

# Bidirectional Dijkstra's Algorithm is Instance-Optimal

Bernhard Haeupler<sup>1,2</sup>, Richard Hladík<sup>1,2</sup>, Václav Rozhoň<sup>1</sup>, Robert E. Tarjan<sup>3</sup>, Jakub Tětek<sup>1</sup>

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SOSA 2025

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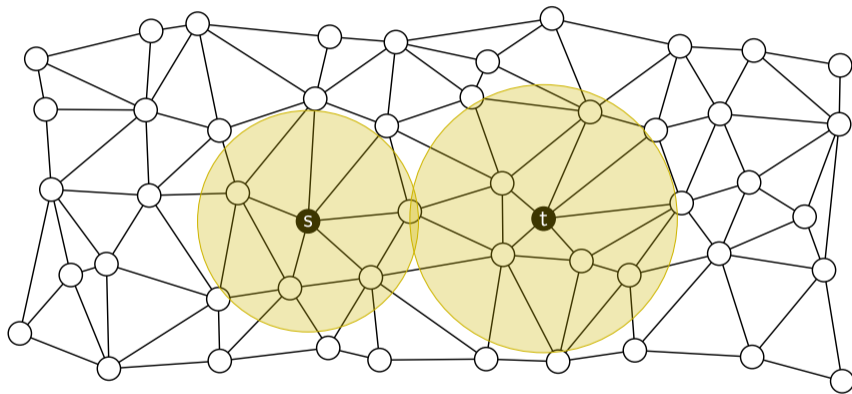
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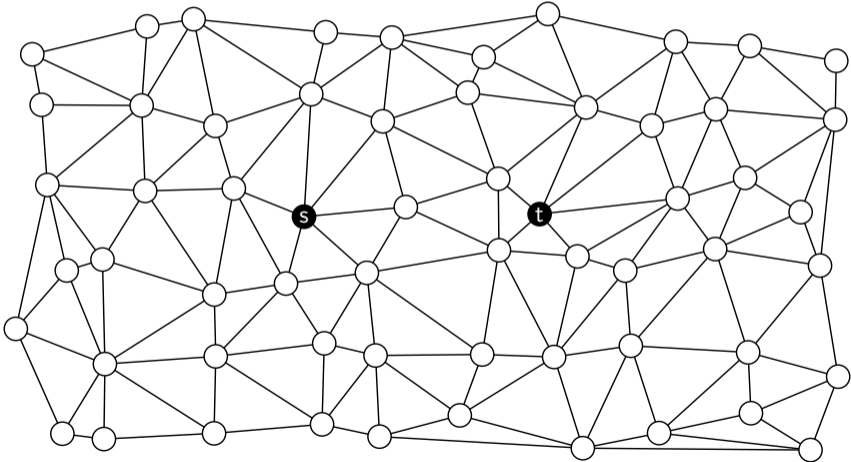
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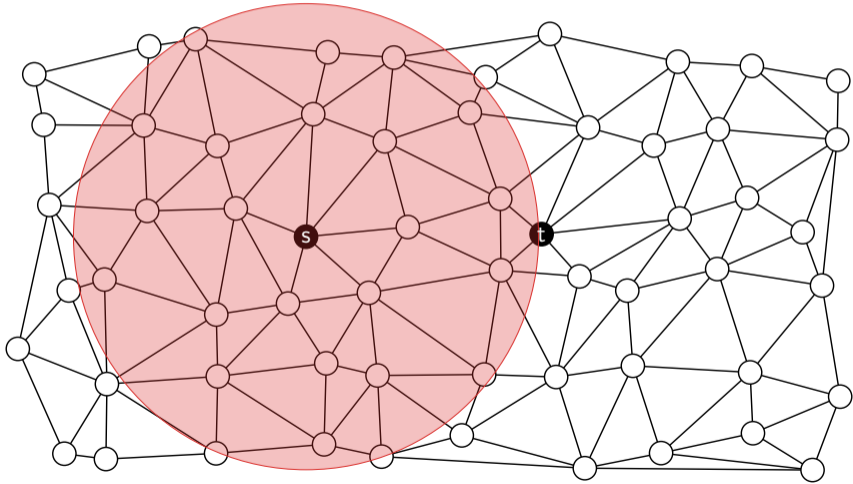
This talk: bidirectional search is optimal on every graph

How quickly can you find the shortest path?

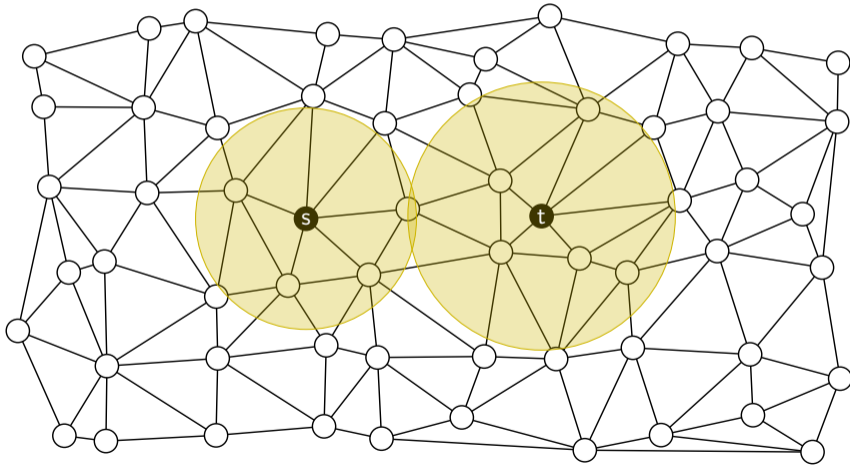
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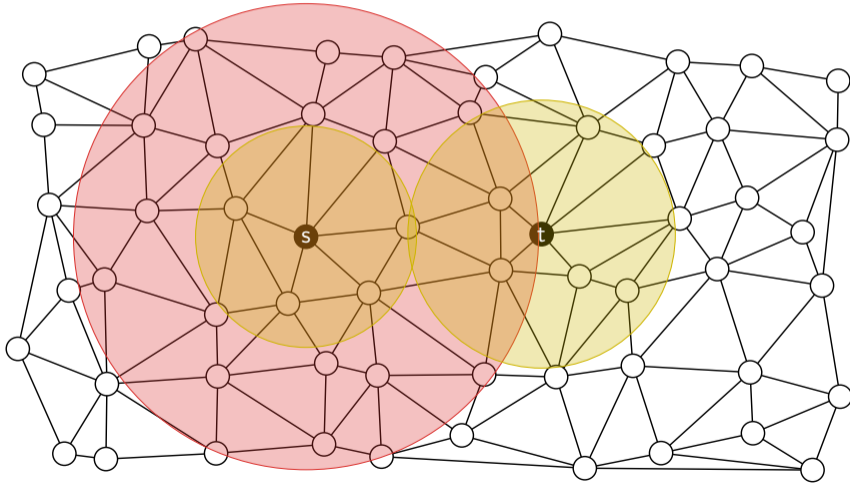
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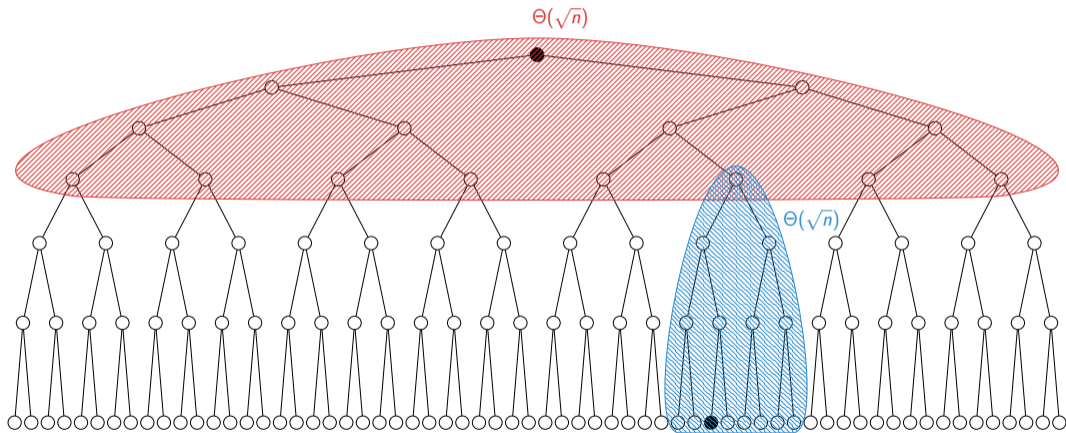
# How quickly can you find the shortest path: bidirectional Dijkstra



# How quickly can you find the shortest path: bidirectional Dijkstra



# Bidirectional search can make a big difference!



# Outline

- ▶ problem statement
- ▶ model
- ▶ instance-optimality
- ▶ the result
- ▶ sketch of the key idea
- ▶ related work & debrief

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(results also hold for time complexity up to a  $\log n$  factor)

## Instance Optimality

$A$  is instance-optimal if:

$$\exists c \forall \text{correct } A' \forall \text{input} \quad \text{Queries}_A(\text{input}) \leq c \cdot \text{Queries}_{A'}(\text{input})$$

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**Our result:** A variant of bidirectional Dijkstra is instance-optimal (w. r. t. query complexity)

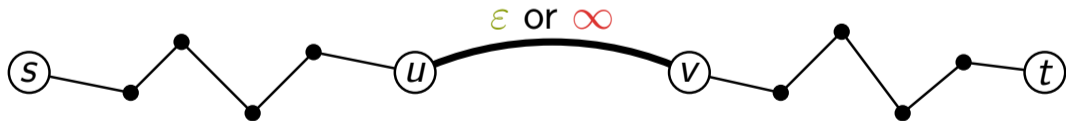
## Proof: Key (but insufficient) idea

**Claim:** Let  $uv \in E$  be such that  $d(s, u) + d(v, t) < d(s, t)$ . Then any correct algorithm needs to see  $uv$ .

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**Claim:** Let  $uv \in E$  be such that  $d(s, u) + d(v, t) < d(s, t)$ . Then any correct algorithm needs to see  $uv$ .

**Proof:** Otherwise it cannot distinguish between  $\ell(uv) = \varepsilon$  and  $\ell(uv) = \infty$  and is incorrect in at least one case.



# Making the proof work

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This fails (bidirectional search sees a lot more edges).  
... but with a stronger claim, we can make it work

## Isn't this all trivial?

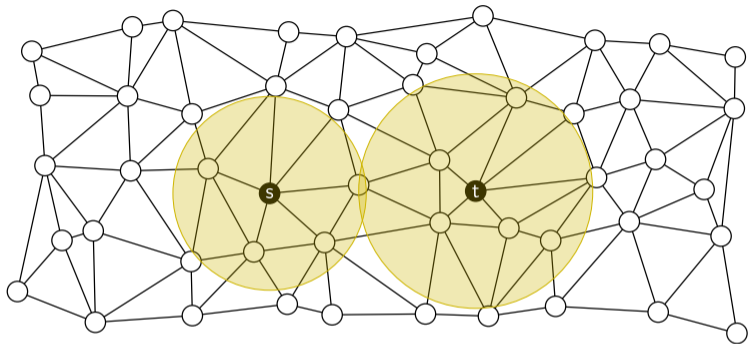
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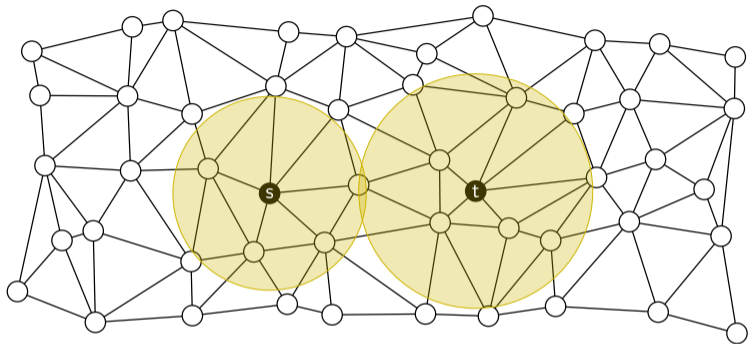
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- ▶ bidirectional search has many variants [Dan63; Nic66; Dre69; Poh69], none of the previous variants is instance-optimal
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- ▶ bidirectional search has many variants [Dan63; Nic66; Dre69; Poh69], none of the previous variants is instance-optimal
  - ▶ we design an instance-optimal variant
- ▶ we also prove that:
  - ▶ allow zero-length edges  $\Rightarrow$  instance-optimality impossible
  - ▶ **un**weighted graphs  $\Rightarrow$  instance-optimality only up to a max-degree factor





Thank you!



# Bibliography I

- [Dan63] George B. Dantzig. *Linear Programming and Extensions*. Princeton: Princeton University Press, 1963. ISBN: 9781400884179. DOI: doi:10.1515/9781400884179. URL: <https://doi.org/10.1515/9781400884179>.
- [Nic66] T Alastair J Nicholson. "Finding the shortest route between two points in a network". In: *The computer journal* 9.3 (1966), pp. 275–280.
- [Dre69] Stuart E Dreyfus. "An appraisal of some shortest-path algorithms". In: *Operations research* 17.3 (1969), pp. 395–412.
- [Poh69] Ira Pohl. *Bi-directional and heuristic search in path problems*. Tech. rep. SLAC National Accelerator Laboratory (SLAC), Menlo Park, CA (United States), 1969.
- [Eck+17] Jürgen Eckerle et al. "Sufficient conditions for node expansion in bidirectional heuristic search". In: *Proceedings of the International Conference on Automated Planning and Scheduling*. Vol. 27. 2017, pp. 79–87.
- [Sha+19] Eshed Shaham et al. "Optimally Efficient Bidirectional Search.". In: *IJCAI*. 2019, pp. 6221–6225.





# Backup Slides



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Intuition:

- ▶ fix  $G, s, t$
- ▶ set all lengths to 0
- ▶ now finding the shortest  $st$ -path is equivalent to finding *any*  $st$ -path
- ▶ there is a set of trivial dumb algorithms that guess what the path is

Banning 0-length edges helps because when a dumb algorithm finds a path, it still has to check if there isn't some shorter one.

## The Bidirectional Search Meta-Algorithm

while the *stopping condition* is not met:

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- ▶ explore one vertex at a time according to some rule [Dan63; Nic66; Poh69]
  - ▶ not instance-optimal!
- ▶ **our rule:** alternate doing  $\mathcal{O}(1)$  work in the forward and backward iterations

