

# Advanced Shortest Paths: Contraction Hierarchies

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Graph Algorithms  
Data Structures and Algorithms

# Outline

- 1 Contraction Hierarchies
- 2 Preprocessing
- 3 Witness Search
- 4 Query
- 5 Query Correctness
- 6 Node Ordering

# Learning Objectives

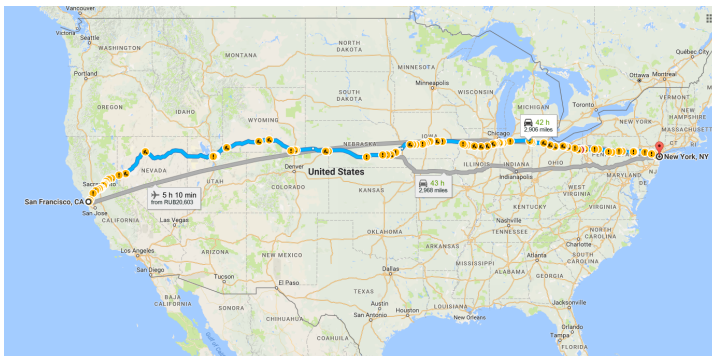
- Bidirectional Dijkstra can be 1000s of times faster than Dijkstra for social networks
- But just 2x speedup for road networks
- This lecture — great speedup for road networks

# Highway Hierarchies

- Long-distance trips go through highways

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# Highway Hierarchies

- Long-distance trips go through highways

42 h (2,906 miles)



via I-80 E

40 h without traffic

⚠ This route has tolls

---

**San Francisco, CA**

USA

> Get on US-101 S

3 min (0.6 mi)

> Follow I-80 E to Holland Tunnel in Jersey City

42 h (2,903 mi)

> Continue on Holland Tunnel. Drive to Steve Flanders Square in Manhattan, New York

10 min (2.8 mi)

**New York, NY**

USA

# Highway Hierarchies

- Long-distance trips go through highways
- To get from  $A$  to  $B$ , first merge into a highway, then into a bigger highway, etc., then exit to a highway, then exit to a street, then go to  $B$

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- Less important roads merge into more important roads



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- To get from  $A$  to  $B$ , first merge into a highway, then into a bigger highway, etc., then exit to a highway, then exit to a street, then go to  $B$
- Less important roads merge into more important roads
- Hierarchy of roads

# Highway Hierarchies

- There are algorithms based on this idea
- “Highway Hierarchies” and “Transit Node Routing” by Sanders and Schultes
- Millions of times faster than Dijkstra
- Pretty complex
- This lecture — “Contraction Hierarchies”, thousands of times faster than Dijkstra

# Node Ordering

- Nodes can be ordered by some “importance”
- Importance first increases, then decreases back along any shortest path
- E.g., points where a highway merges into another highway
- Can use bidirectional search

# Importance Ideas

Many shortest paths involve important nodes



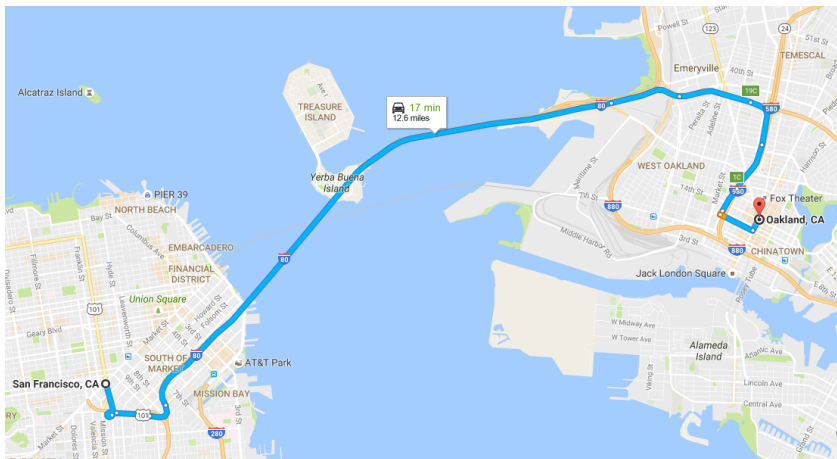
# Importance Ideas

Important nodes are spread around



# Importance Ideas

Important nodes are sometimes unavoidable



# Shortest Paths with Preprocessing

- Preprocess the graph
- Find distance and shortest path in the preprocessed graph
- Reconstruct the shortest path in the initial graph

# Outline

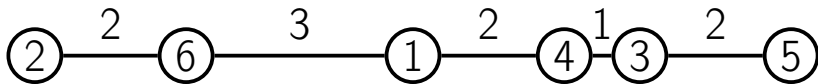
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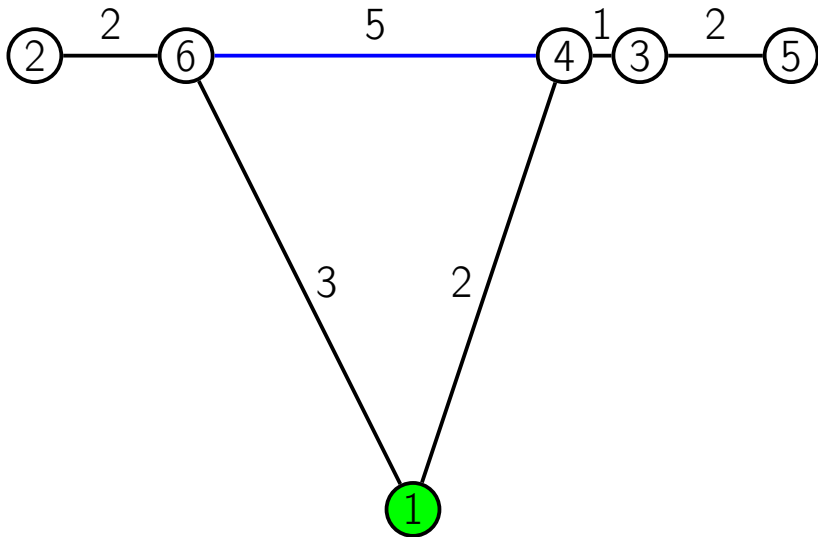
# Preprocessing

- Eliminate nodes one by one in some order
- Add **shortcuts** to preserve distances
- Output: augmented graph + node order

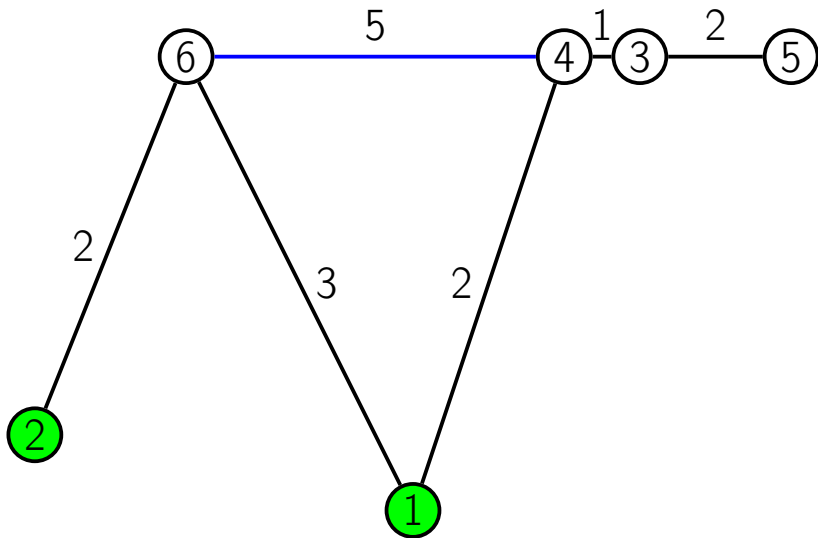
# Node Contraction



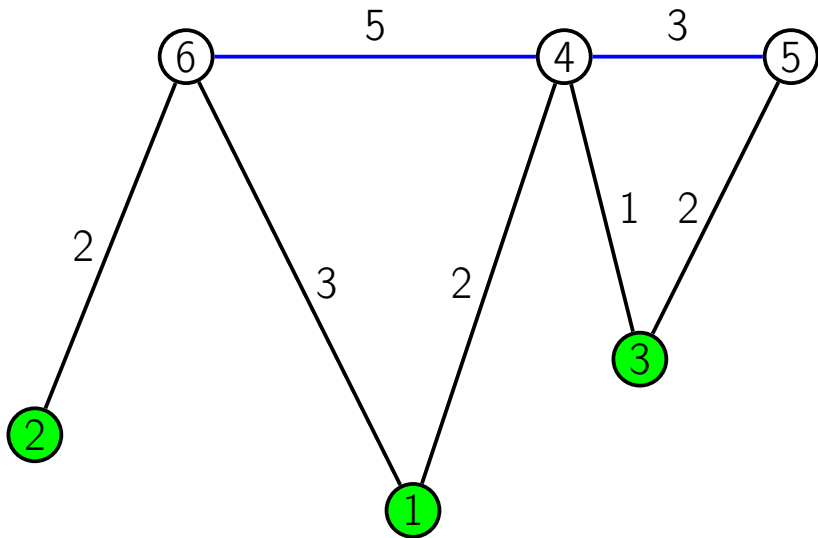
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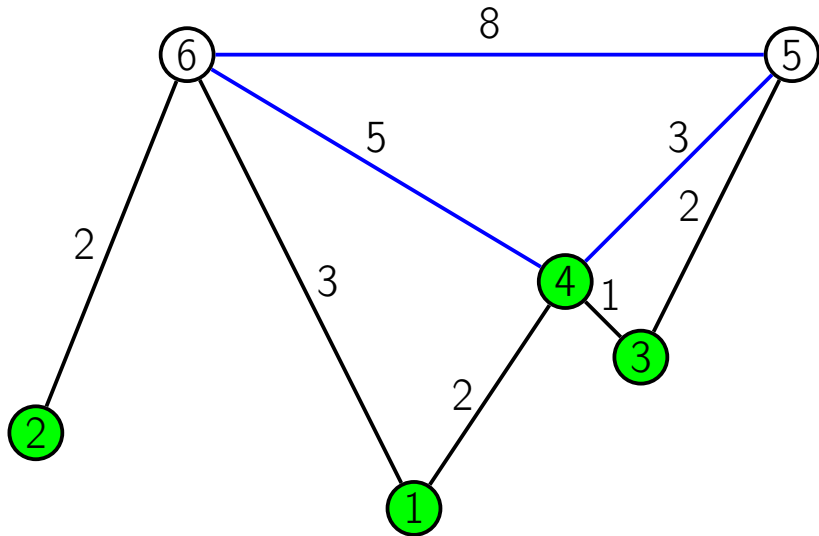
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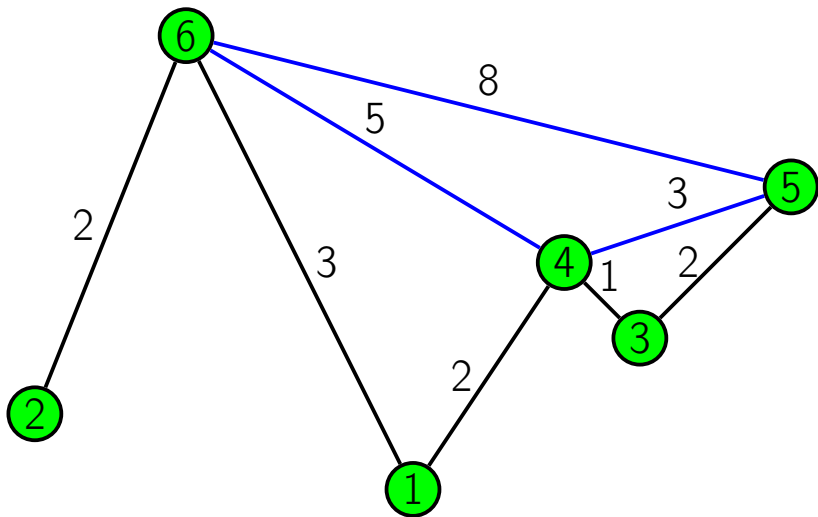
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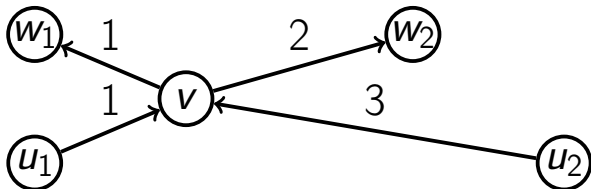


# Node Contraction



# Witness Paths

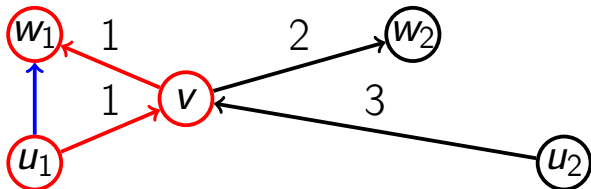
- Contraction of node  $v$





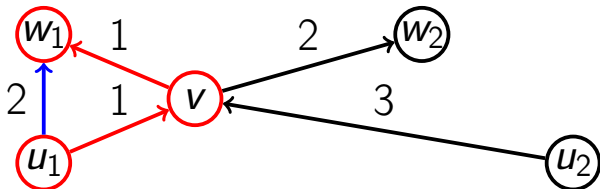
# Witness Paths

- Contraction of node  $v$
- For every pair of edges  $(u, v)$ ,  $(v, w)$  add a new edge  $(u, w)$



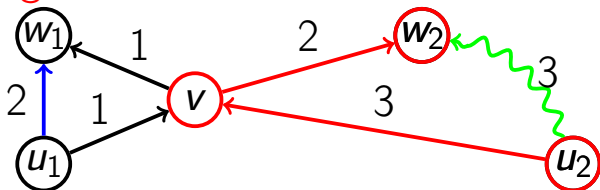
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# Witness Paths

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- For every pair of edges  $(u, v)$ ,  $(v, w)$  add a new edge  $(u, w)$
- $\ell(u, w) \leftarrow \ell(u, v) + \ell(v, w)$
- But only if there is no **witness path**  $P_{uw}$  shorter than  $\ell(u, v) + \ell(v, w)$  and **bypassing**  $v$



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# Witness Search

When contracting node  $v$ , for any pair of edges  $(u, v)$  and  $(v, w)$  we want to check whether there is a **witness path** from  $u$  to  $w$  bypassing  $v$  with length at most  $\ell(u, v) + \ell(v, w) - 1$  — then there is no need to add a shortcut from  $u$  to  $w$ .

## Definition

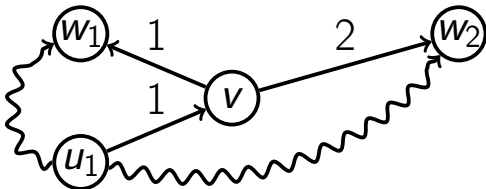
**Witness search** is the search for a **witness path**.

## Definition

If there is an edge  $(u, v)$ , call  $u$  a predecessor of  $v$ . If there is an edge  $(v, w)$ , call  $w$  a successor of  $v$ .

# Witness Search

- For each predecessor  $u_i$  of  $v$ , run Dijkstra from  $u_i$  ignoring  $v$
- Essential for good query performance
- Otherwise the augmented graph will be very dense



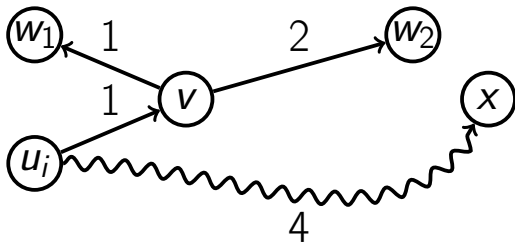
# Witness Search Optimizations

- Stop Dijkstra when distance from the source becomes too big
- Limit the number of hops



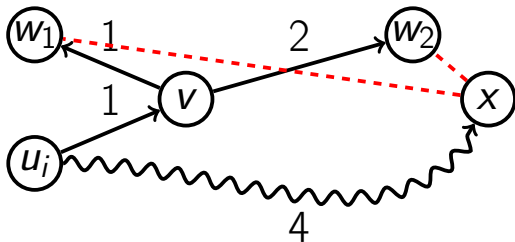
# Stop Dijkstra

- If  $d(u_i, x) > \max_{u, w} (\ell(u, v) + \ell(v, w))$ ,  
there is no witness path going through  $x$



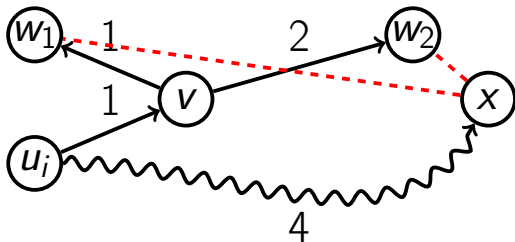
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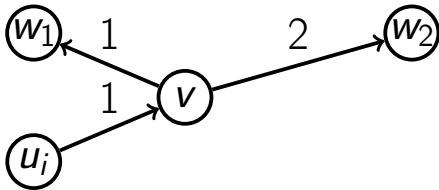
# Stop Dijkstra

- If  $d(u_i, x) > \max_{u,w}(\ell(u, v) + \ell(v, w))$ ,  
there is no witness path going through  $x$
- Limit the distance by  
 $\max_{u,w}(\ell(u, v) + \ell(v, w))$



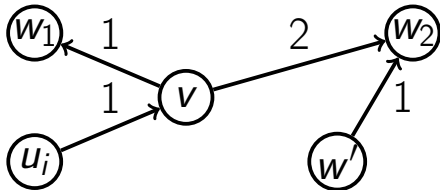
# Stop Dijkstra

- Consider any predecessor  $w'$  of any successor  $w$  of  $v$



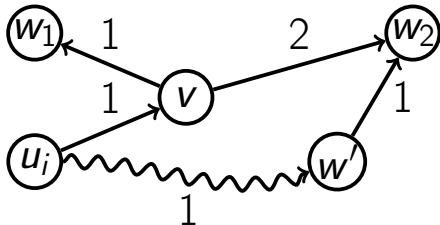
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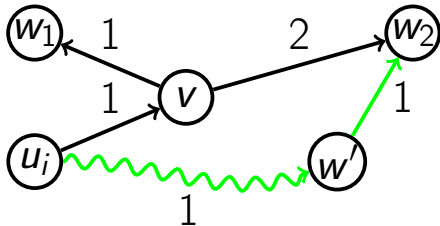
# Stop Dijkstra

- Consider any predecessor  $w'$  of any successor  $w$  of  $v$
- If
$$d(u, w') + \ell(w', w) \leq \ell(u, v) + \ell(v, w),$$
there's a witness path



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# Stop Dijkstra

- If
$$d(u, w') + \ell(w', w) \leq \ell(u, v) + \ell(v, w),$$
there's a witness path
- Limit the distance by
$$\max_{u, w} \max_{(w', w)} (\ell(u, v) + \ell(v, w) - \ell(w', w))$$

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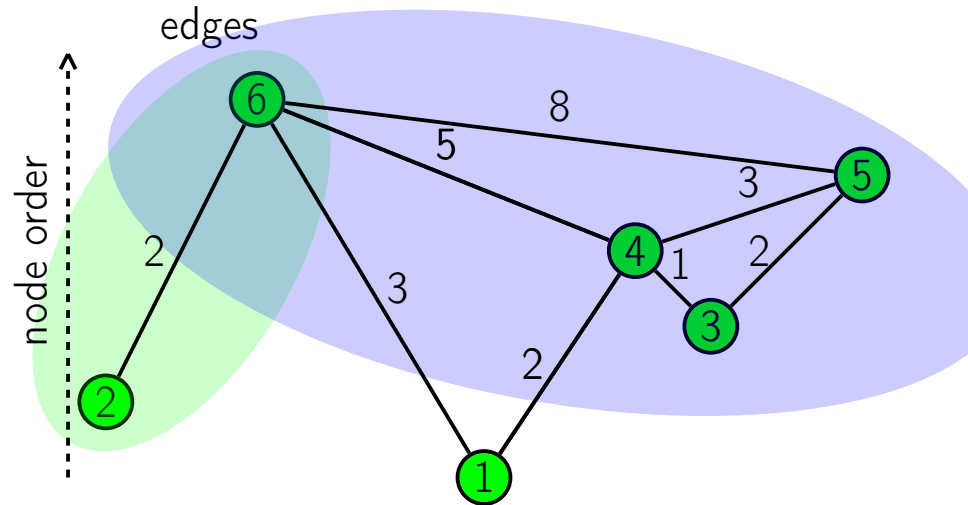
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- Consider only shortest paths from source with at most  $k$  edges
- If witness path not found, add a shortcut
- Tradeoff between preprocessing time and augmented graph size
- E.g., start with  $k = 1$ , increase gradually to  $k = 5$  in the end

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# Bidirectional Dijkstra

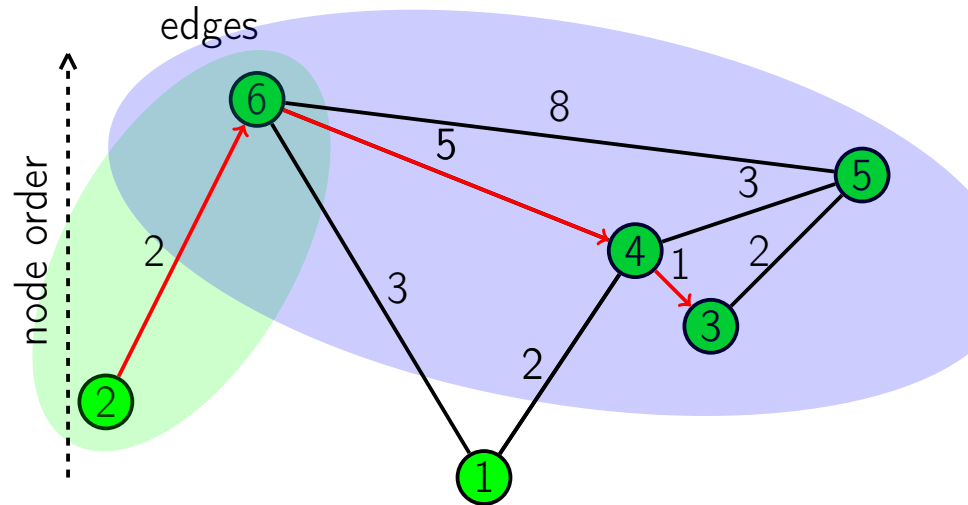
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# Bidirectional Dijkstra

- Bidirectional Dijkstra using only upwards edges
- Don't stop when some node was processed both by forward and backward searches
- Stop Dijkstra when the extracted node is already farther than the target

## ComputeDistance( $s, t, \dots$ )

```
estimate  $\leftarrow +\infty$ 
Fill dist,  $\text{dist}^R$  with  $+\infty$  for each node
dist[s]  $\leftarrow 0$ ,  $\text{dist}^R[t] \leftarrow 0$ 
proc  $\leftarrow$  empty,  $\text{proc}^R \leftarrow$  empty
while there are nodes to process:
    v  $\leftarrow$  ExtractMin(dist)
    if dist[v]  $\leq$  estimate:
        Process(v, ...)
    if v in  $\text{proc}^R$  and dist[v] +  $\text{dist}^R[v] <$  estimate:
        estimate  $\leftarrow$  dist[v] +  $\text{dist}^R[v]$ 
     $v^R \leftarrow$  ExtractMin( $\text{dist}^R$ )
    Repeat symmetrically for  $v^R$ 
return estimate
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- Why is algorithm for query correct?

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# Augmented Graph

## Definition

The **augmented graph**  $G^+ = (V, E^+)$  is the graph on the same set of vertices  $V$  as the initial graph  $G$  and an augmented set of edges  $E^+$  that contains all the initial edges  $E$  of the graph  $G$  along with the shortcuts added at the preprocessing stage.

# Distance Preservation

## Lemma

The distance  $d^+(s, t)$  between any two nodes  $s$  and  $t$  in the **augmented graph**  $G^+ = (V, E^+)$  is equal to the distance  $d(s, t)$  between these nodes in the initial graph  $G = (V, E)$ .

## Proof

- Edges are only added to  $G$ , so
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- Thus  $d^+(s, t) = d(s, t)$



## Definition

The **rank**  $r(v)$  of vertex  $v$  is the position of  $v$  in the node order returned by the preprocessing stage.

## Definition

A path  $P: v_1 \rightarrow v_2 \rightarrow \dots \rightarrow v_k$  in the augmented graph  $G^+$  is called **increasing** if  $r(v_1) < r(v_2) < \dots < r(v_k)$ . Similarly,  $P$  is called **decreasing** if  $r(v_1) > r(v_2) > \dots > r(v_k)$ .

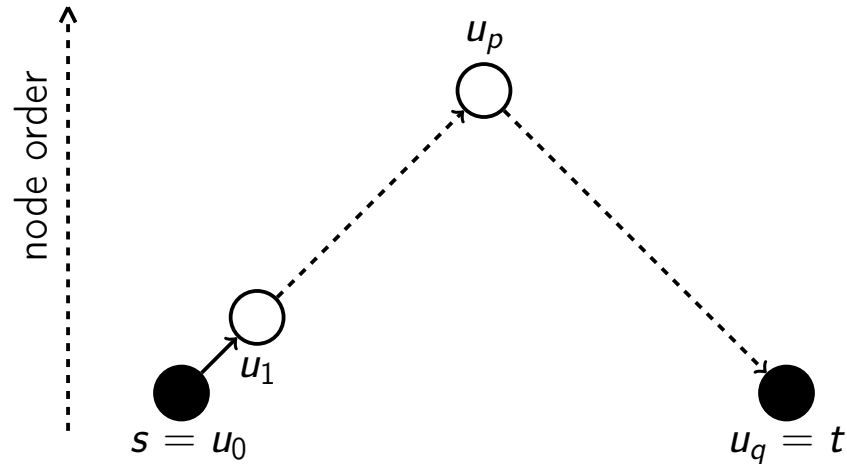
# Justification of Bidirectional Search

## Lemma

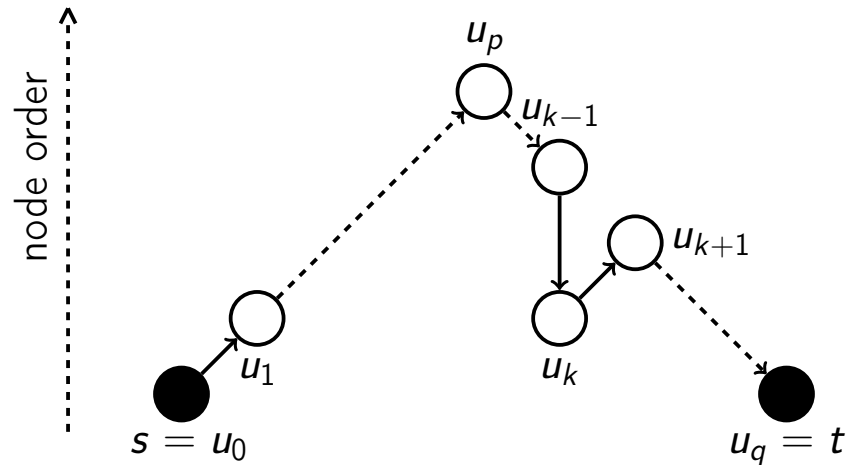
For any  $s$  and  $t$ , the augmented graph  $G^+ = (V, E^+)$  contains a shortest path  $P_{st}$  such that the subpath  $P_{sv}$  is increasing and  $P_{vt}$  is decreasing.



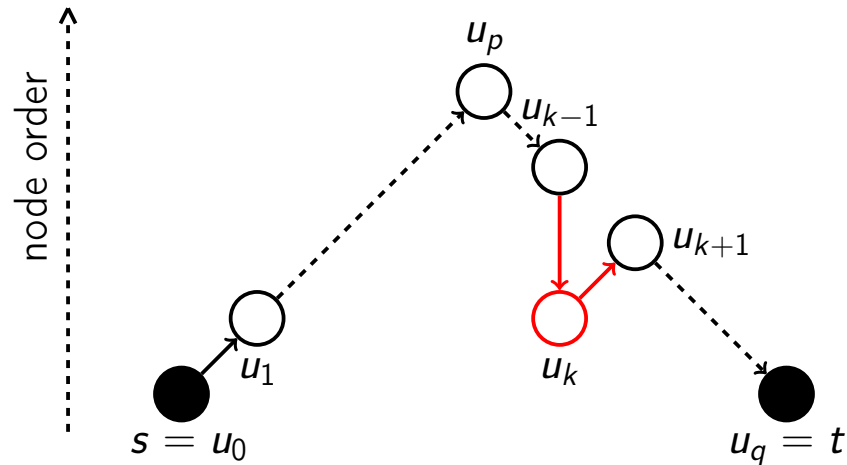
# Proof Idea



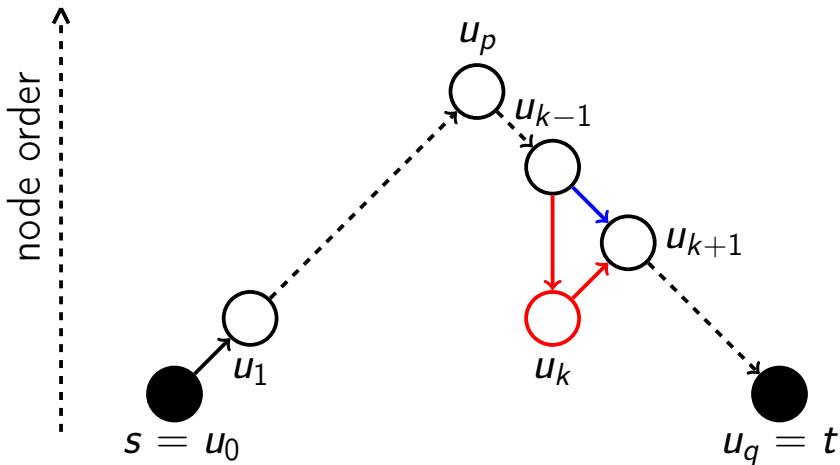
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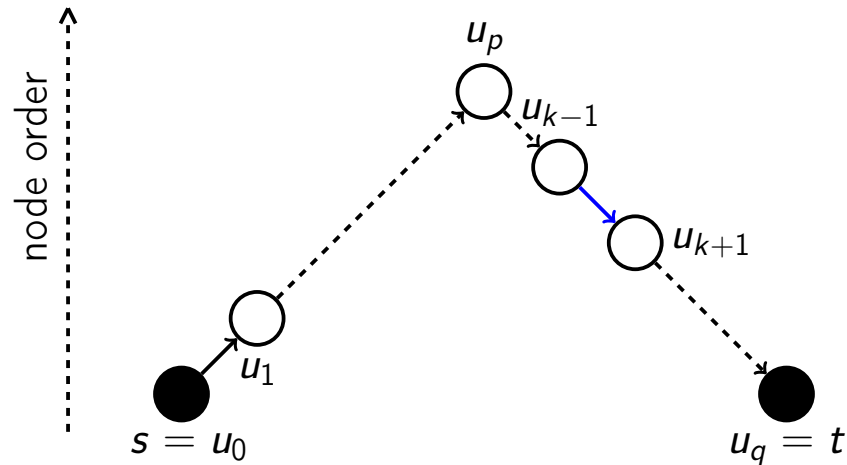
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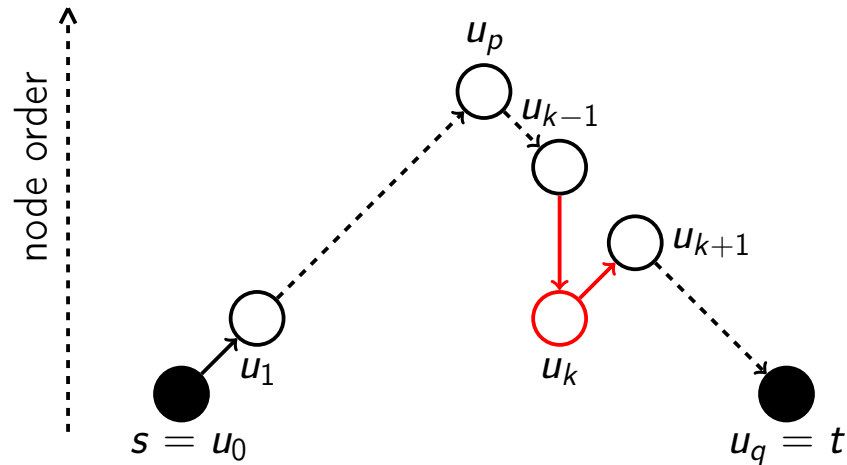
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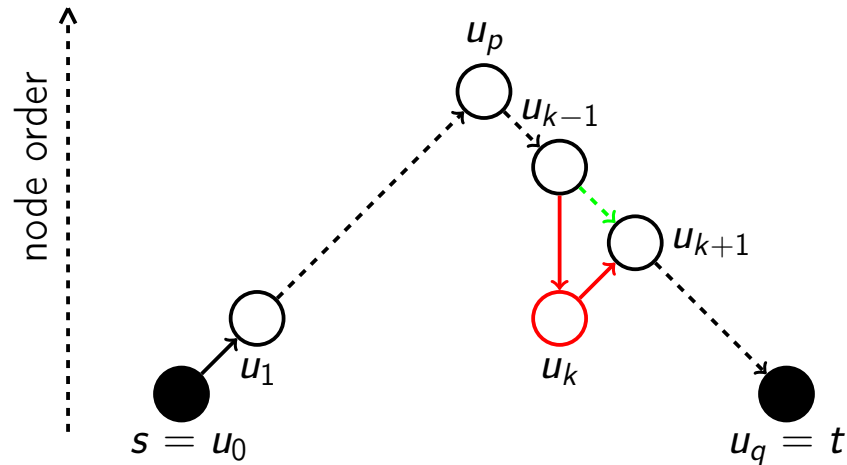
# Proof Idea



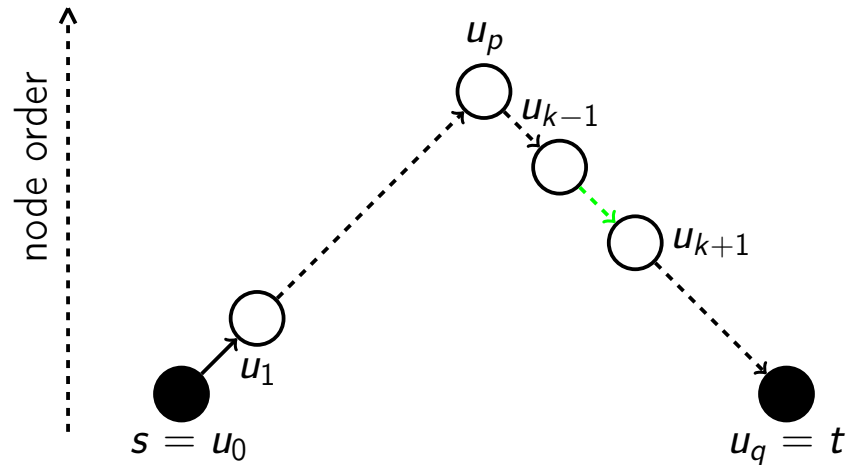
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## Proof

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- Then for any shortest path  $P: s = u_1 \rightarrow u_2 \rightarrow \dots \rightarrow u_k = t$  there is a node  $u_i$ , such that  $r(u_{i-1}) > r(u_i) < r(u_{i+1})$  — call it a **local minimum**

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- For any shortest path  $P$  between  $s$  and  $t$ , denote by  $m(P)$  the minimum rank of a local minimum of this path

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- For any shortest path  $P$  between  $s$  and  $t$ , denote by  $m(P)$  the minimum rank of a local minimum of this path
- Consider the shortest path  $P^*$  with the maximum  $m(P)$ , consider the local minimum  $u_k$  with
$$r(u_{k-1}) > r(u_k) = m(P) < r(u_{k+1})$$

## Proof

- If a shortcut  $(u_{k-1}, u_{k+1})$  was added when  $u_k$  was contracted, there is a shortest path  $P'$  with this shortcut instead of  $u_{k-1} \rightarrow u_k \rightarrow u_{k+1}$ , and  $P'$  doesn't contain  $u_k$ , so  $m(P') > m(P^*) = r(u_k)$  — contradiction with the choice of  $P^*$  with the maximum  $m(P)$

## Proof

- Otherwise, there was a witness path from  $u_{k-1}$  to  $u_{k+1}$  comprised by nodes with rank higher than  $r(u_k)$  (they were contracted after  $u_k$ ) — there is a shortest path  $P''$  with this path instead of  $u_{k-1} \rightarrow u_k \rightarrow u_{k+1}$ , and  $m(P'') > m(P^*)$  — contradiction □

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- Query correctness is proven
- Are we done?
- How to select the node order?

# Outline

- 1 Contraction Hierarchies
- 2 Preprocessing
- 3 Witness Search
- 4 Query
- 5 Query Correctness
- 6 Node Ordering

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- However, preprocessing and query time depend heavily on it
- Minimize the number of added shortcuts
- Spread the important nodes across the graph
- Minimize the number of edges in the shortest paths in the augmented graph

# Order by Importance

- Introduce a measure of importance

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- Introduce a measure of importance
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- Importance can change after that

# Algorithm

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- On each iteration, extract the least important node
- Recompute its importance
- If it's still minimal (compare with the top of the priority queue), contract the node
- Otherwise, **put it back** into priority queue with new priority

# Eventual Stopping

- If we don't contract a node, we update its importance
- After at most  $|V|$  attempts all nodes have updated importance
- The node with the minimum updated importance will be contracted after that

# Importance criteria

- Edge difference
- Number of contracted neighbors
- Shortcut cover
- Node level

# Edge Difference

- Want to minimize the number of edges in the augmented graph
- Number of added shortcuts  $s(v)$ , incoming degree  $in(v)$ , outgoing degree  $out(v)$
- Edge difference
$$ed(v) = s(v) - in(v) - out(v)$$
- Number of edges increases by  $ed(v)$  after contracting  $v$
- Contract node with small  $ed(v)$

# Contracted Neighbors

- Want to spread contracted nodes across the graph
- Contract a node with small number of already contracted neighbors  $cn(v)$

# Shortcut Cover

- Want to contract important nodes late
- **Shortcut cover**  $sc(v)$  — the number of neighbors  $w$  of  $v$  such that we have to shortcut to or from  $w$  after contracting  $v$
- If shortcut cover is big, many nodes “depend” on  $v$
- Contract a node with small  $sc(v)$

# Node Level

- Node level  $L(v)$  is an upper bound on the number of edges in the shortest path from any  $s$  to  $v$  in the augmented graph
- Initially,  $L(v) \leftarrow 0$
- After contracting node  $v$ , for neighbors  $u$  of  $v$  do  $L(u) \leftarrow \max(L(u), L(v) + 1)$
- Contract a node with small  $L(v)$

# Importance

- Use importance

$$I(v) = ed(v) + cn(v) + sc(v) + L(v)$$

- You can play with weights of those 4 quantities in  $I(v)$  and see how preprocessing time and query time change
- Each of the 4 quantities is necessary for fast preprocessing/queries
- Find a way to compute them efficiently at any stage of the preprocessing



# Comparison with Dijkstra

- On a graph of Europe with 18M nodes, on random pairs of vertices Dijkstra works for 4.365s on average
- On the same graph and same random pairs, with the best set of heuristics Contraction Hierarchies work for 0.18ms on average — almost 25000 times faster!

# Conclusion

- Preprocess by contracting nodes ordered *approximately* by importance
- Query by Bidirectional Dijkstra on the augmented graph
- Importance function is heuristic, but works well on road network graphs
- 1000s of times faster than Dijkstra
- Compete on the forums whose solution is the fastest!