

Graph Reconstruction by Discrete Morse Theory

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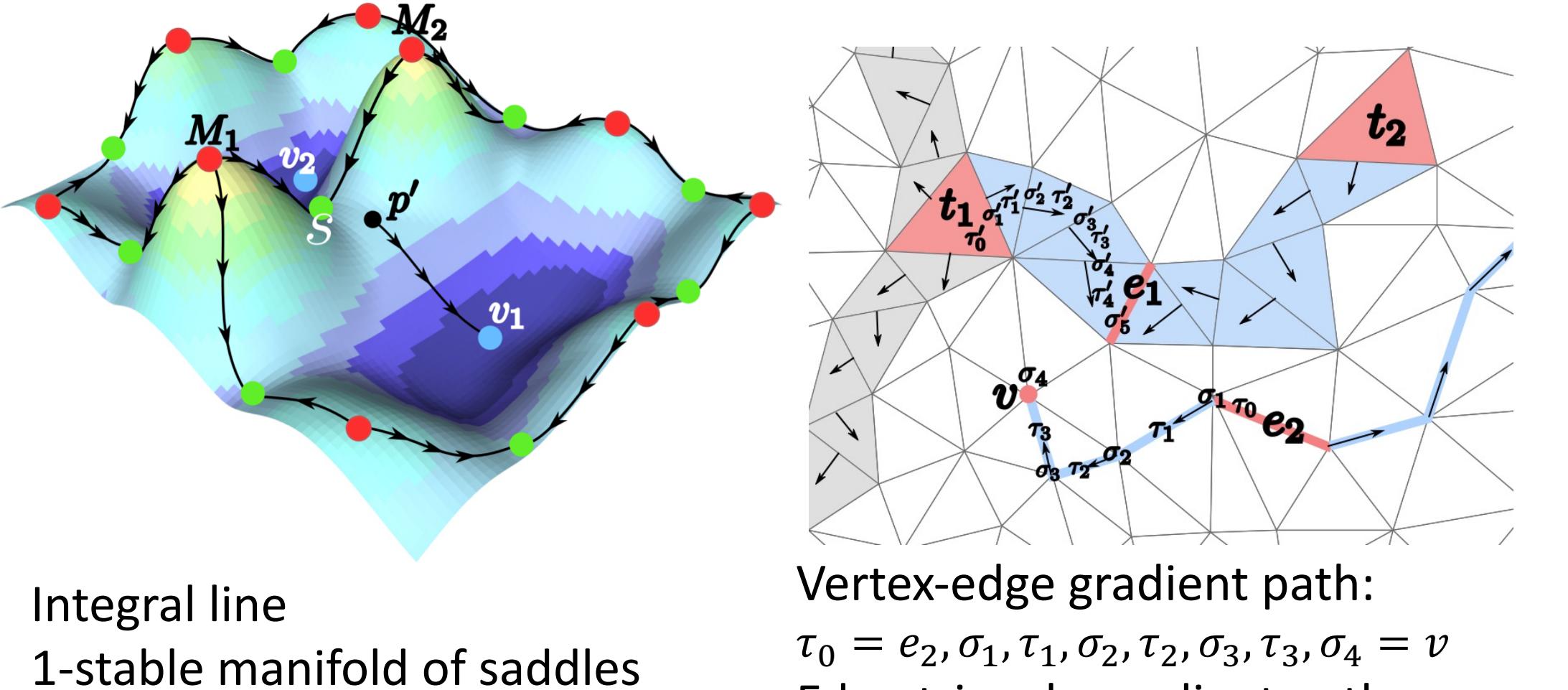
Problem: Recovering hidden structures from potentially noisy data.

Applications include reconstructing the road networks from GPS trajectories, inferring the filament structure of simulated density field of dark matters in universe.

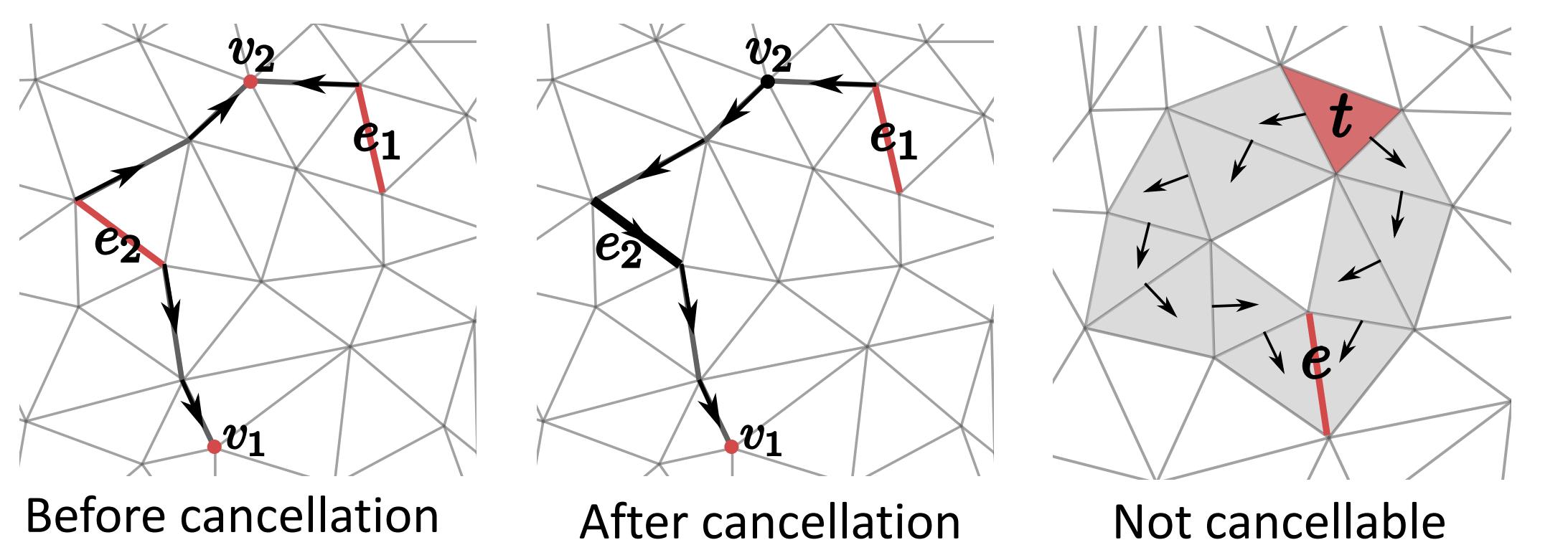
Our contributions:

1. Simplify the existing algorithm.
2. Provide an editing strategy to add missing branches and loops for the 2D case.
3. Give the Geometric and topological reconstruction guarantees under a noise model

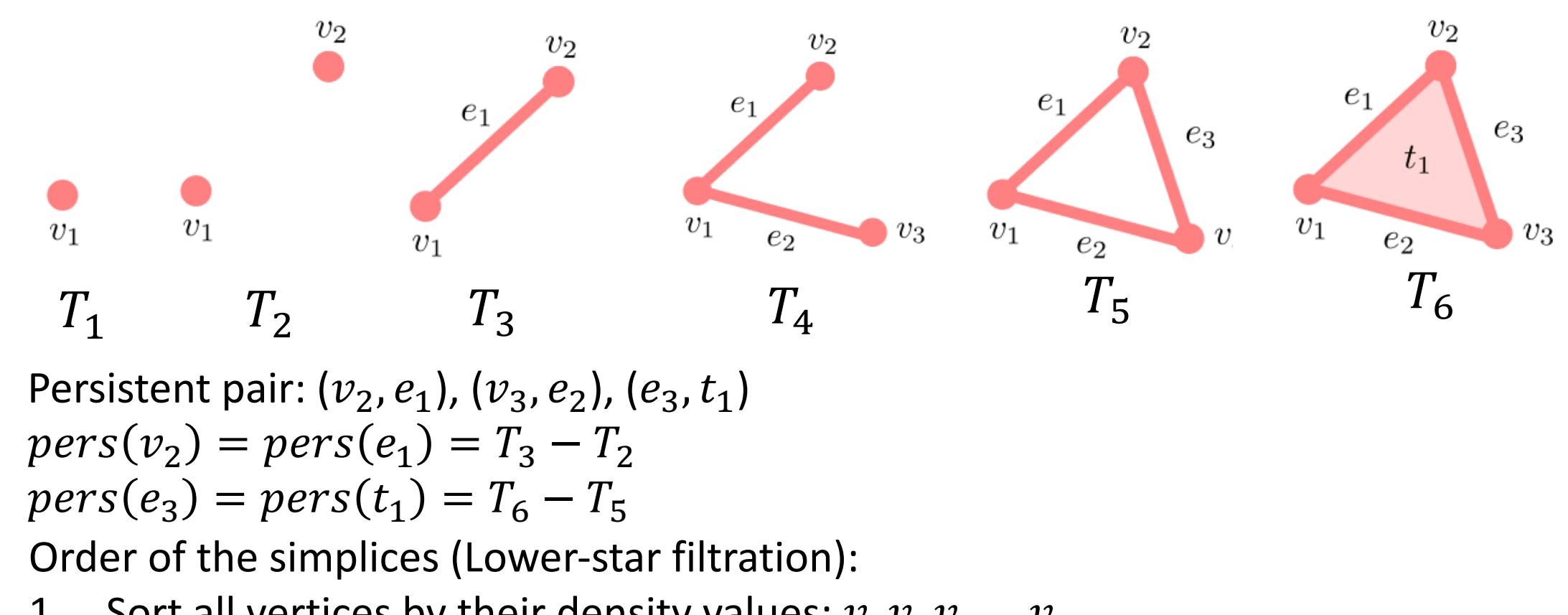
Morse and Discrete Morse theory



Morse cancellation



Persistence pair



Existing algorithm

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Algorithm 1: MorseRecon( $K, \rho, \delta$ )
Data: Triangulation  $K$  of  $\Omega$ , density function  $\rho : K \rightarrow \mathbb{R}$ , threshold  $\delta$ 
Result: Reconstructed graph  $\hat{G}$ 
begin
1 | Compute persistence pairings  $P(K)$  by the lower-star filtration of  $K$  w.r.t  $g_\rho = -\rho$ 
2 |  $M = \text{PerSimpVF}(P(K), \delta)$ 
3 |  $\hat{G} = \text{CollectOutputG}(M)$ 
4 | return  $\hat{G}$ 

Procedure PerSimpVF( $P(K), \delta$ )
1 | Set initial discrete gradient field  $M$  on  $K$  to be trivial
2 | Rank all persistence pairs in  $P(K)$  in increasing order of their persistence
3 | for each  $(\sigma, \tau) \in P(K)$  with  $\text{pers}(\sigma, \tau) \leq \delta$  do
4 |   If possible, perform discrete-Morse cancellation of  $(\sigma, \tau)$  and update the discrete
      gradient vector field  $M$ 
5 | return  $M$ 

Procedure CollectOutputG( $M$ )
1 |  $\hat{G} = \emptyset$ 
2 | for each remaining critical edge  $e$  with  $\text{pers}(e) > \delta$  do
3 |    $\hat{G} = \hat{G} \cup \{\text{1-unstable manifold of } e\}$ 
4 | return  $\hat{G}$ 
```

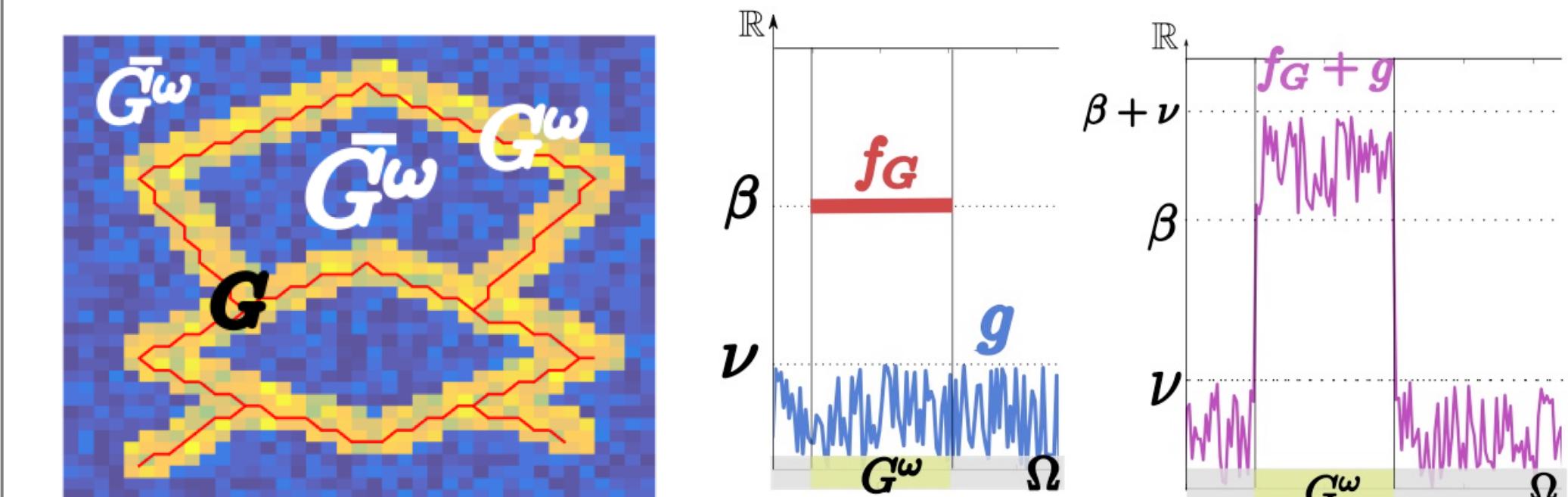
Simplified algorithm

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Algorithm 2: MorseReconSimp( $K, \rho, \delta$ )
Procedure PerSimpTree( $P(K), \delta$ ) /* This procedure replaces original PerSimpVF() */
1 |  $\Pi :=$  the set of vertex-edge persistence pairs from  $P(K)$ 
2 | Set  $\Pi_{\leq \delta} \subseteq \Pi$  to be  $\Pi_{\leq \delta} = \{(v, e) \in \Pi \mid \text{pers}(v, e) \leq \delta\}$ 
3 |  $\mathcal{T} := \bigcup_{(v, \sigma) \in \Pi_{\leq \delta}} \{\sigma = \langle u_1, u_2 \rangle, u_1, u_2\}$ 
4 | return  $\mathcal{T}$ 

Procedure Treebased-OutputG( $\mathcal{T}$ ) /* This procedure replaces CollectOutputG() */
1 |  $\hat{G} = \emptyset$ 
2 | for each critical edge  $e = (u, v)$  with  $\text{pers}(e) \geq \delta$  do
3 |   Let  $\pi(u)$  be the unique path from  $u$  to the sink of the tree  $T_i$  containing  $u$ 
4 |   Define  $\pi(v)$  similarly; Set  $\hat{G} = \hat{G} \cup \pi(u) \cup \pi(v) \cup \{e\}$ 
5 | return  $\hat{G}$ 
```

Illustration of the Simplified algorithm

Noise model



Definition A density function $\rho : \Omega \rightarrow \mathbb{R}$ is a (β, ν, ω) -approximation of a connected graph G if the following holds:

- C-1 There is a ω -neighborhood G^ω of G such that G^ω deformation retracts to G .
- C-2 $\rho(x) \in [\beta, \beta + \nu]$ for $x \in G^\omega$; and $\rho(x) \in [0, \nu]$ otherwise. Furthermore, $\beta > 2\nu$.

Theoretical guarantee

\hat{G} is geometrically close to G :

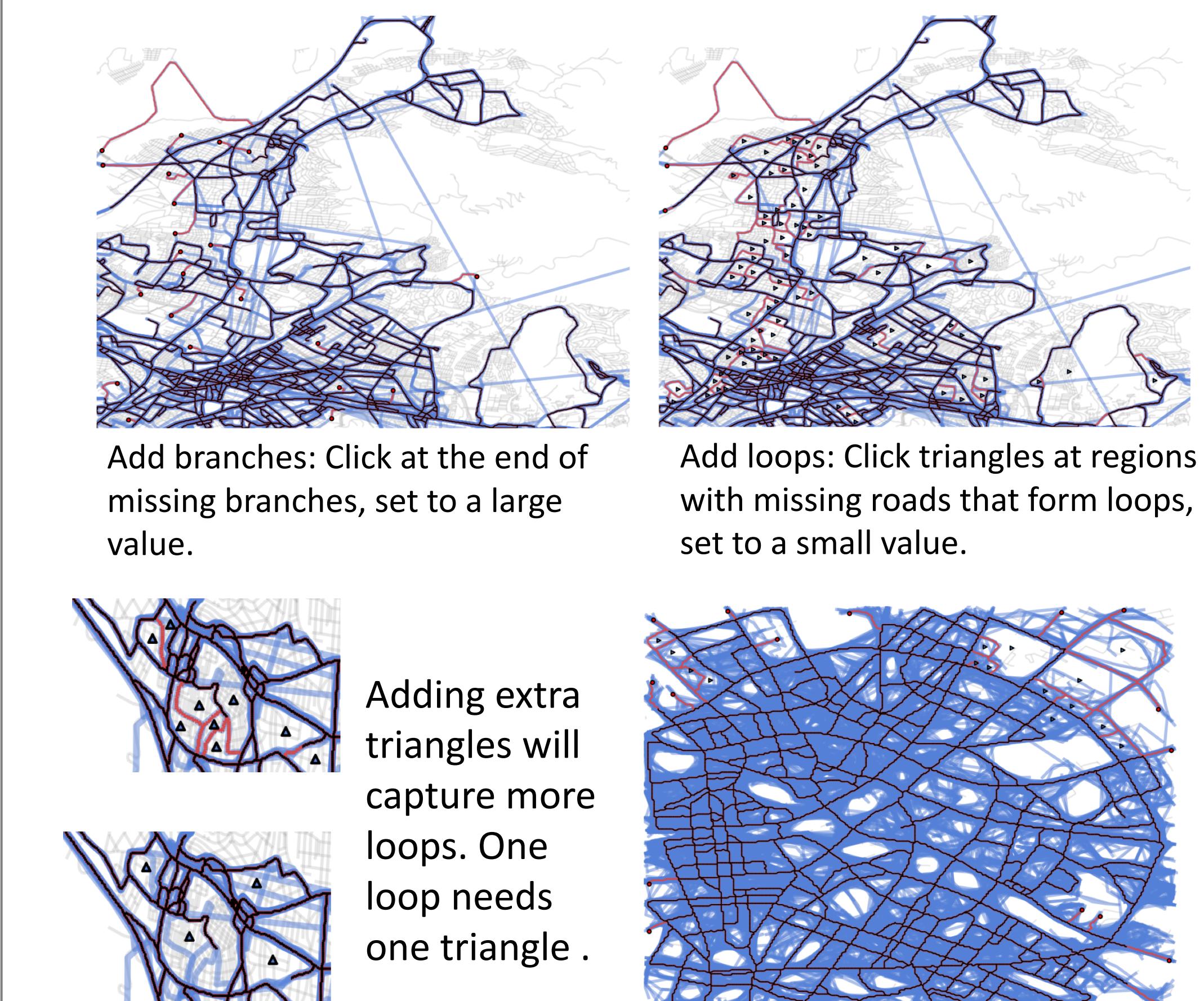
Theorem Under our noise model, the output graph satisfies $\hat{G} \subseteq G^\omega$.

\hat{G} has the same topology with G :

Proposition Under our noise model, \hat{G} is homotopy equivalent to G .

Editing strategy

The idea is to modify the value of the density map at the chosen vertices.



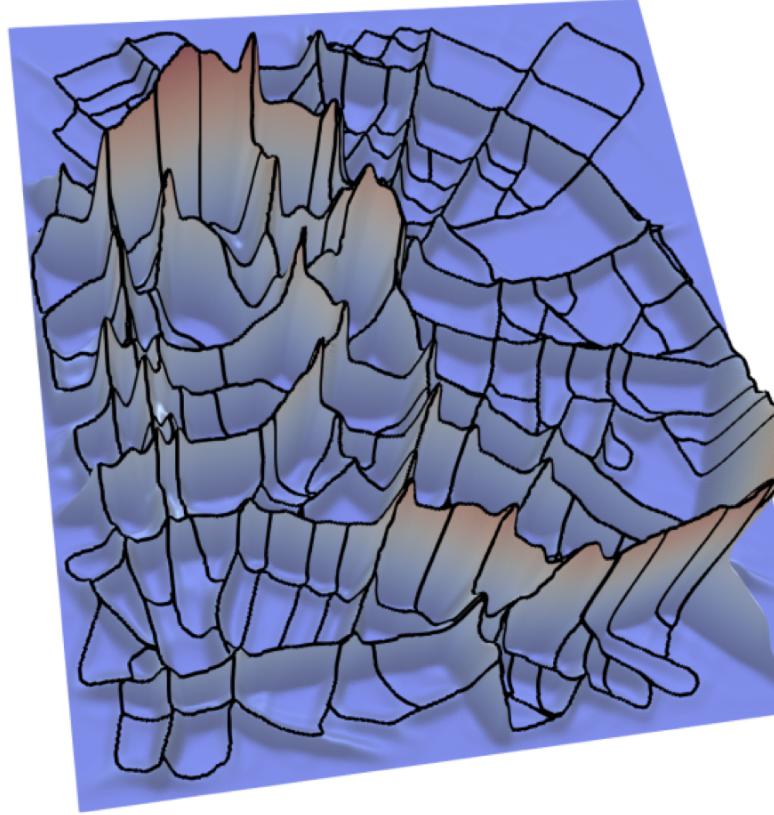
Blue: Input GPS trajectories Black: Original reconstruction Red: Newly added branches

Experiments

2D: GPS TRAJECTORIES



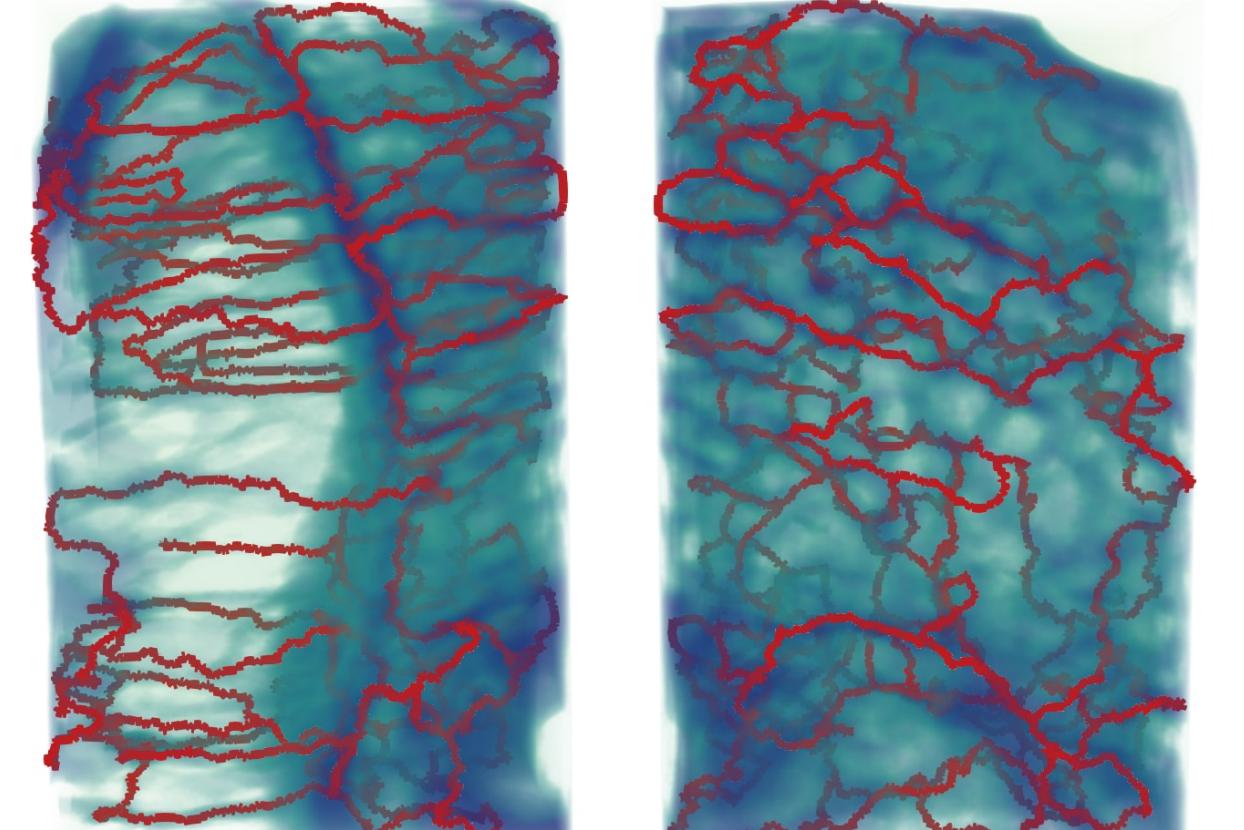
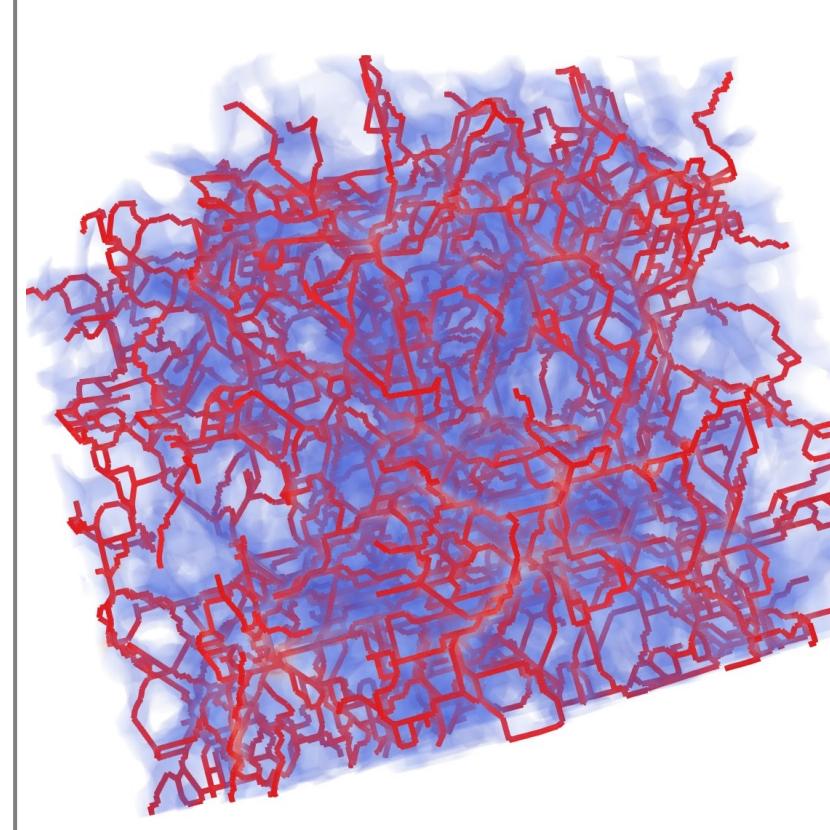
Athens



Berlin

Blue: Input GPS trajectories Black: Original reconstruction

3D DATA



ENZO: Cosmological structure

Micro CT image of a bone

Red: Reconstruction result The volumes show the volume rendering of the input density functions.

Compare with threshold

