

Reconstructing 3D compact sets

F. Cazals; ABS; D. Cohen-Steiner, Geometrica; INRIA Sophia-Antipolis

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Outline

Reconstruction?

The Flow Complex (A Gentle Introduction)

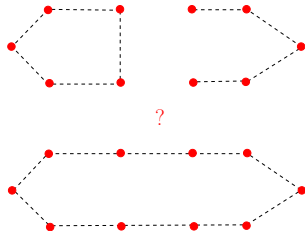
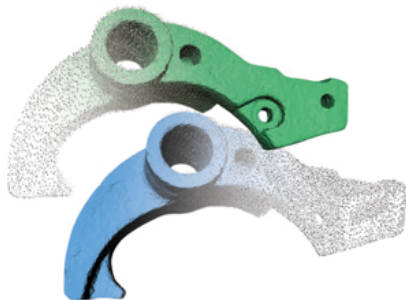
Flow Complex Based Reconstruction

Comments on Related Works

Conclusion - Outlook

Surface Reconstruction

- ▷ Pb.: Compute a (piece-wise) surface from a point cloud
- ▷ Context: Reverse Engineering, Medical data processing, Geology, Cultural Heritage



- ▷ Sc. Challenge(s): Sampling models –minimal amount of information required to reconstruct *faithfully (geometry, topology)*
- ▷ Previous work: heuristics before Amenta - Bern (98): *local Feature Size* —aka *Local Width*

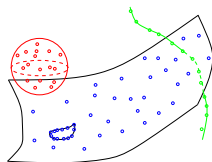
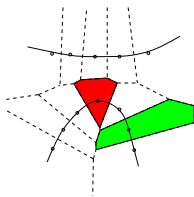
Reconstruction Difficulties

▷ Using an a priori model: classical surface-based reconstruction

- (Compact) Smooth surfaces versus sharps features, boundaries
- Samples may not comply with the model:
 - Unstable estimates for differential geometry based quantities:
normals, principal curvatures
 - The sampling may not be a surface at all
- Unique versus plausible reconstructions

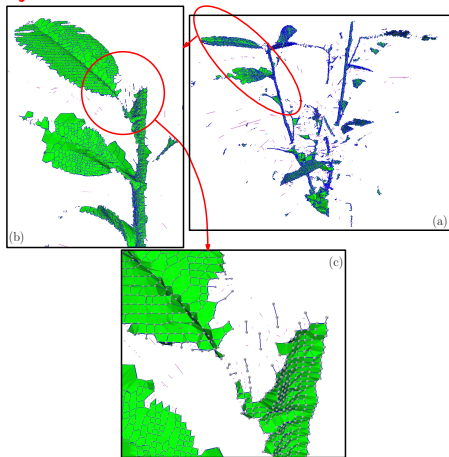
▷ Calls for a general reconstruction strategy:

- No a priori: handling compact sets
- Providing multi-scale reconstruction i.e. plausible shapes



Practically: Reconstructing the Boundary of a Solid?

- ▶ Thin parts might just be too thin



- ▶ Calls for a general reconstruction strategy

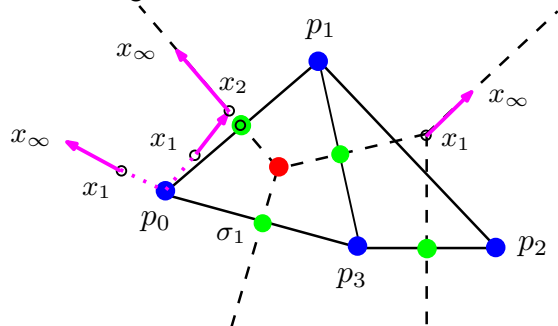
Beech tree reconstruction, with parameters $t_r = 2.2$, $t_p = 1.1$ (a) Overview of this noisy and under-sampled model (b,c) Zoom near a an under-sampled peduncle. The point cloud is courtesy of J-C. Chambelland et al, UMR 547 PIAF - INRA/UBP.

Distance Function to Sample Points

- ▷ Given a collection of points $\{p_i\}_{i=1,\dots,n}$, consider:

$$d(p) = \min_i \|pp_i\|$$

- ▷ Increasing the value of d :

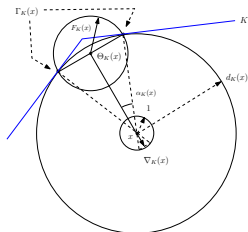


- minimum ●
- saddle ●
- maximum ●

Flow trajectories
(orbits)

Distance function to a compact set K

▷ Distance function to a compact set K



– Distance function :

$$d_K(x) = \min_{y \in K} d(x, y)$$

– Gradient : $\nabla d_K(x) = \frac{x - \Theta_K(x)}{d_K(x)}$

– Its norm:

$$\|\nabla d_K(x)\| = \cos \alpha_K(x)$$

▷ Key properties

∇d_K can be integrated to define a continuous flow (Euler scheme converges)

(Non degenerate) Critical point:
(interior) of the convex hull of $\Gamma_K(x)$

Easy interpretation if $K = \text{point cloud}$:
cf Delaunay-Voronoi diagrams

The Morse Puzzle of the Distance Function

Flow Complex

Critical points

Stable manifolds

Unstable manifolds

Morse-Smale diagram

regions of homogeneous flow:

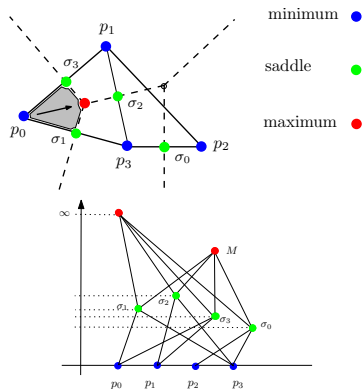
\cap of (un)stable manifolds

Hasse diagram

orbits of the distance function
through consecutive crit. pts

▷ Ref: Giesen, John; ACM SODA; 2003

▷ Ref: Cazals, Pion; ACM SoCG; 2008



Construction of S2 and U1

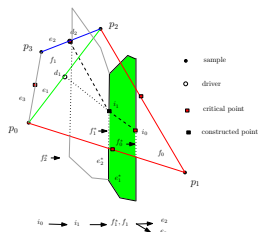
- ▷ Stable manifolds, generically:

Recursive structure

Complicated case S2:

neither in Delaunay nor in Voronoi

- ▷ S2:



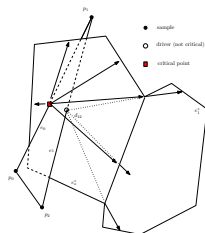
▷Ref: Giesen and John, ACM SODA 03

▷Ref: Giesen, Ramos, Sadri; ACM SoCG 06

▷Ref: Cazals, Parameswaran, Pion; ACM SoCG 08

- ▷ Stable manifolds, generically:

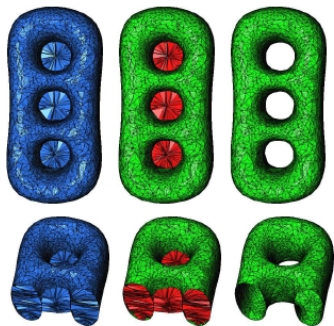
U1: dissection of the two-skeleton



Flow Based Surface Reconstruction: Separating Critical Points

▷ ϵ -samples accommodate a separation of critical points:

Flow complex: 2-skeleton Tagged critical points:
surface vs medial axis Reconstruction



Algo. based on the separation of c.p.:
surface c.p.: angle criterion
MA c.p.: distance criterion wrt poles

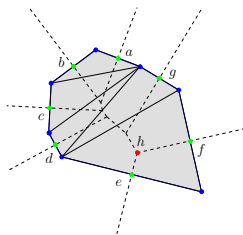
Under mild hypothesis on the sampling:

Reconstruction \hat{S} :
 $\hat{S} \subset \text{tube around } S$
 \hat{S} isotopic to S

▷Ref: Dey, Giesen, Ramos, Sadri; ACM SoCG 2005

Generalization to Compact Sets: Finding Cuts from the Hasse Diagram

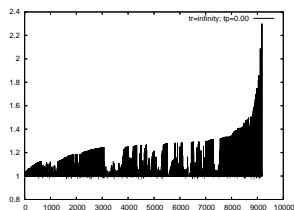
- ▶ **Reconstruction:**
 - Collection of Stable Manifolds (SM) of all indices i.e. 0 to 3
 - Governed by a parameter $t_r > 0$
 - Abusing terminology: inserting a Hasse node \sim inserting its SM
- ▶ **Key ingredient:** ratio $V(b)/V(a)$
between crit. values of incident crit. pts in the Hasse diagram
- ▶ **Initialization:** selected Gabriel edges
- ▶ **Upflow extension:**
 a in reconstruction; $index(b) = index(a) + 1$
 a sponsors b with priority $r_u = V(b)/V(a)$
- ▶ **Regularization:**
 h in reconstruction
 h triggers insertion of a, b, c, d, e, f, g ; $r_r = 1$
- ▶ **Horizontal extension:**
 a in reconstruction; $index(b) = index(a)$
 a sponsors b with priority $r_h = \max(V(a)/V(b), V(b)/V(a))$



The Three Steps: Canonical Ordering of Insertions

Reconstruction Profiles

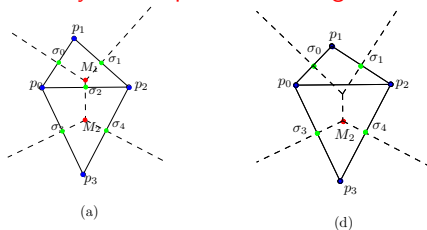
- ▶ **Initialization:** start from selected one-dimensional SM
 - Gabriel edge $e = (v_0, v_1)$ is an init edge iff v_1 is the nearest neighbor of v_0 or vice-versa
- ▶ **Sponsoring as an iterative process:**
 - sponsored nodes placed into a priority queue Q
 - priority : least r_u, r_r, r_h ratio from any sponsor
 - requires insertion and/or updates of nodes already in Q
 - induces a canonical insertion of nodes: take the **easiest** step
i.e. **pop node with least priority provided it is $< t_r$**
- ▶ **Reconstruction profile:** $(r_i)_{i \geq 1}$ for $t_r = \infty$
 - encompasses all possible reconstructions



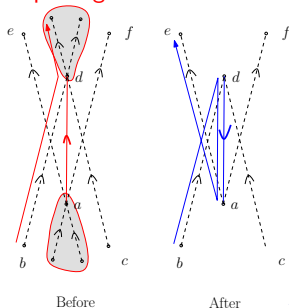
Example recons. profile (more later!)
Note: tail of diagram \sim features
(significant maxima) of the model

Widening the Gaps: Persistence on the Hasse Diagram

- ▷ Insufficient sampling: widen the gaps
- ▷ Cancelling pairs of *nearby* critical points: reverting the flow



- ▷ Hasse simplification: multiplexing and redistribution of SMs

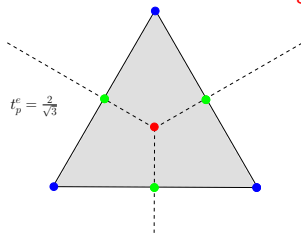


Hasse Simplification Cont'd

▷ Algorithm

- Iterative process based on the **persistence priority** $\rho(a, b) = V(b)/V(a)$
- Incremental cancellation of all pairs up to priority $\leq t_p$

▷ Persistence: the obtuse triangle



$t_p > t_p^e$ cancels edges-triangle

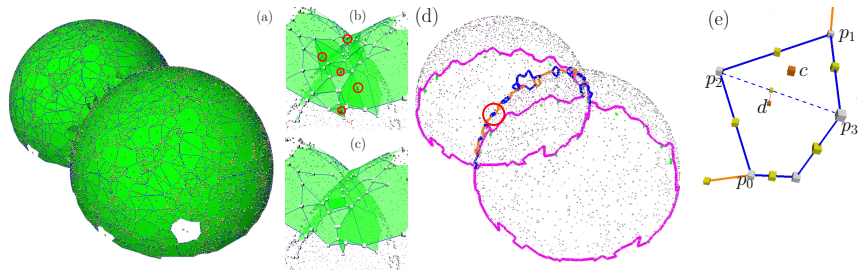
▷ Key steps

- Update of the Hasse diagram
 - remove edges incident on nodes cancelled
 - add edges of bipartite graph
 $\text{In}(b) \times \text{Out}(a)$
- Redistribution of SM

- surface with boundaries:
 - may create leaks of SM
- but not systematic:
 - SM rescued thanks to multiplexing

Non Manifold Reconstruction

▷ Reconstructing two intersecting spheres

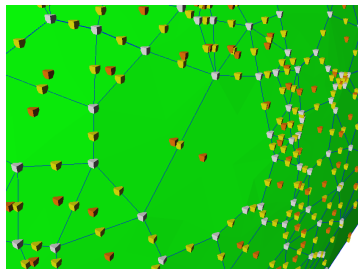
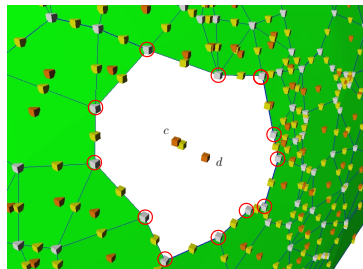


- (a) $t_r = 1.9, t_p = 0$
- (b) Transparent view of (a)
- (c) $t_r = 2.5, t_p = 1.05$:
maxima cancelled

- (d) $t_r = 2.5, t_p = 1.05$,
multiplicity of Gabriel edges:
zero one three four five
- (e) Circled region of Fig. (d):
intersection curve stretched to a disk

Enumerating Plausible Reconstructions

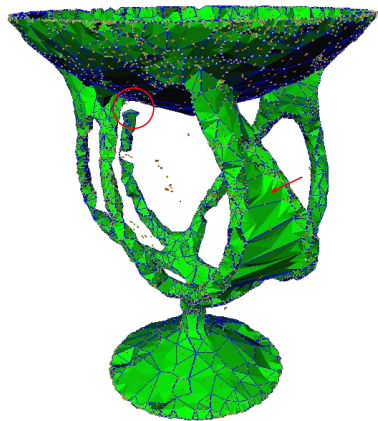
▷ Do the circled points punch a hole or not?



(a) At $t_r = 1.9, t_p = 0$: hole (b) At $t_r = 2.1, t_p = 0$: hole filled

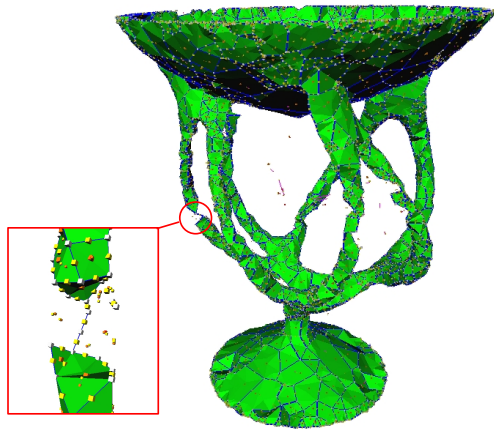
On the Importance of Persistence

▷ Undesirable extensions



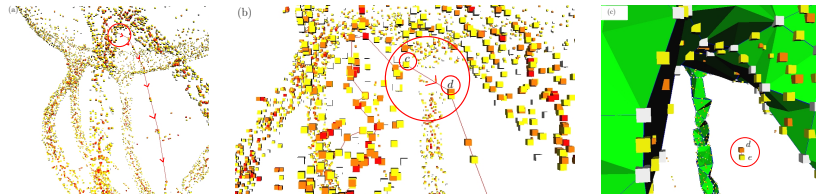
(a) At $t_r = 1.7$, $t_p = 0$: fin

▷ Fixed by persistence



(b) At $t_r = 2$, $t_p = 1.02$: fin is gone

Untangling the Role of Persistence



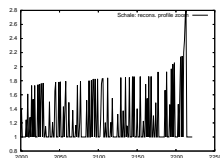
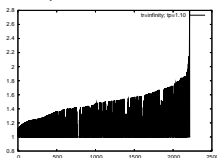
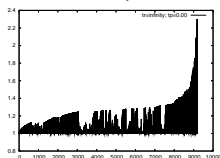
(a,b) The extension path followed by the algorithm—red arrows.

(c) Pairing by persistence: d paired to an index one critical point e

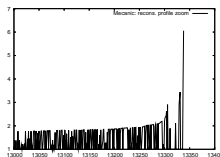
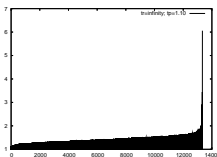
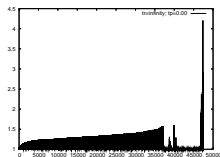
▷ **Note:** offers a local control so as to fix the sampling

Combining Parameters t_r and t_p : the Exple of Vase

- ▷ **Vase:** (a) $t_p = 0, t_r = \infty$ (b) $t_p = 1.1, t_r = \infty$ (c) Tail of (b)



- ▷ **Mechanic:** (a) $t_p = 0, t_r = \infty$ (b) $t_p = 1.1, t_r = \infty$ (c) Tail of (b)



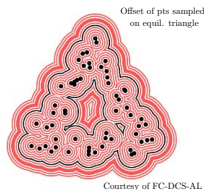
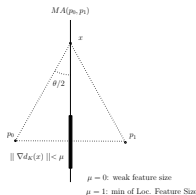
▷ Recommendations

- values of $t_p \in [1.02, 1.2]$ and $t_r \in [1.9, 2.1]$ yield comparable results
- beyond $t_r \in [1.9, 2.1]$: number of features of the model (significant max.)

Reconstruction Guarantees: Background

▷ **Def.** The μ -medial axis of a compact set $K \subset \mathbb{R}^n$ is the set of points $x \notin K$ such that $\|\nabla d_K(x)\| < \mu$. The μ -reach of K , denoted by $r_\mu(K)$, is the minimum distance between a point in K and a point in the closure of its μ -medial axis.

▷ **Def.** Given two non-negative real numbers κ and μ , we say that a compact set $P \subset \mathbb{R}^n$ is a (κ, μ) -approximation of a compact set $K \subset \mathbb{R}^n$ if the Hausdorff distance between K and P does not exceed κ times the μ -reach of K .



▷ **Thm.** Reconstructing with offsets:

- $P \subset \mathbb{R}^n$ is a (κ, μ) -approximation of a compact set K
- Under *suitable* conditions:

P^α is homotopy equivalent to K^η for sufficiently small η .

Reconstruction Guarantees

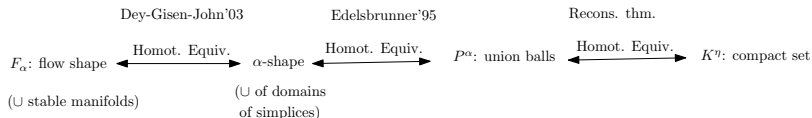
▷ **Def.** Point cloud P is a ρ -uniform approximation of a compact set K if half the distance between the two closest sample points in P is at least ρ times the Hausdorff distance between P and K , where $0 < \rho < 1$.

▷ **Thm.** Let K be a compact subset of \mathbb{R}^3 ;
Assume P is a ρ -uniform (κ, μ) -approximation of K . If

$$\frac{4}{\rho\mu^2} < t_r < \frac{\mu^2}{4\kappa} - 1,$$

then the reconstruction is homotopy equivalent to K^η for small enough η .

▷ **Proof sketch:**



▷ **Note.** Thm uses uniform sampling; algorithm does not.

Union of balls, α -shapes, Flow-shapes

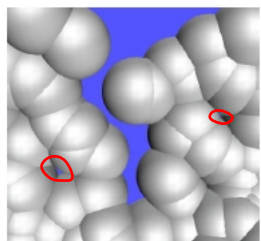
- ▷ **Topology of a union of balls:** union of balls \sim flow shape \sim α -shape

union of balls and α -shapes are homotopy equivalent [Edelsbrunner, 92]

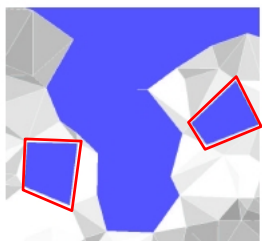
α -shapes and flow shapes are homotopy equivalent [Dey, Giesen, John, 03]

- ▷ **Key differences**

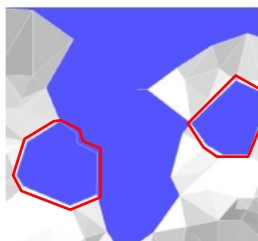
Events triggering addition in α -shape: Gabriel simplices
critical points



Union of balls



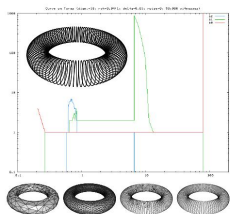
α -complex / α -shape



Flow shape

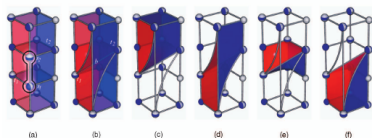
Connexions to Recent Previous Work

▷ Multi-scale reconstruction



▷Ref: Guibas, Dudot; DCG 08

▷ Topological simplification of 3D scalar functions



▷Ref: Gyulassy et al., IEEE TVCG, 06

- Based on the witness complex
- Incremental construction of subset of the Del. triangulation of $L \subset W$, and associated Betti numbers
- complex is the ν -witness complex $C_\nu^W(L)$
- Provably correct reconstruction of curves and surfaces contains a manifold isotopic to the surface

→ approach radically different; we handle non manifold shapes

Morse-Smale diagram simplification

- Operations: cancellations $(m, \sigma_1), (\sigma_1, \sigma_2), (\sigma_2, M)$
- Simplified complex is that of some scalar function: topological constraints respected

→ Our simplified MS diagram is not *realized*

Conclusion - Outlook

- ▶ **Flow complex based reconstruction**
versatile strategy (cf strata)
complex to build, yet tractable for models of (moderate) size
- ▶ **Developments called for**
computing stratifications
handling noise
approximation scheme based on Delaunay (some restricted Delaunay)
gazing on the high dimensional side. . .

