# Route Planning for Bicycles – Exact Constrained Shortest Paths made Practical via Contraction Hierarchy

Sabine Storandt



**ICAPS** 2012



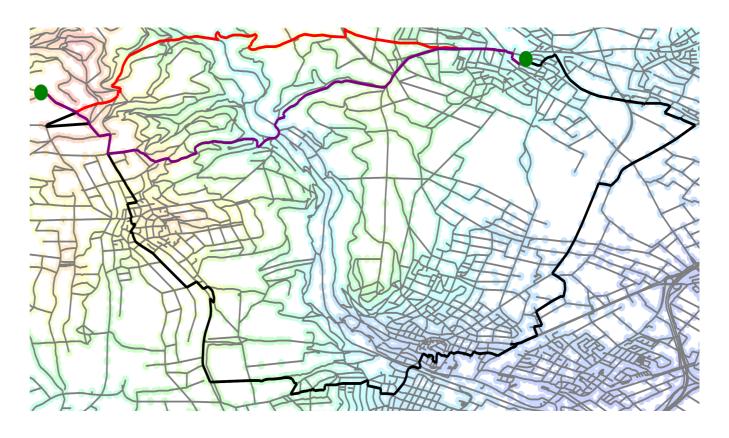
### MOTIVATION

### Bicycle route from A to B

- should be short
- but bear not too much hard climbs

### **Optimization Problem**

Find the shortest path from A to B with a (positive) height difference smaller than H.



length height difference

**RED**: 7.5km 517m

BLACK: 19.1km 324m

PURPLE: 7.7km 410m



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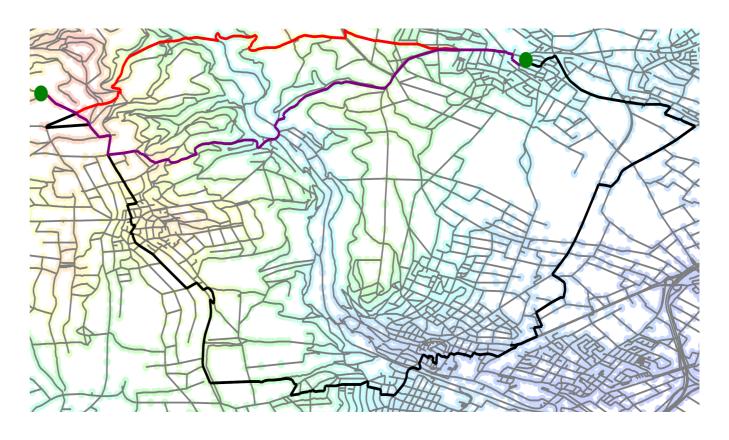
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Constrained Shortest Path(CSP)
NP-hard



len	gth	height	difference

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### FORMAL PROBLEM DEFINITION

#### Given

G(V, E) (street) graph

 $c: E \to \mathbb{R}_0^+ \text{ cost}$ 

 $r: E \to \mathbb{R}_0^+$  resource consumption

#### Goal

for  $s,t\in V$ ,  $R\in\mathbb{R}^+_0$  compute minimal cost path p from s to t whose resource consumption does not exceed R

$$\min c(p) = \sum_{e \in p} c(e)$$
 s.t.  $r(p) = \sum_{e \in p} r(e) \le R$ 

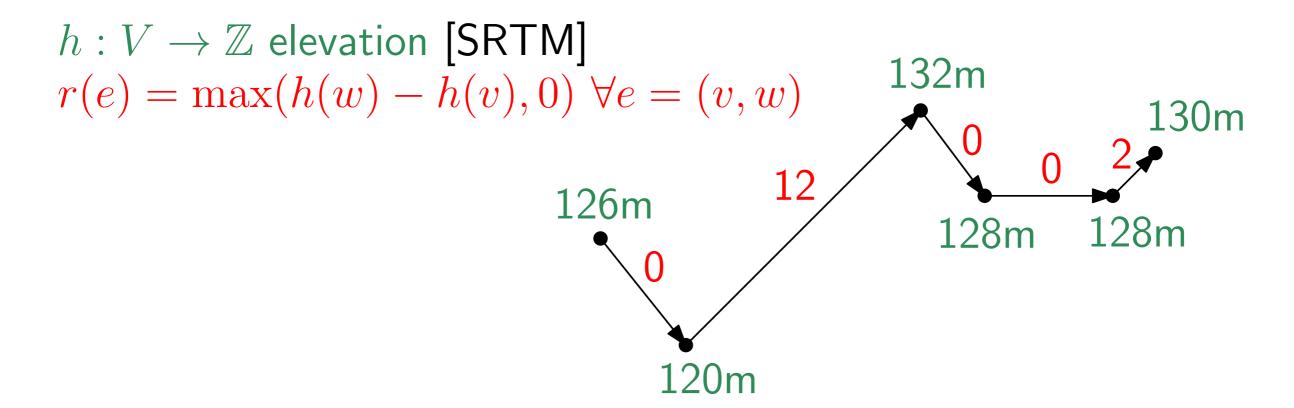


### FORMAL PROBLEM DEFINITION

### **Bicycle Route Planning**

<u>costs:</u> euclidean distance [OSM]

<u>resource</u>: positive height difference





### CONTRIBUTION

**Adaption** of speed-up techniques for the shortest path problem to reduce

- query time
- space consumption

for exact CSP computation in large street networks.

**Focus** Contraction Hierarchy

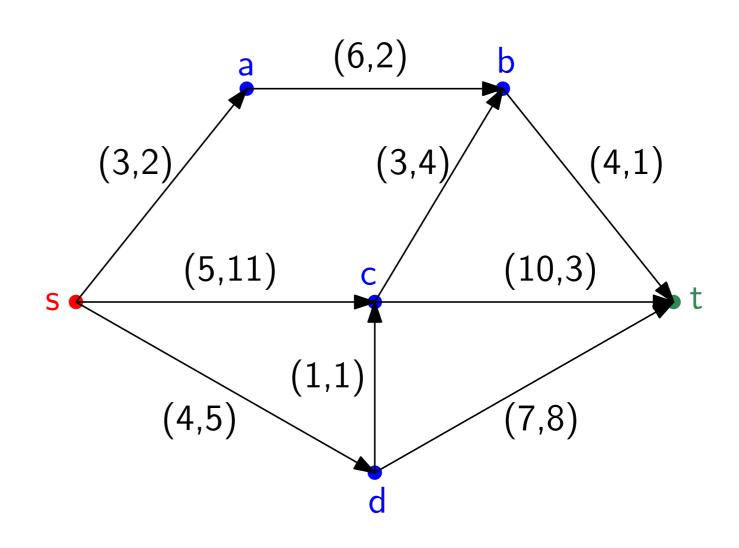


[Aggarwal, Aneja, and Nair 1982]

#### **Approach**

Assign to each node the list of pareto-optimal tuples.

**Pareto-optimal**  $\widehat{=}$  no dominating path exists



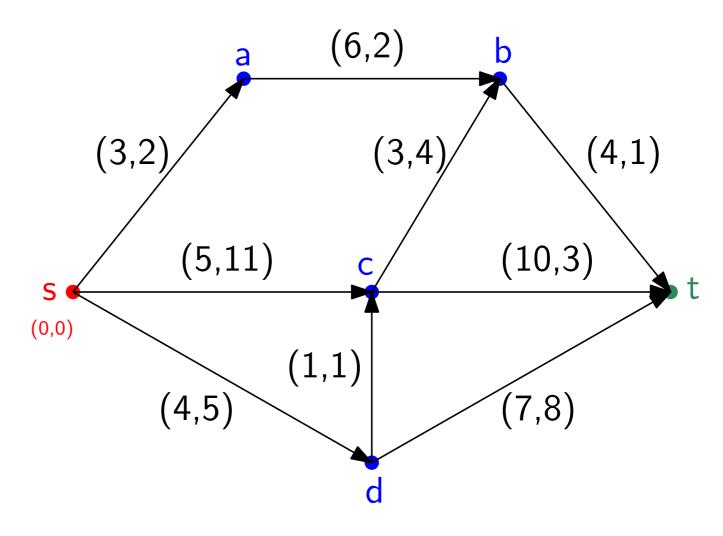


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$$PQ = (0,0,s)$$

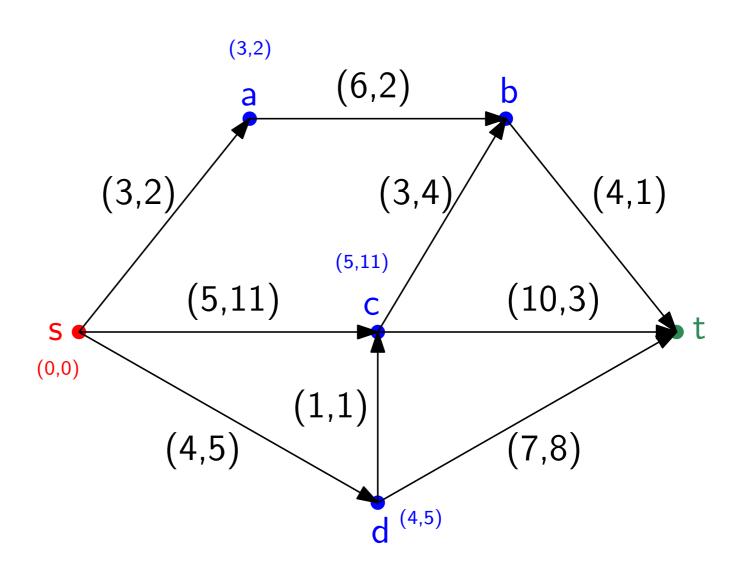


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$$PQ = (3,2,a), (4,5,d), (5,11,c)$$

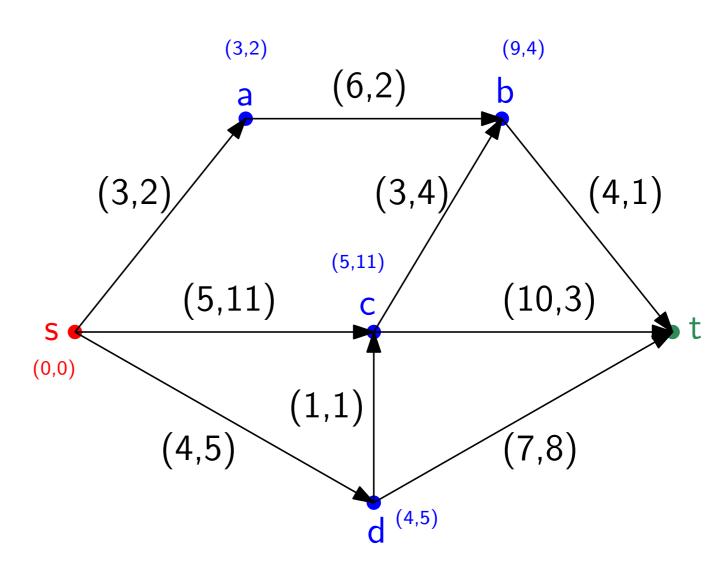


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$$PQ = (4,5,d), (5,11,c), (9,4,b)$$

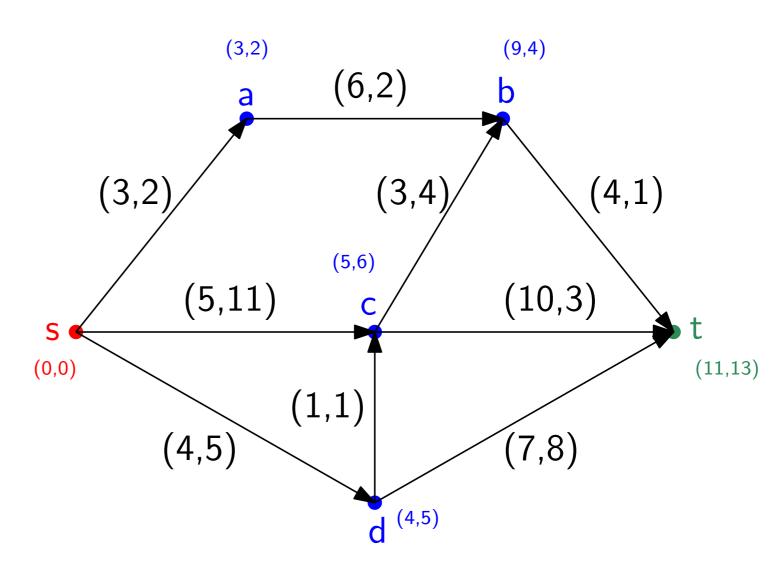


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$$PQ = (5,6,c), (9,4,b)$$

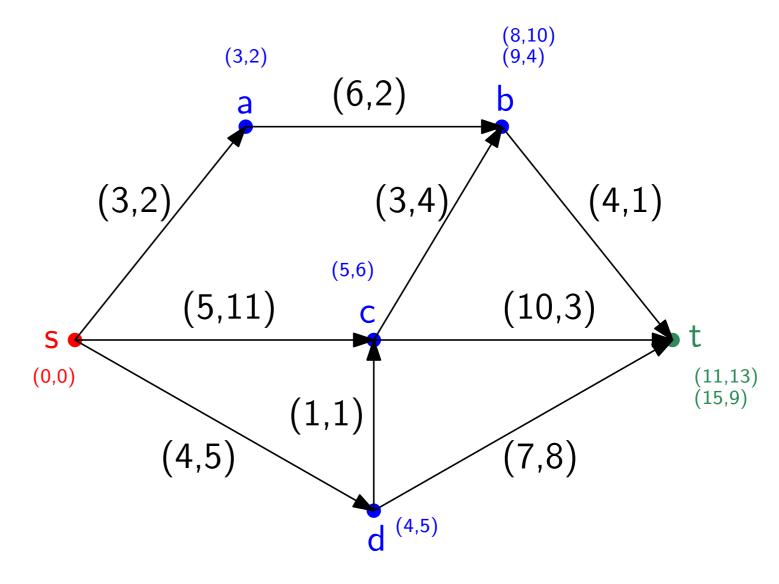


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$$PQ = (8,10,b),(9,4,b)$$

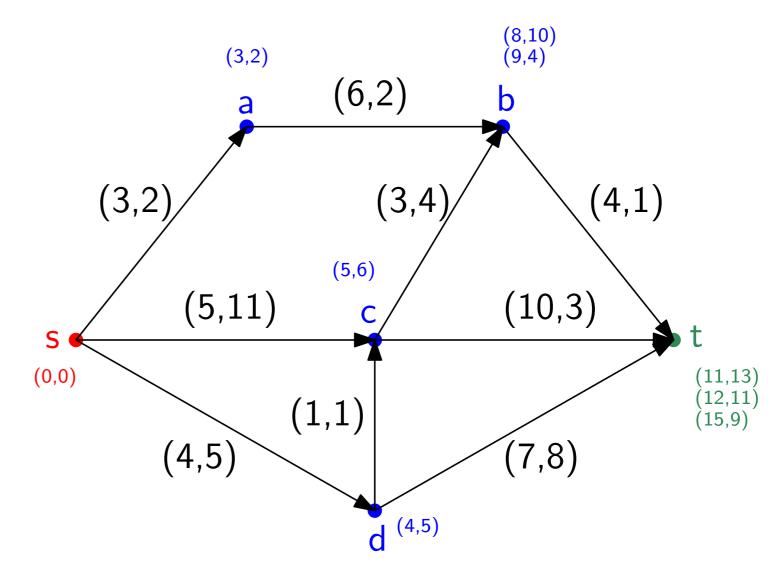


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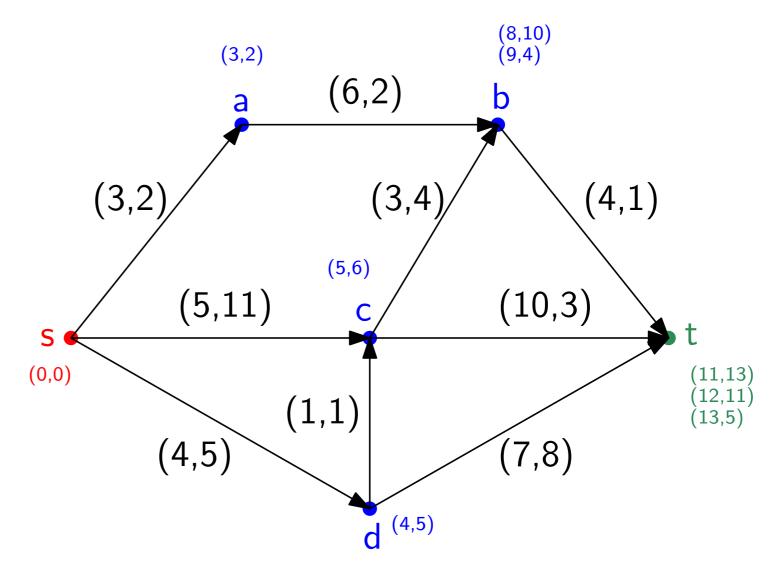


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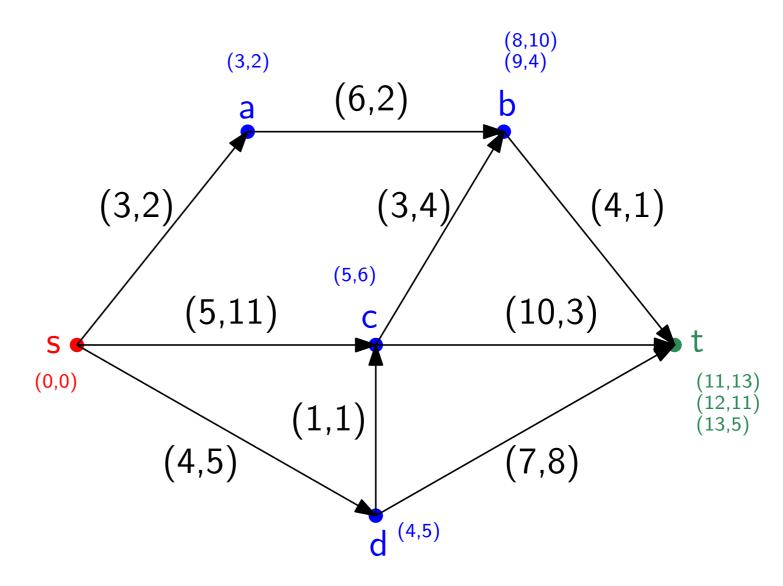
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### Similarities to Dijkstra

- operates directly on the graph
- PQ and edge relaxation
- bidirectional version exists



$$PQ = \emptyset$$



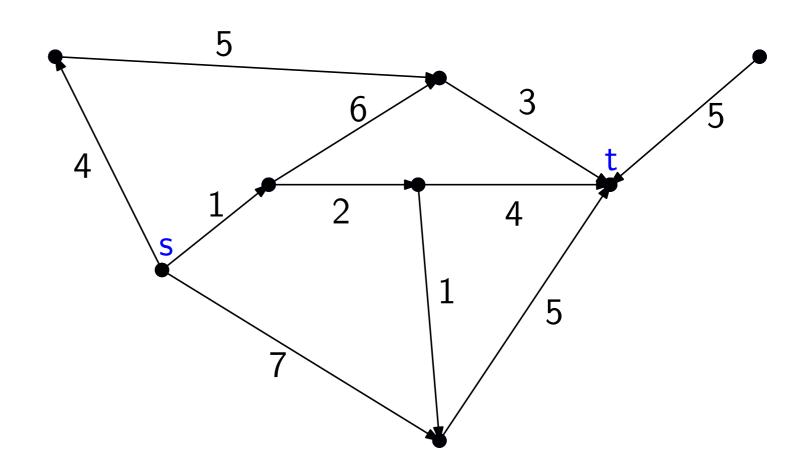
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#### Idea

Consider only resource consumption

 $\forall v \in V$  compute minimal resource consumption  $r_{min}$  for a path  $s, \dots, v, \dots, t$  (via two Dijkstra runs)

Prune all nodes with  $r_{min}(v) > R$ 





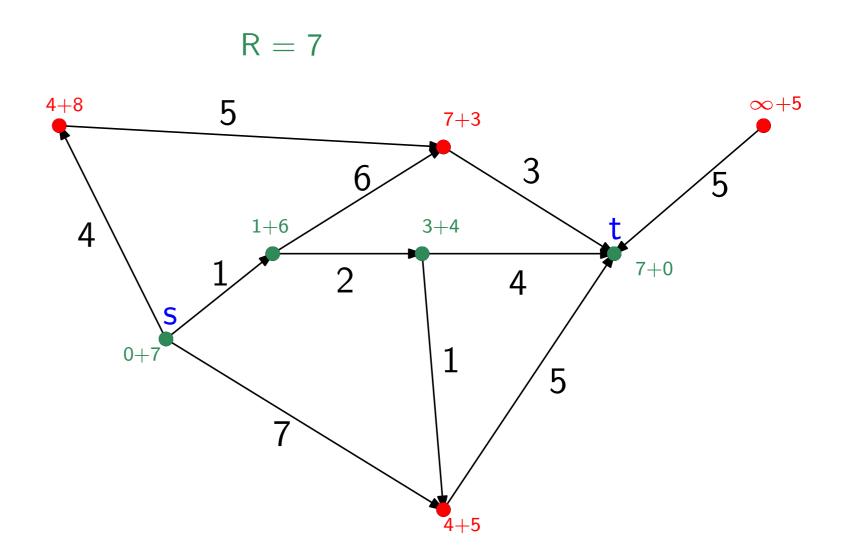
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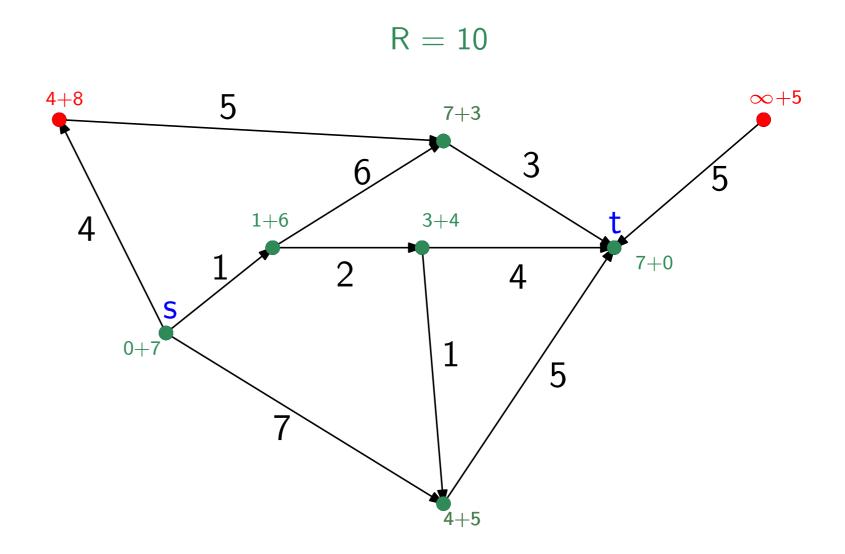
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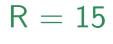
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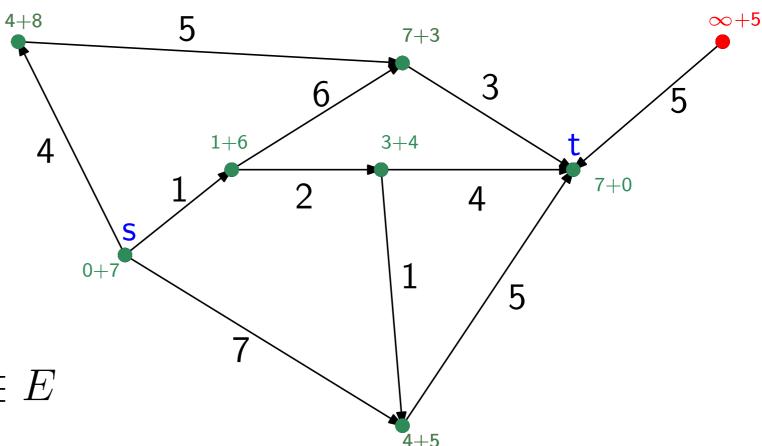
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#### **Problem**

Impact low if

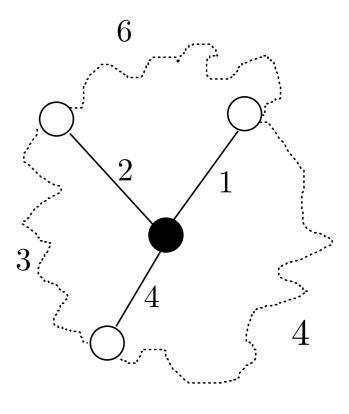
- R is large
- r(e) small for many  $e \in E$





[Geisberger et al. 2008]

### Graph preprocessing method

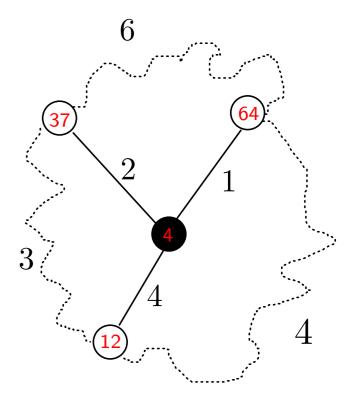




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### Graph preprocessing method

1. Assign distinct importance values to the nodes





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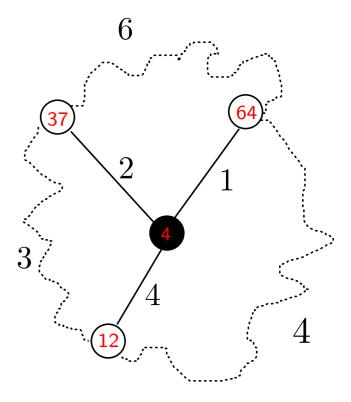
### Graph preprocessing method

- 1. Assign distinct importance values to the nodes
- 2. Remove nodes one by one in order of importance ('contraction')

**Task:** maintain all shortest path distances in remaining graph

Add shortcut if no witness found

Witness: path shorter than reference path





[Geisberger et al. 2008]

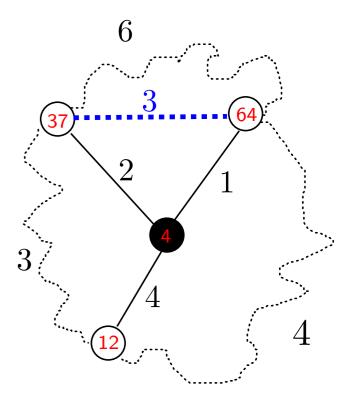
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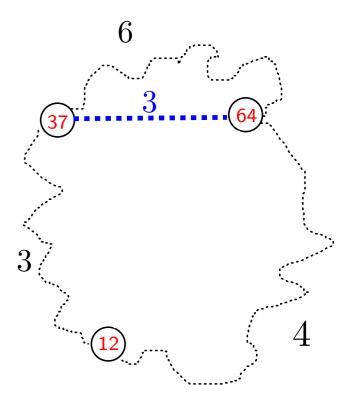
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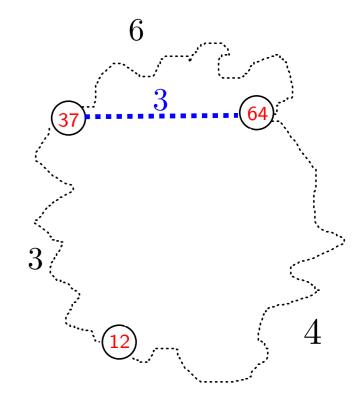
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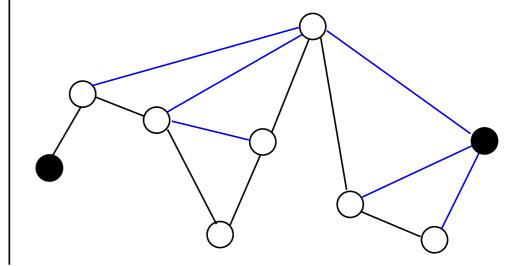
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#### **Query Answering**

bidirectional: only relax edges to nodes with higher importance





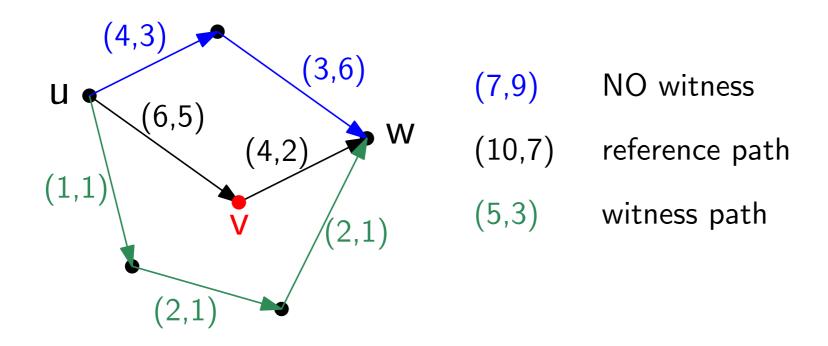


importance

Task maintain all pareto-optimal paths
Witness must dominate reference path

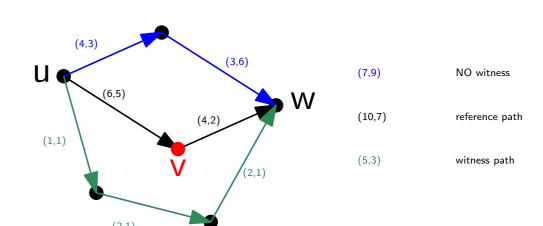


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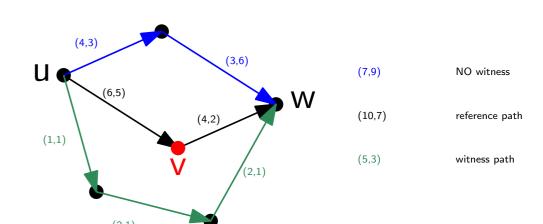
#### **Naive Witness Search**

reference path p = uvw

- start label setting computation(LSC) in u with R=r(p)
- if w receives label with  $c \leq c(p), r \leq r(p)$ , break  $\to$  witness path found
- insert shortcut if no witness was found



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Problem
LSC might be very time and space consuming



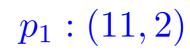
#### **Basic Idea**

Restrict witness search first to paths on the lower convex hull.

### Lower Convex Hull(LCH)

for every v-w-path p: represent (c(p), r(p)) as line segment  $\lambda c(p) + (1-\lambda)r(p)$ ,  $\lambda \in [0,1]$ 

 $p \in LCH(v,w) \Leftrightarrow \exists \lambda \in [0,1]$  for which line segment of p is minimal

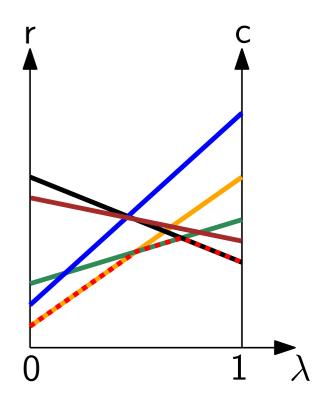




$$p_3:(1,8)$$

$$p_4:(5,7)$$

$$p_5:(4,8)$$





#### Basic Idea

Restrict witness search first to paths on the lower convex hull.

#### **Advantage**

paths on the LCH can be found by a Dijkstra run in  $G^{\lambda}$ 

 $G^{\lambda}$ : edges have single weight  $w(e) = \lambda c(e) + (1 - \lambda)r(e)$ 



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In which cases does exploring the LCH help? What if LCH check procedure is inconclusive?

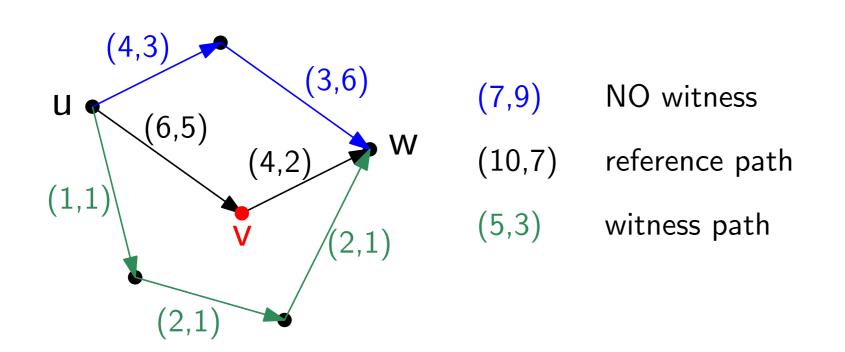


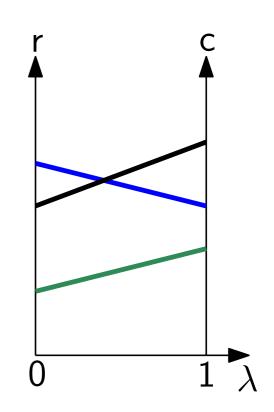
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1. If dominating path is part of the LCH. witness path found, shortcut can be omitted

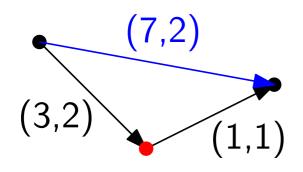






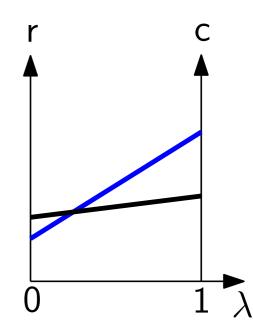
### In which cases does exploring the LCH help?

- 1. If dominating path is part of the LCH. witness path found, shortcut can be omitted
- 2. If reference path is part of the LCH. no dominating path exists, shortcut must be inserted



(7,2)

(4,3) reference path





#### What if LCH check procedure is inconclusive?

#### Reasons

- 1. Neither p nor a possible witness are part of the LCH.
- 2. Number of  $\lambda$  support points too small.

#### **Possibilities**

- Apply LSC on top.
   or
- Add shortcut without further care.



### EXPERIMENTAL RESULTS

#### Test Graphs 10k - 5.5m nodes

#### **Preprocessing**

- t=3 support points led to a conclusive result of the LCH-checker in 62% of the cases
- number of edges in CH-graph about twice the number of original edges (comparable to the conventional case)

### **Query Answering**

- speed-up about two orders of magnitude
- remarkably less space consumption (8GB laptop sufficient, before some queries failed even on a 96GB server)



### CONCLUSIONS

Can answer exact CSP queries in graphs with up to 500k nodes in time less than one second!

#### Also in the paper...

- speed-up via CH for dynamic programming CSP solution
- CSP-variant of arc-flags

#### **Future Work**

- ullet combination with other techniques/heuristics (e.g.  $A^*$ )
- consider other metric combinations and more complicated scenarios, e.g. edge cost functions



# THANK YOU...

... for your attention!

**Questions?** 

