## Advanced Shortest Paths: Bidirectional Dijkstra

Michael Levin

Higher School of Economics

# Graph Algorithms Data Structures and Algorithms

#### Outline

1 Bidirectional Search

2 Bidirectional Dijkstra

#### Shortest Path

Input: A graph G with non-negative edge

weights, a source vertex s and a

target vertex t. Graph can be directional or undirectional

Output: The shortest path between s and t

in G.

Can solve the shortest path problem using the undirectional graphs in "linear time". Such linear time Algo for directional graph is not known we can do better compared to conventional Djikstra's running time

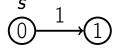
•  $O((|E| + |V|) \log |V|)$  is pretty fast, right?

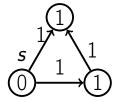
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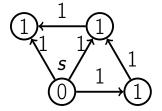
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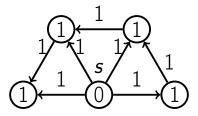
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- For a graph of USA with 20M vertices and 50M edges it will work for several seconds on average
- Millions of users of Google Maps want the result in a blink of an eye, all at the same time
- Need something significantly faster

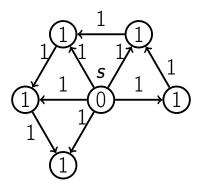


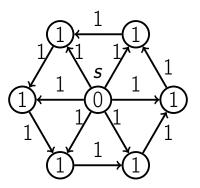


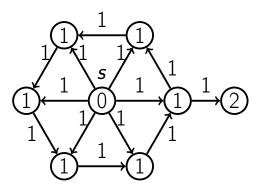


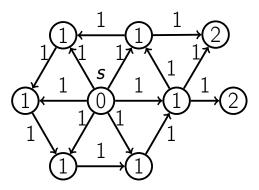


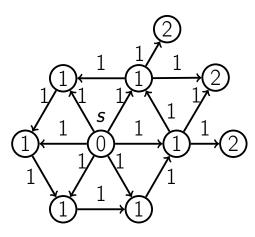


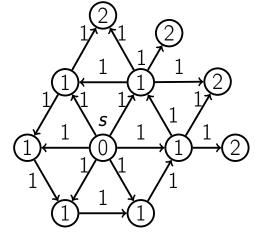


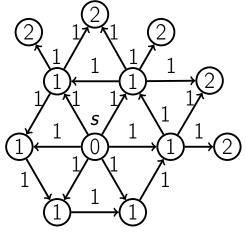


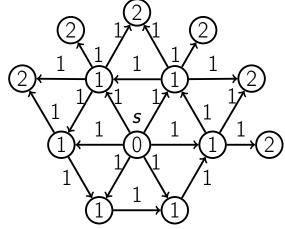


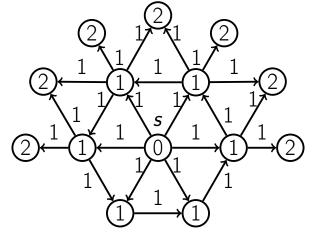


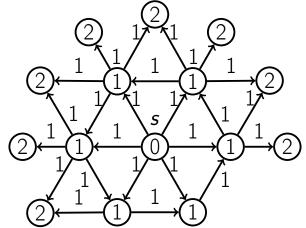


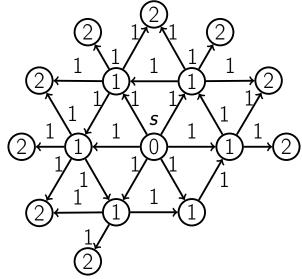


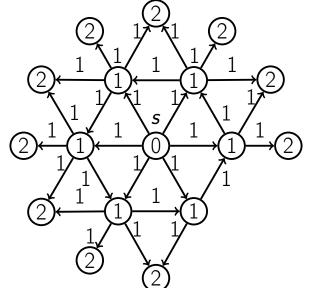


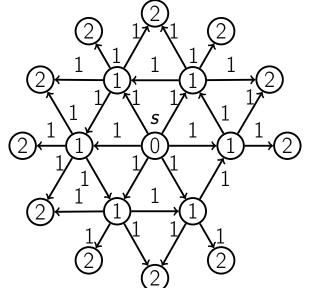


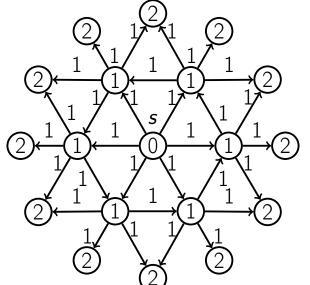












#### Lemma

When a vertex u is selected via ExtractMin, dist[u] = d(s, u).

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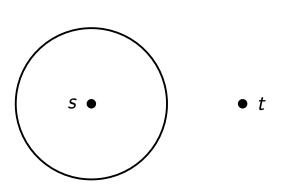
- When a vertex is extracted from the priority queue for processing, all the vertices at smaller distances have already been processed
- A "circle" of processed vertices grows

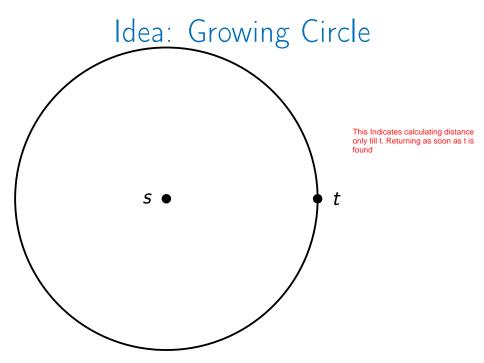
 $s \bullet t$ 



• t







#### Idea: Bidirectional Search





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Instead of growing from s to t, Grow from both the directions. From s grow forwards and from t grow backwards. And as soon as common node is found stop the algo. In Algo we cannot grow paths from s and t simultaneously hence we will alternate growing paths from s and t

#### Idea: Bidirectional Search

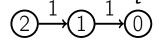
$$0 \xrightarrow{s} 1$$



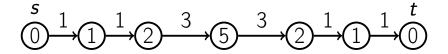
$$0 \xrightarrow{1} 1 \xrightarrow{1} 2$$

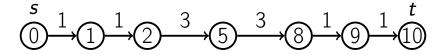


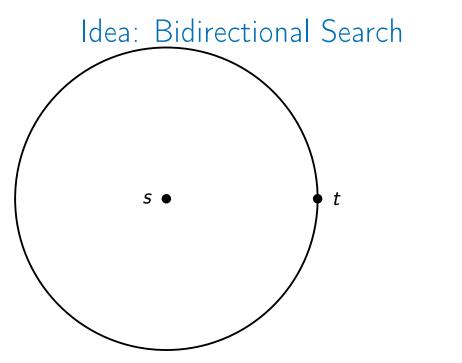
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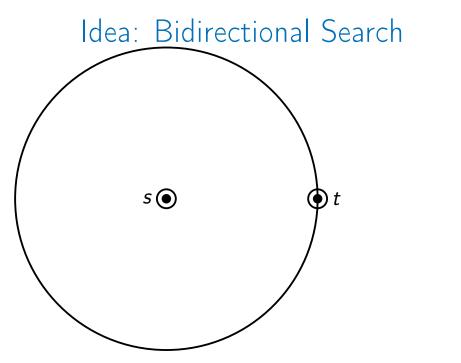


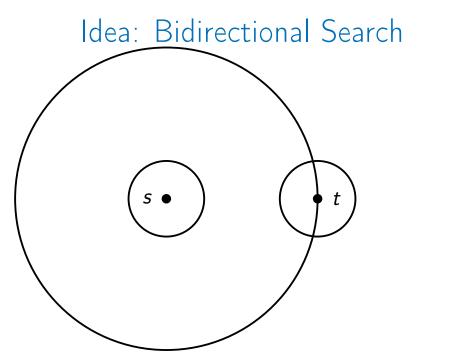
$$0 \xrightarrow{1} 1 \xrightarrow{1} 2 \xrightarrow{3} 5 \qquad 2 \xrightarrow{1} 1 \xrightarrow{1} 0$$

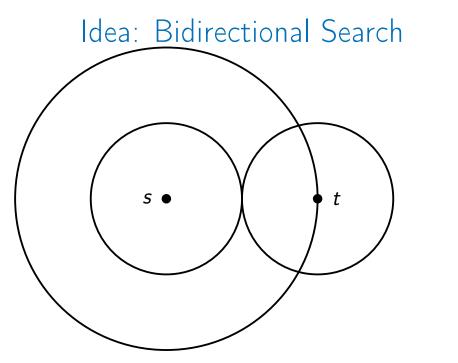


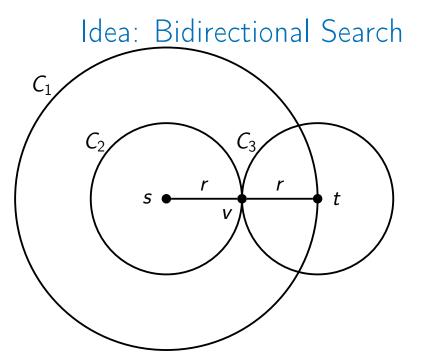












# Idea: Bidirectional Search $\operatorname{area}(C_1) = 4\pi r^2$ $2 \cdot \operatorname{area}(C_2) = 2\pi r^2$ Area covered (Nodes) by Unidirectional twice to that of Bidirectional

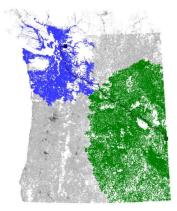


1.6M vertices, 3.8M arcs, travel time metric.



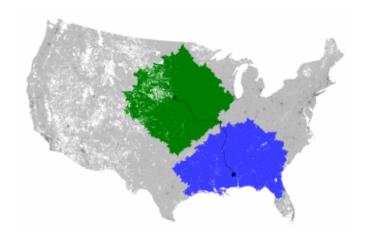
Searched area

Unidirectional Djikstra's



forward search/ reverse search

Bidirectional Djikstra's



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- Let's look at social networks

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- Can pass a message from any person to any person in at most 6 handshakes
- This is close to truth according to experiments and is called a "six handshakes" or "six degrees of separation" idea

Suppose an average person has around
 100 Facebook friends

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- Then 10000 friends of friends

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- Suppose an average person has around 100 Facebook friends
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- Not possible, as there are only about 7 billion people on earth

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- 1M + 1M = 2M people 1000 times

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- Instead of searching for all possible objects, search for first halves and for second halves separately
- Then find "compatible" halves
- Typically roughly  $O(\sqrt{N})$  instead of O(N), including the previous Facebook example

#### Conclusion

- Dijkstra goes in "circles"
- Bidirectional search idea can reduce the search space
- Roughly 2x speedup for road networks

  Meet in Middle Applicable on social network or many other problems and not roads
- Meet-in-the-middle  $\sqrt{N}$  instead of N
- 1000 times faster for social networks
- Next video Bidirectional Dijkstra algorithm

#### Outline

1 Bidirectional Search

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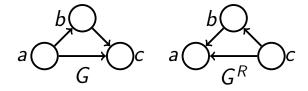
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- Repeat until t is processed

#### Reversed Graph

#### Definition

Reversed graph  $G^R$  for a graph G is the graph with the same set of vertices V and the set of reversed edges  $E^R$ , such that for any edge  $(u, v) \in E$  there is an edge  $(v, u) \in E^R$  and vice versa.



■ Build  $G^R$ 

- $\blacksquare$  Build  $G^R$
- Start Dijkstra from s in G and from t in  $G^R$

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- Stop when some vertex v is processed both in G and in  $G^R$
- Compute the shortest path between s and t

Let v be the first vertex which is processed both in G and in  $G^R$ . Does it follow that there is a shortest path from s to t going through v? Question

Let v be the first vertex which is processed both in the forward search from s in G and in the backward search from t in  $G^R$  in the process of Bidirectional Dijkstra algorithm launched to find the shortest path between s and t in graph G. Can we guarantee then that there exists a shortest path from s to t that goes through v?







See the counterexample in the lecture.



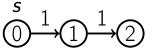




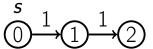


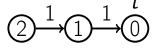


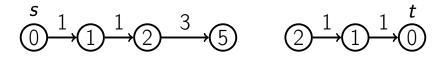


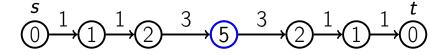


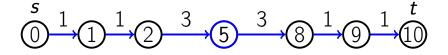


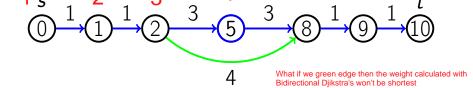


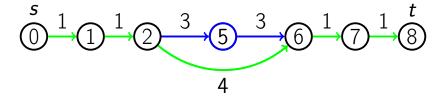






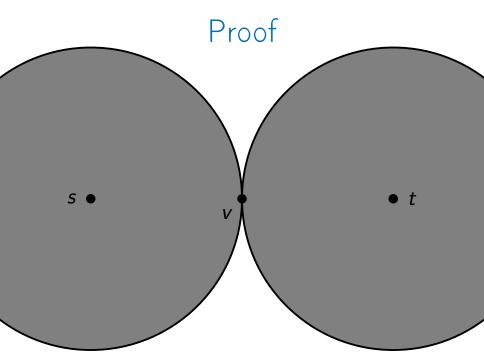


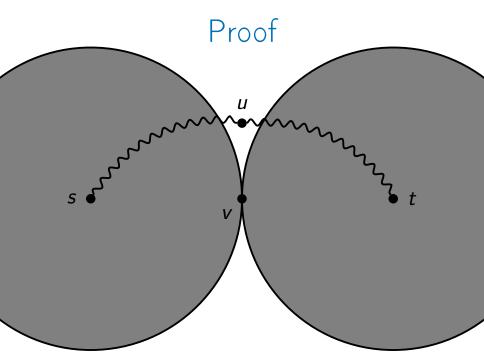


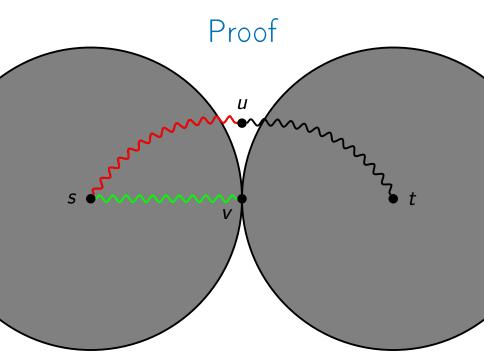


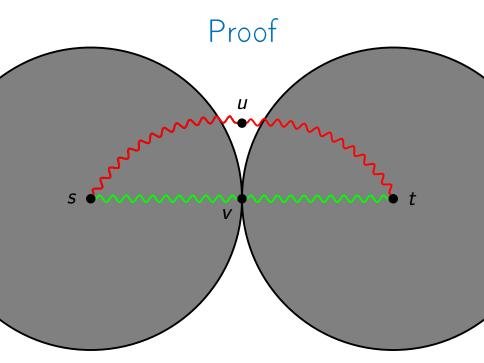
#### Lemma

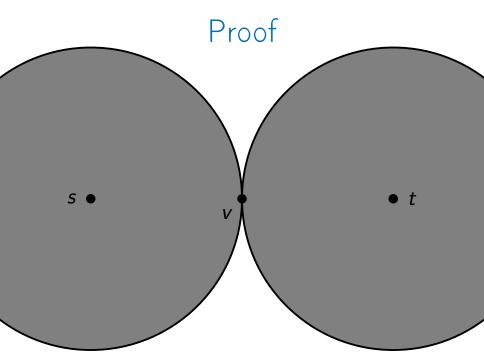
Let dist[u] be the distance estimate in the forward Dijkstra from s in G and dist<sup>R</sup>[u] — the same in the backward Dijkstra from tin  $G^R$ . After some node v is processed both in G and  $G^R$ , some shortest path from s to t passes through some node u which is processed either in G, in  $G^R$ , or both, and  $d(s,t) = \operatorname{dist}[u] + \operatorname{dist}^R[u].$ 

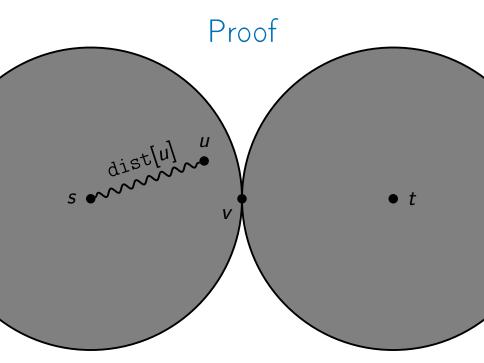


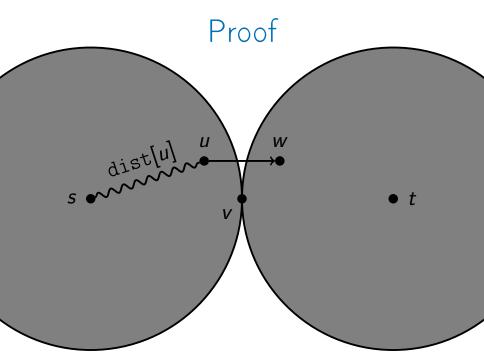


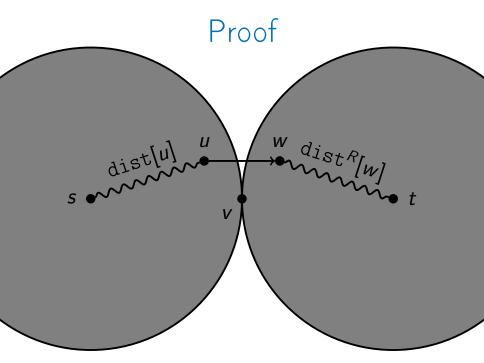


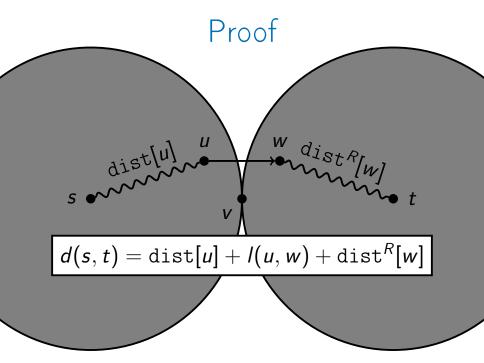


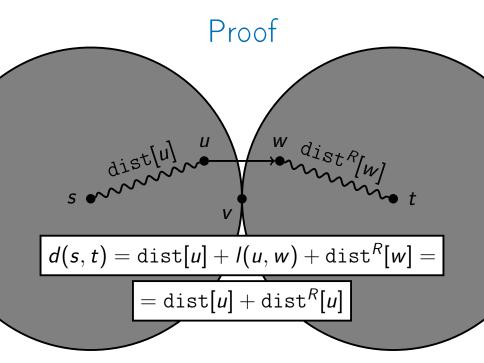












#### BidirectionalDijkstra(G, s, t)

```
G^R \leftarrow \text{ReverseGraph}(G)
Fill dist, dist<sup>R</sup> with +\infty for each node
\operatorname{dist}[s] \leftarrow 0, \operatorname{dist}^{R}[t] \leftarrow 0
Fill prev, prev^R with None for each node
proc \leftarrow empty, proc^R \leftarrow empty
do:
   v \leftarrow \text{ExtractMin(dist)}
   Process(v, G, dist, prev, proc)
   if v in proc<sup>R</sup>:
      return ShortestPath(s, dist, prev, proc, t,...)
   v^R \leftarrow \text{ExtractMin}(\text{dist}^R)
   repeat symmetrically for v^R as for v
while True
```

#### Relax(u, v, dist, prev)

if dist[v] > dist[u] + w(u, v):  $dist[v] \leftarrow dist[u] + w(u, v)$ 

 $prev[v] \leftarrow u$ 

# Process(u, G, dist, prev, proc)

for  $(u,v) \in E(G)$ : Relax $(u,v, ext{dist}, ext{prev})$ 

proc.Append(u)

```
ShortestPath(s, dist, prev, proc, t, dist<sup>R</sup>, prev<sup>R</sup>, proc<sup>R</sup>)
distance \leftarrow +\infty, u_{best} \leftarrow None
for u in proc + proc<sup>R</sup>:
   if dist[u] + dist^R[u] < distance:
       u_{hest} \leftarrow u
       distance \leftarrow dist[u] + dist^R[u]
path \leftarrow empty
last \leftarrow u_{best}
while last \neq s:
   path.Append(last)
   last ← prev[last]
path \leftarrow Reverse(path)
last \leftarrow u_{best}
while last \neq t:
   last \leftarrow prev^R[last]
   path.Append(last)
```

return (distance, path)

#### Conclusion

- Worst-case running time of Bidirectional Dijkstra is the same as for Dijkstra
- Speedup in practice depends on the graph For Road Networks 2X but for social networks it can be 1000 times
- Memory consumption is 2x to store G and  $G^R$
- You'll see the speedup on social network graph in the Programming Assignment