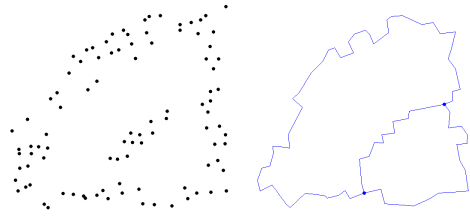


Homologically Persistent Skeleton for a 2D cloud of features

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Motivation: fast and robust recognizing visual markers



- simple pictures like hieroglyphs are easily readable by humans;
- machines should correctly reconstruct skeletons from noisy scans.

Skeletonization problem:

Given only a cloud C of points, find a graph representing the topology of C across all scales.

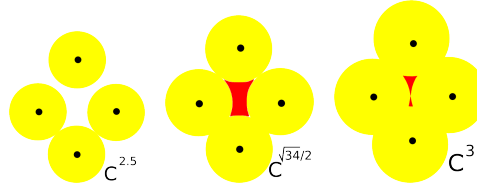
All past methods use extra input parameters: a scale, noise bound.

New solution: Homologically Persistent Skeleton $\text{HoPeS}(C)$

- extends a Min Spanning Tree;
- depends only on the cloud C ;
- has *min length* among all graphs that span C at any scale and also have most persistent 1D cycles;
- has derived subgraphs with the correct topological type of a graph $G \subset \mathbb{R}^2$ given only by a sample C ;
- is *globally stable* under perturbations of C , remains in an offset C^α within a distance α from $C \subset \mathbb{R}^2$.

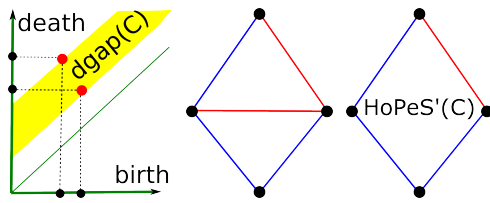
Def: $\text{HoPeS}(C)$ of a cloud C

When the scale α is increasing, the offset C^α is growing, cycles in C^α are born and die (become filled).

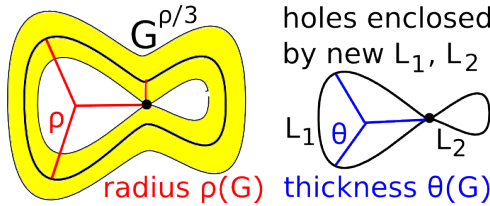


The pairs (birth, death) form the 1D persistence diagram $\text{PD}\{C^\alpha\}$.

Earlier applications of persistence: counting holes in clouds [2], auto-completing closed contours [3].



$\text{HoPeS}(C)$ is a Min Spanning Tree of C plus all red edges giving birth to new cycles in C^α across all α .



If cycle $L \subset G$ encloses a hole, its *radius* ρ is $\min \alpha$ when L^α is contractible. The *thickness* $\theta(G)$ is \max radius of all newborn cycles in G^α .

Reconstruction guarantees

An ε -sample C of a graph $G \subset \mathbb{R}^2$ is a cloud $C \subset G^\varepsilon$ with $C^\varepsilon \supset G$.

Th (VK). Let C be an ε -sample of a graph $G \subset \mathbb{R}^2$ with $\theta(G) \geq 0$ and radii $\rho_1 \leq \dots \leq \rho_m$. If $\rho_1 > 7\varepsilon + \theta(G) + \max_{i=1, \dots, m-1} \{\rho_{i+1} - \rho_i\}$, then

$\text{HoPeS}'(C) \sim G$ is 2ε -close to G .

[4]: extension to a metric space.

Summary and References

input: a noisy point cloud $C \subset \mathbb{R}^2$

output: full skeleton $\text{HoPeS}(C)$ with all derived subgraphs;

running time: $O(n \log n)$ for any n points with real 2D coordinates;

more details: C++ code, papers, examples at <http://kurlin.org>.

[1] H. Edelsbrunner and J. Harer. *Computational topology*. AMS'10.

[2] V. Kurlin. A fast and robust algorithm to count persistent holes in noisy 2D clouds. *CVPR 2014*.

[3] V. Kurlin. Auto-completion of contours in sketches, maps and sparse 2D images. *CTIC 2014*.

[4] V. Kurlin. Homologically Persistent Skeleton of a point cloud in any metric space. *Computer Graphics Forum*, v. 34-5 (2015), 253-262.

1) cloud C of Canny edge points. 2) $\text{PD}\{C^\alpha\}$. 3) $\text{HoPeS}_{1,1}(C)$ for 1st widest gap. 4) $\text{HoPeS}_{2,1}(C)$ for 2nd widest gap.

