth T-junction (right).



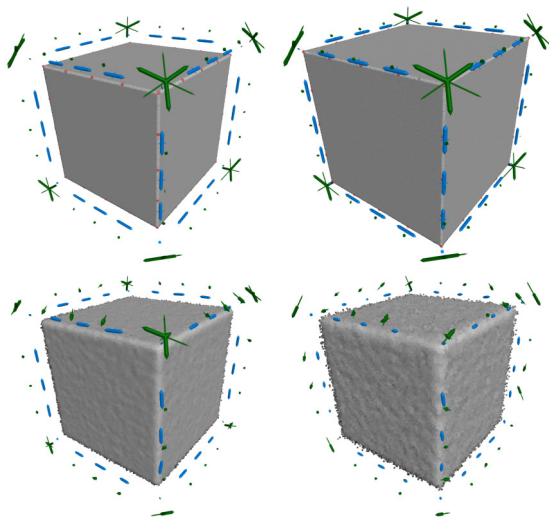
Orders 2 to 7



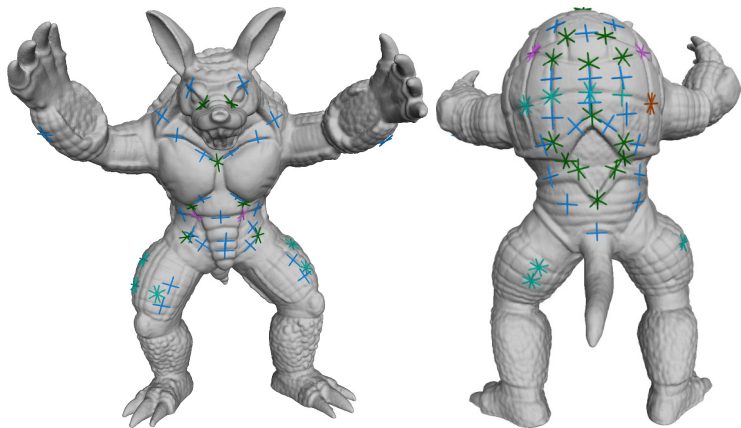
Orders 2 and 3

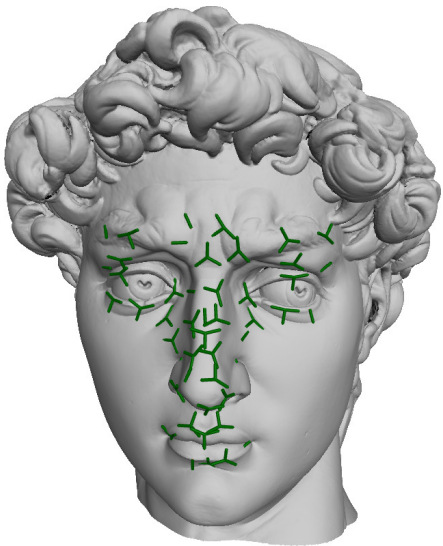
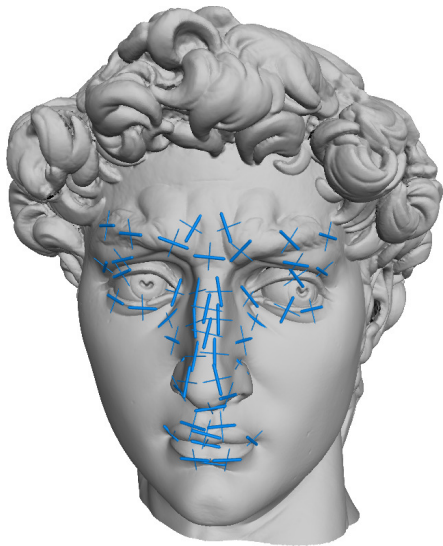


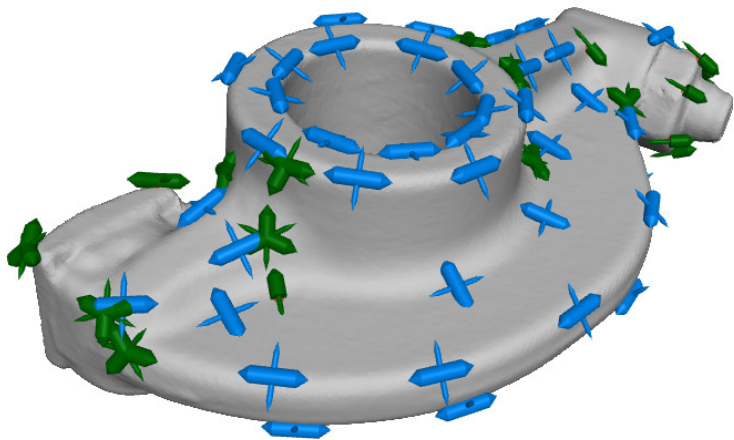
## With noise

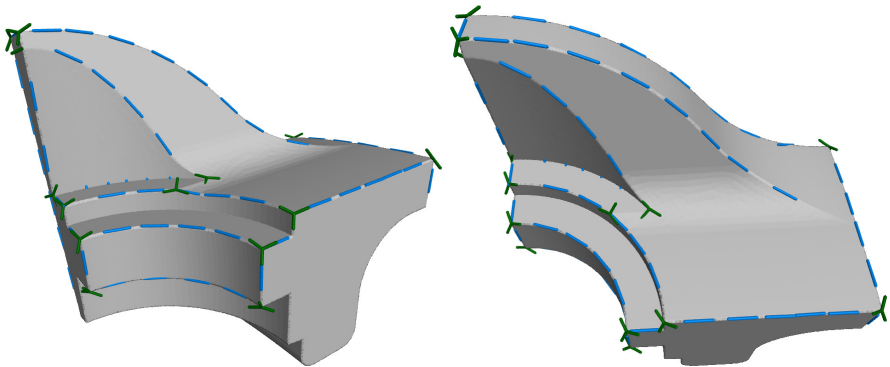


Principal directions of order 2 and 3 computed on a cube with added Gaussian noise on the positions. Top: Noiseless,  $\sigma = 0.01\%$ ; Bottom:  $\sigma = 0.05\%$  and  $\sigma = 0.1\%$

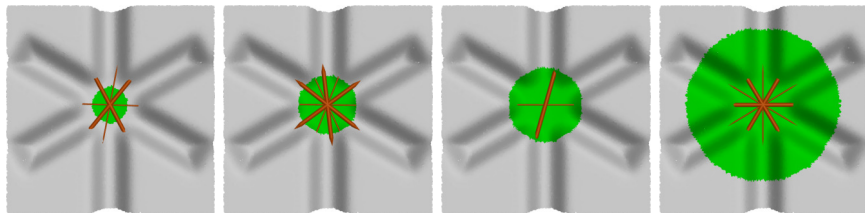








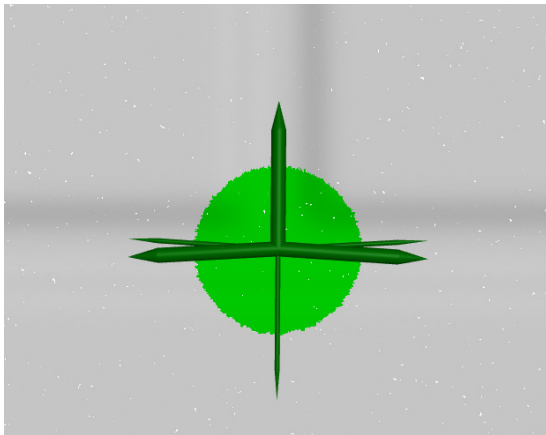
# Dependency on the radius



Estimation with  $r = 50, 80, 100, 200$ .

# Limitations

- ▶ Parameters: radius  $r$ , truncation order  $K$
- ▶ Distribution of the principal directions of a given order are not arbitrary!



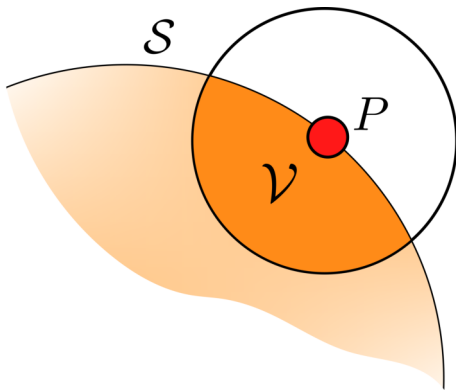
## Applications

Tracing lines on a surface with higher order junctions when directions of lower order are not defined.

# Integral Invariants

## Integral Invariants

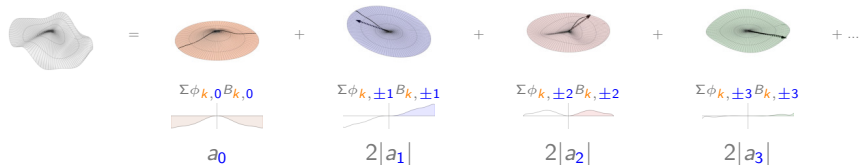
Integral quantities computed locally on a surface that are rotation and translation invariant. [Manay2006, Pottmann2007, Pottmann2009]





# Wavejets

## New integral invariants



- ▶  $V(s)$  : signed volume between surface and tangent plane [Pottman09]
- ▶  $V(s) = \int_0^{2\pi} A(\theta, s) d\theta = 2\pi a_0(s)$

$$A(\theta, s) = \int_0^s \left( \sum_{k=0}^{\infty} \sum_{n=-k}^k \phi_{k,n} r^k e^{in\theta} \right) r dr = \sum_{n=-\infty}^{\infty} a_n(s) e^{in\theta}$$

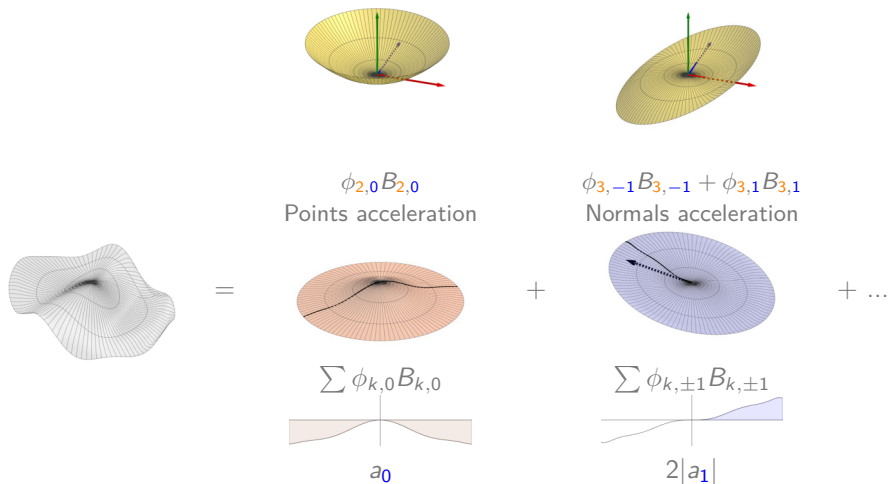
$a_n$

$$a_n(s) = \sum_{k=|n|}^{\infty} \frac{\phi_{k,n} s^{k+2}}{k+2}$$

Each  $|a_n(s)|$  is an integral invariant at scale  $s$

# Application: Detail enhancement

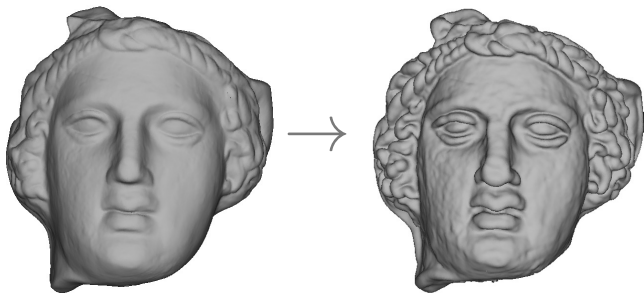
## Principle



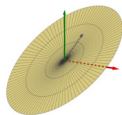
# Position based detail enhancement

- ▶ For each  $(\mathbf{p}, \mathbf{n})$

$$\mathbf{p} \leftarrow \mathbf{p} + 2\pi(1 - \alpha_0)a_0(s)\mathbf{n}$$



# Normal based detail enhancement



- For each  $\phi_{1,\pm 1}$  in point set

$$\phi_{1,\pm 1} \leftarrow \pi(1 - \alpha_{\pm 1})a_{\pm 1}(s)$$

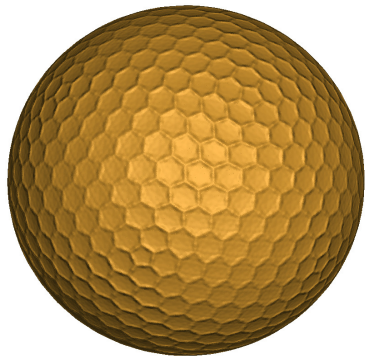
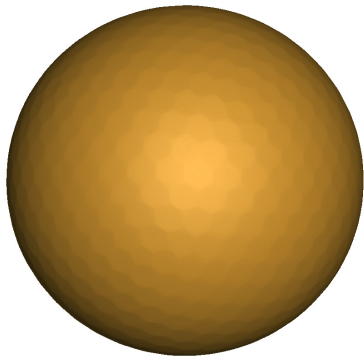
$$\sum \phi_{1,\pm 1} B_{1,\pm 1}$$

Tangent plane error



# Skewing normals

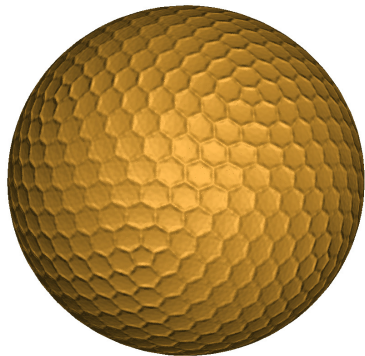
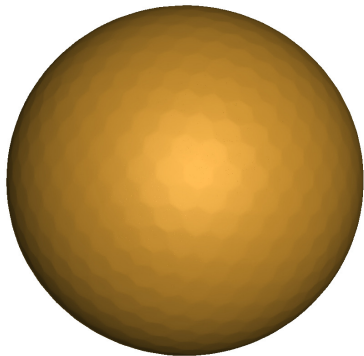
- ▶  $\alpha_{\pm 1} \in \mathbb{C}$  can skew normals orientation



$$\alpha_1 = 24$$

# Skewing normals

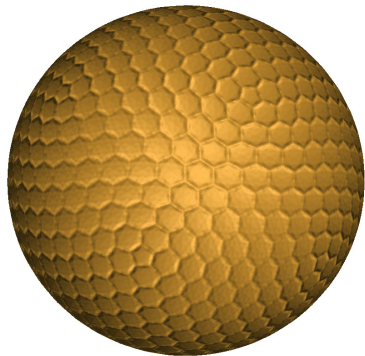
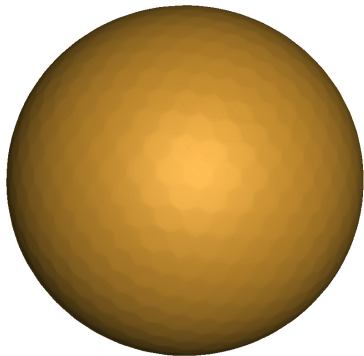
- ▶  $\alpha_{\pm 1} \in \mathbb{C}$  can skew normals orientation



$$\alpha_1 = 24e^{i\frac{\pi}{4}}$$

# Skewing normals

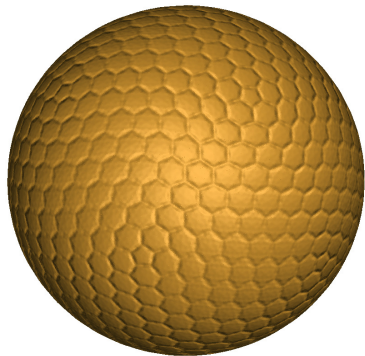
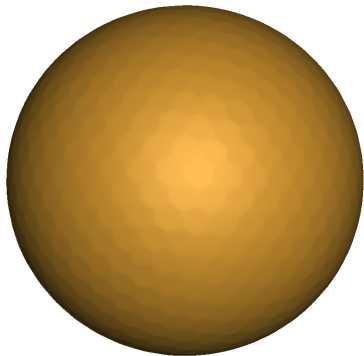
- ▶  $\alpha_{\pm 1} \in \mathbb{C}$  can skew normals orientation



$$\alpha_1 = 24e^{i\frac{\pi}{2}}$$

# Skewing normals

- ▶  $\alpha_{\pm 1} \in \mathbb{C}$  can skew normals orientation

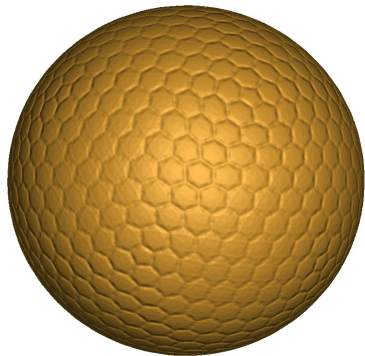
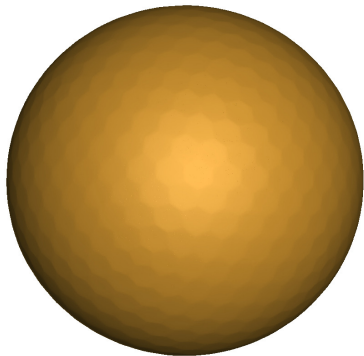


$$\alpha_1 = 24e^{3\frac{\pi}{4}}$$



# Skewing normals

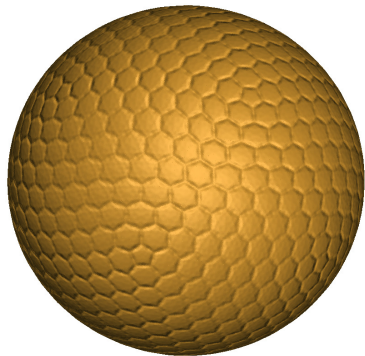
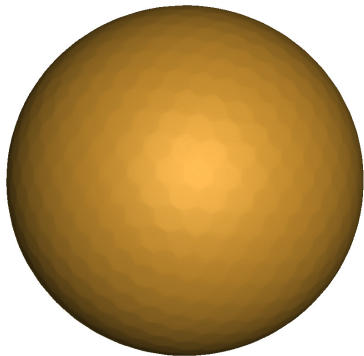
- ▶  $\alpha_{\pm 1} \in \mathbb{C}$  can skew normals orientation



$$\alpha_1 = -24$$

# Skewing normals

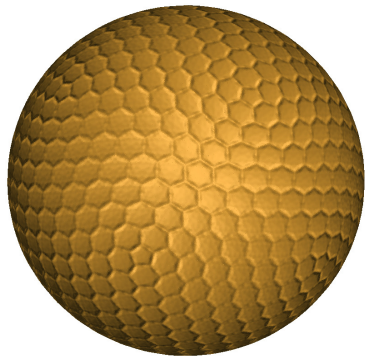
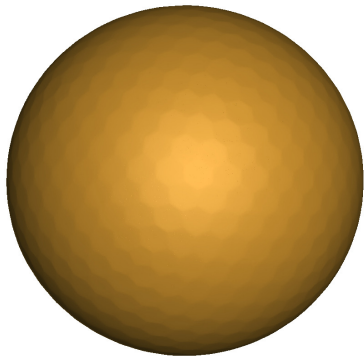
- ▶  $\alpha_{\pm 1} \in \mathbb{C}$  can skew normals orientation



$$\alpha_1 = 24e^{-3\frac{\pi}{4}}$$

# Skewing normals

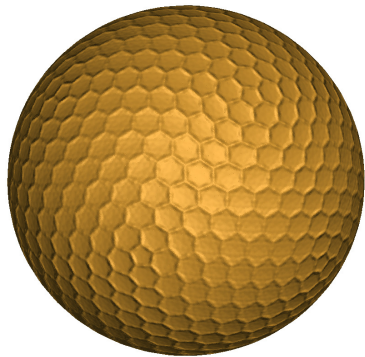
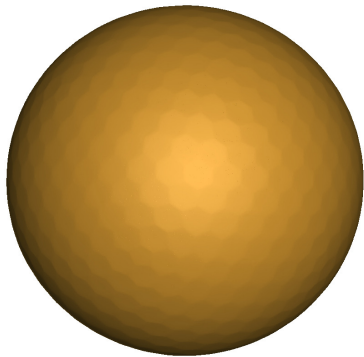
- ▶  $\alpha_{\pm 1} \in \mathbb{C}$  can skew normals orientation



$$\alpha_1 = 24e^{-\frac{\pi}{2}}$$

# Skewing normals

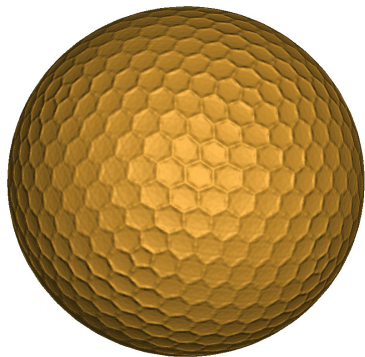
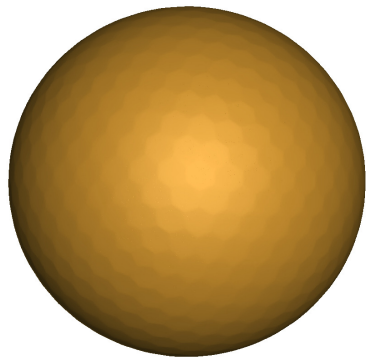
- ▶  $\alpha_{\pm 1} \in \mathbb{C}$  can skew normals orientation



$$\alpha_1 = 24e^{-\frac{\pi}{4}}$$

# Detail inversion

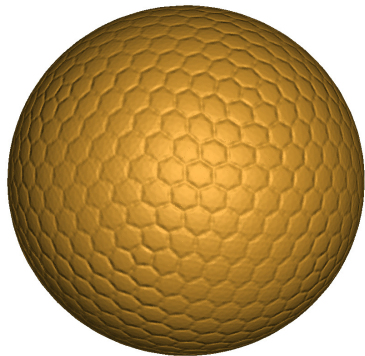
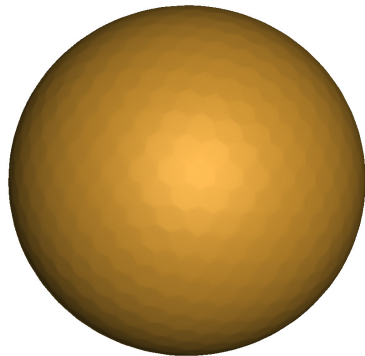
▶  $\alpha_{\pm 1} > 0$



$\alpha_1 = 24$

# Detail inversion

▶  $\alpha_{\pm 1} < 0$



$$\alpha_1 = -24$$

# Results

## Normal vs Position enhancement



Original



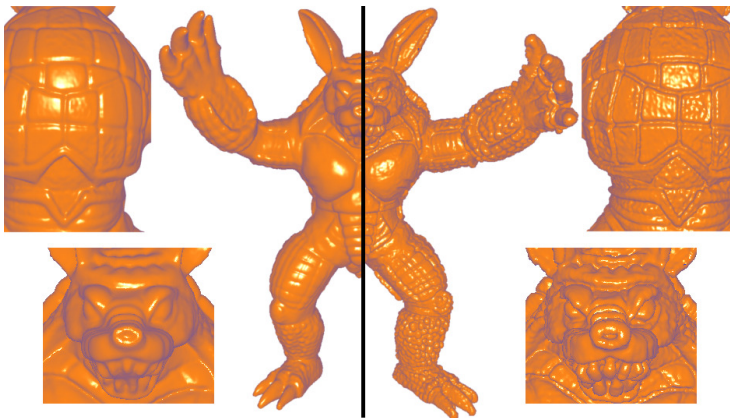
Normal



Position

# Complexity per point

- ▶ Given a set of  $N$  neighbors and a Wavejet order  $K$ :  $O(NK^4)$
- ▶ Once the wavejet is computed, applying the filter amounts to summing  $K$  terms:  $O(K)$
- ▶ 1.5M points, order  $K = 6$ : Decomposition time 1min40s; Filtering time: 0.6s.





# Conclusion

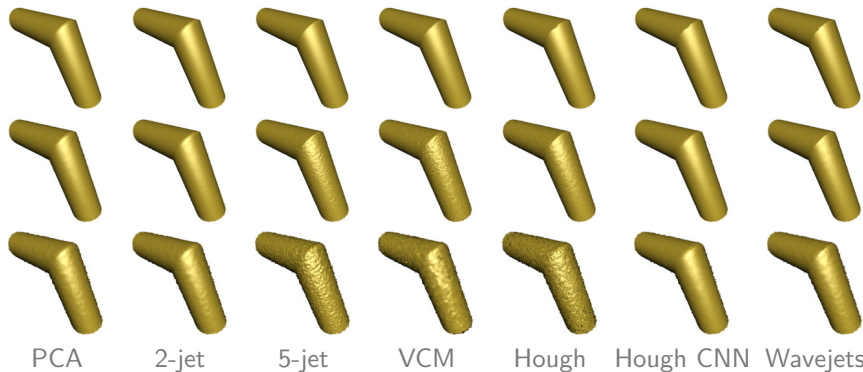
- ▶ Introduction of a new function basis
- ▶ Extension of integral invariants
- ▶ Application to geometry processing, and more to explore

Work funded by ANR Grant PAPS

[https://perso.liris.cnrs.fr/julie.digne/paps/anr\\_paps.html](https://perso.liris.cnrs.fr/julie.digne/paps/anr_paps.html)

Wavejets: A Local Frequency Framework for Shape Details Amplification, Y. Béarzi, J. Digne, R. Chaine 2018.

## Another application: normal correction



Normal estimation on two intersecting cylinders creating a sharp edge. First row : Noise free, Second row : Gaussian noise 1.2% - Third row : Gaussian noise 3.6%