

Algorithmen für Routenplanung

18. Sitzung, Sommersemester 2013

Thomas Pajor | 10. Juli 2013

INSTITUT FÜR THEORETISCHE INFORMATIK · ALGORITHMIK · PROF. DR. DOROTHEA WAGNER

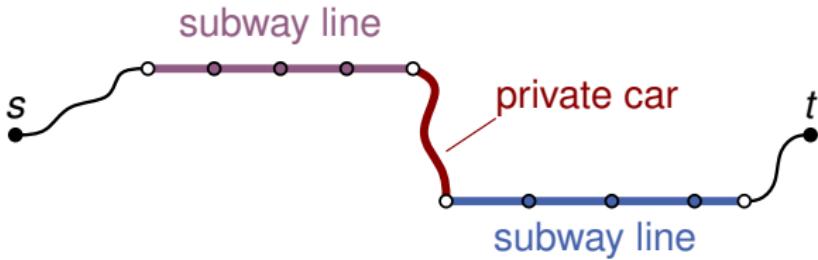


Undesired Modal Transfers

Problem: Unrestricted paths have arbitrary modal transfers.

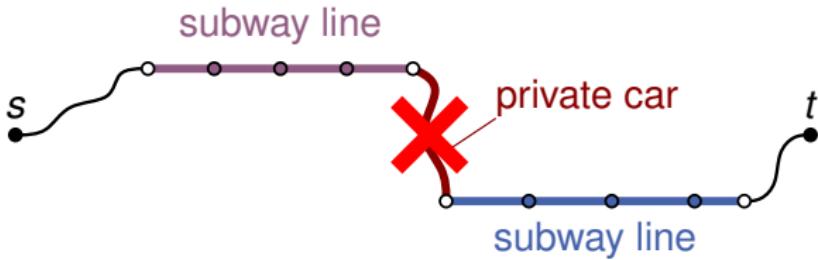
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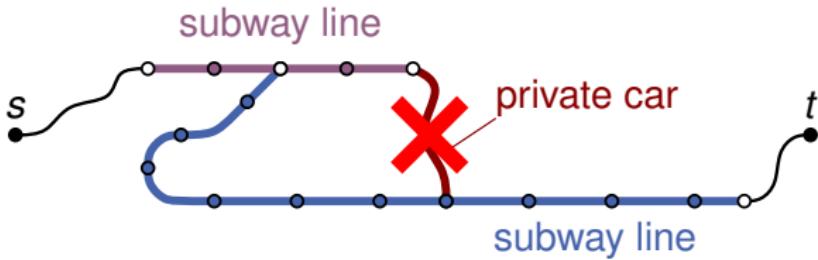
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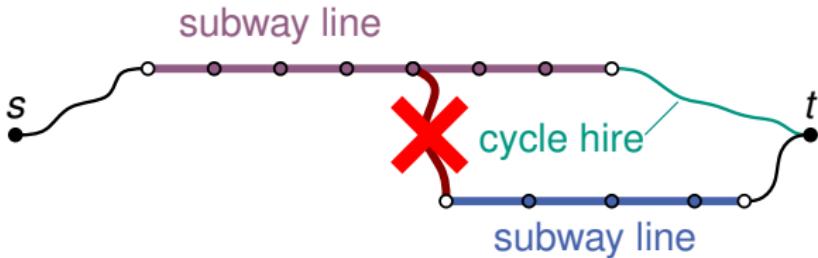
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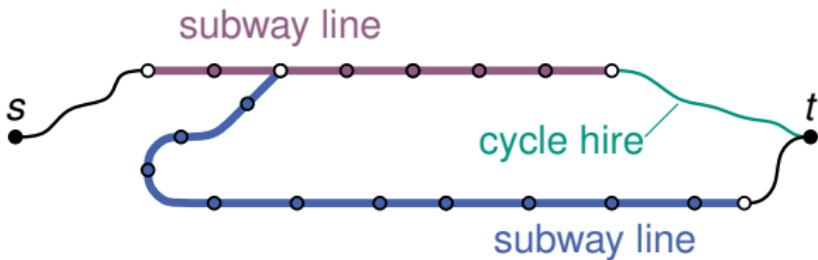
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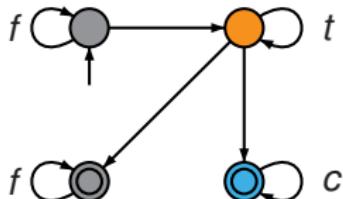


- Not all modal sequences feasible, and
- available/desired modes of transport depend on user.

Existing Solutions

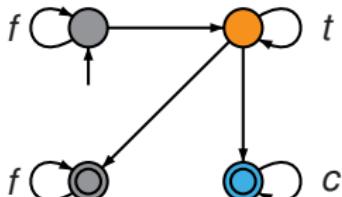
Label Constrained Shortest Path Problem (LCSPP)

- Define alphabet of transport modes.
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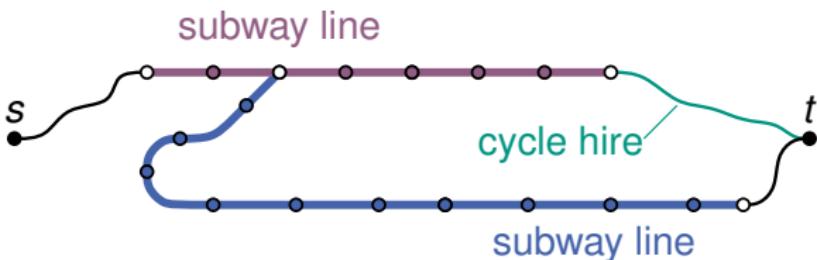
Algorithms for LCSPP

- Dijkstra on product graph with automaton works.
- Speedup techniques: Access-Node Routing, SDALT.
- Constraints as query input: UCCH.

Problem solved?

Existing Solutions

Shortcomings of LCSPP



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Goal: Compute *reasonable set* of multimodal journeys.

Approach

Idea: Compute *multicriteria* multimodal Pareto sets.

- Optimize arrival time plus
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Such as # transfers (public transit), walking duration, taxi cost, etc.

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Criteria: Arrival time, # transfers, walking duration. Sixty-nine solutions.

Known problem: Pareto set significantly grows with # criteria.

Definition (Domination, Pareto Set)

A journey J_1 *dominates* a journey J_2 iff. J_2 is worse (or equal) in all criteria.

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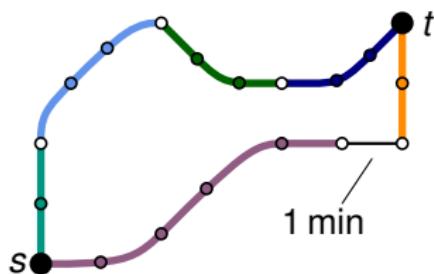
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Nondominated journey may have

- tiny improvement for criterion A,
- even if much worse in criterion B.

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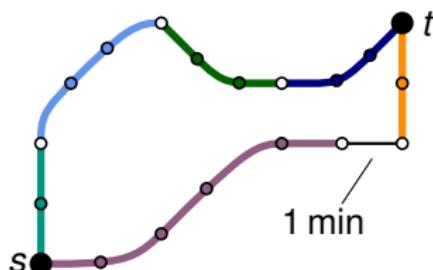
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How do we identify the *significant* solutions of a Pareto set?

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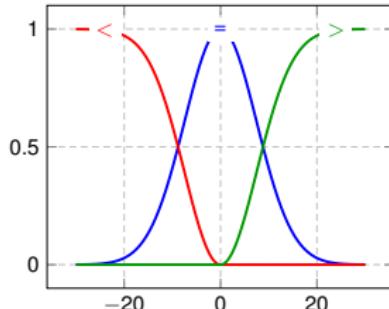
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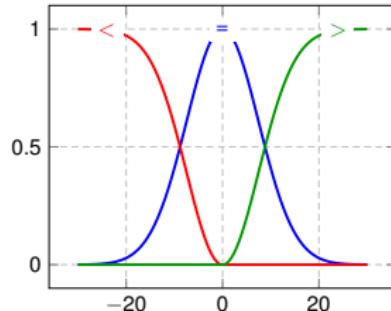
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Journey J_1 may, e. g., 90 %-dominate J_2 , even if J_2 has 1 min less walking.

Identifying Significant Journeys

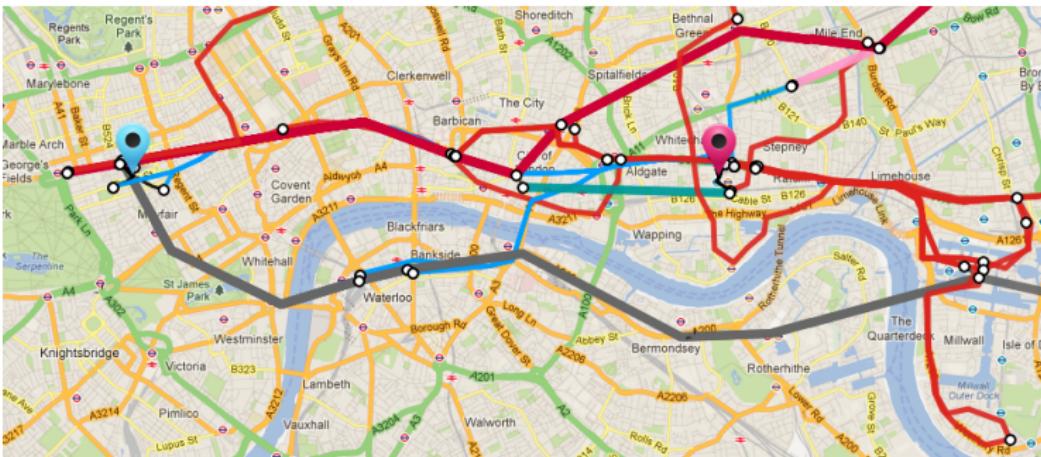
Approach:

- Compute full (exact) Pareto set by multicriteria algorithm.
- Score each journey J by $1 - \max(d(J_1, J), \dots, d(J_n, J))$.
- Then: Higher-scored journeys are more significant.

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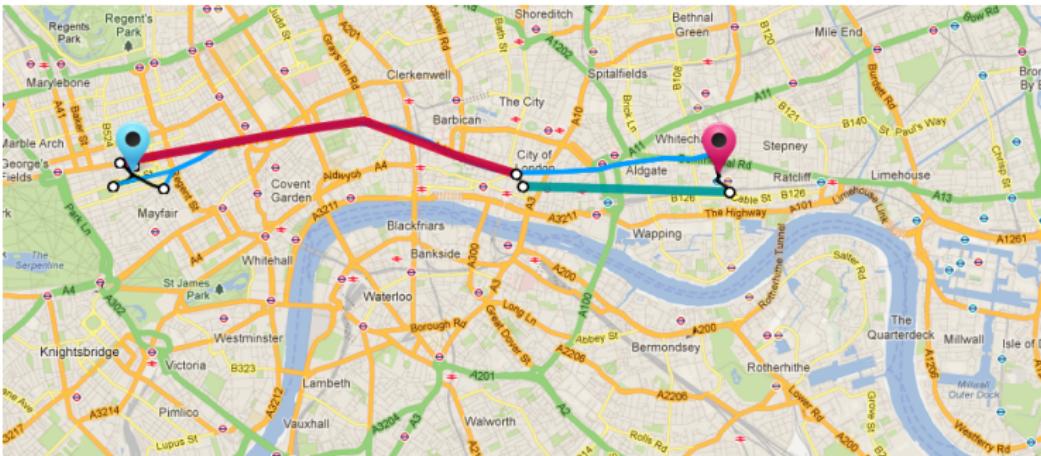
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Three highest-scored journeys.

Observations

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Round-based Framework

- Run one round per trip.
- Maintain Pareto set of labels for each location and round.
- In each round: Run subalgorithm per *transport mode*.
 - Public Transit: RAPTOR.
 - Walking, Taxi, Cycle: LC-Dijkstra & UCCH.
- ... read labels from round $i - 1$, write to round i .

Heuristic Approaches

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- MCR-hf: Fuzzy domination between labels during algorithm.
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- MCR-tx-ry: Max walking x between trips and y at source/target.

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Reduce Number of Criteria

- MR-x: Count every x minutes of walking as trip.

Measuring Quality of Heuristics

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Similarity of two Journeys

- Use fuzzy $=$ -operator on each criterion.
- Total similarity: minimum similarity over all criteria.

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Quality of Solution

- Consider P_{MCR} as *ground truth*.
- Keep only top- k journ. in $P_{\mathcal{A}}, P_{MCR}$.
- Compute (greedy) maximum matching between $P_{\mathcal{A}}$ and P_{MCR} with similarity.

$P_{MCR}: J_1 \quad J_2 \quad J_3 \quad \dots$



$P_{\mathcal{A}}: \quad J_1 \quad J_2 \quad J_3 \quad \dots$

Output average similarity of matched journeys.

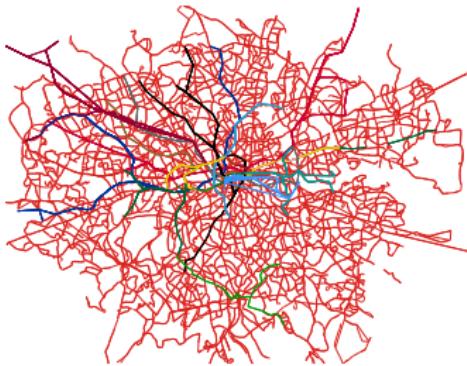
The London Instance

Complete Public Transit Network

- Contains Tube, Bus, DLR, Tram...
- 20 k stops, 2.2 k routes,
- over 5 M daily departures,
- and 564 cycle hire stations.

Road and Pedestrian Networks

- \approx 260 k vertices (27 k uncontracted)
- and 1.4 M edges, each.



Scenario without Taxi

Criteria: Arrival time, # transfers, walking duration.

Algorithm	# Rnd.	# Jn.	Time [ms]	Quality-6	
				Avg.	Sd.
MCR	13.8	29.1	1 438.7	100 %	0 %
MCR-hf	15.6	10.9	699.4	89 %	11 %
MCR-hb	10.2	9.0	456.7	91 %	10 %
MCR-t10-r15	10.7	13.2	885.0	30 %	31 %
MR-10	20.0	4.3	39.4	45 %	29 %

One core of Intel Xeon E5-2670, 2.6 GHz, 64 GiB DDR3-1600 RAM

Scenario with Taxi

Criteria: Arrival time, # transfers, walking duration, cost.

Algorithm	Wlk.	# Rnd.	# Jn.	Time [ms]	Quality-6 Avg.	Quality-6 Sd.
MCR	•	16.3	1 666.0	1 960 234.0	100 %	0 %
MCR-hf	•	17.1	35.2	6 451.6	92 %	6 %
MCR-hb	•	9.9	27.6	2 807.7	92 %	6 %
MCR-hb	○	9.0	11.6	996.4	74 %	12 %

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Wlk. ○: Combines walking duration into cost.

Conclusion

- Multicriteria approach to multimodal route planning.
- Optimizes arrival time plus convenience criteria.
- Fuzzy set theory helps identify significant solutions.
- Heuristics quickly find solutions of good quality,
- if convenience criteria not dropped.



Daniel Delling, Julian Dibbelt, Thomas Pajor, Dorothea Wagner, and Renato F. Werneck.

Computing Multimodal Journeys in Practice.

In *Proceedings of the 12th International Symposium on Experimental Algorithms (SEA'13)*, volume 7933 of *Lecture Notes in Computer Science*, pages 260–271. Springer, 2013.



Marco Farina and Paolo Amato.

A Fuzzy Definition of “Optimality” for Many-Criteria Optimization Problems.

IEEE Transactions on Systems, Man, and Cybernetics, Part A, 34(3):315–326, 2004.