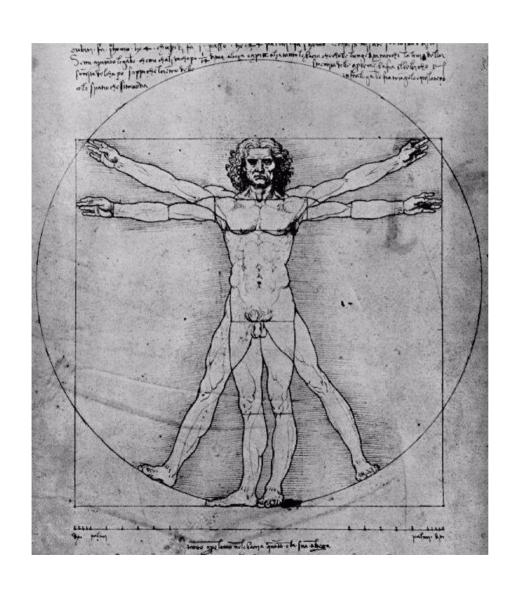
Smart Route Planning for Public Transit

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Theory is when you know something, but it doesn't work.



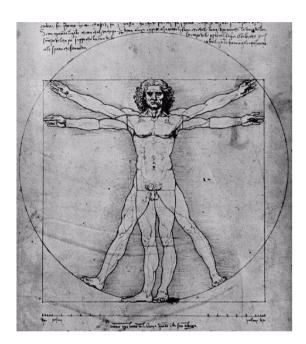
The real world out there...



Practice is when something works, but you don't know why.

Bridging the Gap between Theory and Practice?





Theory is when you know something, but it doesn't work.

Big challenge: combine theory and practice...



...i.e., nothing works and you don't know why.

Practice is when something works, but you don't know why.

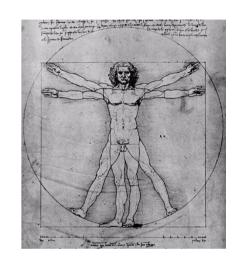


Public Transit Route Planners



Vast majority based on timetabling

Theory: all transit vehicles start their trip at the planned time, no delays throughout their journey



Practice: buses often run behind schedule, and for many unplanned reasons, such as:

- traffic
- road closures
- inclement weather
- sometimes even unrealistic scheduling



Outline



- 1. Are timetabled route planners good enough for public transit?
- 2. Investigate impact of GPS data: run same algorithm with timetabled and GPS data
- 3. Design of GPS-aware algorithms

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- 1. Are timetabled route planners good enough for public transit?
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[Firmani, I., Laura, Santaroni, ATMOS 2013]



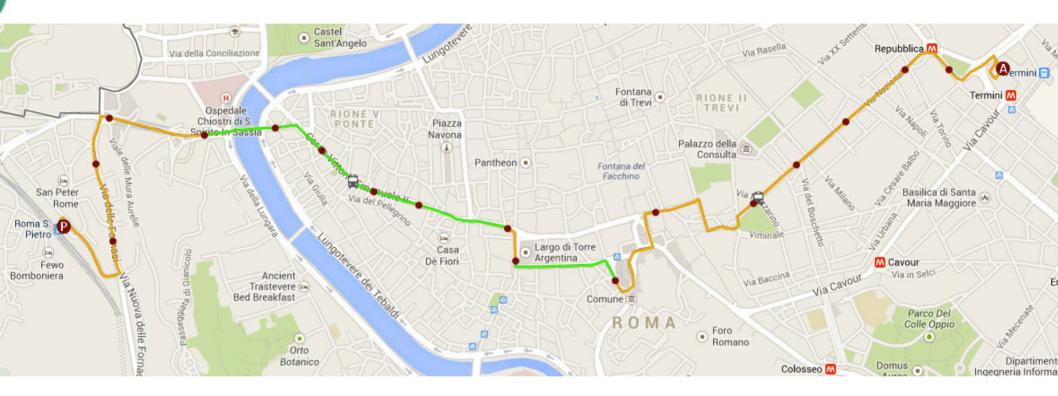
Timetabled Route Planners

 Are timetabling-based routing methods able to deliver "good" solutions in practice?

 Try to measure the quality of timetablingbased solutions in the metropolitan area of a big city.



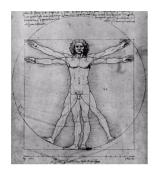


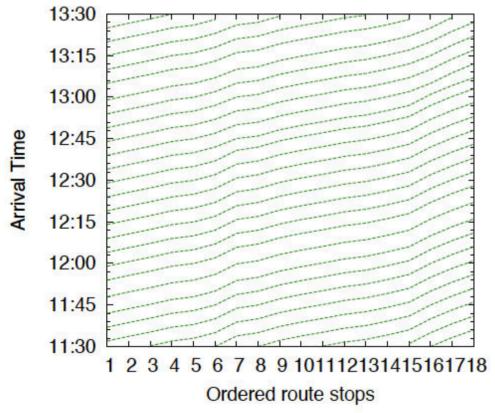


Bus 64, route between S. Pietro to main train station in Rome (Termini)

Connects many important touristic attractions in the city center of Rome



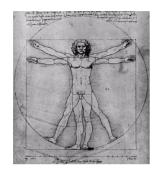




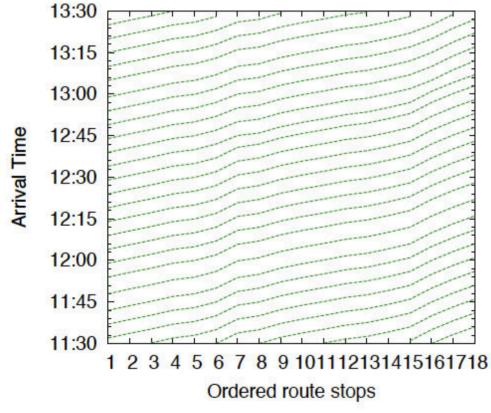
(a) Official timetable (11:30am-1:30pm)

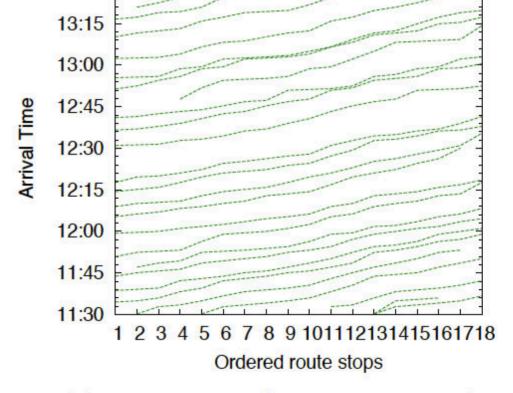


13:30





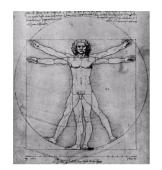




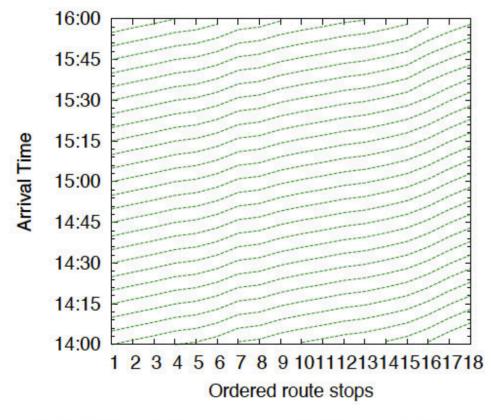
(a) Official timetable (11:30am-1:30pm)

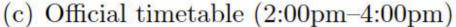
(b) GPS stream (11:30am-1:30pm)

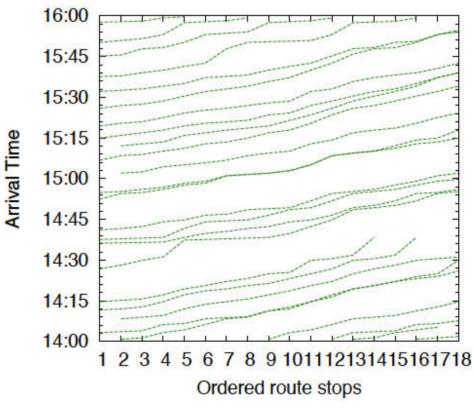






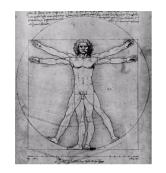




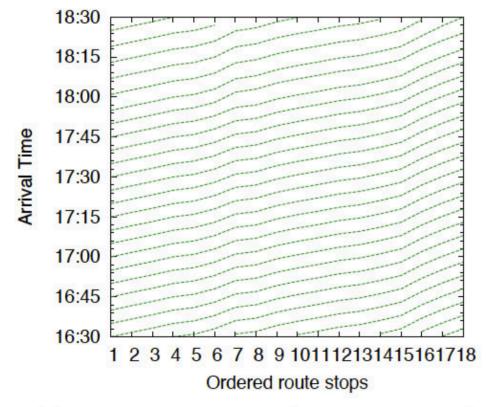


(d) GPS stream (2:00pm-4:00pm)

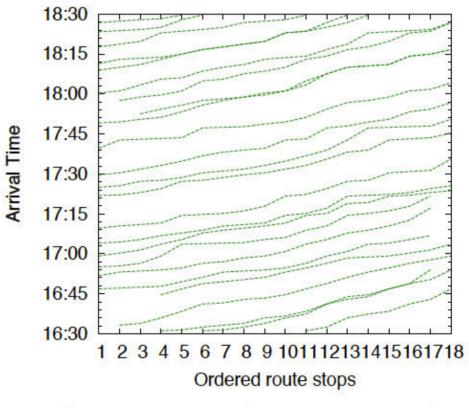












(f) GPS stream (4:30pm-6:30pm)



Experiment (1/2)

How does this affect a route planner?

Submitted queries to Google Transit

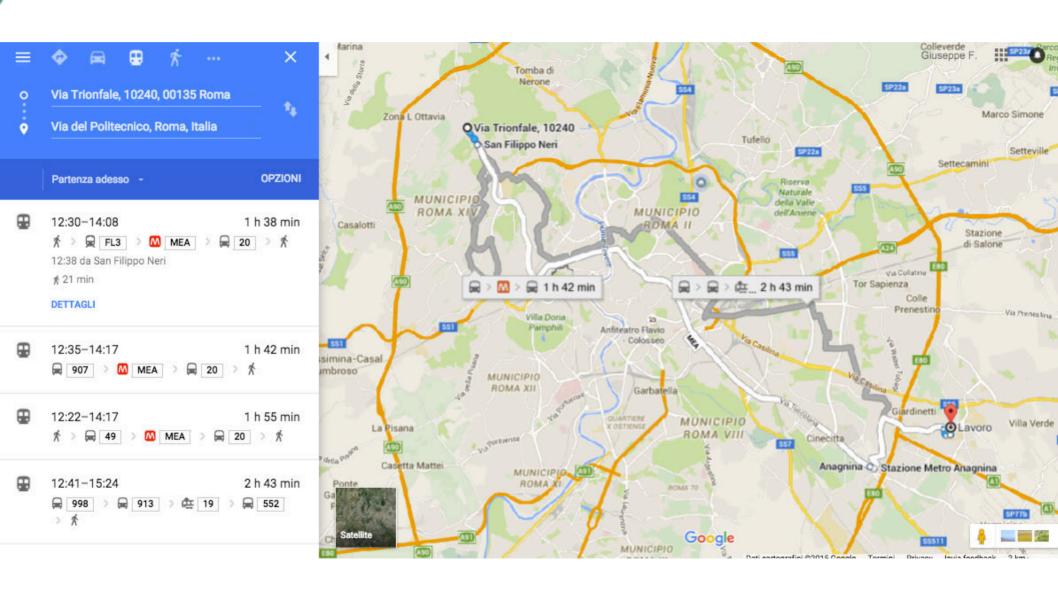
- Origin and destination in metropolitan area of Rome
- Each query has a departure time from origin

Collected suggested journeys and their estimated travel times

 Journeys provided by Google Transit based on timetable data



Google Transit





Experiment (2/2)

"Follow" each journey through GPS data

- GPS data → real-time position of transit vehicles (provided by the very same transit agency)
- Simulate the experience of a user traveling according to a given journey, after leaving the origin at the corresponding time

Compare actual travel times against their original estimates

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Experimental Setup

- Public Transport Network in Rome
 - 309 bus lines and 2 subway lines
 - 7,089 stops (7,037 bus stops and 52 subway stops)
- More than 1,000 queries $q_i = \langle s_i, t_i, \tau_i \rangle$
 - s_i t_i uniformly at random in Rome's metropolitan area
 - τ_i uniformly at random in [7:00am; 9:00pm]
 - On Wed January 8, 2014 (just a typical week day, no major disruptions, constructions works or extreme weather conditions)
- Collected 4,018 journeys
 - For each query, Google Transit returns 4 journeys
 - Possibly less: e.g, one of the journeys might be footpath



Our Simulation

On same day (January 8) submitted queries every minute to the Mobility Agency of Rome

 Obtained (from GPS data) the instantaneous geolocation of all transit vehicles in the network

Simulation

- For each trip in the journey, pick-up and drop-off times computed according to the position of transit vehicles
- If a delayed transit vehicle misses a connection then the next trip of the same line was chosen (similarly, could catch earlier connections)
- Footpath times computed with Google Maps
- No GPS data for subway lines (two lines, usually on schedule)



Experimental Results

For each journey j, compare $t_a(j)$ and $t_e(j)$

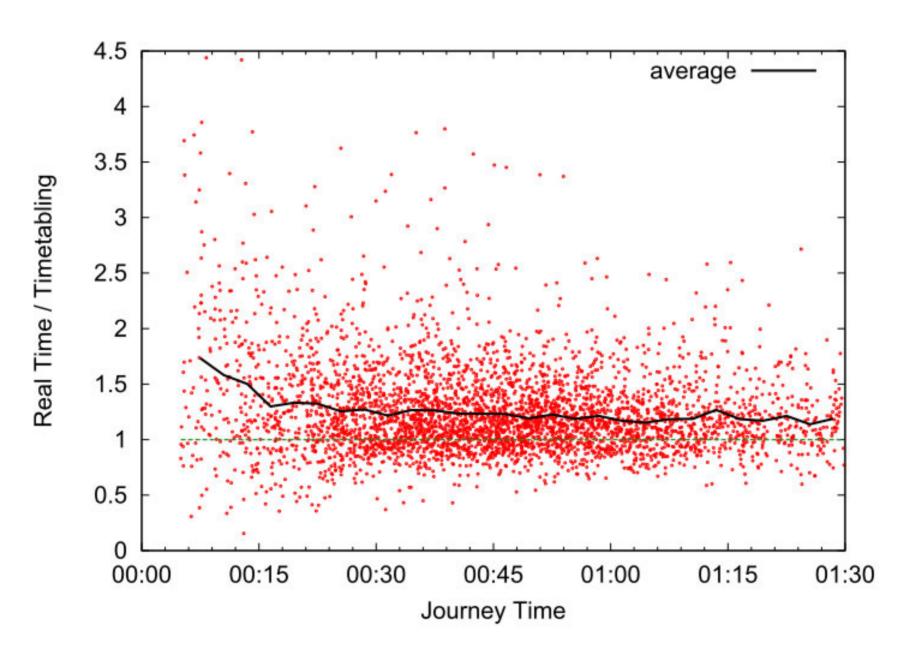
- $t_e(j)$: estimated (planned) time (timetable \leftarrow Google Transit)
- $t_a(j)$: actual time (vehicle real-time positions \leftarrow simulator)

Define error coefficient $t_a(j)/t_e(j)$

- Plot $t_a(j)/t_e(j)$ as $t_e(j)$ increases:
 - **short distance** journeys : $t_e(j) \le 30$ min
 - **medium distance** journeys : $30 \text{min} < t_e(j) \le 60 \text{min}$
 - **long distance** journeys : $t_e(j) > 60$ min
- Group journeys into **time slots** within a 3-minute resolution and plot statistics as $t_e(j)$ increases

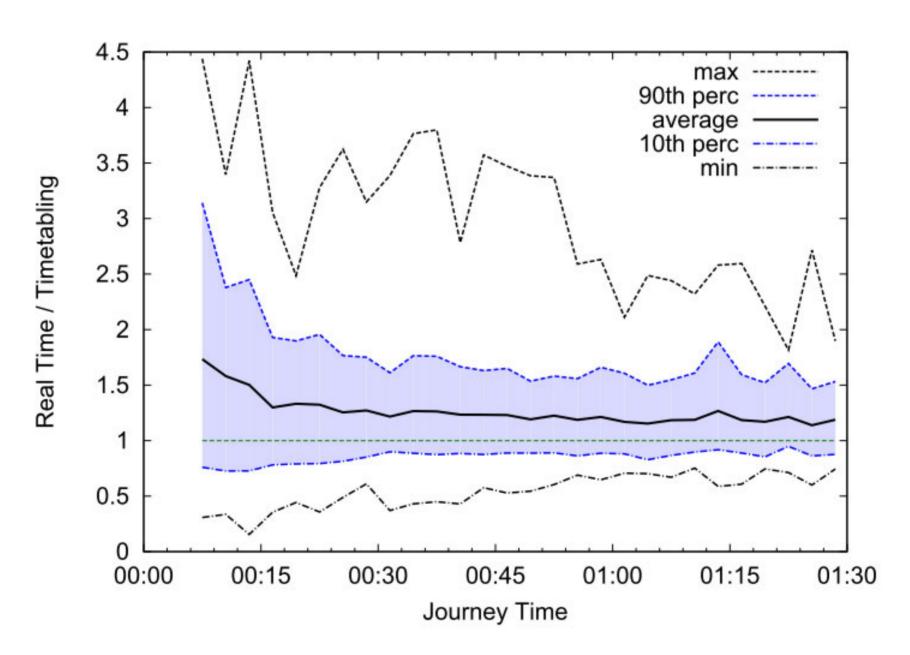


Error Coefficient (1/2)



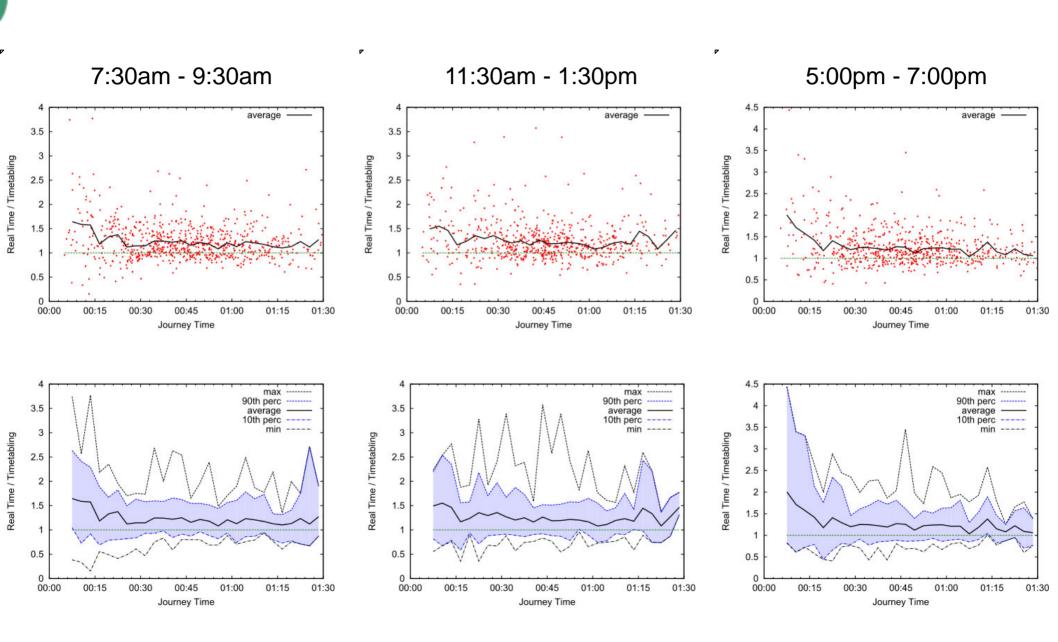


Error Coefficient (2/2)





Different times of day





Correlations in ranking

For each query q_i , take four journeys suggested and compare:

- Relative rankings by **estimated** times $t_e(j)$
- Relative rankings by actual times t_a(j)

Use Kendall tau coefficient

```
(# concordant pairs) - (# discordant pairs)

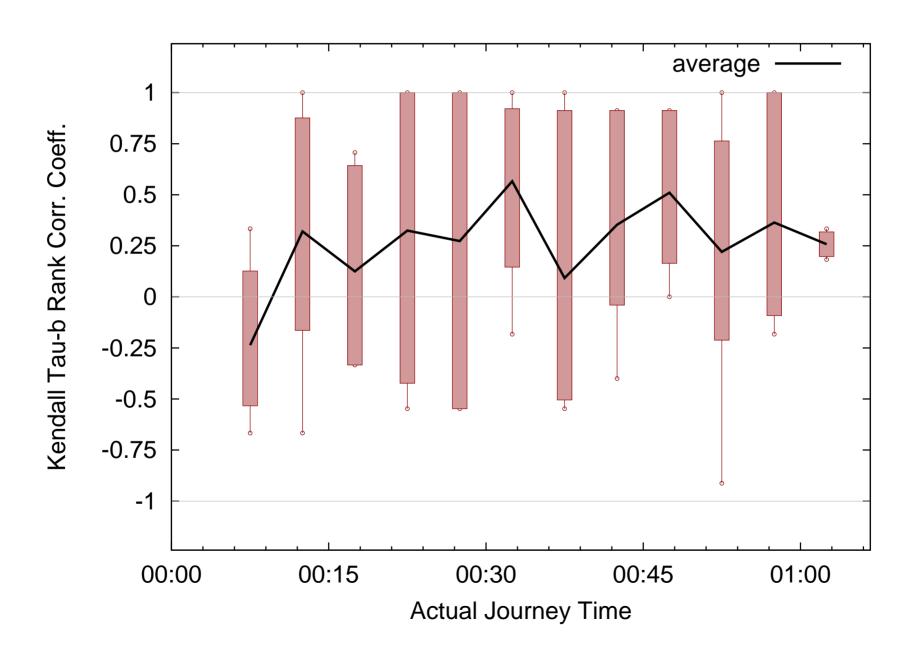
½ n (n-1)
```

- From -1 (perfect disagreement) to +1 (perfect agreement)
- 0 if independent (uncorrelation)

Actually used Kendall tau b (to adjust for ties)



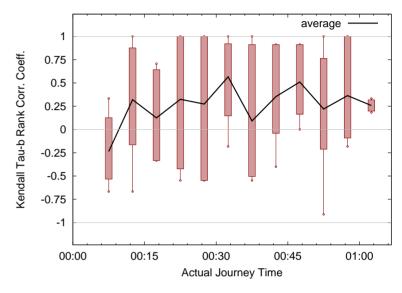
Kendall Tau Coefficient





Kendall Tau Coefficient

Only very mild correlation between *estimated* and *actual* duration of a journey.



Smaller Kendall values for short journeys: they appear more vulnerable to fluctuations in the schedule (thus larger errors on time estimates provided by timetabled routing).

Overall, strong evidence that timetabling fails to deliver good solutions in this scenario















ROMA 2024











Is this just for Rome?

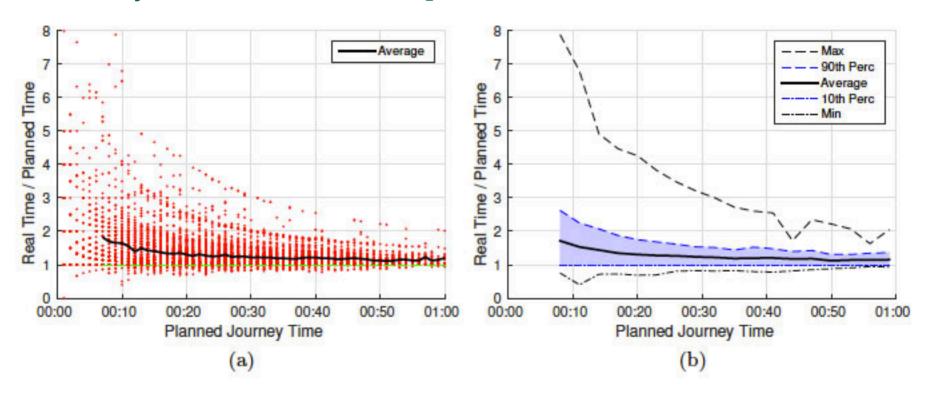
Verbatim from [Böhmová, Mihalák, Neubert, Pröger, Widmayer, ATMOS 2015]:

"Firmani et al. [9] observed in an experimental study on the transportation network of Rome that the timetable information and the real movement of the vehicles (based on GPS data) are only mildly correlated."



Is this just related to Rome?

Verbatim from [Böhmová, Mihalák, Neubert, Pröger, Widmayer, ATMOS 2015]:



"In overall, we observed that the behaviour in Zürich is comparable to the one in Rome."



Takehome message

In several cases, timetable information and real movement of vehicles are only mildly correlated: it's not only a small input perturbation (not only in Italy but also in Switzerland!)

Thus, widely used timetable routing algorithms suffer from many inaccuracies, independently of their own merits. Due to incorrect estimations of the waiting/transfer times at transit stops, might fail to deliver a good solution.

Need smarter data: GPS feeds of transit vehicles?

Outline



- 1. Are timetabled route planners good enough for public transit?
- 2. Investigate impact of GPS data: run same algorithm with timetabled and GPS data
- 3. Design of GPS-aware algorithms

[Delling, I., Pajor, Santaroni, IWCTS@SIGSPATIAL 2014]



Routing Algorithms

Few algorithms that work with timetabled data could be adapted to work with GPS data

- RAPTOR [Delling, Pajor, Werneck.
 Transportation Science, 2014].
- CSA [Dibbelt, Pajor, Strasser, Wagner. SEA 2013]

• . . .

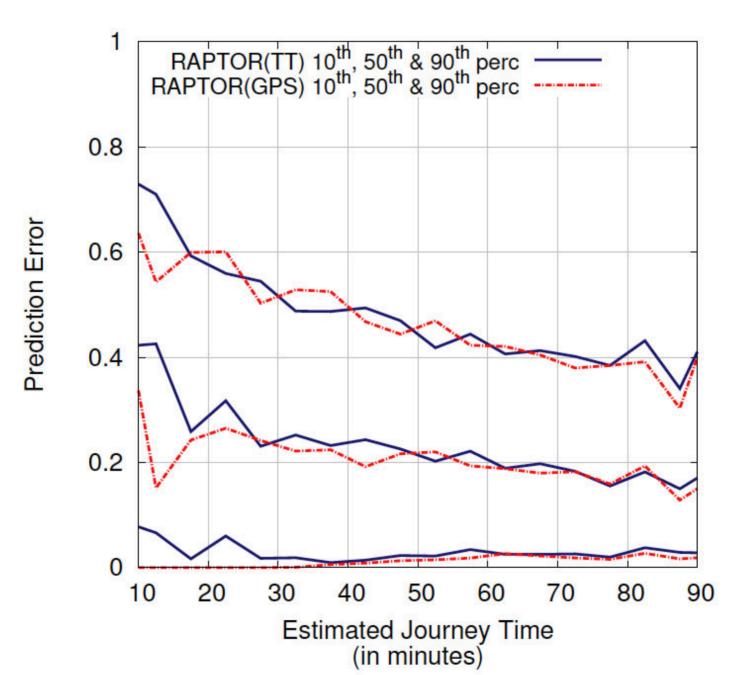


RAPTOR

- Efficiently computes Pareto sets of journeys optimizing arrival time and number of transfers.
- It works in rounds, one per trip, computing in round i earliest arrival times for all stops that can be reached with i trips (i – 1 transfers).
- Our adaptation accepts GPS feeds and timetabled data



Relative Prediction Error





Quality of solutions?

Defining the "best journey" is a very complicated issue

Journey planners try to optimize multiple cost criteria...

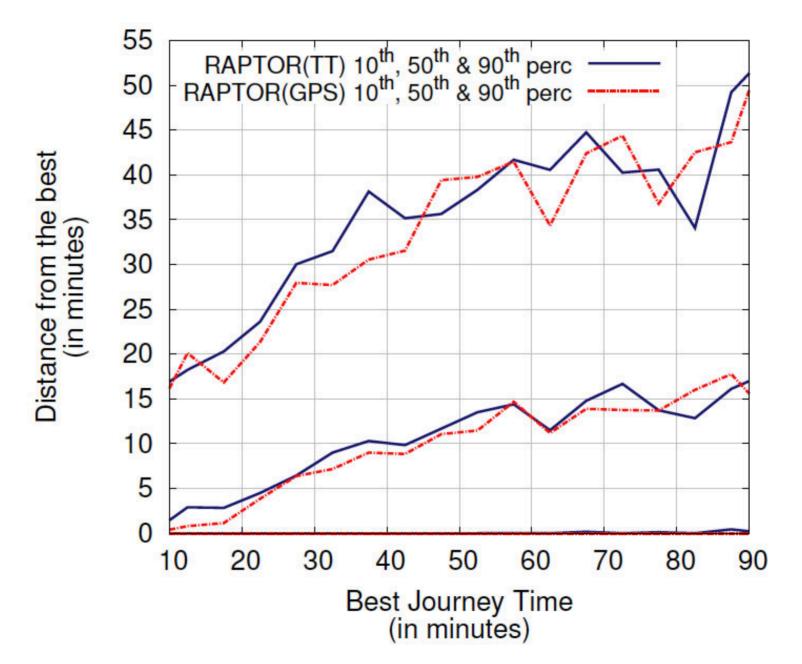
With all the data available we computed a rough proxy: fastest journey (earliest arrival time)

That's our "best journey"

Measured "distance from the best" (slack)

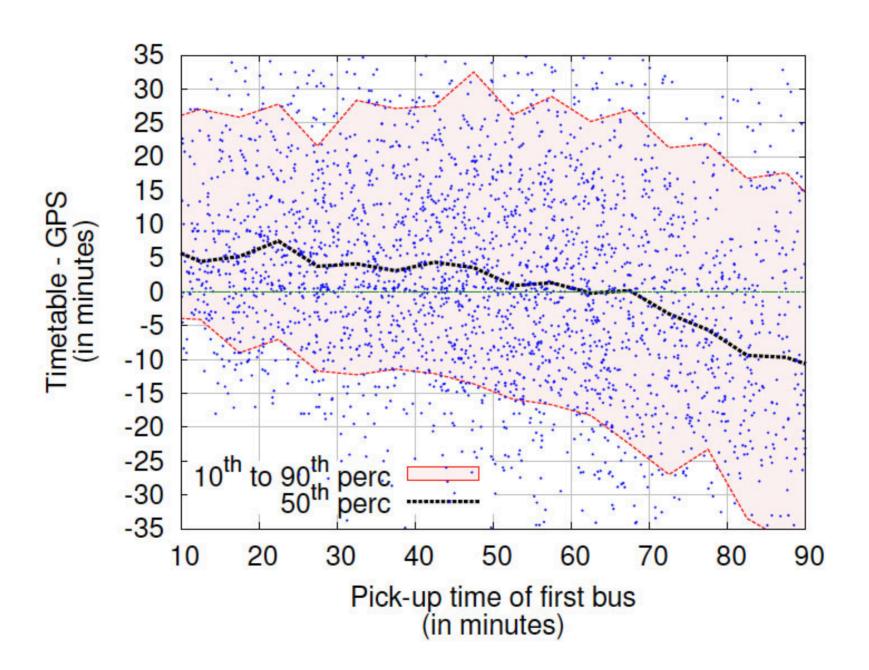
Distance from the best (Slack)





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GPS data useful in short term





Takehome message

Effective exploitation of GPS data seems important to make public transport journey planners smarter.

GPS data helps in short and medium range. Not very useful in the long range.

Main impact of GPS data depends on pick-up time of first bus

Smart data not enough: need smart (i.e., GPS-aware) algorithms

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[Allulli, I., Santaroni, SEA 2014]





http://muovi.roma.it/

Journey planner available for the metropolitan area of Rome

Uses GPS data and historical data to improve the accuracy of time estimates and to predict position of buses in the near future

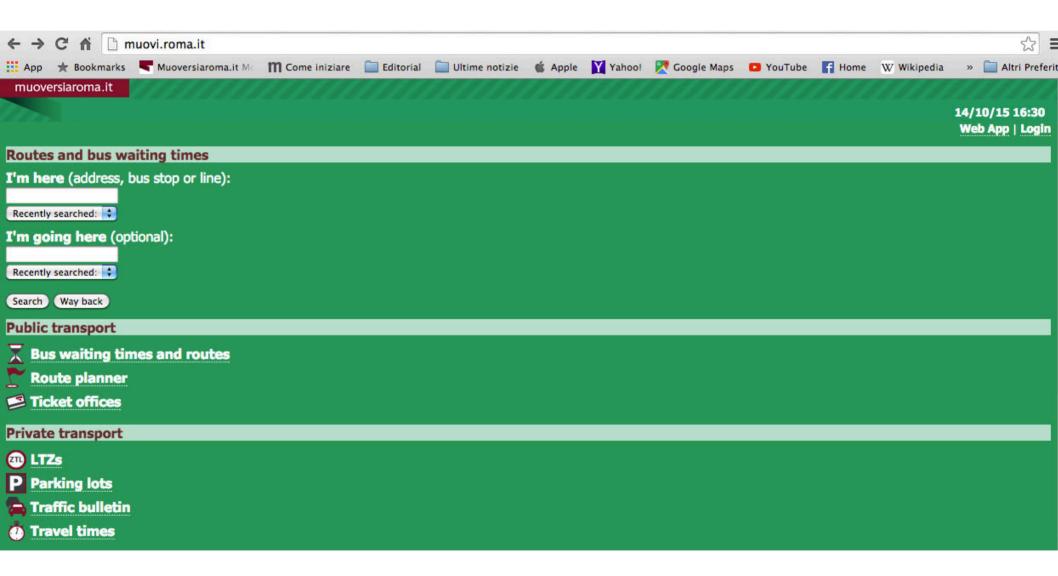
Based on dynamic shortest paths on a time-dependent network

Incorporates some of our own algorithmic ideas

Developed and maintained by the Mobility Agency of Rome (open source / open data)



Muovi Roma



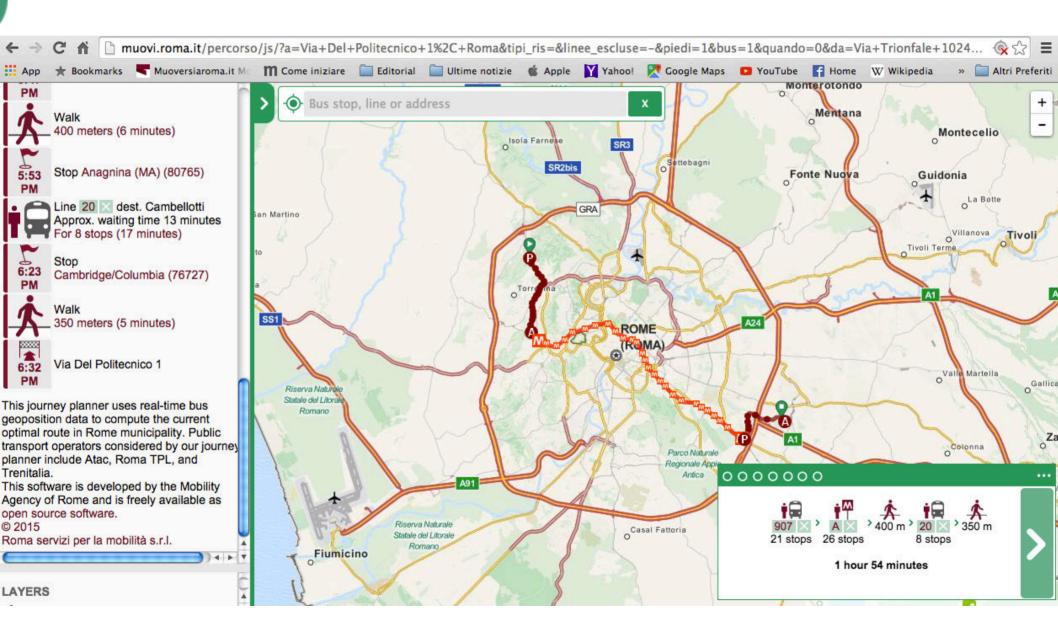


Muovi Roma



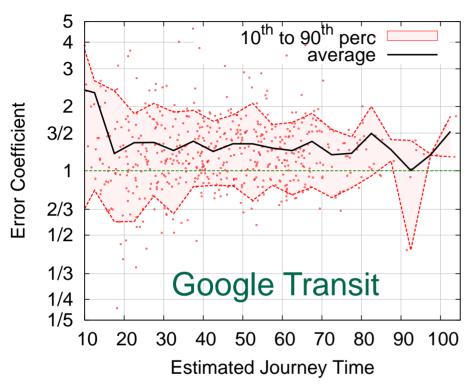


Muovi Roma

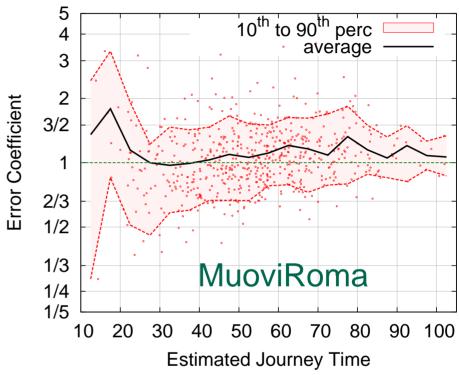




Error Coefficient



useful in a hour from now?)



Error Coefficient = Actual Time / Estimated Time

Short journeys affected more (MR estimates fewer short journeys)

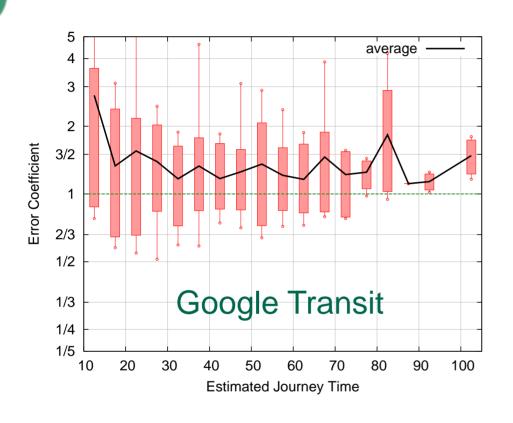
Both err on the optimistic side (actual journeys take longer on avg)

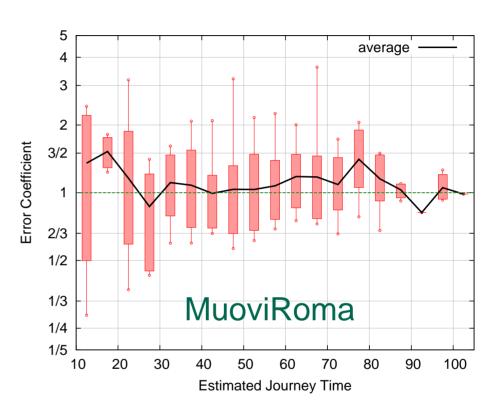
MR much better for medium range journeys

MR slightly better for long range journeys (GPS data not very



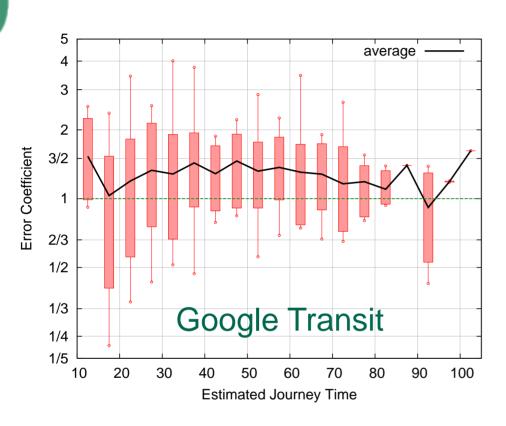
Morning (7:00am-1:30pm)

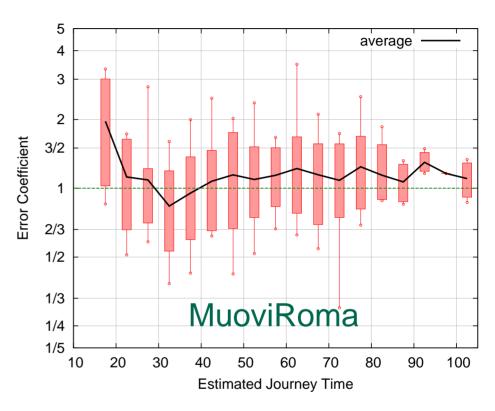






Afternoon (1:30pm-8:00pm)







Quality of solutions

From previous experiments can only conclude that GPS-aware planner (MR) estimates its journeys better than what timetabled planner (GT) does

Accuracy does not say much about quality of solutions provided (fast/best journeys)

MR could estimate better its journeys but could still provide low-quality solutions (slow journeys)



Quality of solutions

Again, computed best journey for query q

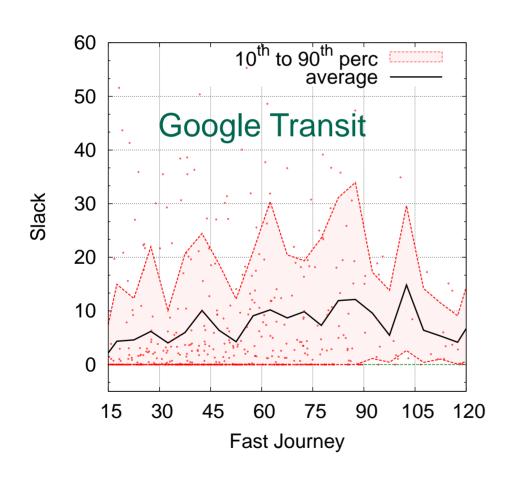
Distance from best (Slack) is distance (actual time) from best journey

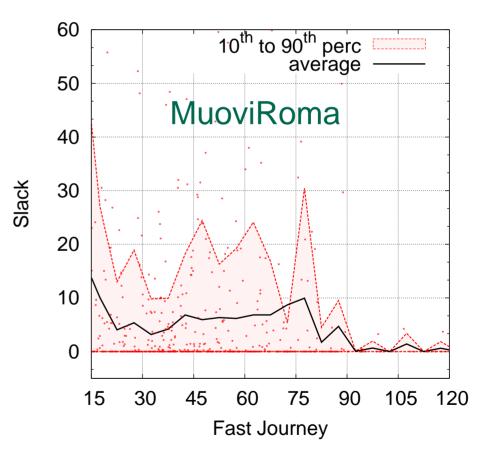
Best journey has slack equal to 0

The smaller the slacks, the better the solutions provided by the route planner



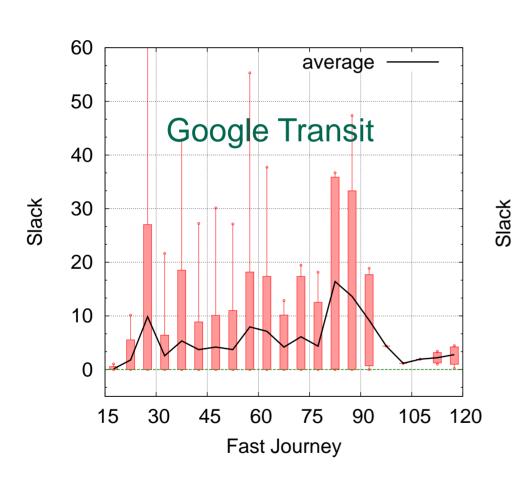
Quality of solutions (slack)

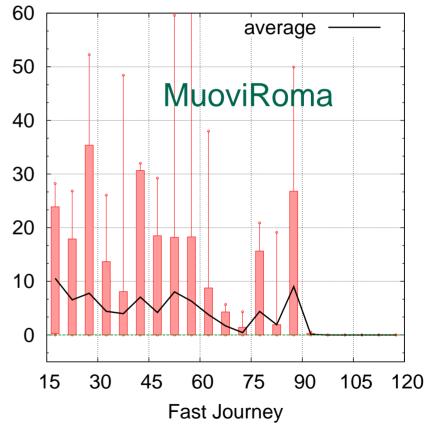






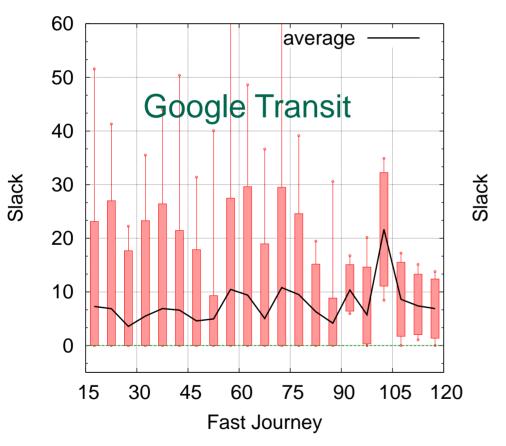
Slacks - Mornings

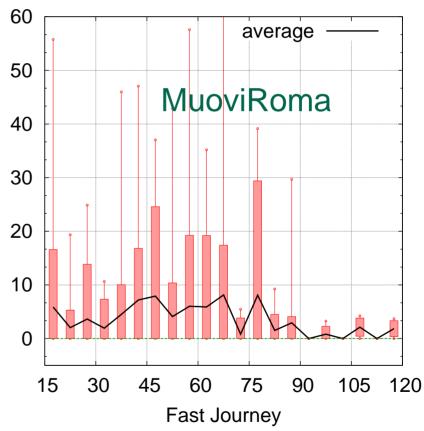






Slacks - Afternoons







Quality of solutions

GPS-aware (MR) able to achieve small slacks and fast journeys, especially for long-range journeys (>60')

How could GPS data be still useful in a hour from now?

Prediction on historical data seems to works well

Long-range journeys likely to involve trips on (less frequent) suburban lines, where missing a connection will cause significant delays

Long-range journeys based on unreliable timetables likely to incur in discrepancies with higher impact on overall travel time.



Conclusions

- Tried to assess empirically impact of GPS data in real public transport network (Rome)
- With significant fluctuations in schedule, timetabling inherently affected. GPS data able to provide better accuracy
- "Traditional" algorithms do not seem to be able to fully exploit GPS data
- We have smart data. Need to work more on smarter algorithms!
- Prediction models are critical issue (what will be the state of the network in 60 minutes?)

Future Work







LOTS OF APPS LET YOU
PLAN YOUR TRIPS USING
REAL-TIME BUS, TRAIN,
AND TRAFFIC DATA.
THEY TRY TO PREDICT WHICH
ROUTE WILL BE FASTER,
BUT AREN'T ALWAYS RIGHT.



INSTEAD OF JUST PLANWING,
MY NEW APP LETS YOU SEND
"GHOST" VERSIONS OF YOU
ALONG DIFFERENT ROUTES,
SIMULATING THEIR TRAVEL
USING THE REAL-TIME DATA



THAT WAY, YOU CAN SEE
WHICH ROUTE TURNED OUT
TO BE FASTER IN PRACTICE.
YOU CAN ALSO RACE
YOUR PAST SELVES.



