Graph-based time-space trade-offs for approximate near neighbors

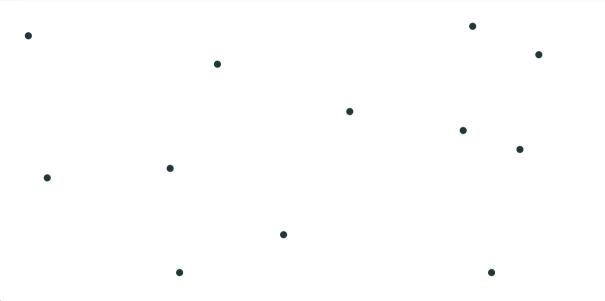
Thiis Laarhoven

mail@thijs.com
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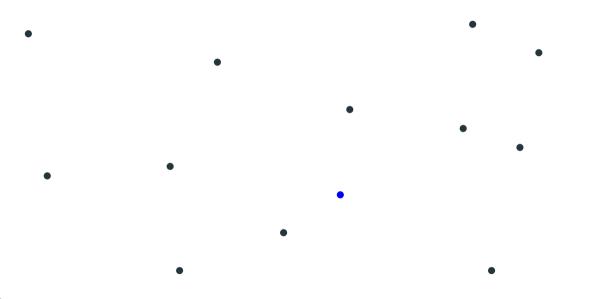
SoCG 2018, Budapest, Hungary (June 13, 2018)

Nearest neighbor searching

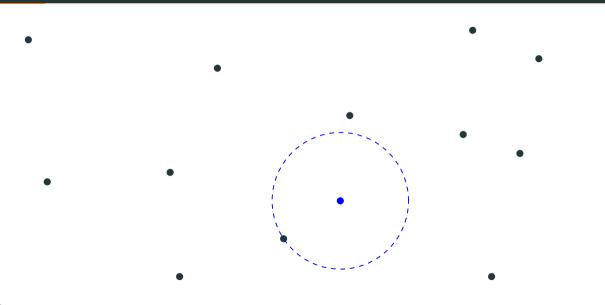
Nearest neighbor problem – Problem description



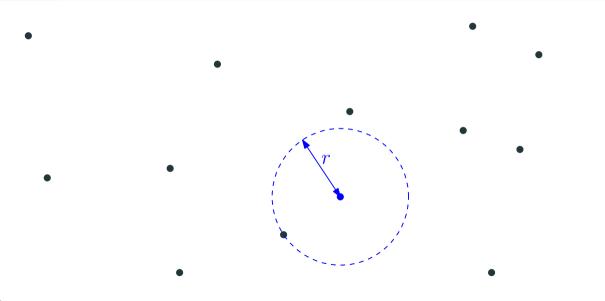
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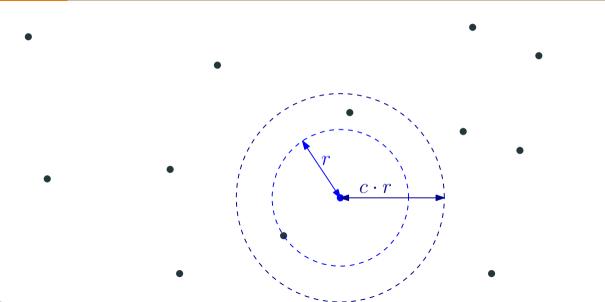
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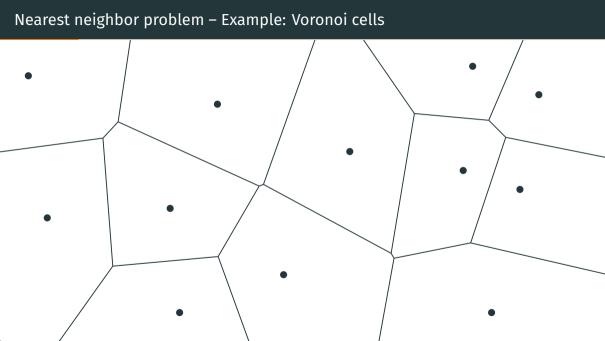


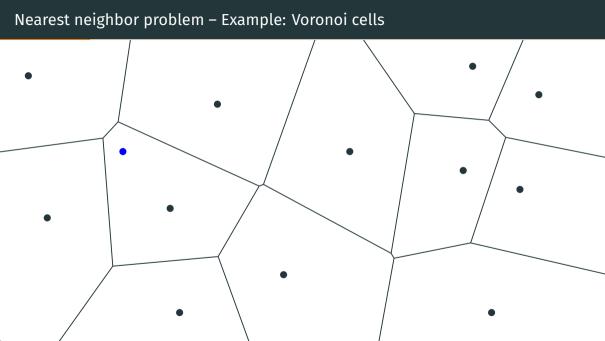
Nearest neighbor problem – Approximate solutions

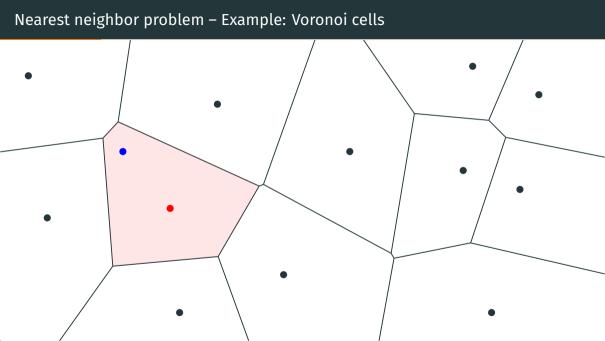


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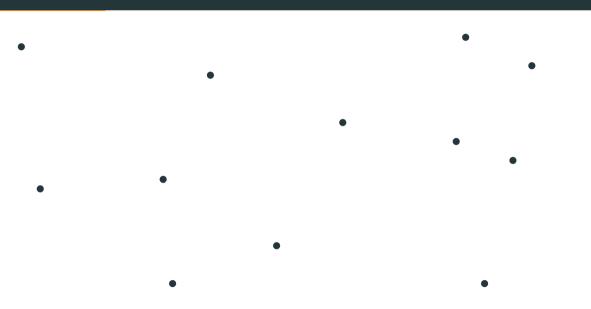


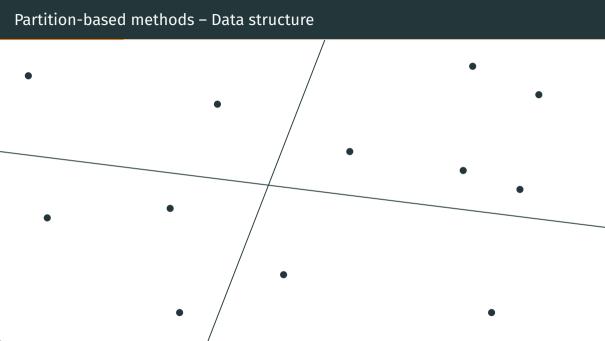


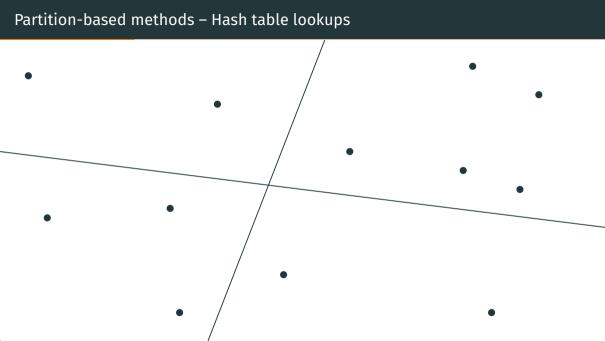


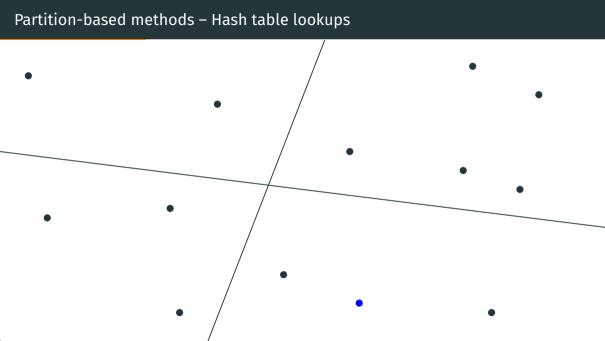
Partition-based methods

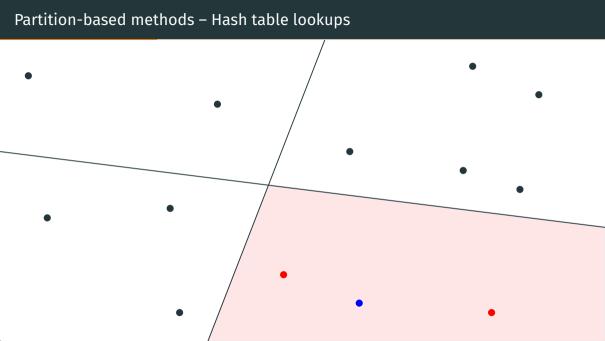
Partition-based methods – Data structure

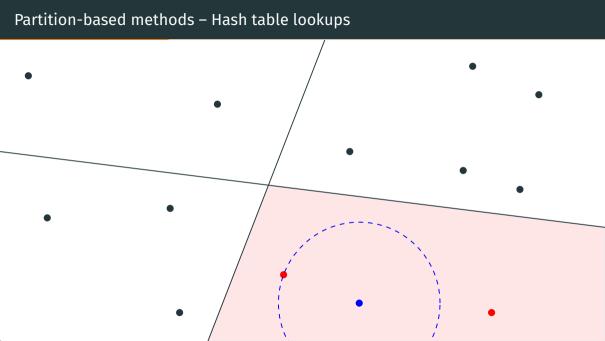


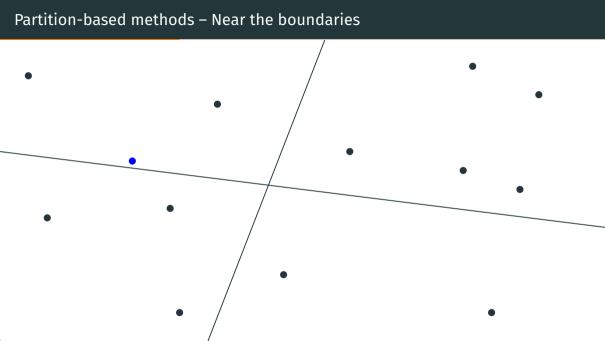


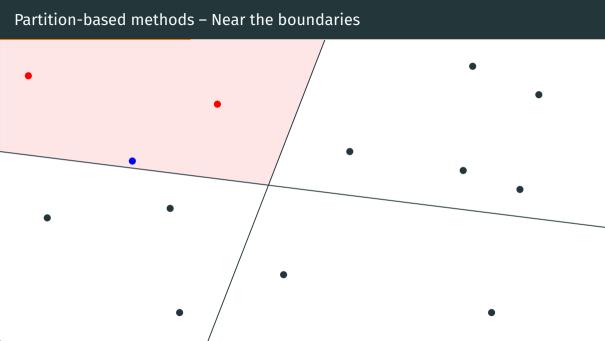




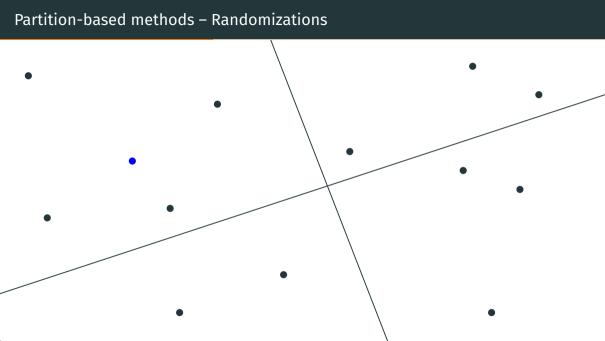








Partition-based methods – Near the boundaries



Partition-based methods – Randomizations

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Main problem: choosing the best types of space partitions.

- · Requires an **efficient decoding** algorithm;
- · Space partitions should have **nice shapes**.

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Real world: approximate ideal solution as best as we can.

- Product of bisections; [Cha03]
- Voronoi cells induced by hypercube; [TT07, Laa16]
- · Random (overlapping) spheres; [AI06, AINR14]
- · Voronoi cells induced by cross-polytopes; [TT07, AIL+15, KW17]
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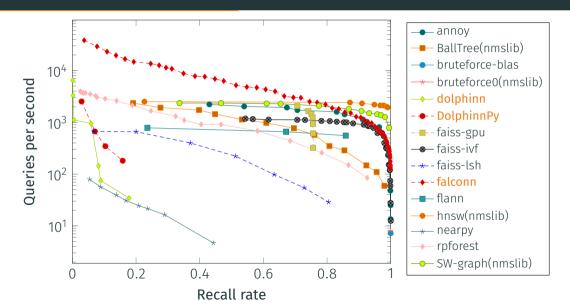
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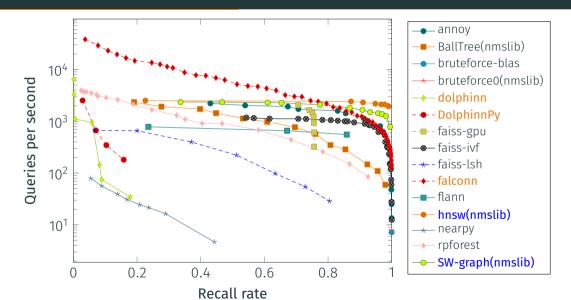
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Best techniques are theoretically optimal as well as practical.

Nearest neighbor methods – Practice (ANN Benchmarks [ABF17])



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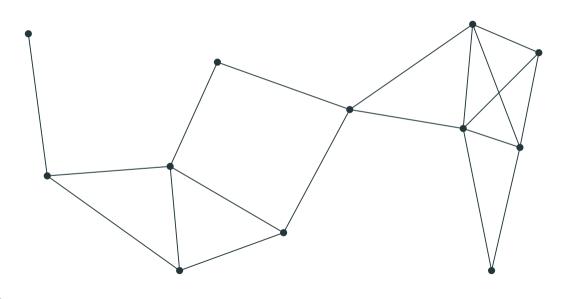


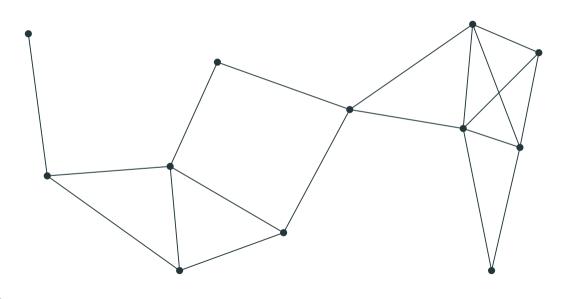
Graph-based methods

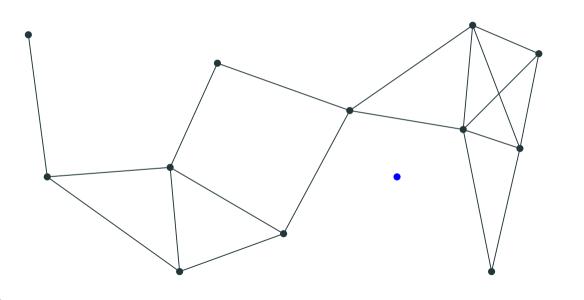
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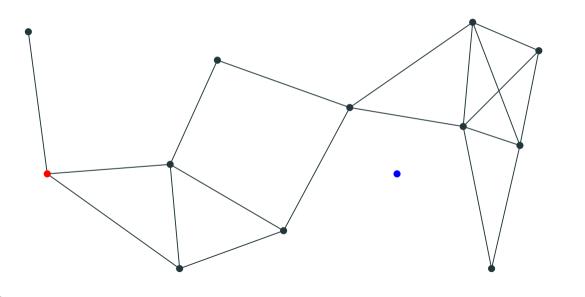


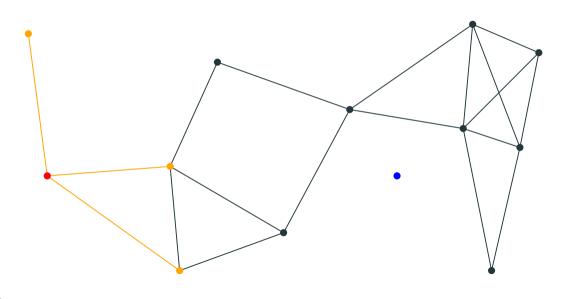
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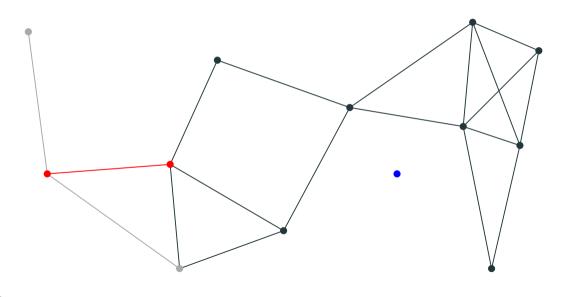


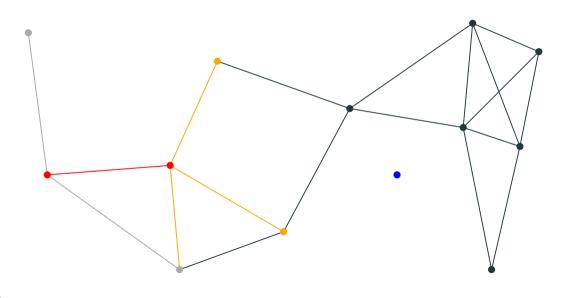


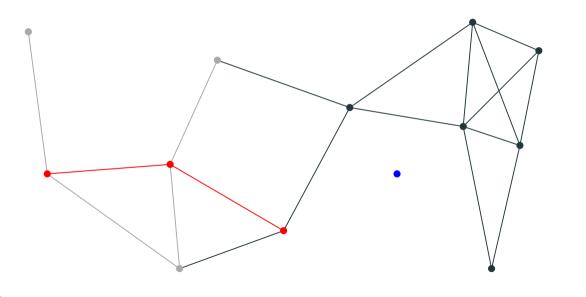


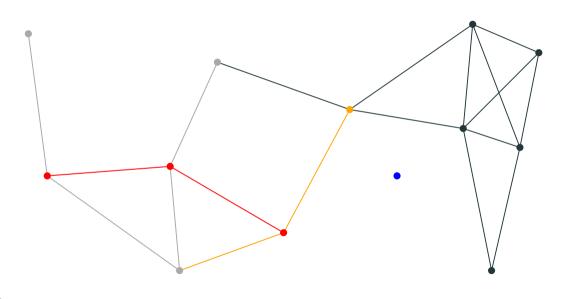


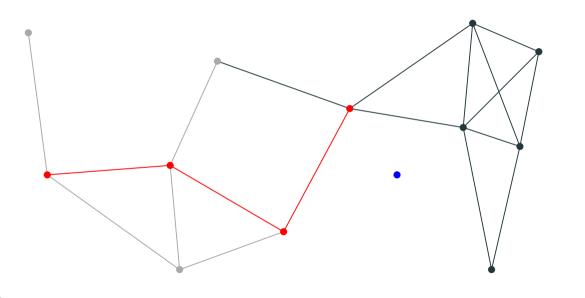


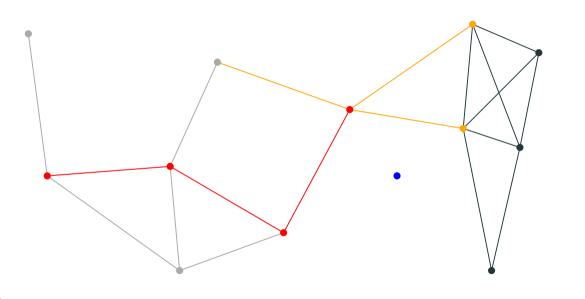


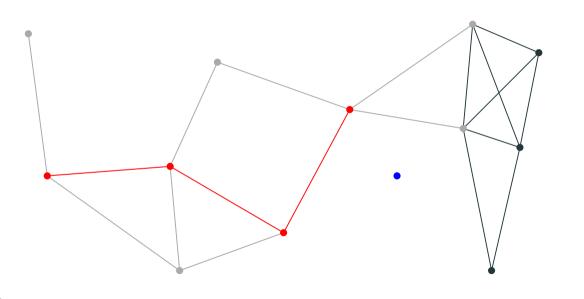


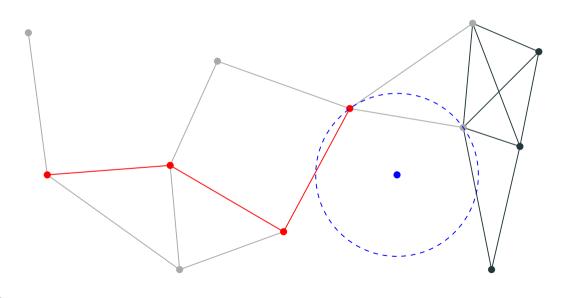


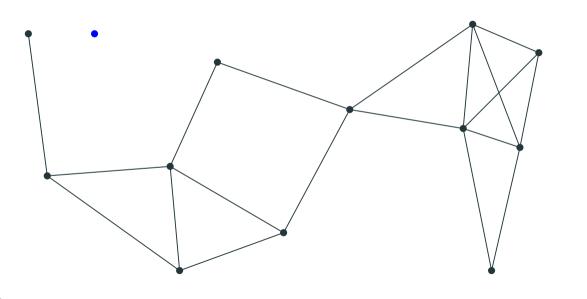


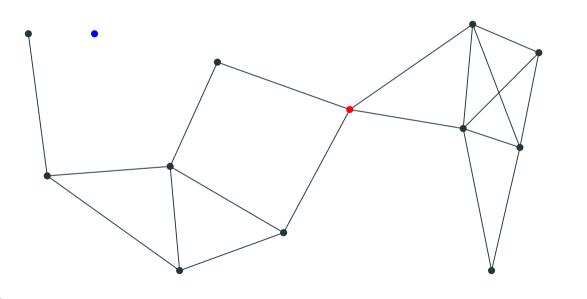


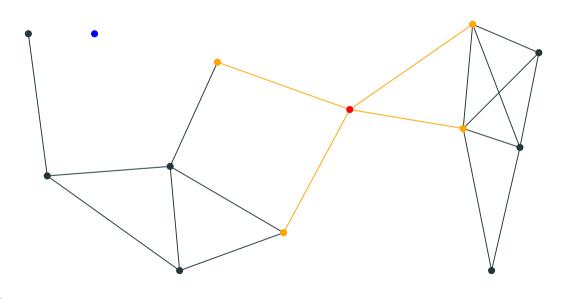


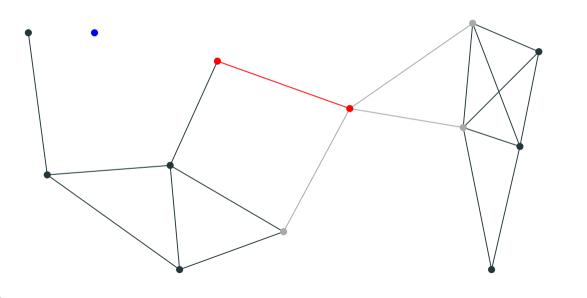


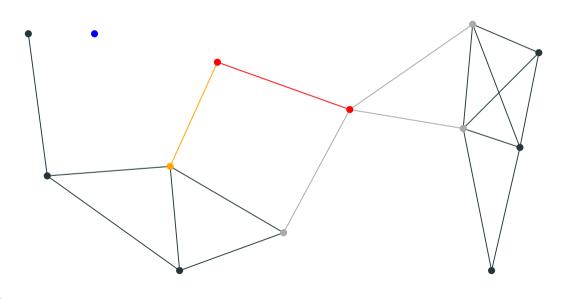


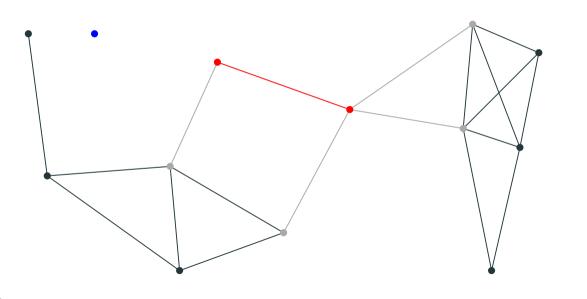


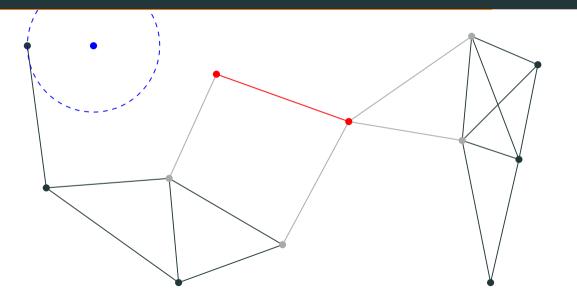


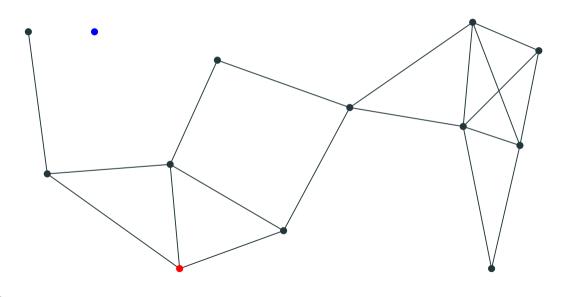


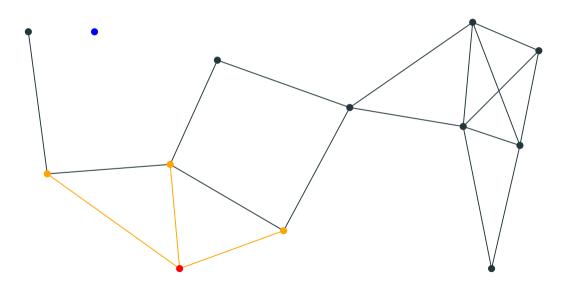


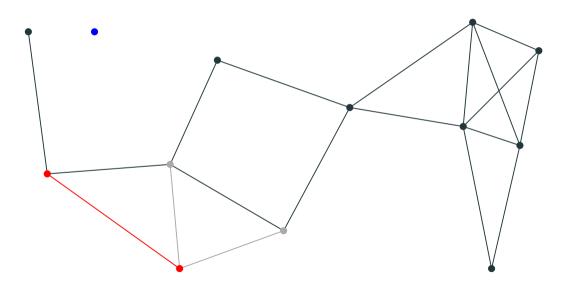


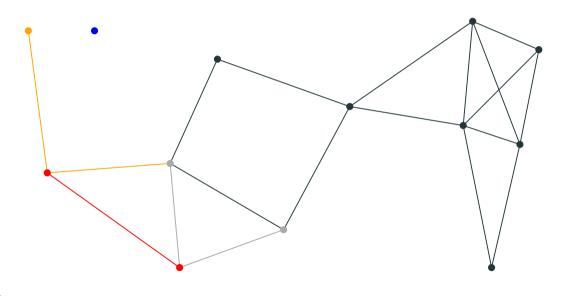


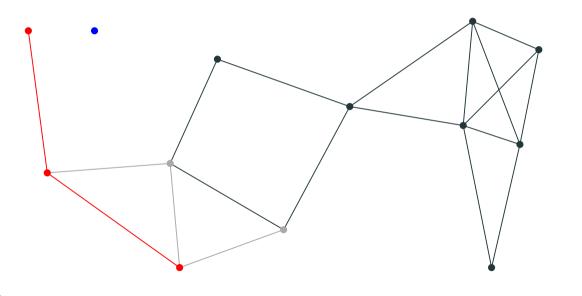


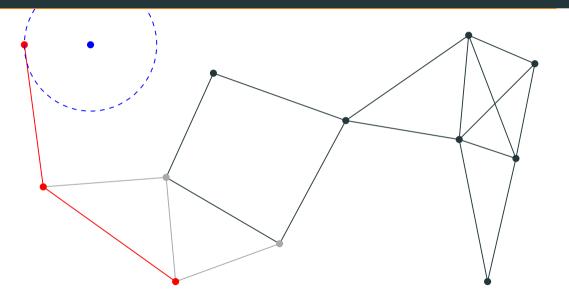












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Graph-based methods - Contributions

Theorem (Main result, informal)

For randomized greedy walks on the near neighbor graph and for "random" data sets, we can solve the approximate nearest neighbor problem on n points with query time $O(n^{\rho_q})$ and space $O(n^{1+\rho_s})$ with $\rho_q, \rho_s \geq 0$ satisfying

$$(2c^2-1)\rho_q+2c^2(c^2-1)\sqrt{\rho_s(1-\rho_s)}\geq c^4.$$

Graph-based methods – Contributions

In the most common regime of $c \approx 1$ (high recall rate) and $\rho_s \approx 0$ (near-linear space), this scales equivalently as the best partition-based trade-offs: [ALRW17]

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Negative result: (analysis of) this algorithm is not competitive for $c \gg 1$ or $\rho_s \gg 0$.

Graph-based methods – Open problems

Various **open problems** remain:

- · Current **analysis** may not be sharp can it be tightened?
- Does adding long edges lead to better theoretical guarantees?
- · What can theoretically be said about hierarchical approaches?
- · Can we obtain **lower bounds** showing limitations of graph-based methods?

