Dynamic Time-Dependent Routing in Road Networks Through Sampling

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# Abstract

Studying the earliest arrival in road networks with:

- TIME DEPENDENT functions
- DYNAMIC UPDATES
- It compute with SMALL ERROR
- Doesn't suffer from MEMORY LACK on large instances
- This algorithm is the only that is able to answer to queries below 50 ML

# Introduction

The road network as a weighted graph, directed graph:

Nodes: positions

**Edges**: road segments

Weight: travel time a car need to traverse the road segments.

A common assumption: travel times are time independent

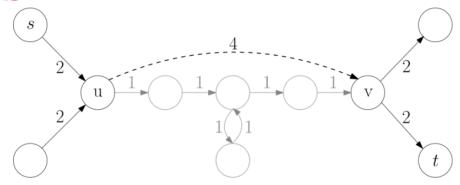
Result: problem will be a classical shortest path problem

Problem: road networks are huge and Dijkstra algorithm is slow!

Solution: contraction hierarchy (CH)

#### contraction Hierarchies

Contraction hierarchies (CHs) are a well-known algorithm for fast and exact route planning on road networks with constant travel costs



To find a path from s to t the algorithm can skip over the Grey vertices and use the dashed shortcut instead. This reduces the number of vertices the algorithm has to look at.

#### Overview on contraction Hierarchies

#### contraction Hierarchies:

- skip over unimportant vertices
- road networks are highly hierarchical. for example highway, are "more important" and higher up in the hierarchy than for example a junction leading into a dead end. the algorithm doesn't have to consider the full path at query time.

#### PROBLEM DEFINITION

The input of earliest problem:

NODES > S AND T

Departure time > s

The task is consist of computing **ST- path** with minimum arrival time.

We want a routing system that accounts for predicted and real time congestion (dynamic time dependent routing )

TD-S: Solve the earliest arrival problem with predicted congestion

TD-S+P: solve profile problem

TD-S+D: take real time congestion into account

## Computing Errors

When taking predicted congestion into account, many proposed algorithms compute paths with an error

A shortest path with respect to the prediction is not necessarily shortest with respect to reality.

Let d denote the length of the computed path and dopt the length of a shortest path

Absolute error: |d-dopt|

Relative error: |d-dopt| / dopt.

# A solution to the time-independent routing problem

### Free flow Heuristic

assumption

**No Congestion** 

The min value of e's travel time function of fe

**Step** 

Find the shortest time independent **path H** 

Compute **time** dependent travel time along **H given departure time** 

The first step can be fasten using any time independent speed up: such as CH

Avg flow heuristic: the same but it use average travel time

## Free flow Heuristic

Sampling means the variety of departure time in free flow

Free flow never reroutes based on the current traffic situation

**Steps** 

Compute sub graph H

Run time dependent extension of Dijkstra algorithm on sub graph

Compute the sub graph using a sampling approach

Define a constant number of **K** time intervals
Better k <=10

For every interval : obtain a time – independent **graph**!

within each time intervals average the time dependent travel times.

IN STEP 3, THEY USE A SAMPLING APPROACH, AVG THE TIME DEPENDENT TRAVEL TIME WITHIN EACH INTERVAL FOR EX: 7:10, 7:15, 7:50

For every interval : obtain a time – independent **graph**!

It s time independent Because we calculate avg

For every graph we compute a shortest time independent **path** 

The **union** of these paths is the sub graph H

They use Dijkstra here!

They compute for each time-interval a time-independent CH

The first step of TD+S executes k CH-queries.

# Computing profile: TD-S+P

**Steps** 

Compute sub graph H

Just for example For a sampling rate 10 minute

Sample at regular intervals the travel time by running the time-dependent extension of Dijkstra's algorithm restricted to the sub graph H.

The sub graph computation step is the same as for TD-S.

#### TD-S+P

For a sampling rate of 10min,

- 1- algorithm first computes the sub graph H,
- 2- then runs Dijkstra's algorithm with the departure times 0:00, 0:10, 0:20...23:50 restricted to H.

Denote by a1, a2, a3...a144 the computed arrival times.

For example, the computed arrival time for departure time 0:07 is 0.3 a1+ 0.7 a2

#### TD-S+P

TD-S+P is faster than iteratively running TD-S as the sub graph is only computed once.

Further, Dijkstra's algorithm iteratively runs on the same small sub graph H.

The first run loads H into the cache and thus all subsequent runs incur nearly no cache misses

## TD-S+P

#### **Dynamic TD-Routing Through Sampling**

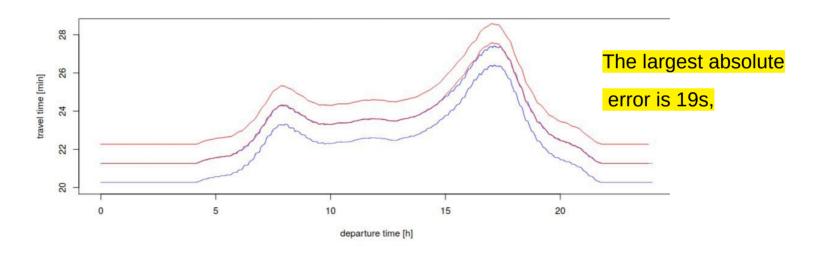


Figure 2

Example profile over 24h. The red curve (top) was computed with TD-S+P, while the blue one (bottom) is the exact solution. The middle overlapping curves are the actual profiles. To improve readability

#### TD-S+D

However, in many applications we must also take real time congestion into account.

- 1- Adapting TD-S by modifying the computation of the sub graph H yielding TD-S+D.
- 2-The second step is left unchanged

#### TD-S+D

The sub graph H is the union of k paths

TD-S+D adds a shortest st-path according to the current real time traffic as k+1-th path to the union.

A efficient solution to the routing problem with real time congestion is needed. They use CCH algorithm .

#### TD-S+D

In the pre processing step, TD-S+D computes a CCH in addition to the k CHs of TD-S.

At regular time intervals, such as every 10s, TD-S+D updates the CCH edge weights to reflect the real time traffic situation

A TD-S+D query consists of running k CH queries and one CCH query.

The sub graph H is the union of the k+ 1 shortest paths. Finally, Dijkstra's algorithm is run to the sub graph H.

## Simulating Traffic

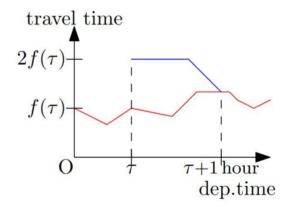
Unfortunately, they do not have access to good measured real time traffic. they simulate real time congestion to study the performance of TD-S+D

## Simulating Traffic

- For an earliest arrival time query from s to t with departure time τ, we first compute the shortest time-dependent path P with respect to the historic travel times
- On P we **generate three traffic congestion** by picking three random start edges.
- From each of these edges we follow P for 4min, yielding three sub paths

- 1- Travel time function f of e
- 2- by doubling the travel time at  $f(\tau)$  and assume that it remains constant for some time
- 3- The congestion should be gone at  $\tau$ + 1h.
- 4- FIFO-property:e modified function must have slope of -1 before  $\tau$ + 1before joining the predicted travel time
- 5- ON THIS  $2f(\tau) < f(\tau + 1h)$ , we do not generate a congestion.

# Simulating Traffic



**Table 2** Number of exact time-dependent queries and absolute and relative errors for Freeflow, TD-S+4, and TD-S+9. "Q99" refers to the 99%-quantile and "Q99.9" the 99.9%-quantile.

The number of exactly
solved queries decreases
with instance size as the
paths lengths grow with size

as most queries are answered exactly.

TD-S+4 has larger errors as it uses fewer time windows.

the query running times and the memory footprint of TD-S+4 are lower.

Graph	Algo	Exact	Relative Error [%]				Absolute Error [s]				
		[%]	Avg	Q99	Q99.9	Max	Avg	Q99	Q99.9	Max	
Lux	Freeflow	80.0	0.244	5.1	11.5	28.1	5.0	106	235	356	
Lux	Avgflow	81.0	0.123	2.5	6.4	19.4	2.5	49	143	329	
Lux	TD-S+4	97.7	0.008	0.2	1.5	4.9	0.2	4	30	141	
Lux	TD-S+9	99.6	< 0.001	0.0	0.1	1.7	< 0.1	0	3	27	
Ger	Freeflow	67.9	0.085	1.5	3.1	12.4	11.1	200	417	825	
Ger	Avgflow	69.2	0.044	0.8	1.9	10.3	5.9	113	284	587	
Ger	TD-S+4	94.6	0.005	0.1	1.0	3.0	0.8	17	159	474	
Ger	TD-S+9	98.2	0.001	< 0.1	0.4	3.0	0.3	1	76	374	
OGer	Freeflow	60.7	0.140	2.0	4.7	12.4	15.9	219	465	1 104	
OGer	Avgflow	68.8	0.050	0.9	2.2	6.5	5.7	96	227	619	
OGer	TD-S+4	96.4	0.002	0.1	0.4	2.0	0.3	6	47	333	
OGer	TD-S+9	98.5	0.001	< 0.1	0.2	2.0	0.1	1	24	276	
CEur	Freeflow	54.9	0.089	1.4	2.7	10.8	26.4	428	833	1 477	
CEur	Avgflow	55.8	0.048	0.8	1.7	6.6	14.2	235	507	1 069	
CEur	TD-S+4	91.1	0.006	0.2	0.7	3.8	1.8	47	226	547	
CEur	TD-S+9	96.8	0.001	< 0.1	0.3	1.2	0.5	6	109	397	

## Analysis

TD-S with two selections of time windows:

TD-S+4 uses the windows0:00–5:00, 6:00–9:00, 11:00–14:00, and 16:00–19:0

TD-S+9 uses the windows 0:00–4:00,5:50–6:10, 6:50–7:10, 7:50–8:10, 10:00–12:00, 12:00–14:00, 16:00–17:00, 17:00–18:00

Table 3 Average preprocessing and running times and memory consumption of various algorithms.

the query running times and the memory footprint of TD-S+4 are lower.

TCH is lacking from memory .they couldn't run on 128 gig

TD-S+4 only needs about 2.4 times the memory required by the input. TD-S+9 needs 4.1 times the memory.

TD-S+4has an about a factor 10 lower pre processing time than TCH

Graph	TD-Dijkstra	Freeflow	Avgflow	TD-S+4	TD-S+9	TCH
	A	verage Que	ry Running	g Time [ms]		
Lux	4	0.02	0.02 0.11 0.26		0.18	
Ger	1 116	0.19	0.20	0.99	3.28	1.81
OGer	813	0.12	0.14	0.97	2.09	1.12
CEur	4 440	0.42	0.29	3.83	6.85	OOM
		Max. Qu	ery Memor	ry [MiB]		
Lux	13	17	17	29	47	328
Ger	1 550	2132	2130	3 630	6127	42857
OGer	461	855	854	1880	3589	8 153
CEur	4 980	7058	7053	12411	21336	> 131072
	Tota	al Preproces	ssing Runn	ing Time [n	nin]	
Lux	_	- <0.1		< 0.1	0.1	0.6
Ger	_	1.6	2.2	7.6	14.7	86.2
OGer	_	1.5	1.6	5.9	16.4	26.8
CEur	_	8.6	8.1	33.9	70.7	381.4

Graph	Algo	Exact [%]	Relative Error [%]				Absolute Error [s]			
			Avg	Q99	Q99.9	Max	Avg	Q99	Q99.9	Max
Lux	Predict.P	1.6	17.228	56.1	75.0	93.8	323.0	739	826	997
Lux	TD-S+D4	94.7	0.017	0.5	2.3	6.2	0.6	15	93	231
Lux	TD-S+D9	95.0	0.016	0.5	2.2	6.2	0.5	14	89	231
Ger	Predict.P	55.1	1.2	17.9	36.9	79.3	78.5	552	741	1 001
Ger	TD-S+D4	90.9	0.032	1.0	2.6	7.0	3.5	116	233	474
Ger	TD-S+D9	93.4	0.026	0.9	2.5	6.2	2.8	99	216	469
OGer	Predict.P	52.3	1.352	18.7	38.3	65.8	84.9	563	738	934
OGer	TD-S+D4	91.5	0.031	1.0	2.6	5.4	3.2	108	224	462
OGer	TD-S+D9	92.9	0.028	0.9	2.5	5.4	2.9	102	219	462
CEur	Predict.P	72.6	0.392	7.0	25.9	81.9	41.0	443	653	1870
CEur	TD-S+D4	89.5	0.015	0.5	1.6	5.2	3.3	106	244	547
CEur	TD-S+D9	94.0	0.011	0.3	1.4	5.2	1.9	69	205	397

Table 6Number of exact dynamic, time-dependent queries and absolute and relative errors for the predicted path, TD-S+D4, and TD-S+D9.

On the Luxembourg instance only 1.6% of the queries are solved

TD-S+D reduces these

instances, the minimum

number of optimally solved queries is

optimally.

92.9%.

errors. Over all

# Dynamic Time-Dependent Routing

In the dynamic scenario, we consider two types of congestion:

- (a) the predicted congestion.
- (b) the real time congestion.

The Predicted Path heuristic (Predict. P) as baseline, TD-S+D4, and TD-S+D9.

The Predicted Path heuristic computes a shortest path P with respect to only the predicted congestion NOT REAL TIME CONGESTION

P then is then evaluated with respect to both congestion types TD-S+D4, and TD-S+D9.

#### Predicted Path P

Free flow and Predicted Path are similar in spirit.

Free flow solves the time-dependent routing problem by ignoring predicted congestion.

Similarly, Predicted Path solves the dynamic time-dependent routing problem by ignoring real time congestion.

The Free flow heuristic produces surprisingly small errors

#### Conclusion

Introducing TD-S, a simple and efficient solution to the earliest arrival problem with predicted congestion on road graphs.

We extend it to TD-S+P which is the only algorithm to solve the profile variant in at most 50ms on all test instances.

Further, we demonstrated with TD-S+D that additional real time congestion can easily be incorporated into TD-S.