

Approximate nearest neighbor search using the Hierarchical Navigable Small World (HNSW) algorithm

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Outline

1 Theoretical foundations

- Voronoi diagram
- Delaunay graph
- Greedy NN search using Delaunay graph

2 HNSW algorithm

- Navigable small world (NSW)
- Hierarchical navigable small world (HNSW)
- Nearest neighbor search using HNSW

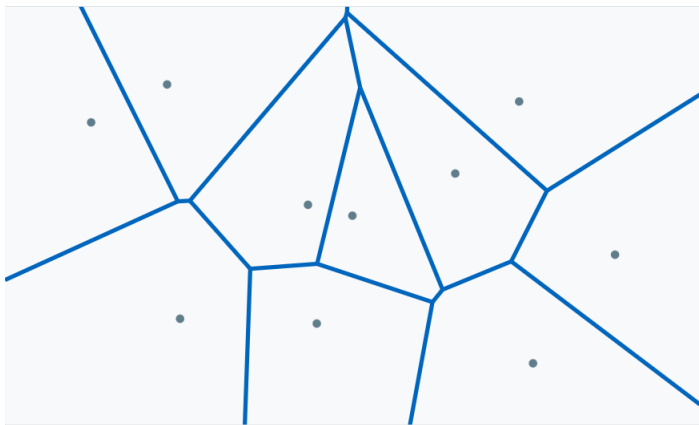
3 Performance

- Search accuracy
- Build time

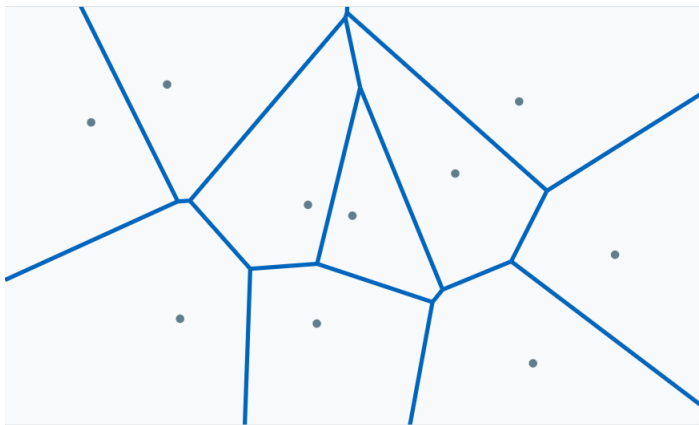
Voronoi diagram for a set of points



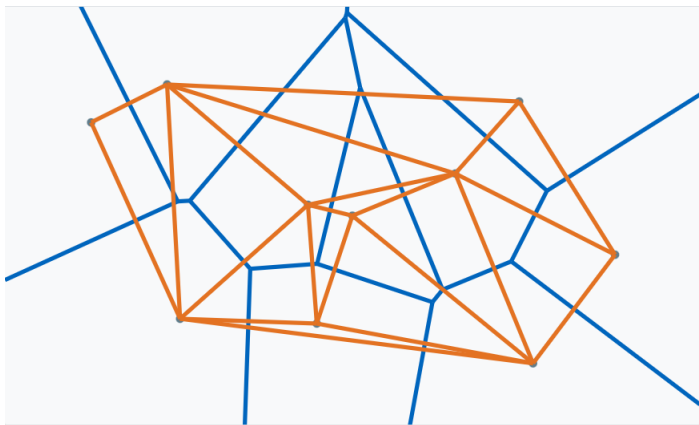
Voronoi diagram for a set of points



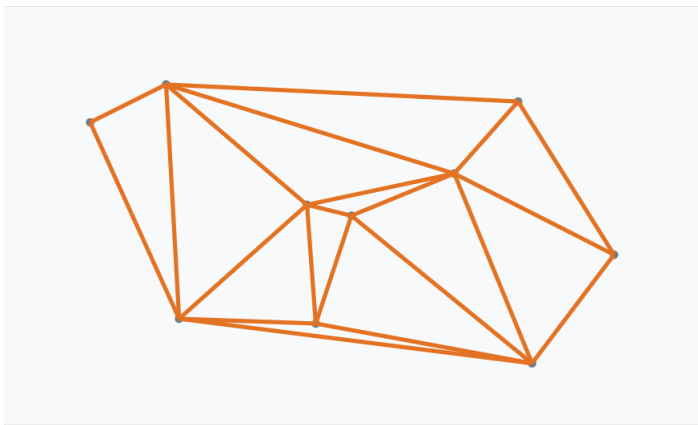
Voronoi diagram to Delaunay graph



Voronoi diagram to Delaunay graph



Delaunay graph



Greedy NN search algorithm

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- 1 Select any graph node as entry node

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- 2 Calculate distance from query to current node and from query to all neighbors of current node

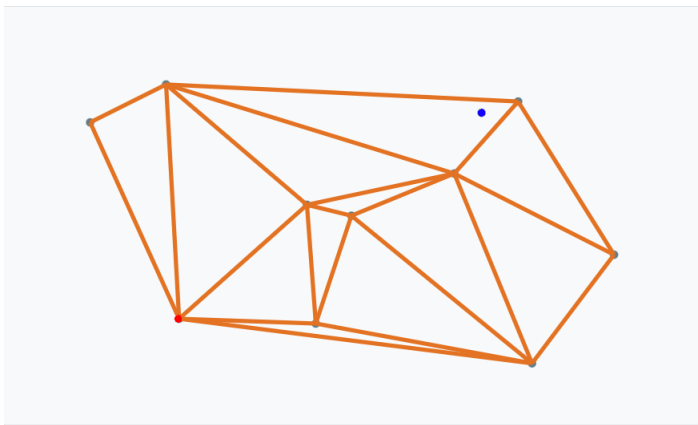
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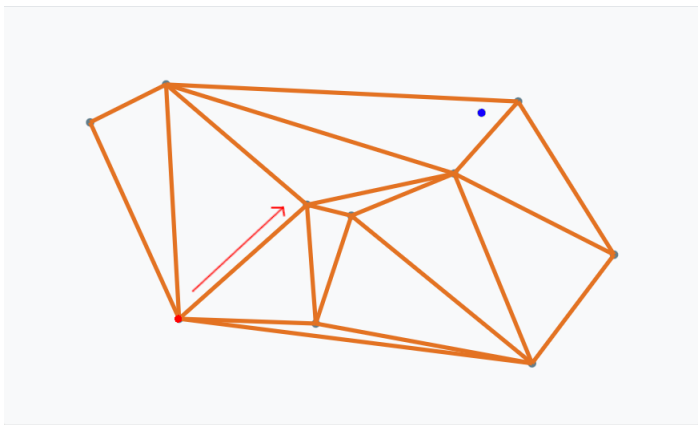
Greedy NN search algorithm

- 1 Select any graph node as entry node
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- 4 Repeat 2 and 3 until no neighbor is closer to query than the current node

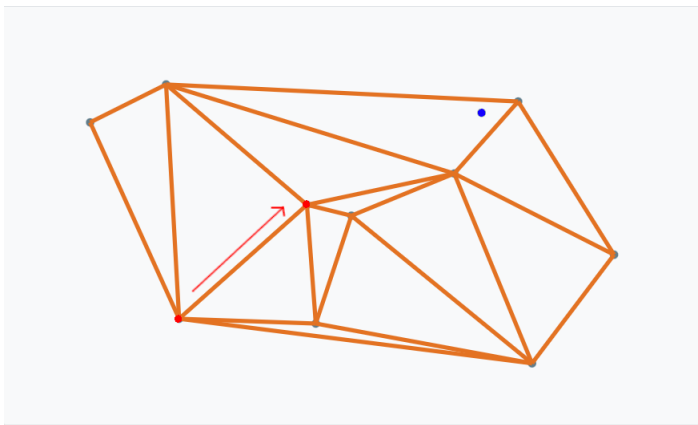
Greedy NN search start - Query and entry node



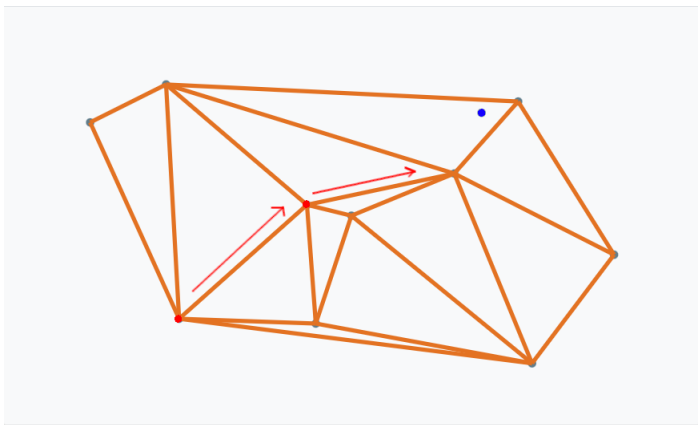
Greedy NN search - iteration



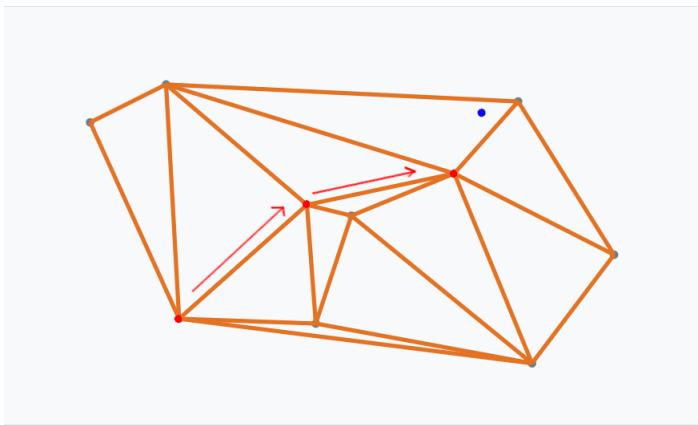
Greedy NN search - iteration



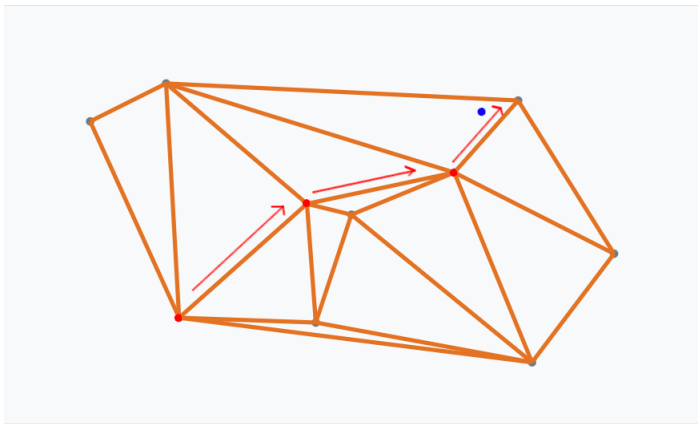
Greedy NN search - iteration



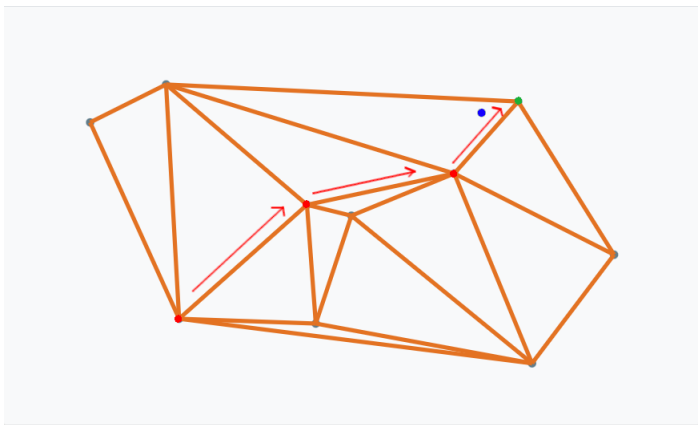
Greedy NN search - iteration



Greedy NN search - iteration



Greedy NN search done!



Drawbacks

- Delaunay graph intractable to construct for large, high-dimensional data sets
- Greedy search might require a lot of steps if graph is large

Navigable small world (NSW) graph

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■ Small world graph

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■ Navigability

- Greedy search algorithm has logarithmic scalability

Why is an NSW useful for nearest neighbor search?

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- Logarithmic distance allows us to get anywhere in the graph quickly

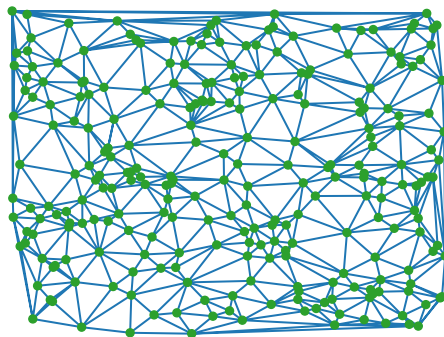
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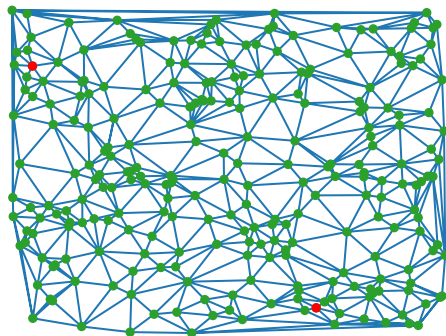
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- Navigability ensures that the greedy algorithm finds the logarithmic path
- High clustering coefficient lets us zoom in on the actual correct node when we're in the right area

Making Delaunay graph navigable

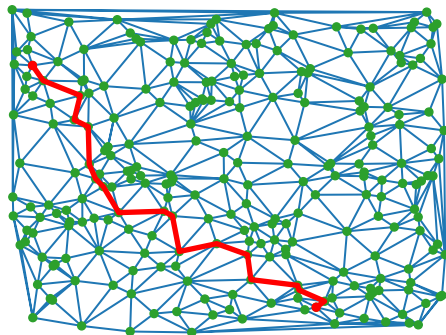


256 nodes

Making Delaunay graph navigable

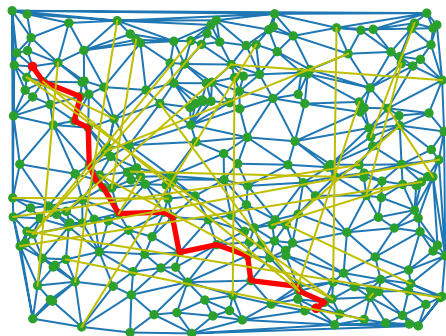


Making Delaunay graph navigable



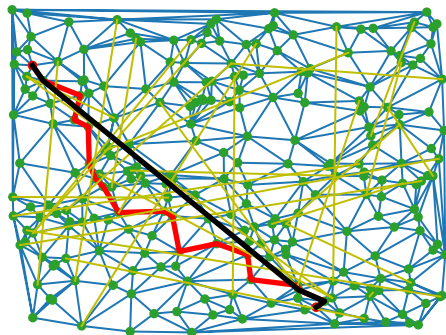
Length of path: 19

Making Delaunay graph navigable



32 random edges added

Making Delaunay graph navigable



Length of path: 5

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- Thus the greedy algorithm doesn't always return the actual nearest neighbor
- Ok since we're doing approximate nearest neighbor search!

Constructing NSW graph

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- Goal: Construct a graph that has the Delaunay graph as a subgraph, but also has longer connections to make it navigable

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- Approximation of Delaunay graph is sufficient

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- 1 Randomize order of data points

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- 4 Repeat 2 and 3 until all data points have been added

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- Adding enough nearest neighbor edges approximates Delaunay graph

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- Adding enough nearest neighbor edges approximates Delaunay graph
- The edges added for the early nodes give long-range connections, enabling navigability

kNN search using NSW graph

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kNN search using NSW graph

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- To improve results we can redo the search m times from different start nodes

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- Greedy search may get stuck in local minimum
- Algorithm scales polylogarithmically in general (logarithmic scaling in both steps and degrees of nodes)
- Performance degrades on high-dimensional data
- Insertion order must be random

References

- *Efficient and robust approximate nearest neighbor search using Hierarchical Navigable Small World graphs (Malkov et al.*
<https://arxiv.org/abs/1603.09320>
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- *Hierarchical Navigable Small Worlds (HNSW) (Pinecone blog)*
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