

A Heuristic for Multi-modal Route Planning

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Introduction





A journey from A to B

Walk, Car, Bike, Public Transport



Introduction





A journey from A to B

Walk, Car, Bike, Public Transport

Carpooling? Carsharing? Bikesharing? Ride-hailing? Bus-on-demand? An arbitrary combination? Personalization?



Difficulties





Car runs on Street Network Every point is reachable



Public Transport runs between Stops Only certain «transfer nodes» are reachable





More Difficulties

- Richer user profiles lead to **computationally expensive** graphs and edge weights.
- Weight coefficients are not well-suited for representing hard restrictions.
- Traditionally, **all possible paths** taken into consideration by the routing algorithm.
- System adaption to new requirements difficult.





Transport Layers



Link / Map to nodes and areas (areas are important for, e.g., carpooling, because pickups can happen in an area)





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Difficulties



A large dynamic graph of heavily interconnected nodes





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Two Steps





Use dynamic «transfer graph» to compute possible multi-modal routes from A to B 2 Compute feasibility of trip legs using a conventional route planner for the respective mode





Rule Base for Transport Modes

Rules (cf. agent planning) O[condition] → M[condition] → D[condition]: [outcomes]

Example: Drive a Bike

A[user[bikeLocation] = A] → BIKE[context[weather] != "rain"] → B[bikeParking = true]: user[bikeLocation] = B, user[distBiked] += dist(A, B)



Algorithm





Algorithm





Algorithm

Directly

2



\Rightarrow

Algorithm

3



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Algorithm

3)



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Algorithm

3)

5



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Algorithm

3

5

6



Implemented Rules

Rules for Walking, Biking, Driving, using Public Transport, Carsharing, Carpooling, Bikesharing

Integrating personalized constraints, contextual factors, mode-dependent restrictions Table 1 A selection of rules implemented in our prototype system

Mode	Rule	Description
Walk	$\begin{array}{l} d_w = user[distWalked] + dist(A, B)\\ A[\emptyset] \rightarrow WALK[(d_w < user\\ [maxDist]) and (NOT context\\ [rainyWeather] OR (d_w < user\\ [maxDistRain])) and (context\\ [currentTime] IN user\\ [acceptableTimeIntvlWalk])]\\ \rightarrow B[\emptyset]:\\ user[distWalked += dist(A, B)],\\ context[time += time(A, B)]\end{array}$	Every node provides walking, however, a user can only walk up to a maximal distance (which gets decreased if it is raining), and if the current time is within an accepted time interval for walking. As a result of walking, the total walked distance is updated as well as the context
Bike	$\begin{array}{l} A[user[bikeLocation] = A] \rightarrow BIKE\\ [(NOT context[rainyWeather]) AND\\ (context[currentTime] IN user\\ [acceptableTimeIntvlWalk])] \rightarrow B\\ [bikeParking = true]: user\\ [bikeLocation] = B, user[distBiked]\\ += dist(A, B), context[time += time\\ (A, B)] \end{array}$	A user can only take the bike, if her bike currently is at the location. Further, the destination needs to have a bike parking spot available. Concerning contextual variables similar to walking
Car	$\begin{array}{l} A[user[carLocation] = A] \rightarrow CAR\\ [\phi] \rightarrow B[\#parkingSpots > 0]: user\\ [carLocation] = B, context[time\\ += time(A, B)] \end{array}$	Taking the car is only possible from the location where the user currently has parked her car to locations with a parking spot available. As a result, the car is at location B
Train	$\begin{array}{l} A[connectsLineX = true] \rightarrow TRAIN\\ [\phi] \rightarrow B[connectsLineX = true]:\\ [\phi], context[time += time(A, B)] \end{array}$	Similar to bus
Tram	$\begin{array}{l} A[connectsLineX = true] \rightarrow TRAM \\ [\phi] \rightarrow B[connectsLineX = true]: \\ [\phi], context[time += time(A, B)] \end{array}$	Similar to bus
Carshare	$\begin{array}{l} A[carSharing = true, \\ \#cars > 0] \rightarrow CARSHARE[\phi] \rightarrow B \\ [\#parkingSpots > 0]: A[\#cars \\ - = 1], context[time += time(A, B)] \end{array}$	Carsharing is possible from carsharing locations, where enough cars are available. The destination needs to have free parking spots
Carpool	$\begin{array}{l} A[intersects(A, \\ C) = true] \rightarrow CARPOOL[\phi] \rightarrow A \\ [intersects(B, D) = true]: [\phi] \end{array}$	Carpooling is possible from locations that intersect with a spatio-temporal corridor of a carpooler
Bike-Share	$\begin{array}{l} A[bikeSharing = true, \\ \#bikes > 0] \rightarrow BIKESHARE \\ [context[weather] ! = "rain"] \rightarrow B \\ [bikeParking = true]: A[\#bikes \\ - = 1], user[distBiked] += dist(A, \\ B), context[time += time(A, B)] \end{array}$	Bikesharing is possible from bikesharing locations, where enough bikes are available



Benefits

- 1. The heuristic only has to consider a graph consisting of transfer nodes (mostly public transport stops).
- 2. For validation of individual route segments, only the respective sub-graphs have to be queried.



Raw

Output



Example Outputs

ürich Affoltern Zürich Affoltern Seebacherstrasse EBACH Seebacherstrasse **CEEBACH** FFOLTERN. AFFOLTERN 17 17 Hagenholz Hagenholzst Byozmühlestrasse Biczmühlestrasse Zürich Oe ich Oerlikon Regensbergstrass Regensbergstrass Wallise Walliselle DERLIKON Sterbe aferbe, 4. U -U 0.0 U Toti-Strasse Tiechestrasse Ton-strasse Trèchestra se Wasser Hardturmstrasse PKINGEN W PKINGEN rmstrassr UNTERSTRASS Zürich UNTERSTRASS Zürich Wipkingen Wipkingen Germaniastrasse INDUSTE EQUARTIER Germaniastrasse INDUSTRIE JUARTIER ALL OBERSTRASS AIL OBERSTRASS AUS ERSIHL AUSSERSIHL FLUNT FLUNTER Zürich Selnau . Zürich Selnau . ALTSTADT 17 Ó ALTSTADT 17 c h h WIEDIKON WIEDIKON

Planned Route





Example Outputs



Normal Weather

Rainy Weather



Example Outputs



Afraid during the night



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Conclusions

We can bulid a dynamic graph «on the fly», and use it to compute multi-modal route options.

We can integrate personalized preferences, and quickly update the respective rules.

The heuristic can be expanded to include for example points of interest, as intermediate stops.



Questions

F1

Multi-Modal Routing