

# Vehicle Routing With Time-Windows

## Presentation

Anshul Sawant<sup>1</sup>   Catalin-Stefan Tiseanu<sup>2</sup>

<sup>1</sup>University of Maryland,  
asawant@cs.umd.edu

<sup>2</sup>University of Maryland,  
ctiseanu@cs.umd.edu

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# Outline

- 1 Introduction
- 2 Point-to-Point Orienteering
- 3 Routing with Time-windows

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# Problem Definition

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- Let  $G = (V, E)$  be the graph. Then  $r_i$  is the reward,  $R_i$  is the release time and  $D_i$  is deadline for node  $i$ .
- $D_{max}$  is the maximum deadline.

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# Related Problems

- **k-TSP**. Visit  $k$  nodes on a graph. Minimize total distance travelled. 2-approx by **Garg**.
- **Minimum Excess Path**. Visit  $k$  cities on a path  $P$  from  $s$  to  $t$ , minimize the difference  $d_P(s, t) - d(s, t)$ .  $2 + \epsilon$  approximation by **Blum et al.**. Uses k-TSP as a sub-routine.
- **Point-to-Point Orienteering**. Find a path from  $s$  to  $t$  with  $d_P(s, t) \leq D$  which maximizes the number of cities visited. Uses Minimum Excess Path as a sub-routine.



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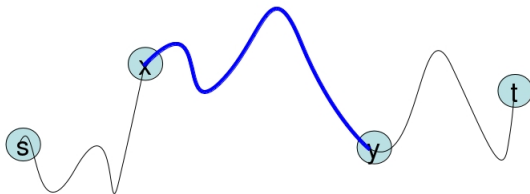
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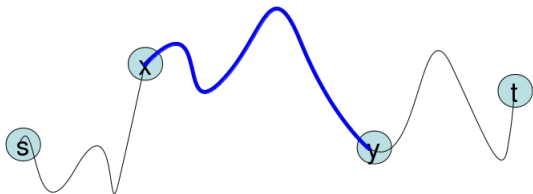
# Orienteering Intuition

- Let reward be evenly spread among segments  $s - x$ ,  $x - y$  and  $y - t$ .
- In one of the segments, we can increase the excess 3 times if we minimize excess in the other segments.
- If we have a 3-approximation for min-excess problem we can get all the reward from segment  $x - y$ .



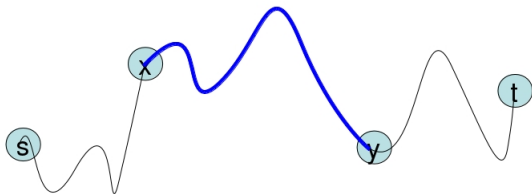
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# Algorithm

- For each pair of nodes  $(x, y)$  and value  $k$ , we compute a minimum excess path from  $x$  to  $y$  which visits  $k$  nodes.
- We then select the triple  $(x, y, k)$  with the maximum  $k$ , such that the computed path has excess  $D - d(s, x) - d(x, y) - d(y, t)$  or smaller.
- We return the path which travels from  $s$  to  $x$  via shortest path, then  $x$  to  $y$  via the computed path, then  $y$  to  $t$  via shortest path.

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# Time-windows Intuition

- If we can start after release time, we don't have to worry about release time. Problem is reduced to deadline-TSP.
- If some nodes above a deadline  $d$  are captured by time  $d$ , then we can collect the same reward by reducing deadlines of all such nodes to  $d$ .
- If deadlines are constant for all vertices, then deadline-TSP can be solved using point-to-point orienteering.

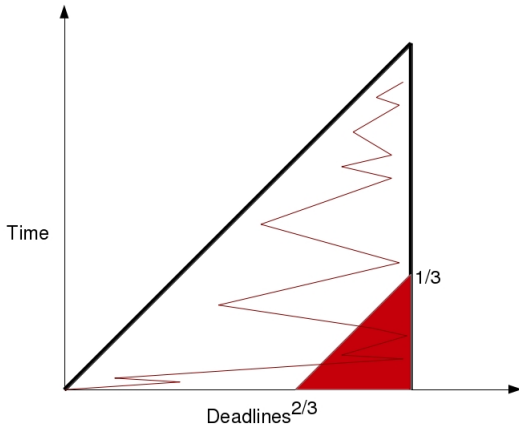
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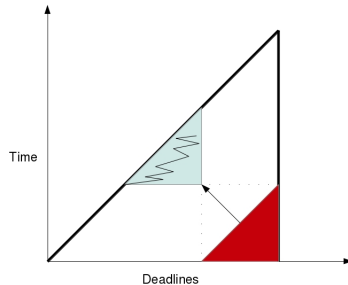
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# Optimal Path



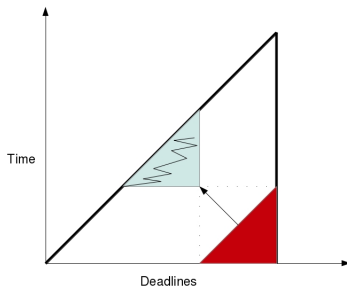
# A Path That Does Well

- Starts after release time of nodes in red area.
- Ends at at time equal to least deadline in the red area.
- Gets a constant factor of reward by using Point-to-Point Orienteering.



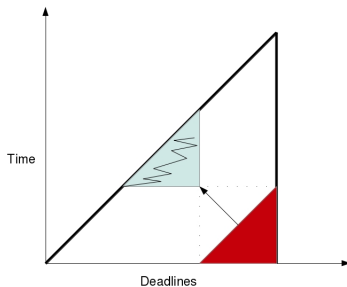
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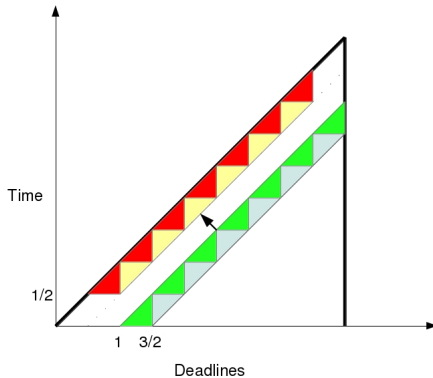
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# Partitioning the Time-Deadline Space

- We want to get a constant factor of reward from each segment using DP.





## Outline of Algorithm for a Segment

- Let the segment start at deadline  $d$  and let its width be  $k$ .
- For  $i = 0$  to  $j$  such that  $d + k * j \leq D_{max}$ 
  - Let  $S_i$  be the set of all vertices with deadlines in  $[d + i * k, d + (i + 1) * k)$  and release times  $\leq (i + 1) * k$ .
  - Assign all vertices in  $V - S_i$  a reward of 0. Let other rewards be the same as in original instance.
  - With the above reward assignment, let  $S(G, x, y, l)$  be the approximate solution to p2p orienteering problem on graph  $G$ , with path length  $l$ , starting point  $x$  and end point  $y$ .
  - $A[i, x, y, l] = S(G, x, y, l) \forall x, y \in S_i, \forall l \in \{1, \dots, k\}$

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