

DISCRETE MORSE THEORY AND APPLICATIONS

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MOTIVATION

A reasonably “good” smooth function reveals the [Combinatorial Description](#) of a smooth manifold.

Let M^n be a smooth manifold and $f : M \rightarrow \mathbb{R}$ be a smooth function.

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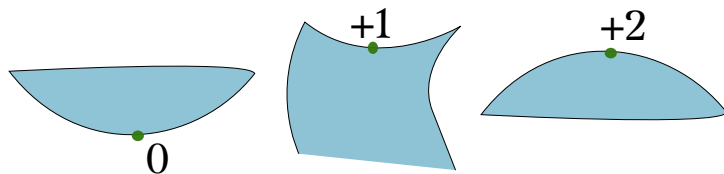


FIGURE: critical points with index

MORSE FUNCTION

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p has index **0**. w has index **+2**. Others have index **+1**.

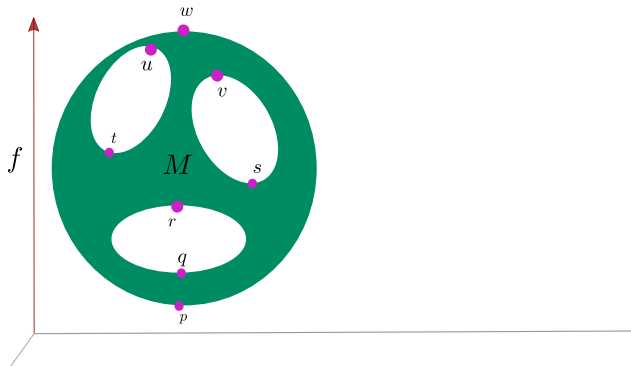


FIGURE: Morse Function

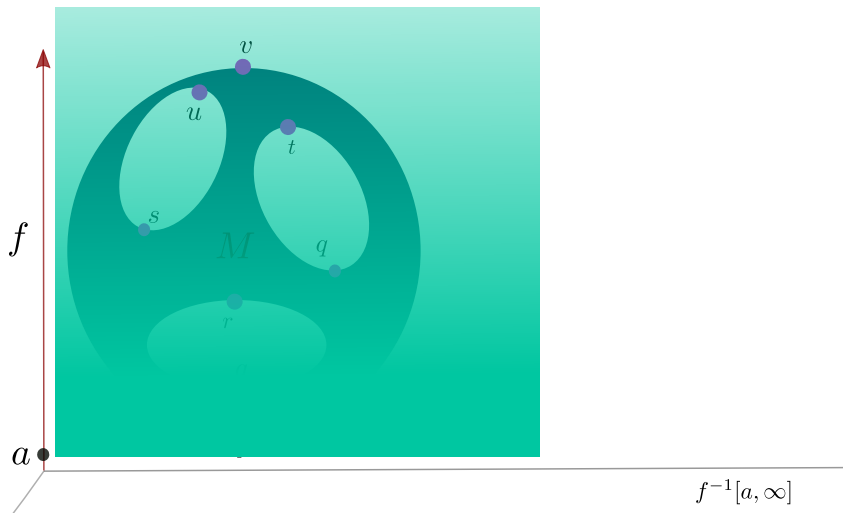


FIGURE: Sub-level Set

CLASSICAL MORSE THEORY

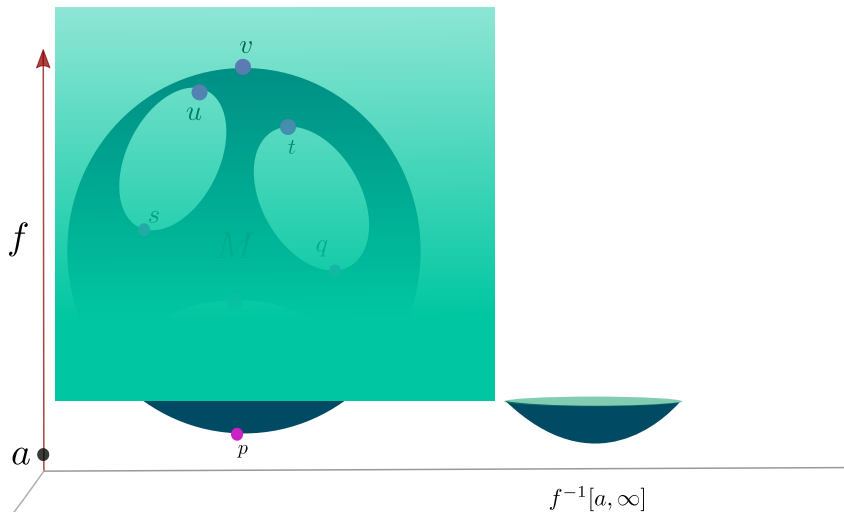


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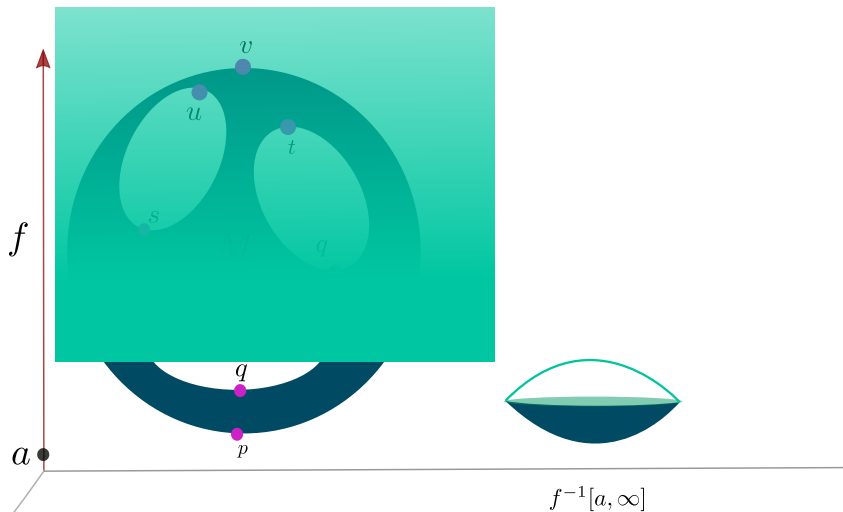


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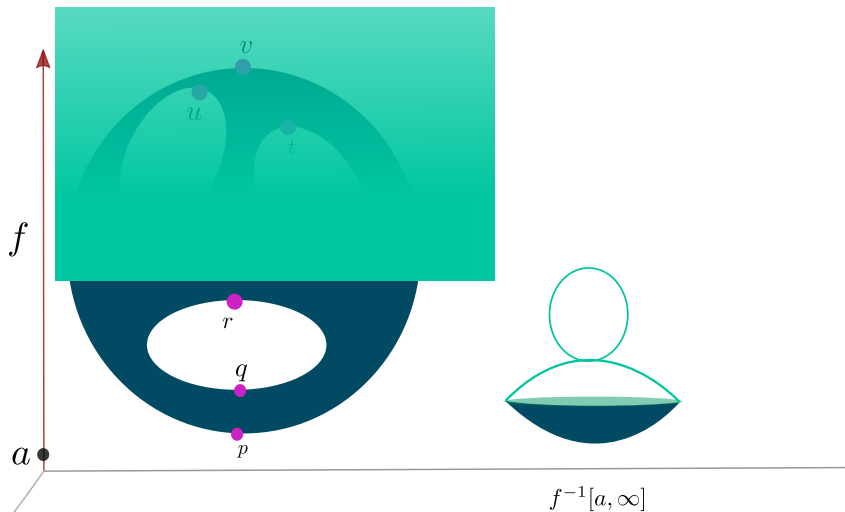


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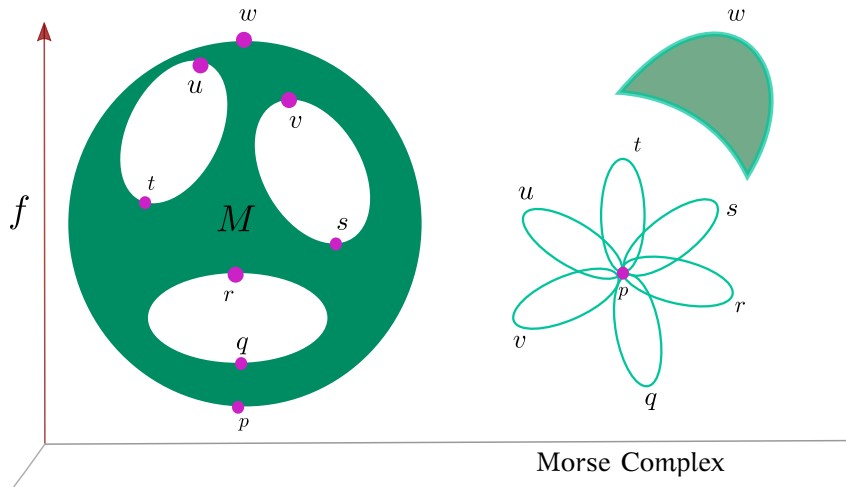


FIGURE: Morse Complex

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If f is Morse on M , then M is *homotopy equivalent* to a CW-complex having a d -cell for each critical point of f of index d .

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MORSE INEQUALITY

critical d -index critical points $\geq H_d(M)$.

GRADIENT OF MORSE FUNCTION

The gradient vector field of a smooth function f

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UNSTABLE MANIFOLD

The **unstable manifold**

$$W_u(p) = \{x \in M \mid \lim_{t \rightarrow -\infty} \Phi_t(x) = p\}$$

OBSERVATIONS

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- 1 critical points are the equilibrium points of ∇f .
- 2 f (strictly) decreases along the flow-lines.
- 3 no limit cycles.

PROBLEM STATEMENT

Given a (noisy) sample S taken around a (hidden) embedded graph G , how one can “reconstruct” the topology and geometry of G from S .

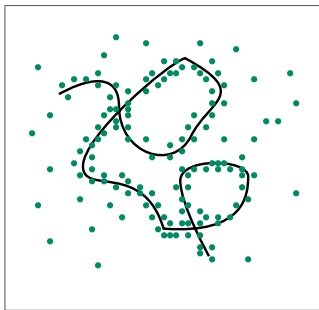


FIGURE: Sample around an embedded graph

MAP RECONSTRUCTION FROM GPS TRACES

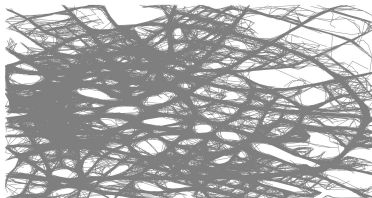


FIGURE: GPS traces of Berlin (mapreconstruction.org)

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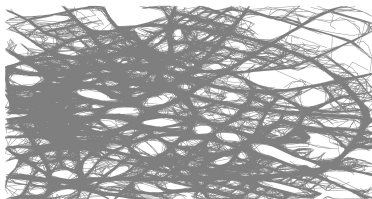


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FIGURE: A reconstruction

NOISE MODELS

- 1 Hausdorff noise

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- 1 Topology ([homotopy type](#))

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Given a discretized domain \tilde{D} and a sample S around G .

① Compute density f of S over \tilde{D} .

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- Kernel Density Estimate

$$K(x, y; \tau) := \exp\left(\frac{-\|x - y\|^2}{2\tau^2}\right)$$

$$f(x) = \frac{1}{2\pi n \tau^2} \sum_{X_i \in S} K(x, X_i; \tau)$$

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③ heuristic pruning methods are applied this super-level set to approximate G .

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LIMITATIONS

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- ② Theoretical guarantees on the topological/geometric correctness is not proved.
- ③ The output is often a thick region around the hidden graph.

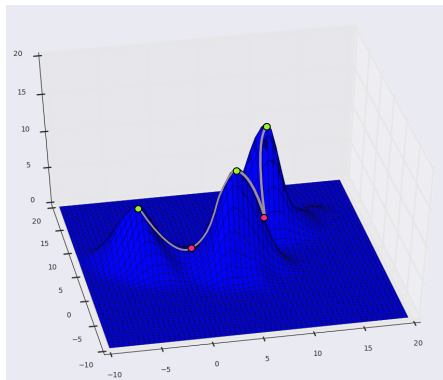


FIGURE: KDE

If the samples are concentrated around a graph, then the mountain ridges on the graph of the density function are expected to capture it.

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Let K be a simplicial complex. A function $f : K \rightarrow \mathbb{R}$ is a **discrete Morse function** if for every $\alpha^{(p)} \in K$

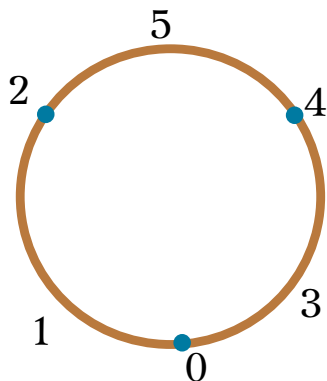
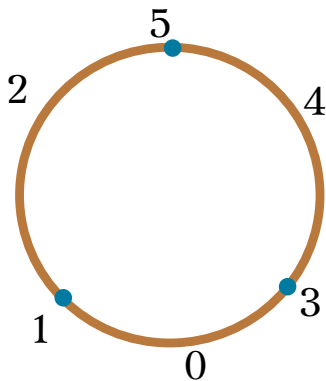
- ① $\#\{\beta^{(p+1)} > \alpha \mid f(\beta) \leq f(\alpha)\} \leq 1,$
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CRITICAL SIMPLEX

A simplex $\alpha^{(p)}$ is critical if

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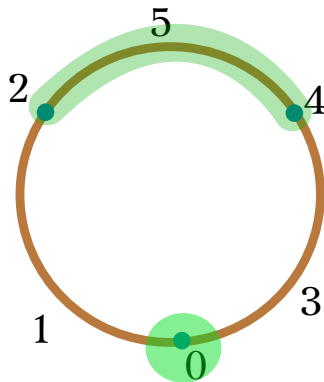


FIGURE: Discrete Morse Function

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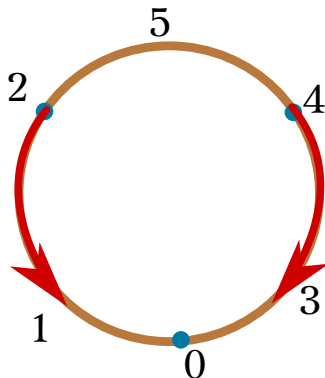


FIGURE: Simplicial Collapse

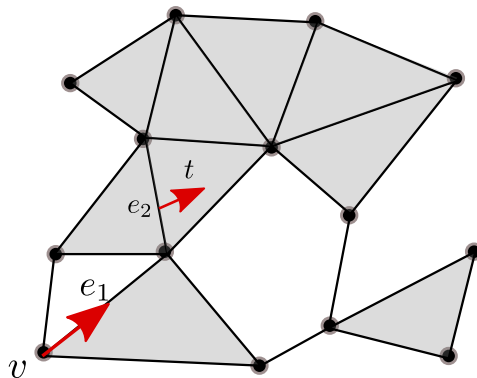


FIGURE: Simplicial Complex K and discrete vector

DISCRETE VECTOR FIELD

(σ, τ) is a **discrete vector** in K if $\tau < \sigma$.

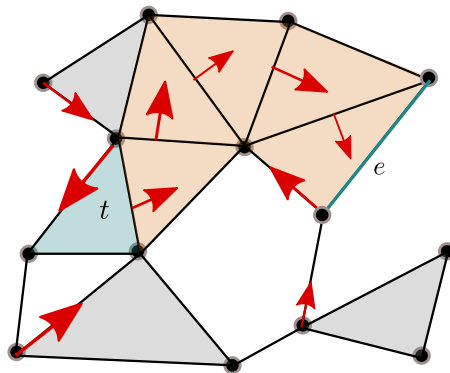


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DISCRETE VECTOR FIELD

(σ, τ) is a **discrete vector** in K if $\tau < \sigma$. A discrete vector field is a collection of discrete vectors such that every simplex of K is head/tail of at most one vector.

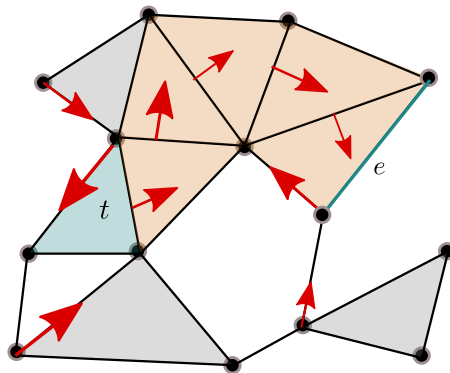


FIGURE: V-path

V-PATH

$$\sigma_0, \tau_0, \sigma_1, \tau_1, \dots, \sigma_{l+1}$$

where (σ_i, τ_i) is a vector and $\tau_i < \sigma_{i+1}$.

BACKGROUND: MORSE CANCELLATION

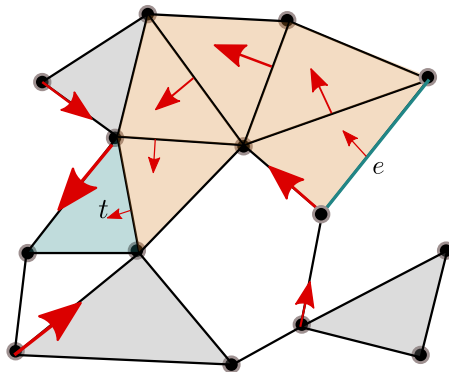


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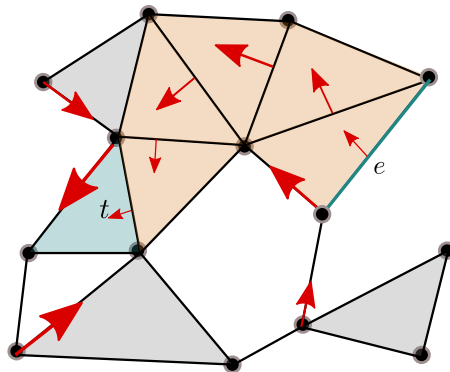


FIGURE: Morse Cancellation

STABLE MANIFOLD

For a critical edge e , its **stable manifold** is the set of all V-paths ending at the boundary of e .

$(\omega, \beta_1, \beta_2, \nu)$ -APPROXIMATION OF \mathcal{G}

$$f(x) \in \begin{cases} [\beta_1, \beta_1 + \nu], & x \in V^\omega \\ [\beta_2, \beta_2 + \nu], & x \in E^\omega \\ [0, \nu], & \text{otherwise} \end{cases}$$

Input: The discretized domain K , the density function f , the threshold δ

Output: The reconstructed graph \hat{G}

- 1 Initialize V as the trivial vector field on K and $\hat{G} = \emptyset$.
- 2 Run persistence on the super-level set filtration of f to get the persistence pairs $P(K)$.
- 3 For each $(\sigma, \tau) \in P(K)$ with $Pers(\sigma, \tau) < \delta$
Try to perform a Morse cancellation for the pair and update V .
- 4 For each $(v, e) \in P(K)$ and $(e, t) \in P(K)$ with $Pers(v) \geq \delta$, $\hat{G} = \hat{G} \cup \{\text{stable manifold of } e\}$.
- 5 output \hat{G}

THEOREM

If G is a connected, embedded planar graph in a cubical complex K and f is an $(\omega, \beta_1, \beta_2, \nu)$ -approximation then the output \hat{G} has the same homotopy type as G . Moreover, $d_H(G, \hat{G}) < \omega$.

- 1 How to circumvent the heavy persistence computation.

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- 2 What condition on the density function gives up a small Fréchet distance between the edges of the output and the edges of the graph.

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- 2 What condition on the density function gives up a small Fréchet distance between the edges of the output and the edges of the graph.
- 3 Extend the result to higher dimensions.

Thanks



Mahmuda Ahmed, Brittany Terese Fasy, Matt Gibson, and Carola Wenk, *Choosing thresholds for density-based map construction algorithms*, Proceedings of the 23rd SIGSPATIAL International Conference on Advances in Geographic Information Systems (New York, NY, USA), SIGSPATIAL '15, ACM, 2015, pp. 24:1–24:10.



Tamal K. Dey, Jiayuan Wang, and Yusu Wang, *Graph reconstruction by discrete Morse theory*, 34th International Symposium on Computational Geometry, 2018, pp. 31:1–31:15.