# Algorithm Engineering of Timetable Information 

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## How To Travel?



## Public Transport =

## Schedule-Based Travelling



## Public Transport =

## Schedule-Based Travelling



- My own experience based on cooperation with Deutsche Bahn AG,
- so most of my examples will be about railways
- but application area covers all of public transport


## From Printed Schedule Books to Fully

 Electronic Timetable Information(10. AMTLICHES


## Timetable Information Systems: The Classical Use

## Pre-trip information and selling:

- Selling of tickets without a timetable information system is nowadays impossible.
- Used by all distribution channels: at ticket counter, by ticket vending machines, by travel agencies, for Internet ticket sales.


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- Used by all distribution channels:
at ticket counter, by ticket vending machines, by travel agencies, for Internet ticket sales.
- Quality of recommended connection is crucial:

Connections which the search engine does not offer will not be sold.
Examples:

- special fares, campaigns
- extra trains
- night trains
- trains with seat reservation


## Usage of Timetable Information Systems

## Commercial state-of-the-art:

- (train) timetable information in Europe HAFAS: computes more than 60 million connections per day [www.hacon.de/hafas]
- $\rightarrow$ high relevance
- very fast, but
- only heuristic solutions
often suboptimal, even with respect to travel time


## Algorithmic Engineering enters ...

The driving question in 1999 was:
Can we do exact timetable information? And can we do it efficiently enough?

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Seminal paper by Frank Schulz, Dorothea Wagner, and Karsten Weihe, WAE 1999:
"Dijkstra's Algorithm On-Line: An Empirical Case Study from Public Railroad Transport."

## Event-Activity Networks Time-Expanded Network Model

Small excerpt at some station:


## Properties:

- arrival/departure events correspond to vertices
- feasible connections correspond to directed paths


## Earliest Arrival Problem

## Earliest arrival problem

Input: source station, destination station, start time, event-activity network
Task: Find the connection with earliest arrival time at destination
In other words:

- have to solve an $s$ - $t$-shortest path problem in acyclic digraph
- solvable in linear time $O(m+n)$ ( $m$ number of arcs, $n$ number of vertices)


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- solvable in linear time $O(m+n)$ ( $m$ number of arcs, $n$ number of vertices)
- but graphs are fairly large (several millions of vertices)
- we definitely need sublinear algorithms
- $\rightarrow$ development of speed-up techniques


## Success Story of Algorithm Engineering: Classical s-t Shortest Paths

## Road networks:

- millions of nodes (continental size)
- query times of few microseconds achievable
- many new techniques: overlay graphs, contraction hierarchies, shortcuts and arc-flags, transit node routing, hub labeling, ...


## Timetable information in public transport:

- considerable speed-ups, although much less effective
- but achievable milliseconds are fine from a practical point of view


## Example

Query: from Halle (Saale) Hbf to Stuttgart Hbf at 08:00 on Wednesday September 12, 2012

Earliest arrival connection:

| Halle (Saale) Hbf | departure | $8: 00$ |
| :--- | :--- | ---: |
| Naumburg (Saale) Hbf | arrival | $8: 29$ |
| Naumburg (Saale) Hbf | departure | $8: 35$ |
| Fulda | arrival | $10: 42$ |
| Fulda | departure | $10: 47$ |
| Stuttgart Hbf | arrival | $13: 08$ |

Earliest arrival time: 5h 08 minutes

## Example: Real Query



## Example: Real Query



## Example: Real Query



## What is Wrong?

- Too simple model:
- start time interval
- further search attributes
- missing footpaths
- too space-consuming model
- only single-criterion search
- no alternatives
- no delays


## Overview

(1) Realistic models
(2) Multi-criteria search
(3) Realtime information
(1) Reliable connections (robustness, stochastic forecasts)
(5) Extensions

- train disposition
- multi-modal search


## Towards More Realistic Models

## Start time intervals:

- easy extension (standard trick in network flows)
- just add an artificial source vertex $s$
- add arcs, connecting the source with all departure vertices within the departure interval
- give these arcs zero length
- do shortest path computation from $s$
- Warning:
some subtleties in combination with multi-criteria search!


## Towards More Realistic Models

## Further search attributes:

- bicycle transportation, on-board restaurant, wheel chair access, seat reservation possible, ...
- again an easy extension possible
- equip each travel arc with a bit vector of the attributes it serves
- during search simply ignore all arcs which don't have the required attributes


## Alternative Graph Models

- time-expanded graph model uses one vertex per event

Can we do better if we use only one vertex per station (stop, airport)?

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Can we do better if we use only one vertex per station (stop, airport)?
(Simple) Time-Dependent Graph Model:

- every vertex represents a station
- two vertices are connected by an arc if the corresponding stations are connected by a direct non-stop connection
- lengths on the arcs are determined "on-the-fly": the length of an arc depends on the time in which the particular arc will be used


## Time-Dependent Graph Model

Small excerpt:


- different node types:
train route nodes (Rx), station node (S), foot-node (F)
- train edges have variable length
(depending on the moment of time when they are used)


## Brief Discussion of Models

## Time-expanded model:

- pro: easier to extend to realistic setting
- contra: higher memory + slower performance


## Time-dependent model:

- pro: smaller memory + better performance
- contra: fairly complicated extension to multi-criteria

For more details see survey by M.-H., Schulz, Wagner, Zaroliagis 2007.

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## Multi-Criteria Search

## Typical criteria:

- travel time
- number of transfers
- fare
- reliability (maximize minimum buffer time)


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## Typical criteria:

- travel time
- number of transfers
- fare
- reliability (maximize minimum buffer time)


## Difficulties with advanced criteria:

- non-linear or non-additive (for example, fares)
- expensive to evaluate black-box optimization (for example, fares)
- strong negative correlation (for example, travel time vs. reliability)


## Fare Zones in London

## Fare-optimal paths:

- bit-vectors indicate which zones are used
- apply dominance rules during search [DPW12]

Tube map


## Pareto-Optimality

- Standard notion of dominance:

Definition: A dominates $B$ if $A$ is at least as good as $B$ in all criteria and strictly better in at least one.

- Pareto-optimal solutions are the non-dominated ones.
- Basic algorithmic problem with several objectives:

Find all Pareto-optimal solution values.
or

> Find all Pareto-optimal paths.

## Size of Pareto Set

- In general, the Pareto set can be exponentially large.
- In practice, depending on the kind of criteria, it is often manageable [M.-H., Weihe 2006].
- Some Pareto optima irrelevant for practical purposes.
- But what is an appropriate size?

Who wants to see 17 alternatives?

## Attractive Alternatives

- Pareto-optimality excludes "near-optimal" paths which often are very reasonable and attractive alternatives.
- Consequence: we need some kind of relaxed Pareto-dominance. [M.-H., Schnee 2004]
- Further consequence: using relaxed Pareto-dominance restricts the set of applicable speed-up techniques.


## Multi-Criteria Dijkstra Algorithm

- generalization of Dijkstra's algorithm to several criteria
- uses multi-dimensional labels (representing partial solutions)
- efficiency depends on effective dominance rules



## Multi-Criteria Dijkstra Algorithm

## Remarks:

- Can also be adapted to time-dependent network model
- Main difficulty: need to ensure "substructure optimality"
- otherwise, domination goes wrong!
- See [Annabell Berger, M.-H. 2008]


## A Dilemma

- preprocessing tries to compute additional information (at most linear extra memory) which can be used to reduce the search space for queries
- most speed-up technique work only well if we look for a single optimal solution
- the more aggressive the speed-up technique is able to reduce the search space, the less likely it is that it can find good alternatives


## Techniques that do not work as well as expected ...

We tested several preprocessing techniques in a multi-criteria setting for time-dependent networks: [BDGM09]

- short-cuts
- arc-flags
$\rightarrow$ they yield only comparably small speed-ups


## Transfer Patterns (Bast et al., ESA 2010)

## What is special in public transportation networks?

- Even when you take a very long trip, the number of transfers is almost always a very small number
- And more than that, for a given source and destination, there is only a very limited number of "patterns" where it makes sense to transfer
- "Pattern" = sequence of stations where a transfer occurs


## Transfer Patterns (Bast et al., ESA 2010)

Idea: for each pair of stations, precompute all transfer patterns of all optimal paths (at all times) and store them

Query algorithm: do a time-dependent Dijkstra computation on this so-called query graph, where each arc evaluation is again a shortest path query, but restricted to direct connections

Such direct-connection queries are easy

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## Remarks:

- Preprocessing is quite expensive, but doable
- successful tests with networks in Switzerland, New York area, and a large part of North America
- (You may try yourself: http://www.google.com/transit)


## Round-Based Search

## RAPTOR - Round-bAsed Public Transit Optimized Router (Delling, Pajor, Werneck, ALENEX 2012)

## Idea:

- routes $=$ lines (sequence of stops served by the same means of transport)
- operates in rounds, one per transfer
- round $k$ computes the fastest way of getting to every stop with at most $k-1$ transfers
- computes arrival times by traversing every route at most once per round


## Round-Based Search

RAPTOR - Round-bAsed Public Transit Optimized Router (Delling, Pajor, Werneck, ALENEX 2012)

## Advantages:

- simple data structures and excellent memory locality
- easy to parallelize
- no preprocessing necessary

Tested successfully for London's public transport system

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- train disposition


## Delays and Cancellations

Various causes


## Realtime Train Information

Considers current traffic situation:

- train cancellations, extra trains
- delay messages for all trains
- dispositions of the central train traffic management (connection train will wait/will not wait)


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Considers current traffic situation:

- train cancellations, extra trains
- delay messages for all trains
- dispositions of the central train traffic management (connection train will wait/will not wait)
"On-Trip Scenario":
You are already on the way How to find the best continuation to your destination?

chaos on a strike day


## Secondary Delays

- Decision "train A waits for train B" generates cascade of secondary delays for other trains
- Depending on waiting policy possibly many additional delays
- Revocation of wait decisions requires undo process



## Static Delay Propagation

## Event-Activity-Based Dependency Model

[M.-H., Schnee 2009]

"Realtime" Timestamp of event $=$ maximum over all incoming dependencies

## Massive Delay Streams



## Typical Delay Message (simplified)

```
<Paket TOut="20120602195644533">
    <ListNachricht>
        <Nachricht>
            <Ist>
                <Service Id="80031551" IdZNr="1744" IdZGattung="ICE"
                    IdBf="DH" IdBfEvaNr="8010085" IdZeit="20120602142300">
                <ListZug>
                        <Zug Nr="1744" Gattung="ICE">
                        <ListZE>
                        <ZE Typ="Ab">
                        <Bf Code="HB" EvaNr="8000050" Name="Bremen\sqcupHbf" />
                        <Zeit Soll="20120602195500" Ist="20120602195600" />
                        </ZE>
                </ListZE>
                </Zug>
                </ListZug>
                </Service>
        </Ist>
        </Nachricht>
    </ListNachricht>
</Paket>
```


## Delay Propagation

Our prototype: (M.-H., Schnee 2009)

- graph size: $1,000,000$ nodes (German train schedule)
- can handle massive data streams (6 million messages per day)
- time per operation: < 1ms per update


## Conclusion:

Handling (static!) delay information is not the computational bottleneck

## MOTIS (developed with Mathias Schnee)

- MOTIS is an abbreviation for Multi-Objective Traffic Information System
- is fully realistic
- its core includes a Dijkstra-based multi-criteria search algorithm


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- MOTIS is an abbreviation for Multi-Objective Traffic Information System
- is fully realistic
- its core includes a Dijkstra-based multi-criteria search algorithm
- minimizes exactly travel time and number of interchanges
- delivers many attractive alternatives using our concept of relaxed Pareto optimality
- MOTIS is easily extendable to further objectives (like train reservation, buffer times between train changes, ...)
- > 100k lines of C++ code


## MOTIS - Example



## MOTIS - Example



## Applications of Realtime Train Information

Application I: at service points

- Used by experienced staff to guide passengers.

Application II: feasibility check and rerouting

- Service provider constantly checks feasibility of planned connections.
- Of course, this assumes that a customer is willing to tell his travel plans to the provider.

- If some planned connection becomes infeasible (for what reason so ever) the customer is informed by a short message (SMS).
- The message does not only tell this fact but
 gives a recommendation for alternative routes.
- This recommendation can be tailored to the preferences of the customer.


## MOTIS - Example



## Transfer into Practice

## DB BAHN



## Transfer into Practice

Beta-version of realtime information of German Railways

## Disclaimer

Your connections were calculated on the basis of the current traffic situation, which can change at any time.

Please be aware that your ticket (e.g. with the requirement to use a specific train) does not necessarily include the use of all alternatives.

## Transfer into Practice

## DB BAHN



## Example: Real Query



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## Robust Timetable Information

[ATMOS 2011, jointly with Marc Goerigk, Marie Schmidt, Anita Schöbel (Göttingen) and Martin Knoth (MLU)]

- Goal: plan your journey such that you will reach all your transfers regardless of potential delays
- Scenarios: we have to define scenario sets specifying potential delays (uncertainty sets)
- Strictly robust optimization: determine the fastest route such that all included transfers will be reached for every scenario


## Uncertainty Sets

- scenario described as a vector $d$ of dimension \# travel and waiting arcs
- each entry specifies by how much the travel or waiting activity is delayed (primary delays)
- our model: very few large delays (parameter $K$ ), arbitrarily many small delays
- small delays: $\leq \epsilon_{a}$
- large delays on arc a: between $\epsilon_{a}$ and $d_{a}^{\max }$
- 

$$
\begin{aligned}
U:=U_{\epsilon}^{K}:=\left\{d \in \mathbb{R}^{\left|A_{\text {wait }} \cup A_{\text {drive }}\right|}:\right. & 0 \leq d_{a} \leq d_{a}^{\max } \text { for all } a \in A_{\text {wait }} \cup A_{\text {drive }} \\
& \left.\left|\left\{a \in A: d_{a}>\epsilon_{a}\right\}\right| \leq K\right\}
\end{aligned}
$$

## Results

Given: a set of delay scenarios and an event-activity network (i.e., the schedule)

Goal: identify all transfer activities which will never fail in the given scenario set (i.e., "strictly robust transfers")
Afterwards: Do timetable information subject to the network restricted to strictly robust transfers (i.e., forbid all other transfers)

## Main results:

- To decide whether a transfer arc is strictly robust is NP-hard
- Dynamic programming can be used to solve the problem heuristically (may classify some transfer arcs erroneously as non-robust)


## Strict Robustness

Average travel time increase for $A=20 \mathrm{~min}$.

$$
\mathrm{K}=0 \rightarrow \mathrm{~K}=1 \not \mathrm{~K}=2 \leftarrow \mathrm{~K}=3
$$



Way too restrictive!

## Price of "Light Robustness"

Goal: minimize the number of "non-robust" transfers subject to an upper bound on the travel time: earliest arrival time $+x$ minutes (in the figure: $x=$ TAA)

increase of strictly robust paths (in \%, left), increase of average travel time (center) and increase of minimum buffer time for transfers in the chosen light robust path in comparison to the nominal scenario (right)

## Stochastic Delay Propagation

[ATMOS 2011, with Annabell Berger, Andreas Gebhardt and Martin Ostrowski]
Drawbacks of static propagation:

- Assumes constant travel times (as scheduled!)
- Fluctuations and catch-up potential ignored (different speeds, track conflicts ...)
- Stochastic online model



## Main achievement:

- stochastic forecasts possible within few seconds,
- but further investigations with empirical delay data necessary


## Model Assumptions

General scenario:

- arbitrary discrete distributions
- stream of online messages about the delay status of trains
- "is"-messages about what has been realized


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General scenario:

- arbitrary discrete distributions
- stream of online messages about the delay status of trains
- "is"-messages about what has been realized

Assumption 1: With respect to status messages, a train can arrive or depart at any time after the planned arrival or departure time, respectively.

Assumption 2: With respect to our forecasts of arrival and departure time distributions, no train departs before its scheduled time or arrives at a station before its planned arrival time.

## Model Assumptions

Assumption 3: We assume that the distributions of arrival times of all feeder trains of a given train are stochastically independent.

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- standard assumption used throughout in all previous work
- makes the computation tractable from a practical point of view
- is not so unrealistic as it seems:
- suppose train $A$ is heavily delayed (primary delay) and
- train B and train C both catch a secondary delay from train A (that means, their arrival and departure distributions are dependent)
- but this dependency is reflected in our predictions since the information about A's primary delay is fully used, as soon as it becomes known


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- train B and train C both catch a secondary delay from train A (that means, their arrival and departure distributions are dependent)
- but this dependency is reflected in our predictions since the information about A's primary delay is fully used, as soon as it becomes known
- what about track conflicts?
here is indeed a problem - but the required data is not available


## Model Assumptions

Assumption 4: Waiting rules are defined for any pair of arriving and departing trains for which a transfer arc is defined.

## Remark:

degree of freedom to define "planned transfer arcs"
for simplicity, we did not implement new transfer possibilities due to other delayed trains

## Experiment: Predictions over time

Setup: all delays before 11:59 a.m. incorporated in predictions
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average distance of our predicted expectation values in minutes from the realized ones

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(1) Realistic models
(2) Multi-criteria search
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(5) Extensions

- train disposition
- multi-modal search


## General Goal: Passenger-Oriented Train Disposition

Basic question (train disposition): Shall a train wait for delayed feeder trains or not?

Why passenger orientation?

- increase of passenger satisfaction and therefore indirectly of the attractiveness of trains as means of transport
- avoidance of reimbursements or repays to customers according to customer charta in case of delays

'. . . die größte Freud', ist doch die Zufriedenheit.'
'no joy is found like contentment on earth's round!'
(Wilhelm Busch)


## Multi-Criteria Optimization

## Possible criteria:

(1) Sum of delay (in minutes) at destination, number of cases above 60 minutes, weighted according to customer groups (business travellers or others)
(2) number of missed transfers
(3) costs for reimbursements to customers (taxi costs, hotel fares, ...)
(9) strength of deviation from published schedule (additional expenses for staff and resources; number of track changes, ...)

## Status Quo

## Current situation in disposition centers

 for passenger trains (long-distance and regional):- only local view on effects of decisions (in particular on consequences for passengers)
- (in Germany, due to European regulations)
only partial view on potential track conflicts

- repeated and time-consuming decision processes


## Local Decisions - Global View

## Our vision of an optimized disposition

## The dispatcher decides

- upon waiting conflicts (shall a train wait?) in his regional disposition center, but
- under consideration of global consequences on the whole train network



## Our task:

development of tools for decision support

## Passenger Flows

Theoretical ideal: each planned connection for each individual passenger is known (on a daily basis)


Two problems:

- data on this detailed level not available
- data volume fairly large:
seems infeasible to represent 4 million passengers (and their routes!) individually

Our model: group passengers with same destination together, i.e.
for each train we assume to know for each driving section (approximately) how many passengers are heading to which destination

## What is needed?

## Our goals:

- decision support for train dispatchers
- decisions shall be based on effect on passenger flow
- real-time capability


## Algorithm Engineering Challenges:

- modeling issues:

How to model the objective function? $\rightarrow$ conflicting goals

- dynamic update of passenger flows (kind of multicommodity flow) in real-time


## Previous Work on Delay Management

- delay management is hard (Gatto et al.)
- integer linear programming (ILP) models (Schöbel and co-workers)
- static view (complete delay scenario is known)
- periodic schedules
- computational studies on comparatively small subnetworks

We consider online-scenario where the newest delay information is revealed step by step.

## Passenger-Oriented Train Disposition

[ESA 2011, joint work with Annabell Berger, Christian Blaar, Andreas Gebhardt and Mathias Schnee)]

Achievements: efficient prototype with building blocks:
(1) routines for the permanent update of our graph model subject to incoming delay messages,
(2) routines for forecasting future arrival and departure times,
(3) the update of passenger flows subject to several rerouting strategies (including dynamic shortest path queries), and
(9) the simulation of passenger flows.

## Module 1: Timetable Update

This tool

- updates the timetable with respect to a steady stream of delay messages
- always keeps predictions for future departure and arrival times

```
timetable
update
```

- determines critical transfers -
- with respect to standard waiting rules (of German Railways)


## Module 2: Passenger Flow Update

This tool

- updates every minute the number of passengers towards their destinations for each train
- passengers who miss their planned connecting train are rerouted.

- A kind of multi-commodity flow problem (one commodity for each destination).

Recall: We do not consider individual passengers, but groups of passengers with the same destination

## Rerouting Strategies

We apply the following rerouting strategies if necessary (in this order):

- Rule 1: Reroute passenger to the very next train towards his destination.
- Rule 2: Apply a dynamic timetable query to calculate a fastest alternative connection. Take the new connection, if acceptable.
- Rule 3: If neither rule 1 nor rule 2 apply,
 we send the passengers to "Nirvana".

This means: Such passengers have no reasonable alternative to reach their destination!

## Module 3: Optimization

Note: there is always a dilemma

- "waiting decision" usually means: many passengers catch a small delay while few "transfer passengers" are happy
- "do not wait decision" usually means: few "transfer passengers" are heavily delayed, all others are not delayed


## Our interpretation of passenger-friendly disposition:

- minimize sum of delays at destinations
- minimize deviation from planned passenger flow
- minimize number of non-arriving passengers
$\rightarrow$ we take a weighted combination as our objective
$\rightarrow$ weight factors model dependence of the time horizon


## Module 3: Optimization

Simple optimization loop:

- order critical transfers with respect to time horizon (most urgent ones come first; consider also number of affected passengers)
- select the most important one
- evaluate objective function for alternative waiting decisions
- propose the most promising ones to dispatcher
- once the current case is settled, proceed to the next


## Train Disposition - Overview

published schedule $\downarrow$
delay messages, cancellations

## Train Disposition - Overview



## Train Disposition - Overview



## Train Disposition - Overview



## Visualization of Alternatives



Disposition alternatives and their impact on passenger flows

## Train Disposition - Overview



## Test Instances and Environment

## Test instances:

- German train schedule
- 8800 stations, about 40000 trains, one million events per day
- stream of delay messages for whole traffic days

Test environment:

- $\mathrm{C}++$ code, compiled with $\mathrm{g}++4.4 .3$ and option -O3 under ubuntu linux 10.04.2 LTS
- PC Intel(R) Xeon(R) $2.93 \mathrm{GHz}, 4 \mathrm{MB}$ cache, 47 GB main memory, used only one core


## Simulation of Passenger Flows

- Problem: no actual passenger flow data available
- our solution: a SIMULATOR for passenger flows
- in this study: a fairly simple,
 but extendable prototype


## Extensions:

- day-time dependence (like rush hours)
- representative passenger countings

- origin-destination matrices


## Flow Updates



Scenario: 20 "light" delays of 5 minutes each

## Flow Updates



## Scenario: 20 "heavy" delays of 30 minutes each

## Passenger-friendly Train Disposition

How many disposition decisions are needed?


## Summary

- train disposition is a fairly complex, highly dynamic multi-criteria optimization problem
- real-time update of schedule and passenger flows can be achieved
- simple optimization fast enough to handle typical amount of disposition decisions on ordinary traffic days


## Future Work on Train Disposition

- more realistic passenger flows (currently we use a simple simulation)
- more advanced optimization
- improved wide-ranging predictions of delay forecasts
- more efficient rerouting of passengers
- capacity restrictions


## Final Thoughts

## Success story of Algorithm Engineering

Public transport timetable information has achieved quite some progress. But many challenges remain!

## Transfer into practice

There is a slow, but steadily improving impact of our research on solutions in industry.

[^0]
## Future Work - Challenges

## Multi-modal timetable information:

- integration of several modes of transport
- exact point-to-point queries
- hybrid models for combinations of "road" and "public transport"



## Future Work - Challenges

Pre-trip planning: "recovery robust" timetable information

Goal: compute a travel plan which has an acceptable alternative "backup connection" for every delay scenario
acceptable: with some guarantee on the
 worst-case arrival time

## Future Work - Challenges

## Real-time timetable information

- Improve forecasting models for the spreading of delays
- How can we exploit historical delay data?
- How to cope with volatile situations?

Dilemma: when to inform passengers what to do?

## Empirical Delay Distributions



## The End?

"When people think, one is done, one has really to start working"
"Wenn die anderen glauben, man ist am Ende, so muss man erst richtig anfangen."


Konrad Adenauer


[^0]:    Industry should be more cooperative
    Research in our field is often hindered by lack of cooperation from industry side. In particular, availability of test data is a crucial issue.

