

Efficient Generation of Geographically Accurate Transit Maps

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¹ University of Freiburg

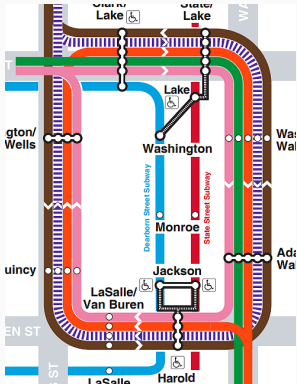
² LMU Würzburg

26th ACM SIGSPATIAL - Seattle, Washington, USA

November 7, 2018

Motivation

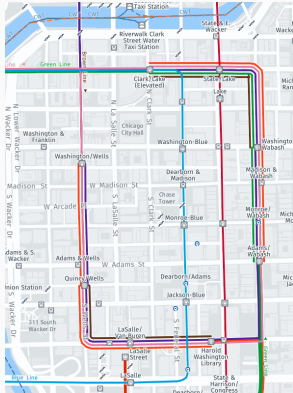
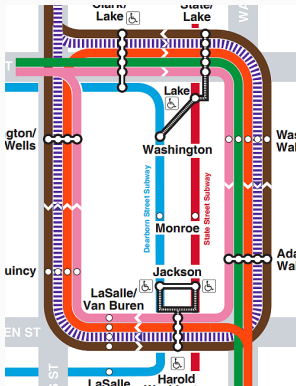
Official CTA map



Motivation

Official CTA map

HERE

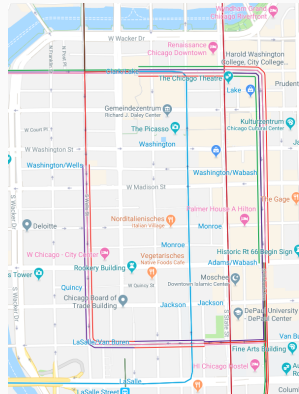
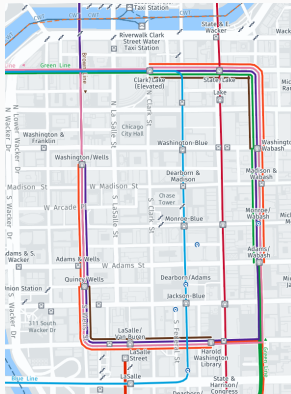
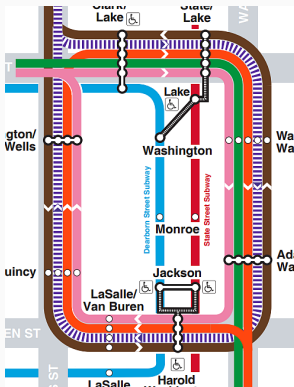


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HERE

Google

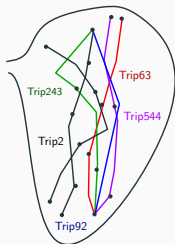


Goal

Goal: Generate these maps automatically, in high quality

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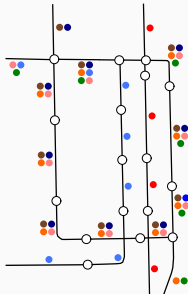
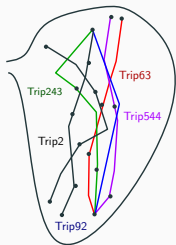
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"Bag of trips"
(GTFS)

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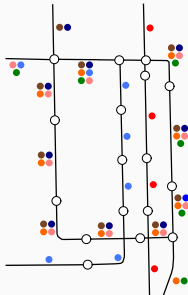
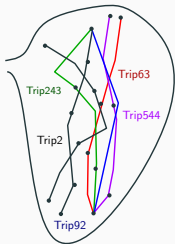
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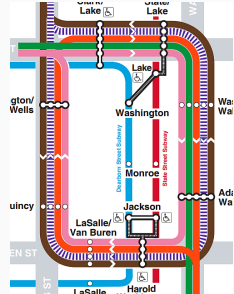
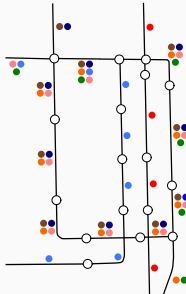
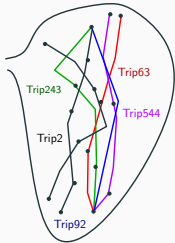


"Bag of trips"
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Line graph

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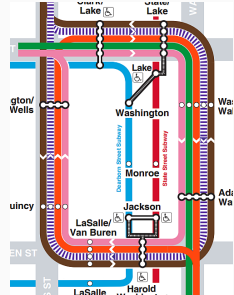
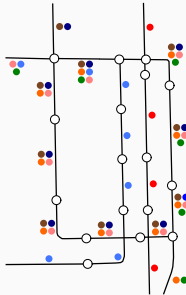
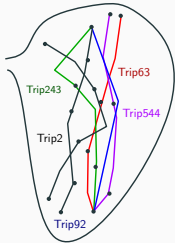


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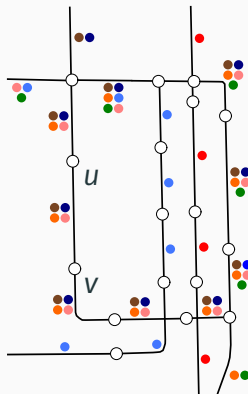
Final map

Line graph construction

Line graph:

- Undirected labeled graph

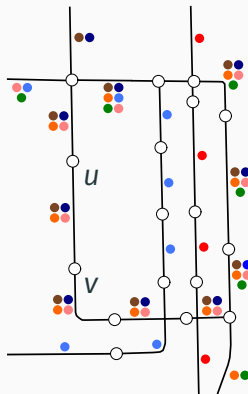
$$G = (V, E, L)$$



Line graph construction

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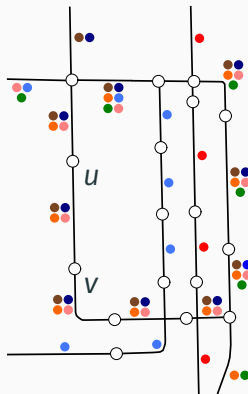
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 $(L(e) \subseteq \mathcal{L})$



Line graph construction

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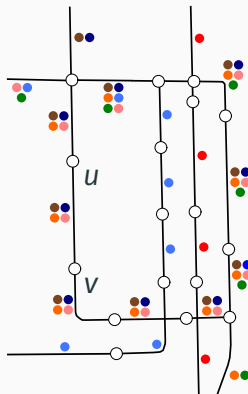
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Line graph construction

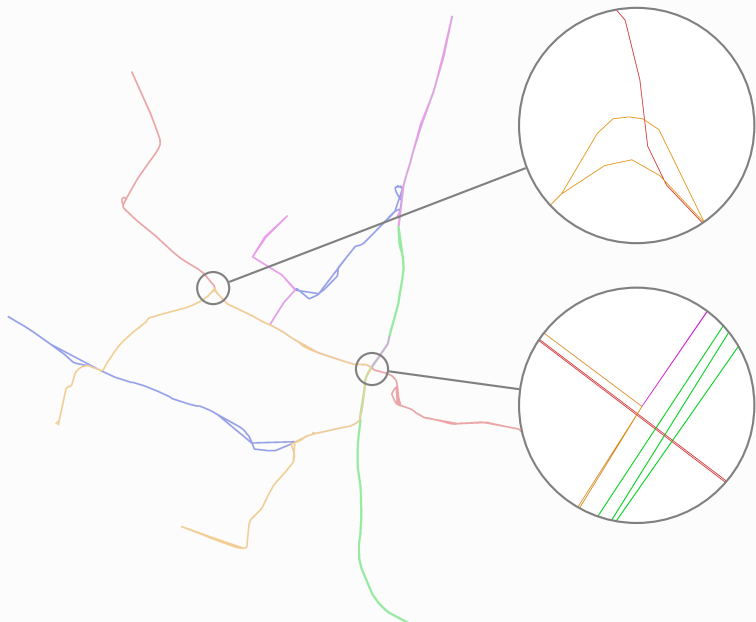
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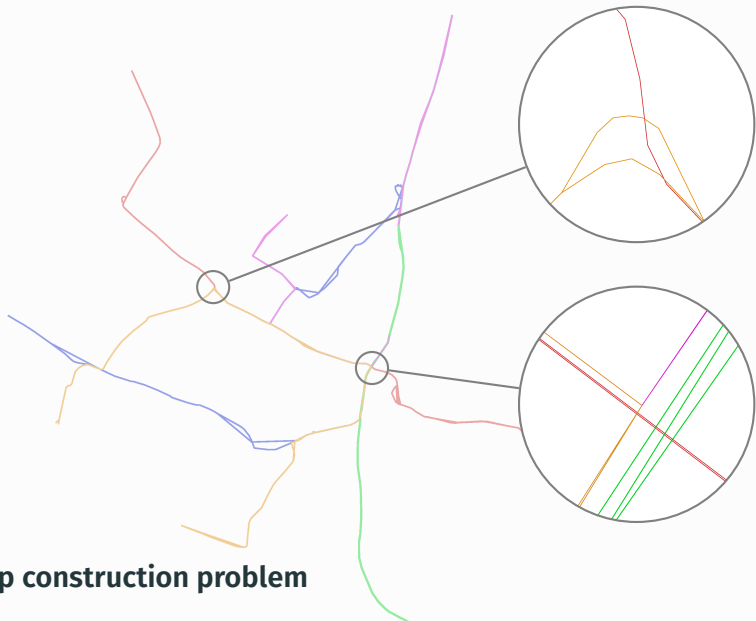


Example: $\mathcal{L} = \{\text{red, blue, orange, pink, green, brown}\}$, $L((u, v)) = \{\text{brown, blue}\}$

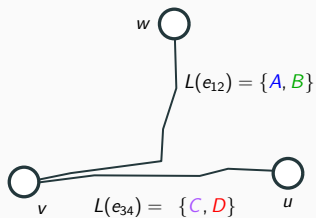
Line graph construction - Input data



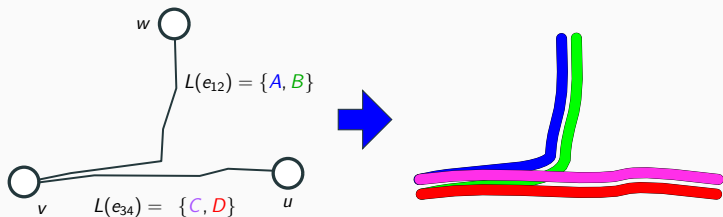
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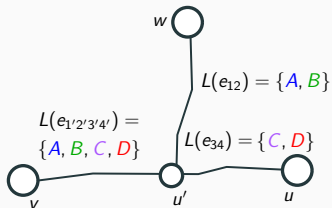
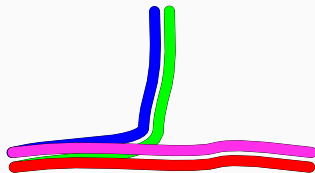
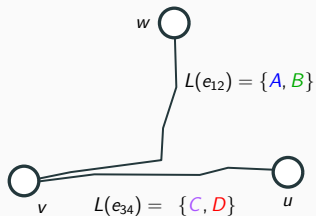
Line graph construction - Non-station nodes



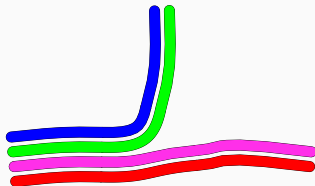
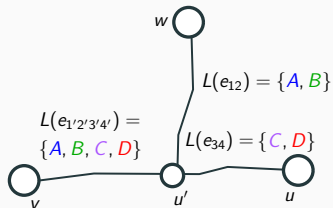
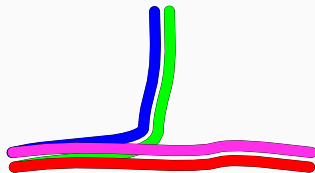
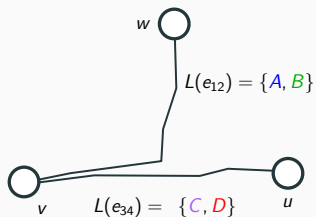
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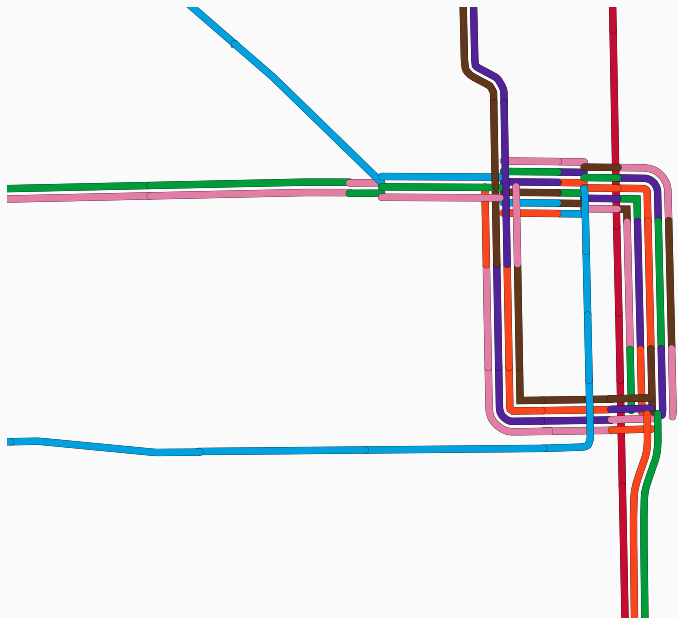
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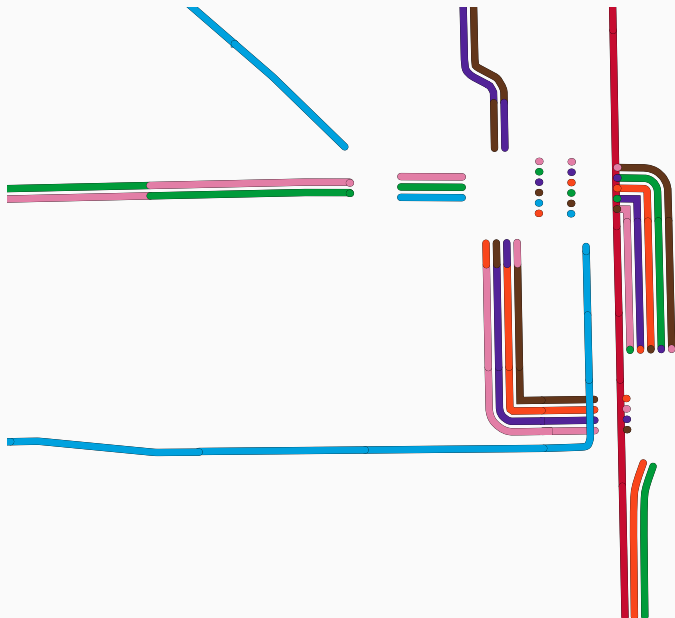
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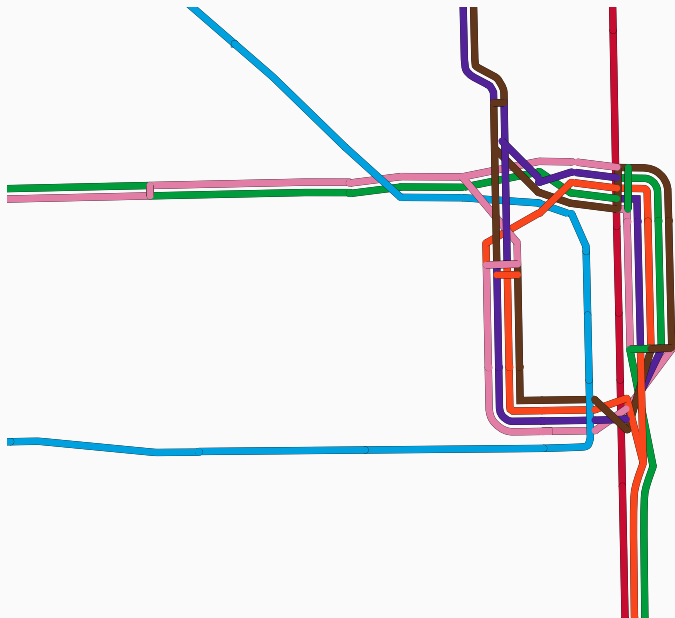
Results so far (1)



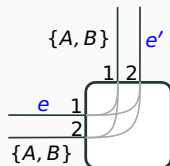
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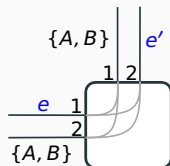


Line-ordering optimization - Baseline ILP



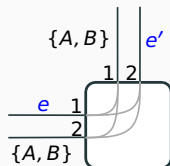
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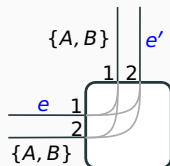
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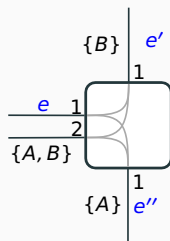
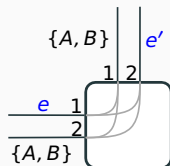
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- Constraint: all x_{elp} have to sum up to 1 for a single line l on a single edge e , and for a single p on a single edge

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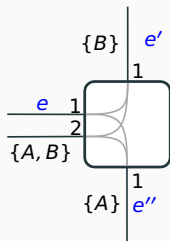
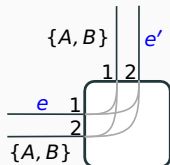
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$\Rightarrow \mathcal{O}(|E|M^2)$ variables, $\mathcal{O}(|E|M^6)$ constraints

- **Observation:** we only need to check if $p_e(A) < p_e(B)$ (or vice versa) for both types of crossings

Line-ordering optimization - Improved ILP

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- But we explicitly enumerate all possible line positions of A and B on e

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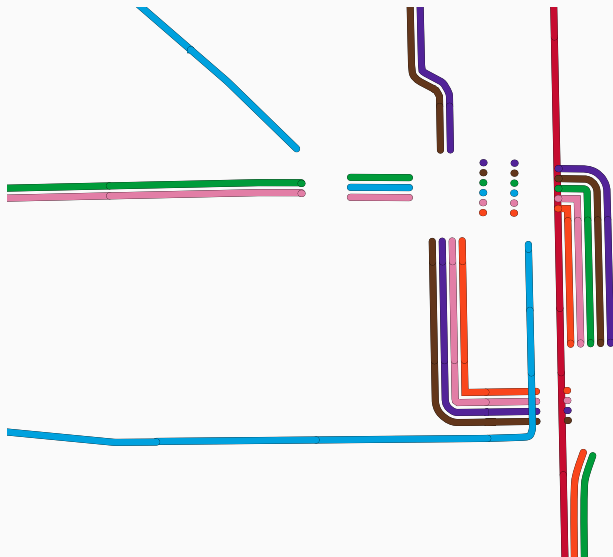
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- **Basic idea:** introduce binary variables $x_{eA < B}$ and $x_{eB < A}$ which can be efficiently checked

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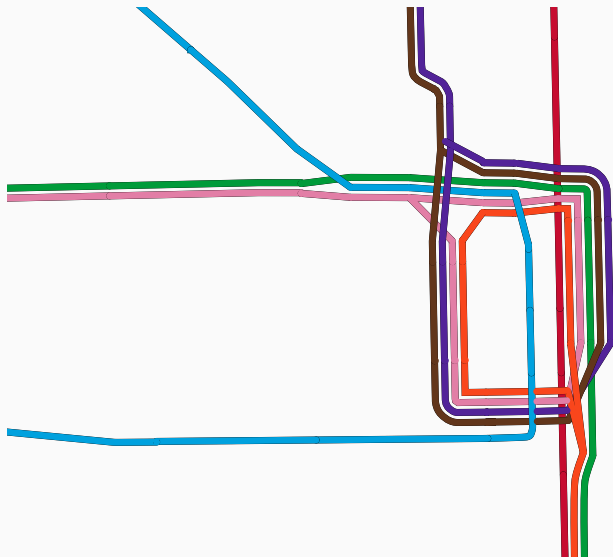
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$\Rightarrow \mathcal{O}(|E|M^2)$ variables, $\mathcal{O}(|E|M^2)$ constraints

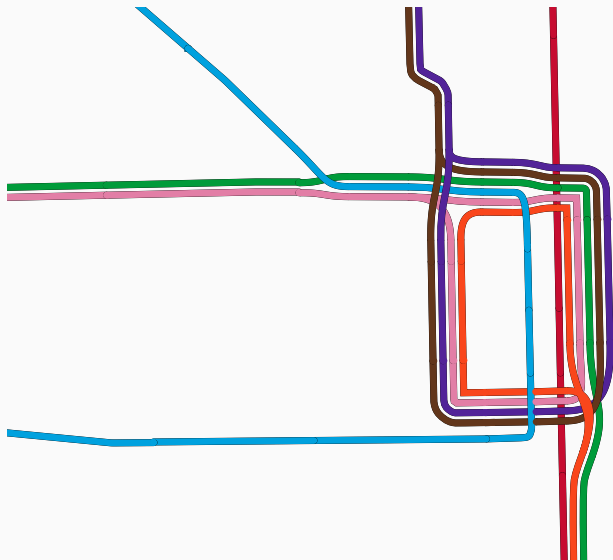
Results so far (2)



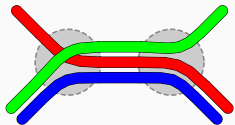
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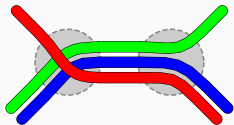
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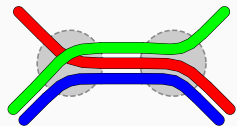
Line-ordering optimization - Line separations



VS.

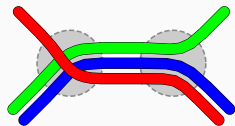


Line-ordering optimization - Line separations



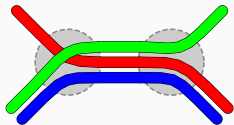
1 crossing

vs.



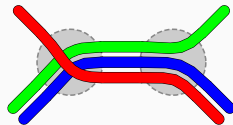
2 crossings

Line-ordering optimization - Line separations

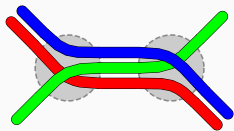


1 crossing

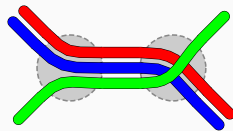
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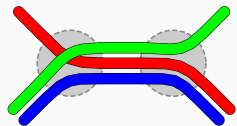
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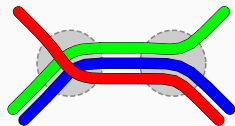


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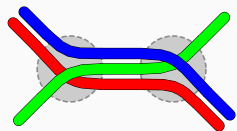


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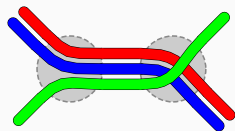


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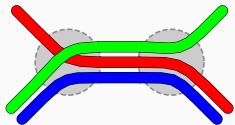
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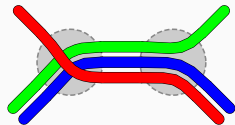
2 crossings

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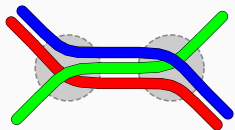


1 crossing
1 separation

vs.

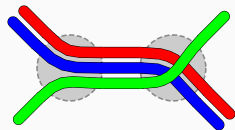


2 crossings



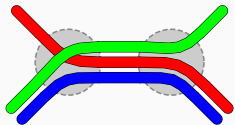
2 crossings
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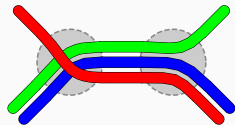
Line-ordering optimization - Line separations



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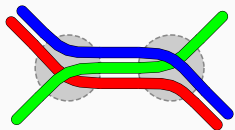
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vs.



2 crossings

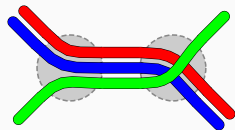
0 separations



2 crossings

1 separation

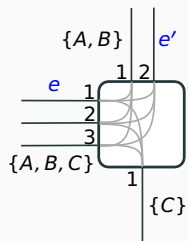
vs.



2 crossings

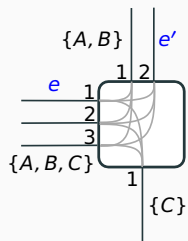
0 separations

Line-ordering optimization - Line separations (ctd.)



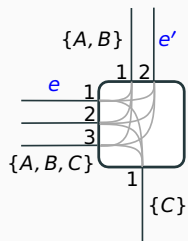
- **Idea:** If two lines A, B continue from e to e' , set a binary separation variable $x_{ee'A||B} = 1$ if they are next to each other in e , but no in e'

Line-ordering optimization - Line separations (ctd.)



- **Idea:** If two lines A, B continue from e to e' , set a binary separation variable $x_{ee'A||B} = 1$ if they are next to each other in e , but no in e'
- Add $x_{ee'A||B}$ to the objective function

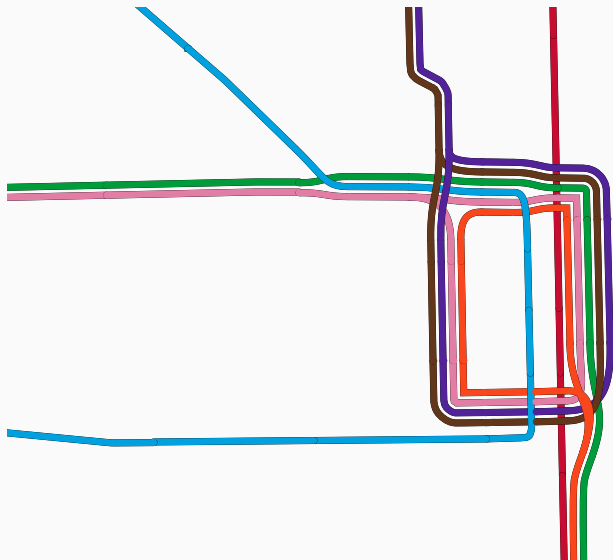
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\Rightarrow Still $\mathcal{O}(|E|M^2)$ variables, $\mathcal{O}(|E|M^2)$ constraints

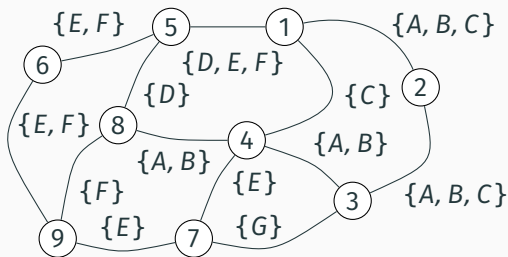
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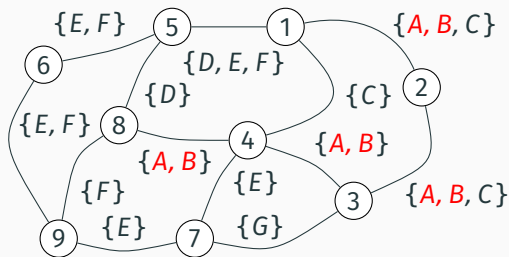
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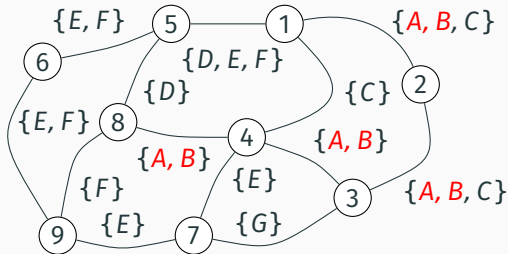
Line-ordering optimization - Core optimization graph



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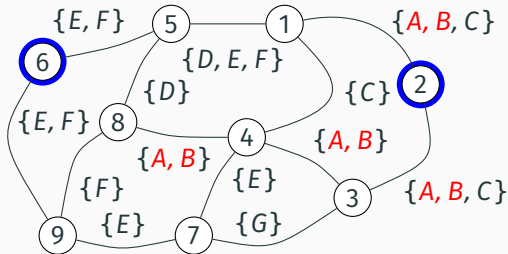


Line-ordering optimization - Core optimization graph



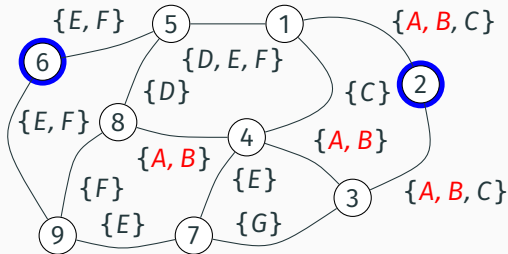
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Line-ordering optimization - Core optimization graph



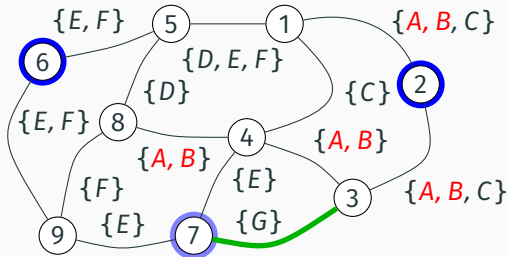
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Line-ordering optimization - Core optimization graph



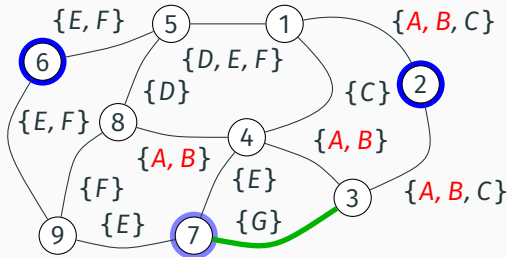
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Line-ordering optimization - Core optimization graph



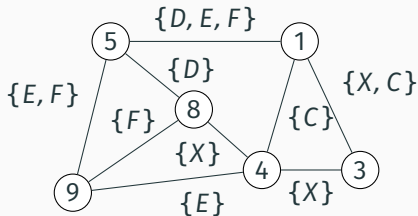
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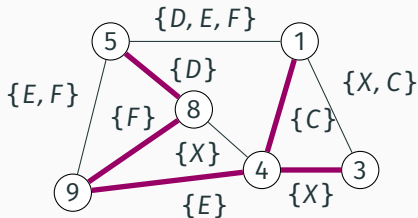
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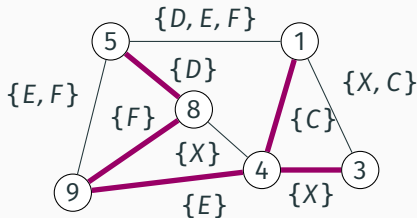
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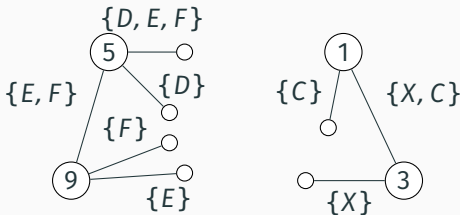
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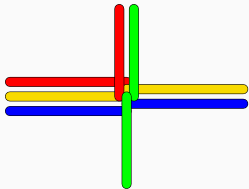
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Line-ordering optimization - Core optimization graph



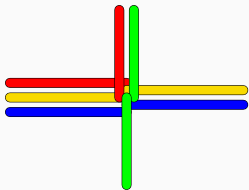
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Rendering

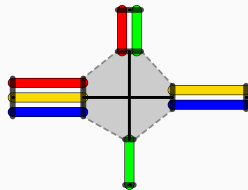


1. Render parallel lines

Rendering

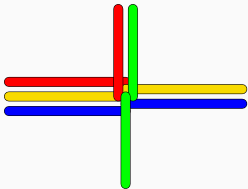


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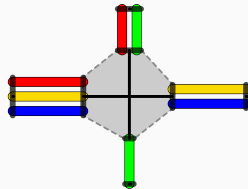


2. Free node space

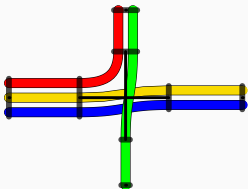
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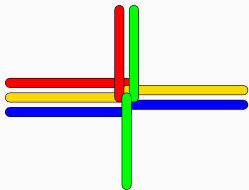


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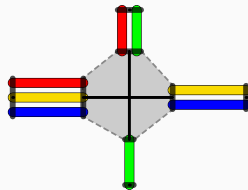


3. Render inner node connections

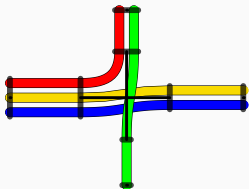
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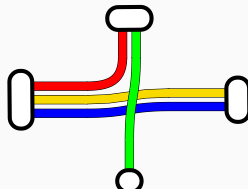
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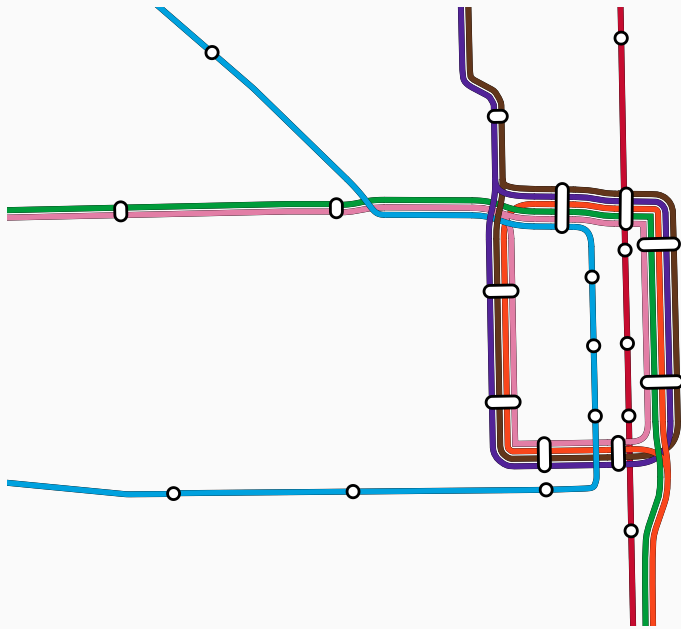


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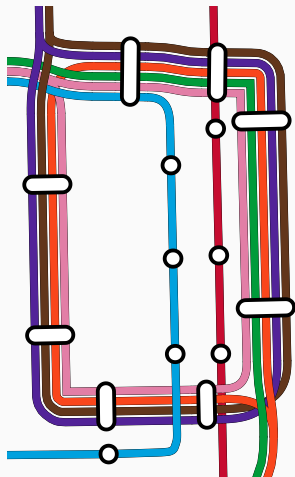


4. Render stations

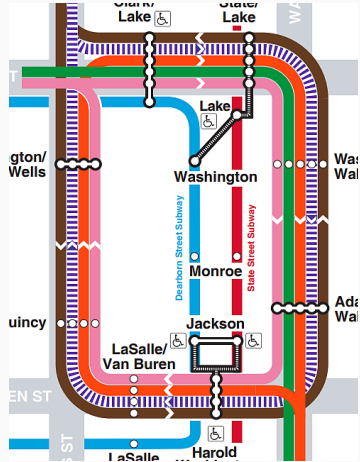
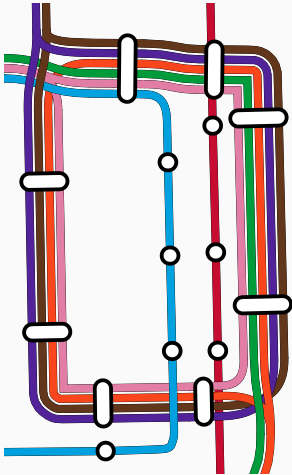
Results so far (4)



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Results so far (4)



Evaluation - ILP Solution times

ILP solution times for Chicago, on **baseline** graph

| | rows × cols | GLPK | CBC | GU | × | |
|--------|-------------|------------|------------|-------------|----|-----|
| Base | 41k × 861 | — | — | — | 22 | 4-7 |
| Impr. | 1.4k × 982 | 9s | 1s | 41ms | 22 | 4-7 |
| + Sep. | 1.9k × 1.2k | 47m | 19s | 1.8s | 27 | 0 |

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ILP solution times for Chicago, on **core** graph

| | rows × cols | GLPK | CBC | GU | × | |
|--------|-------------|-------------|-------------|-------------|----|-----|
| Base | 8.2k × 266 | — | 47m | 2m | 22 | 4-7 |
| Impr. | 394 × 285 | 0.8s | 0.1s | 10ms | 22 | 4-7 |
| + Sep. | 505 × 338 | 23s | 3.8s | 0.3s | 27 | 0 |

- Additional rules for core graph reduction
(work in progress)

Future work

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- Faster construction times of line graph
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Future work

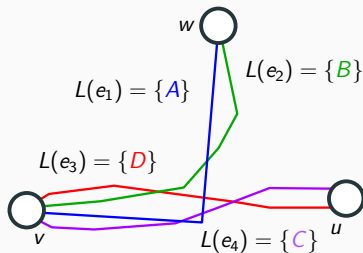
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- Octilinearize line graph for (non-overlay) schematic
metro maps (work in progress)



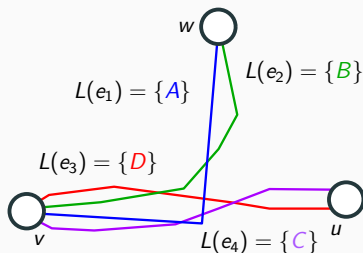
Thank you!

<http://loom.informatik.uni-freiburg.de>

Line graph construction - Shared segment collapsing

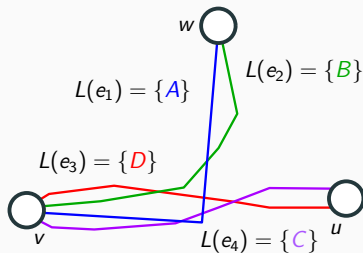


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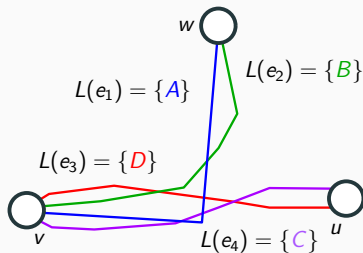
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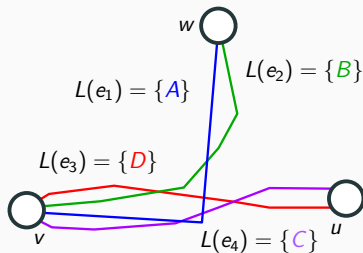
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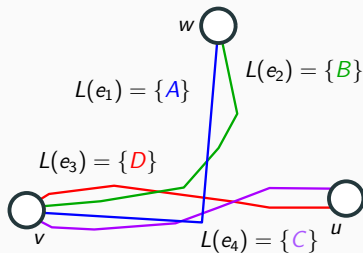
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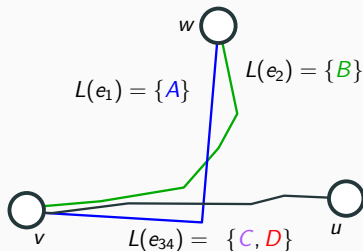
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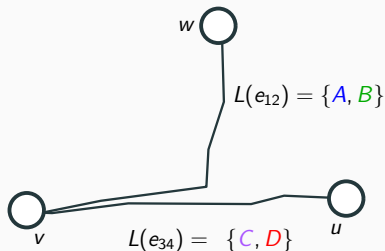
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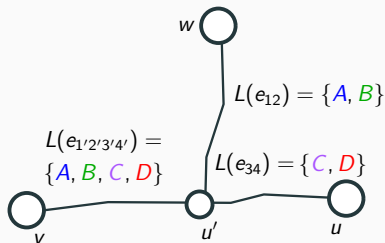
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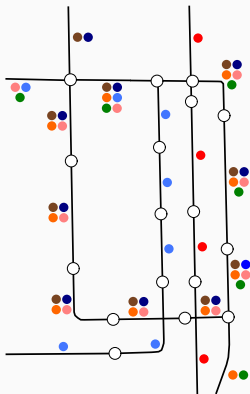
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Evaluation - Line Ordering

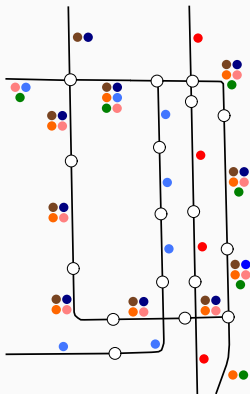
T = number of (consecutive) **line swaps** necessary to transform official map into our map

| | <u>Off. map</u> | | <u>Our map</u> | | |
|-----------|-----------------|---|----------------|---|----------|
| | × | | × | | T |
| Freiburg | 7 | 1 | 7 | 0 | 2 |
| Dallas | 3 | 1 | 3 | 0 | 1 |
| Chicago | 26 | 0 | 27 | 0 | 1 |
| Stuttgart | 65 | 5 | 64 | 2 | 4 |

Line-ordering optimization - Exhaustive approach

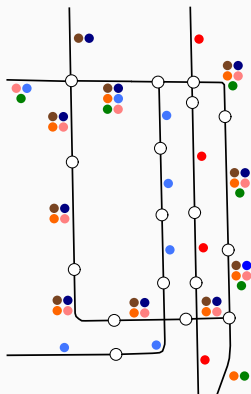


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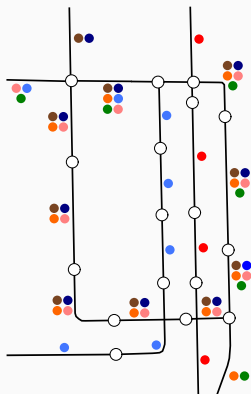
- 23 edges

Line-ordering optimization - Exhaustive approach



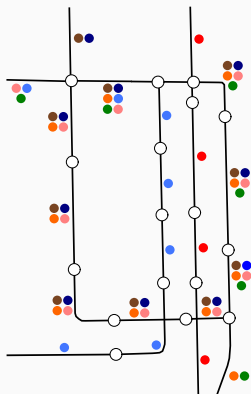
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- Each edge e has $|L(e)|!$ possible line permutations

Line-ordering optimization - Exhaustive approach



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- Possible configurations for the graph on the left: $> 2 \times 10^{17}$

Line-ordering optimization - Exhaustive approach



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- Possible configurations for the graph on the left: $> 2 \times 10^{17}$

⇒ Naive exhaustive search infeasible

Baseline ILP - Details

Each line must only be assigned **one** position:

$$\forall l \in L(e) : \sum_{p=1}^{|L(e)|} x_{elp} = 1.$$

Each position must only be assigned once:

$$\forall p \in \{1, \dots, |L(e)|\} : \sum_{l \in L(e)} x_{elp} = 1.$$

Constraints for ensuring that $x_{ee'AB} = 1$ if a crossing occurs:

$$x_{eA1} + x_{eB2} + x_{e'A2} + x_{e'B1} - x_{ee'AB} \leq 3$$

$$x_{eA2} + x_{eB1} + x_{e'A1} + x_{e'B2} - x_{ee'AB} \leq 3$$

...etc

Stadtbahn-Liniennetz



Dataset dimensions

| | t_{extr} | $ S $ | $ V $ | $ E $ | $ \mathcal{L} $ | M |
|-----------|-------------------|-------|-------|-------|-----------------|-----|
| Freiburg | 0.7s | 74 | 80 | 81 | 5 | 4 |
| Dallas | 3s | 108 | 117 | 118 | 7 | 4 |
| Chicago | 13.5s | 143 | 153 | 154 | 8 | 6 |
| Stuttgart | 7.7s | 192 | 219 | 229 | 15 | 8 |
| Turin | 4.9s | 339 | 398 | 435 | 14 | 5 |
| New York | 3.7s | 456 | 517 | 548 | 26 | 9 |

Core graph dimensions

| | $ V $ | $ E $ | $ \mathcal{L} $ | M |
|-----------|-------|-------|-----------------|-----|
| Freiburg | 20 | 21 | 5 | 4 |
| Dallas | 24 | 24 | 7 | 4 |
| Chicago | 23 | 24 | 8 | 6 |
| Stuttgart | 50 | 58 | 15 | 8 |
| Turin | 91 | 124 | 14 | 5 |
| New York | 110 | 138 | 23 | 9 |

Challenges - Detail

Official



HERE



Google



I. Avoid line overlaps

Challenges - Detail

Official

HERE

Google



I. Avoid line overlaps



II. Match line orderings

Challenges - Detail

Official

HERE

Google



I. Avoid line overlaps



II. Match line orderings



III. Clearly indicate line continuations