## Graph Approximations to Geodesics on Metric Graphs

## MOTIVATION

## Proximity graphs are often used to approximate geodesics

Used for various dimensionality reduction methods

- Laplacian Eigenmaps
- Locally Linear Embedding
- ISOMAP

- Maximum Variance Unfolding
- Local Tangent Space Alignment
- ...

Used as intermediate representations for topological inference [1, 2, 3]


## THE PROBLEM

Existing results for approximating geodesic distances [4]

- Only apply to smooth submanifolds of $\mathbb{R}^{n}$
- No singularities (such as bifurcations) allowed
- Conditions under which approximated and true geodesic distances are close are stringent
- Must approximate all local patterns well
- However, global patterns may be approximated well without local patterns





## REFERENCES

## OUR APPROACH

## New geometric characteristics for metric graphs

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Definition 3. Given a connected metric graph \(M\). For any \(\epsilon>0\) we define the branch separation \(s_{\epsilon}(M) \in \mathbb{R}^{+} \cup\{\infty\}\) of
\(M\) at resolution \(1 / \epsilon\) as
\(s_{\epsilon}(M):=\sup \left\{s \in \mathbb{R}:\|x-y\|<s \Longrightarrow d_{M}(x, y) \leq \epsilon\right\}\)
Definition 4. Given a connected metric graph M. For any \(0 \leq \epsilon^{\prime}<\epsilon\) we define the linearity of \(M\) between resolution \(1 / \epsilon\) and \(1 / \epsilon^{\prime} \in \mathbb{R}^{+} \cup\{\infty\}\) as
\(\lambda_{\epsilon^{\prime}, \epsilon}(M):=\sup \left\{\lambda \in \mathbb{R}: \epsilon^{\prime} \leq d_{M}(x, y) \leq \epsilon \Longrightarrow\right.\) \(\left.\lambda d_{M}(x, y) \leq\|x-y\|\right\}\)
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- More flexibility (account for data at hand)
- Is also necessary to extend to singularities (proof in paper)
Theorem 3. Let $M$ be a connected metric graph in $\mathbb{R}^{d}$ and
let $X$ be a finite set of data points in $M$. Suppose a graph
$G=(X, E)$ is given, defining the following three thresholds:

1) $\|x-y\| \geq \epsilon^{\prime}$ for all $\{x, y\} \in E$,
2) $\|x-y\|<s_{\epsilon}(M)$ for all $\{x, y\} \in E$, with $\epsilon>0$,
3) for all $x, y \in X$ with $\|x-y\| \leq \tau$, we have $\{x, y\} \in E$.
If for $0<4 \delta<\tau$, $X$ satisfies the $\delta$-sampling condition, i.e.,
for every $m \in M$ there is $x \in X$ with $d_{M}(m, x) \leq \delta$, then
for all $x, y \in M$
$\lambda_{\epsilon, \epsilon^{\prime}}(M) d_{M}(x, y) \leq d_{G}(x, y) \leq(1+4 \delta / \tau) d_{M}(x, y)$. (1)


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Representations," Journal of Machine Learning Research, 21(215):1-68, 2020
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[3] R. Vandaele. "Topological Inference in Graphs and Images," Doctoral thesis, Ghent University, 2020
[4] M. Bernstein et al, "Graph approximations to geodesics on embedded manifolds," 2000
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