

Train self-organisation for traffic management decisions

Vito Trianni

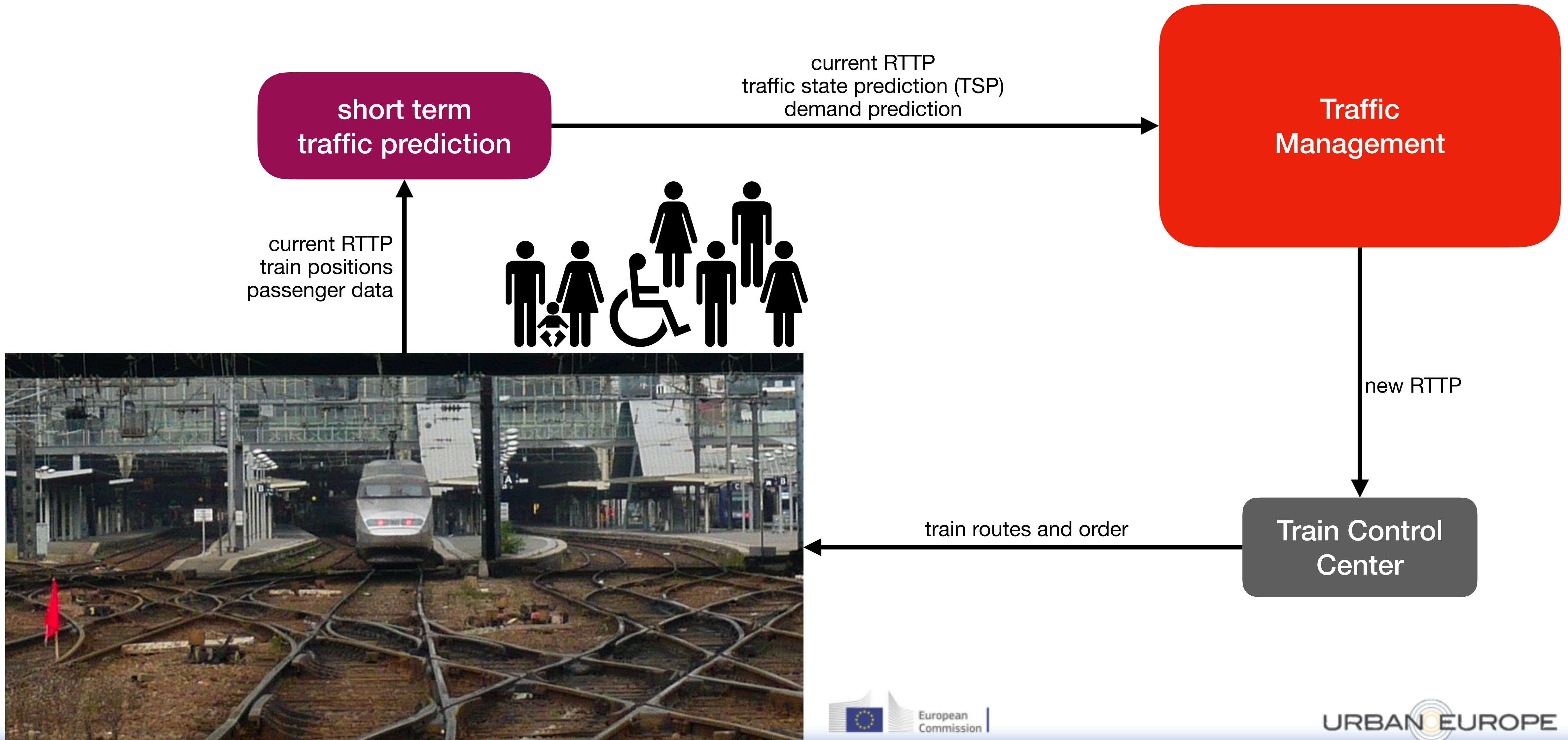
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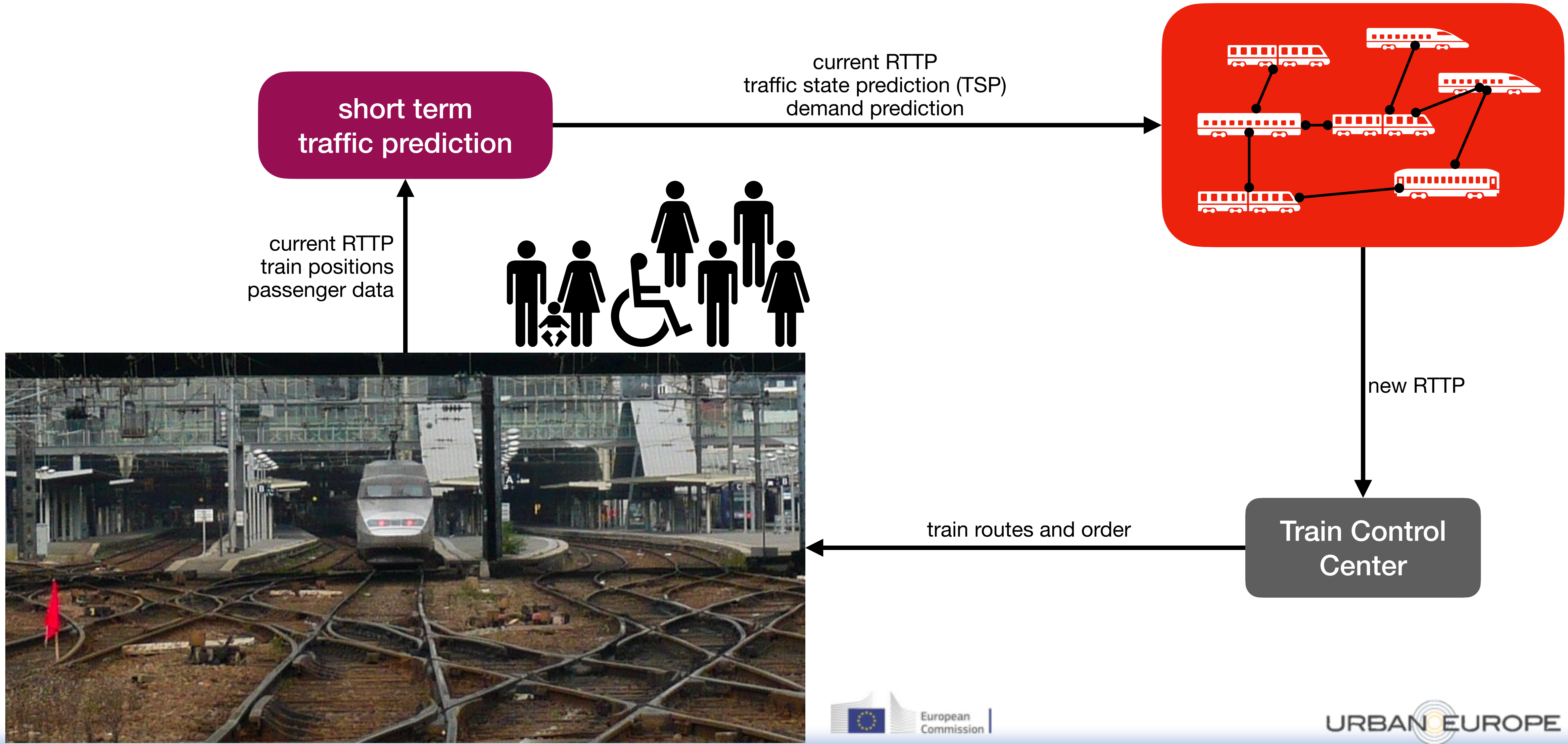
SORTEDMOBILITY Final Event

Paris, May the 13th, 2024

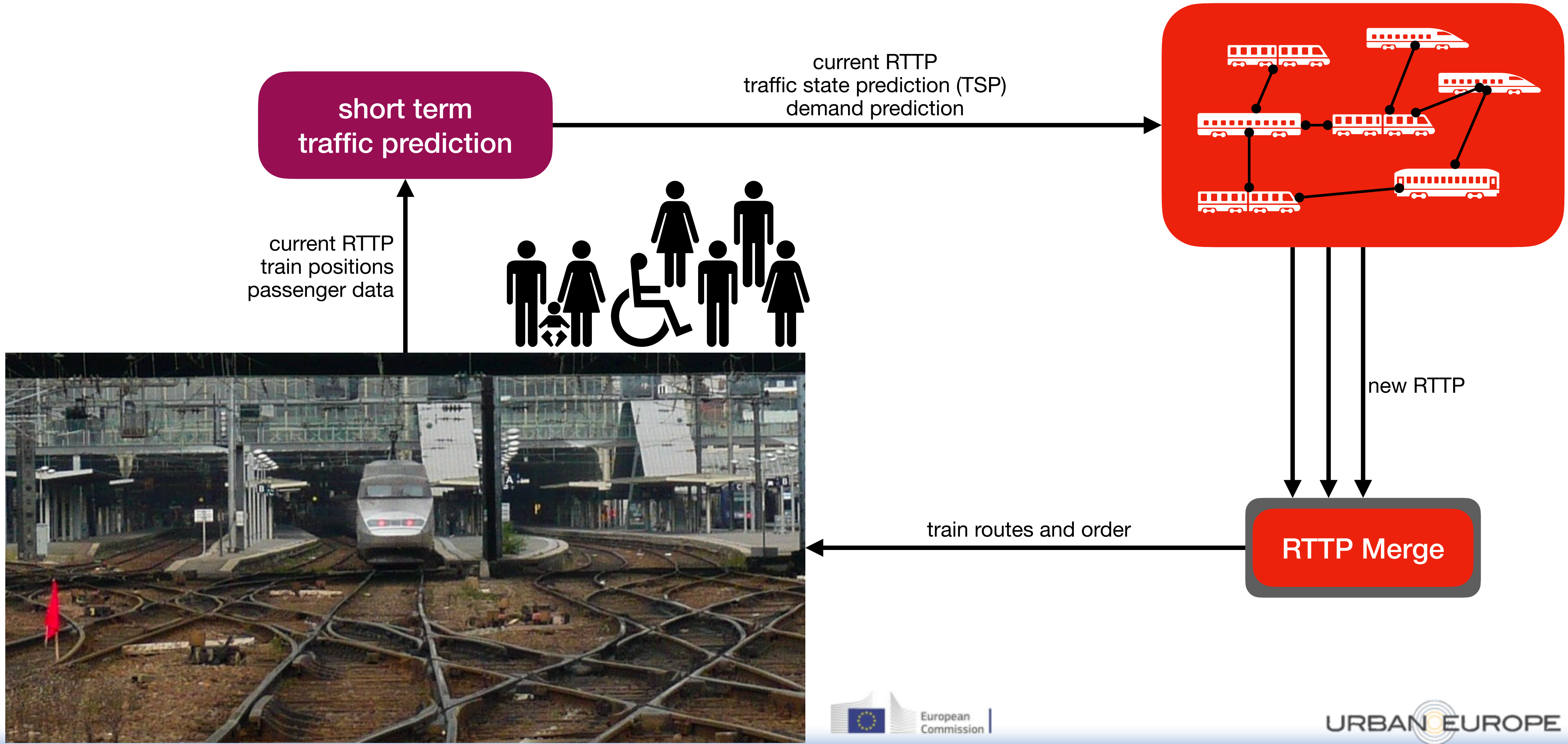
Closed-Loop Traffic Management

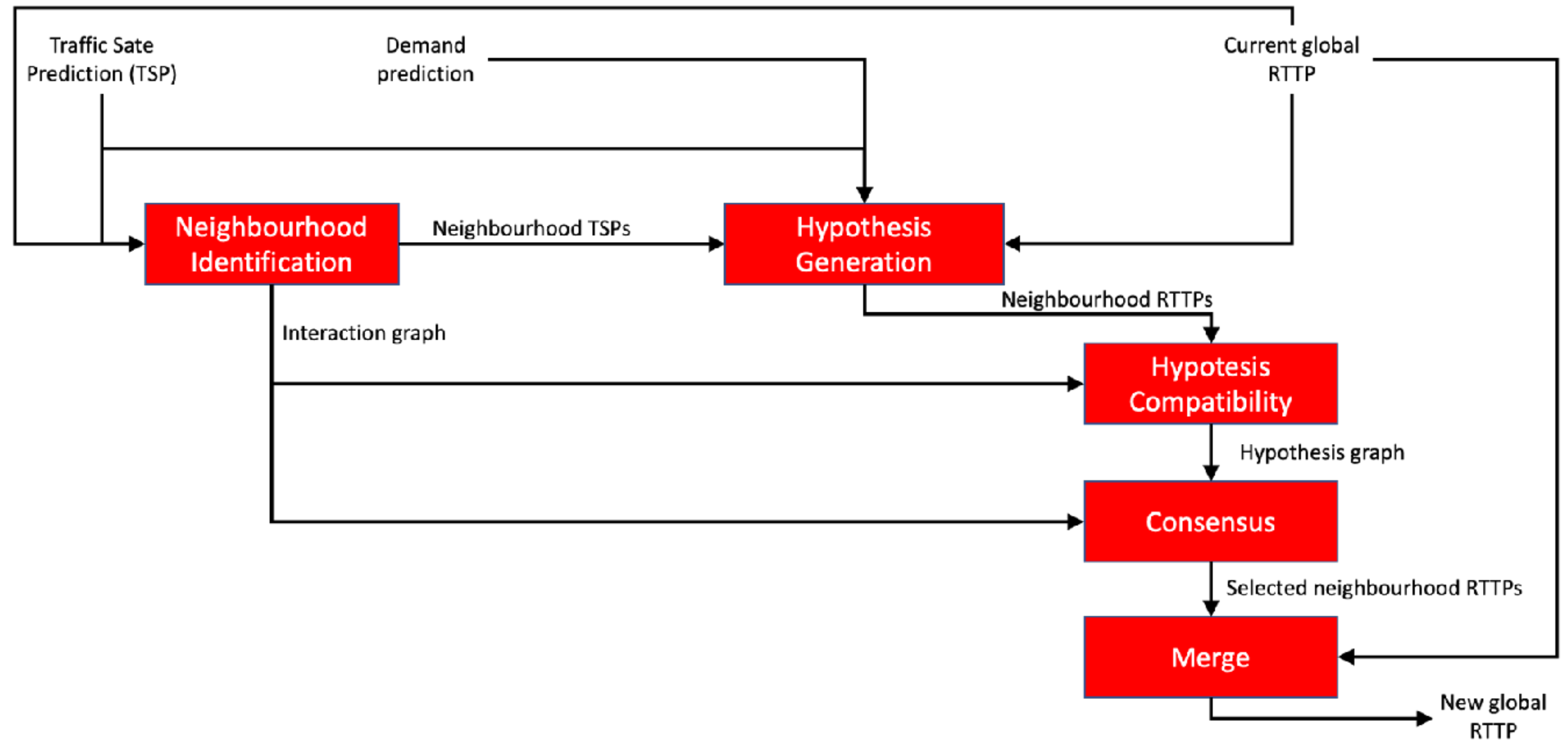


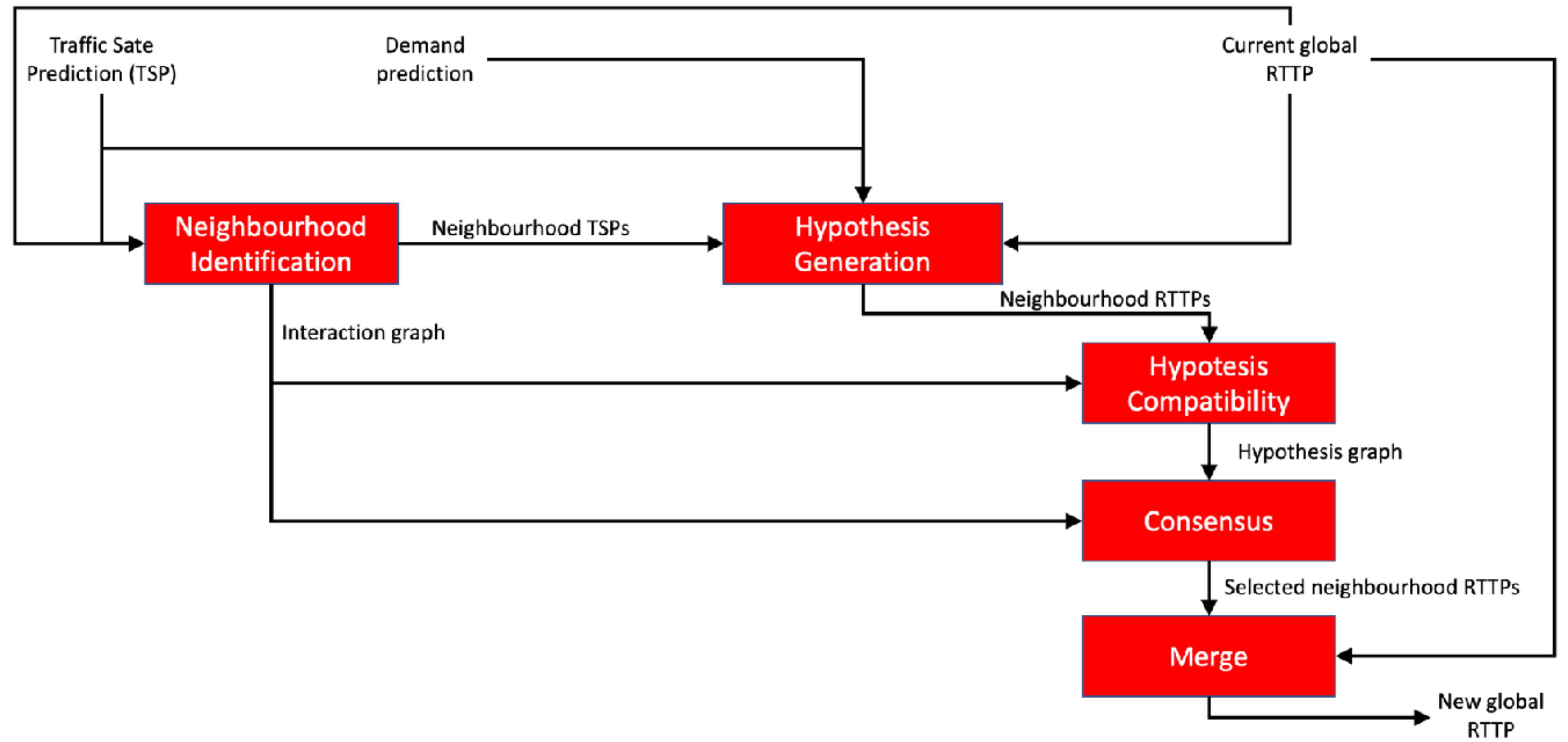
Closed-Loop Traffic Management

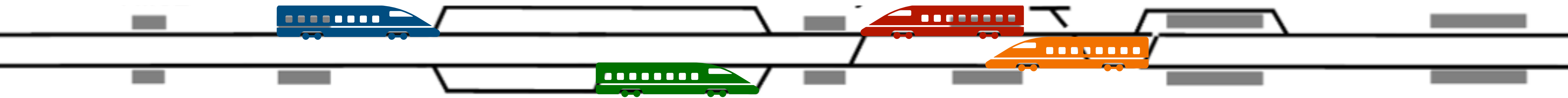
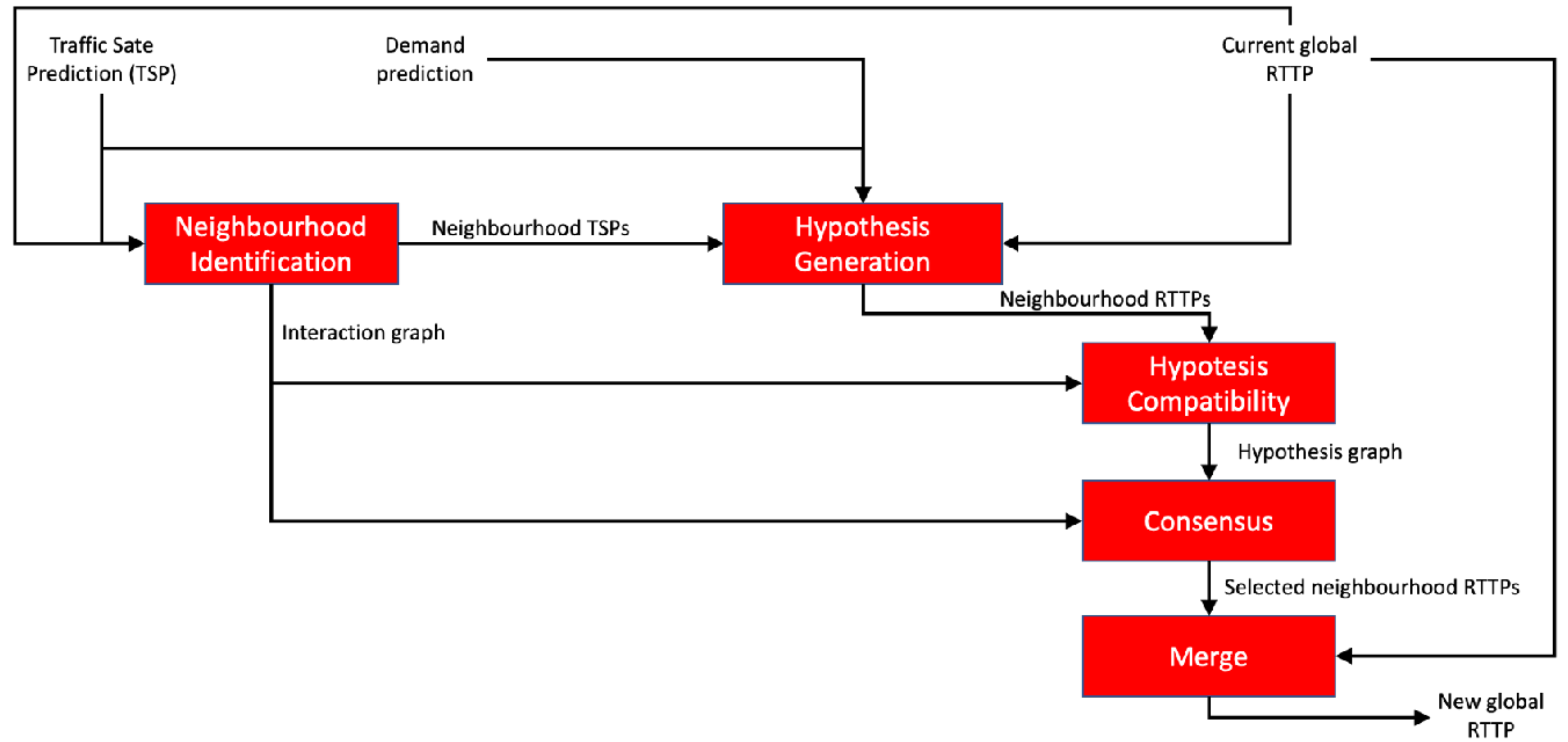


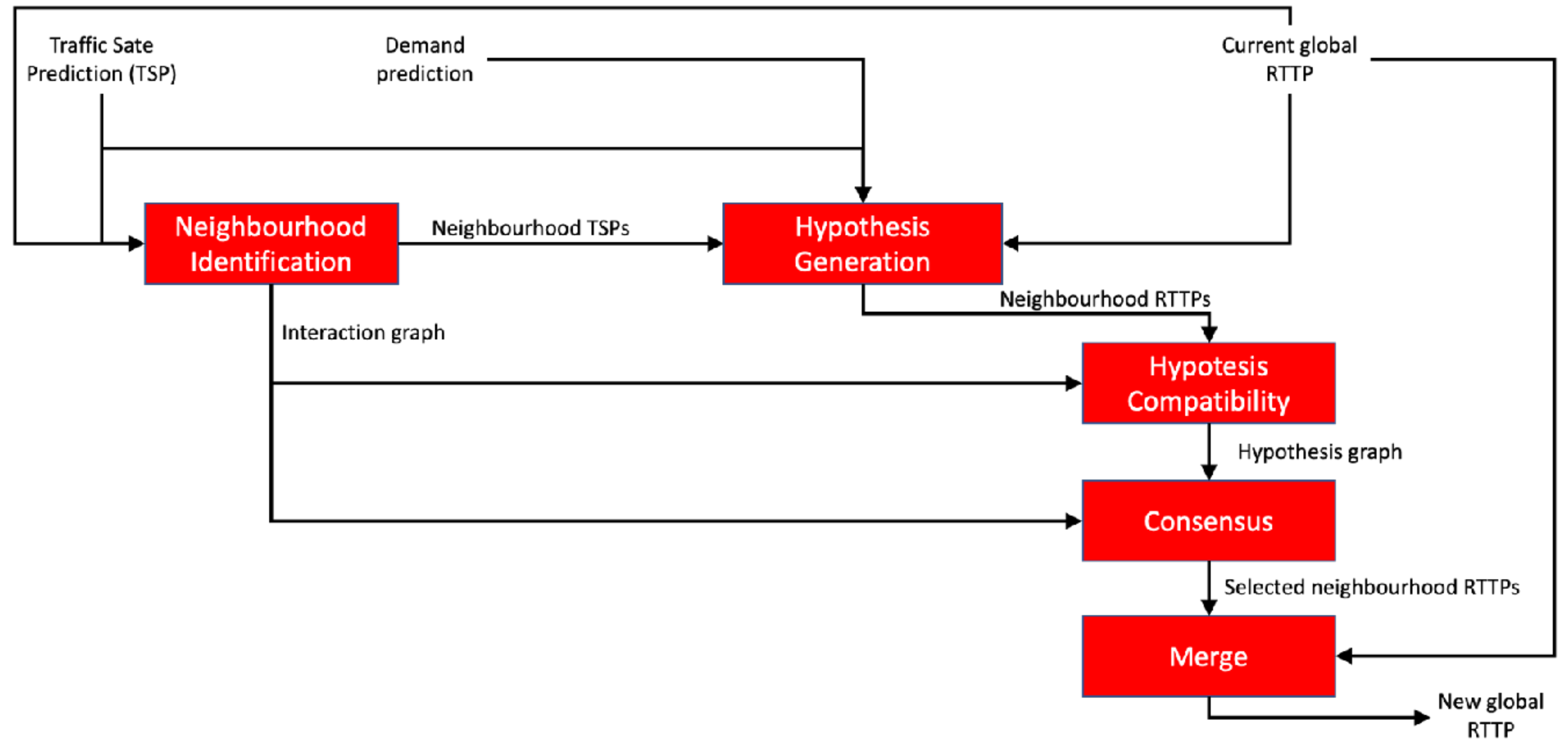
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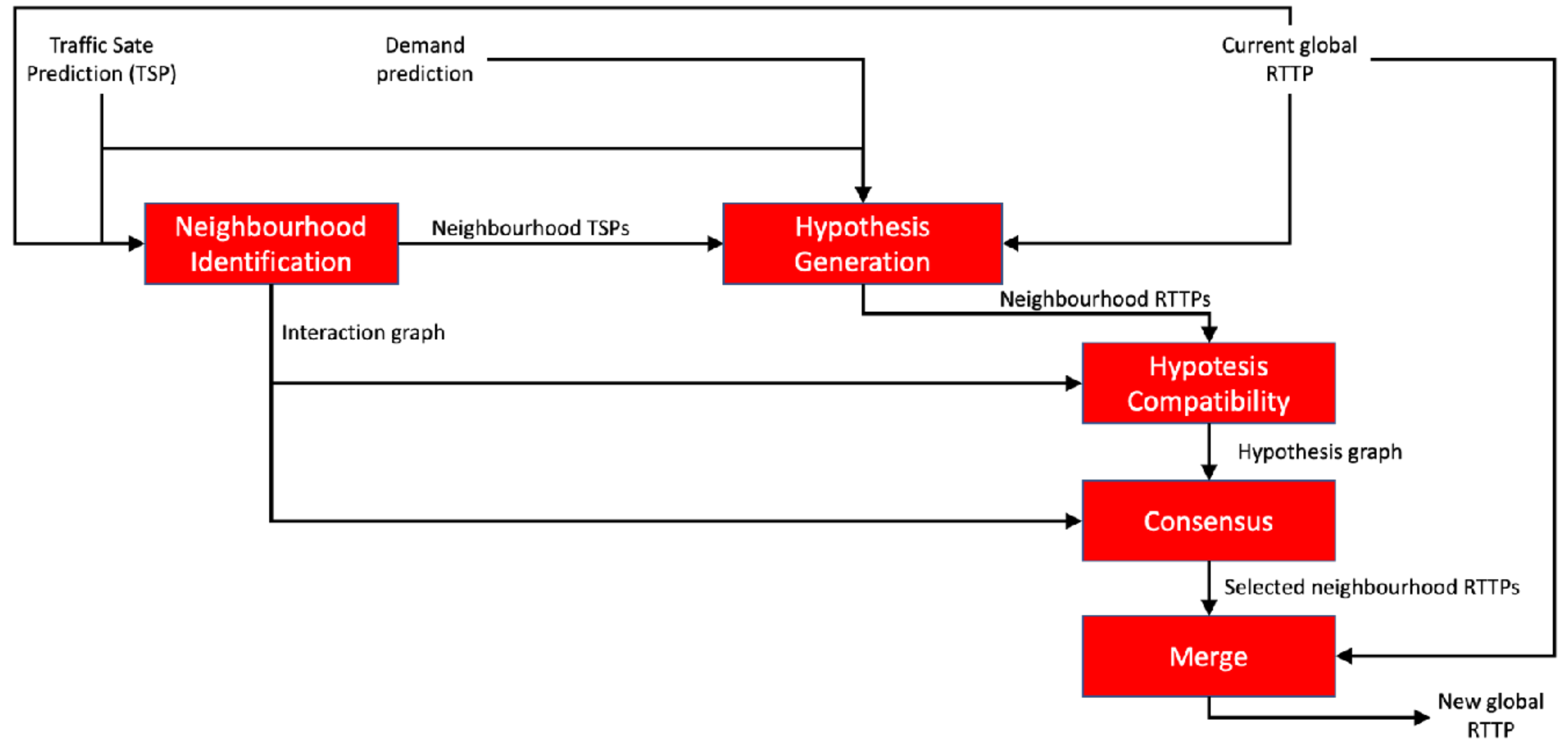




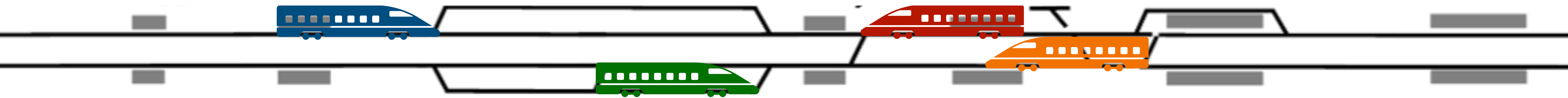
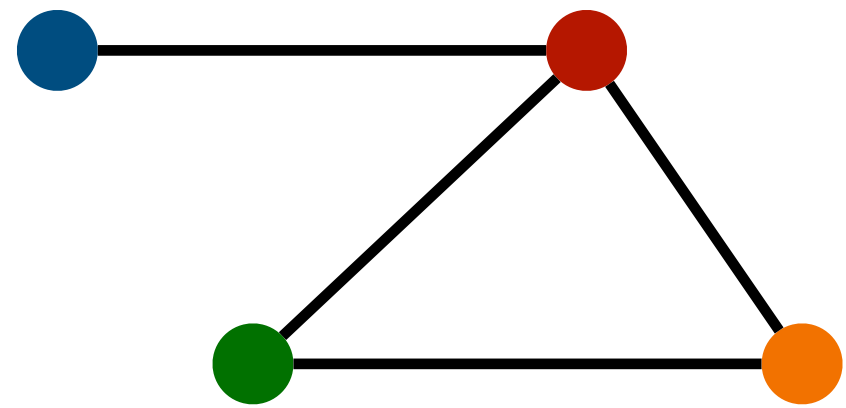


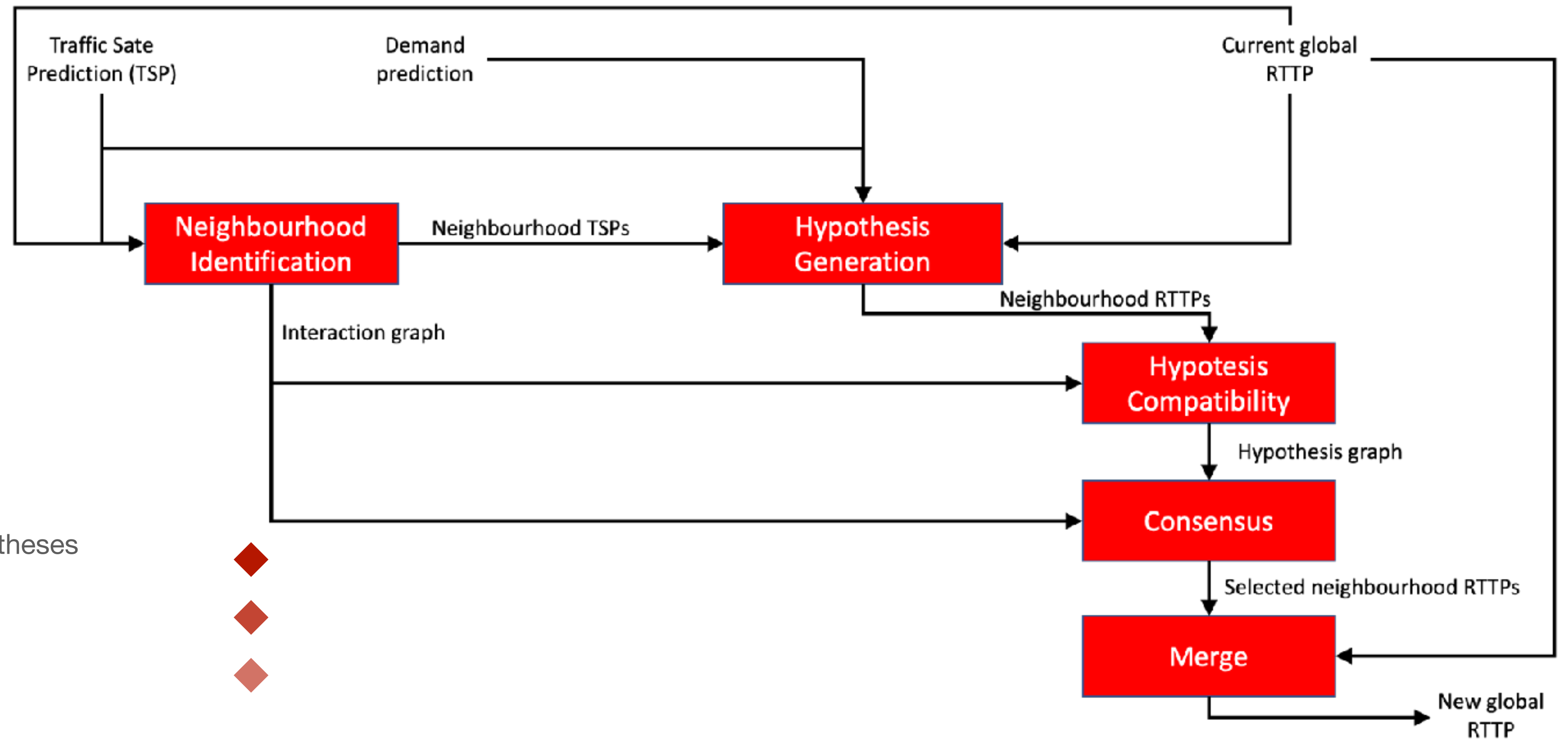




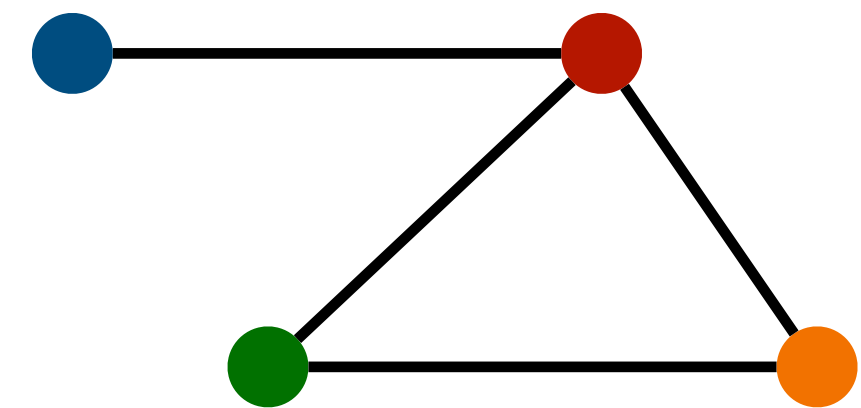


Interaction Graph



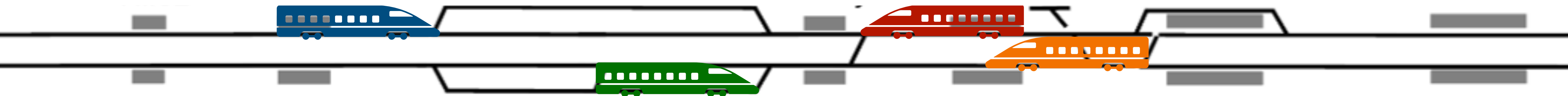
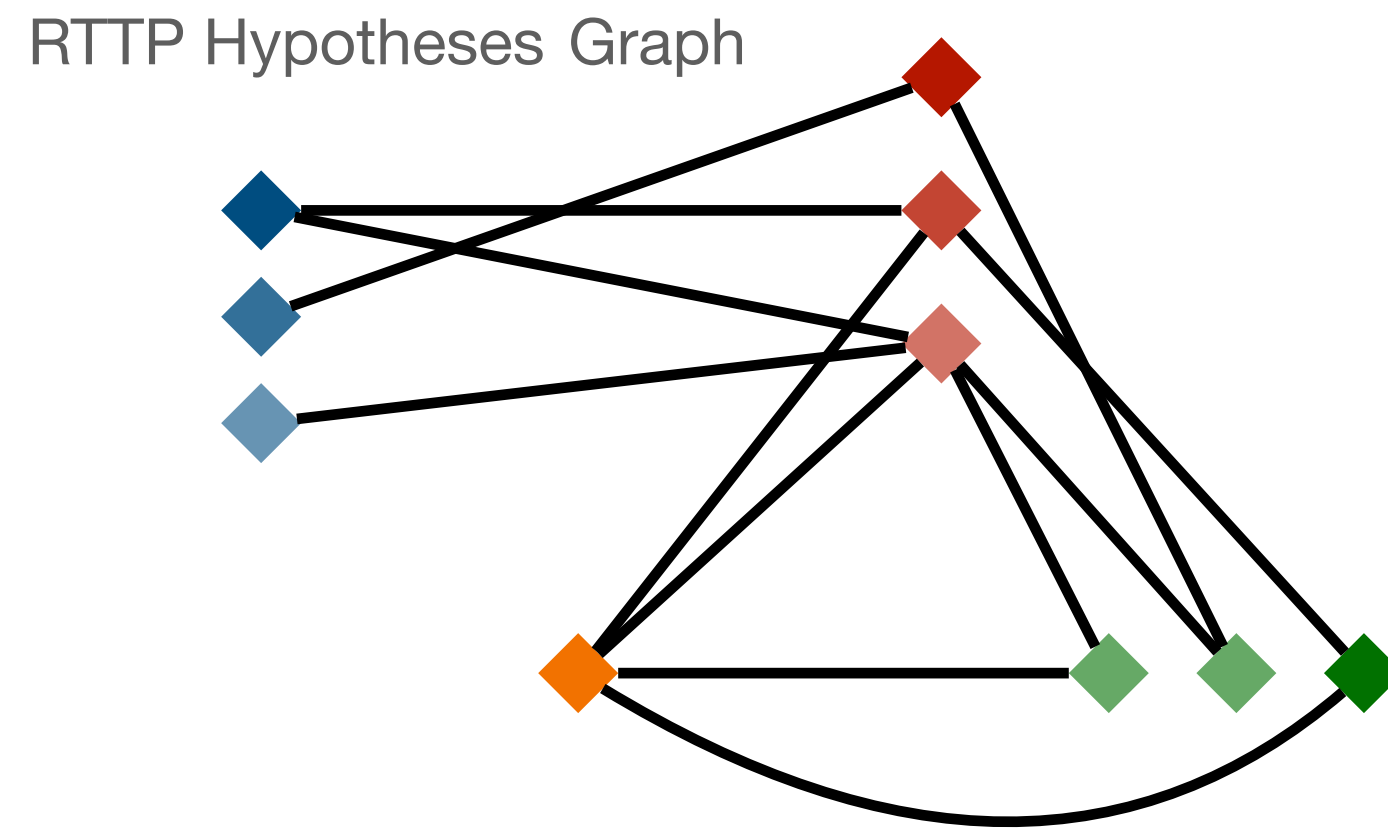
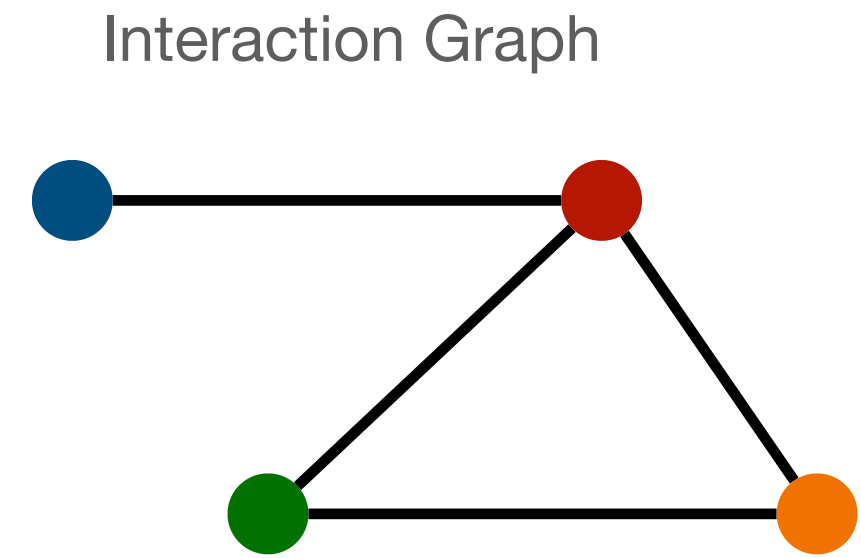
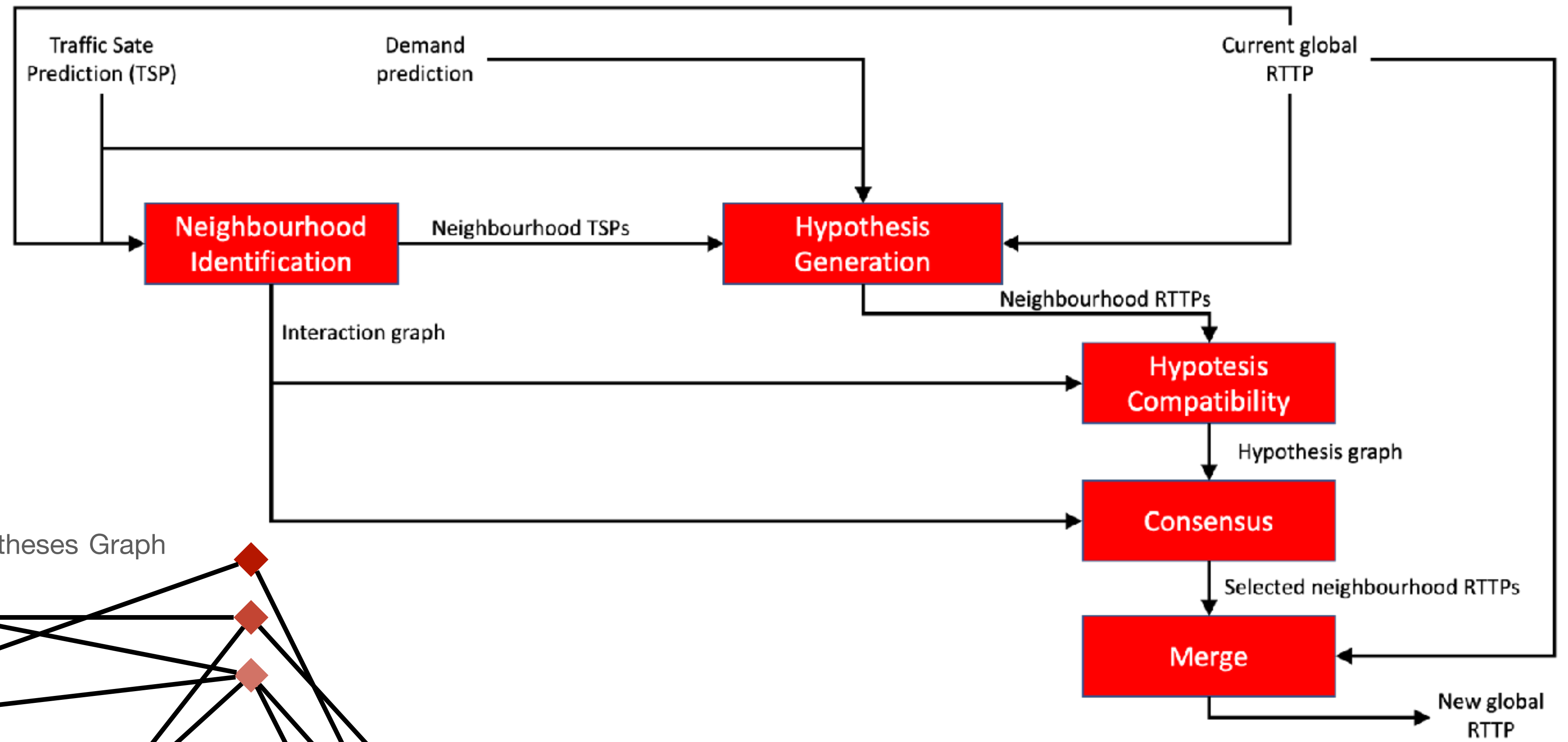


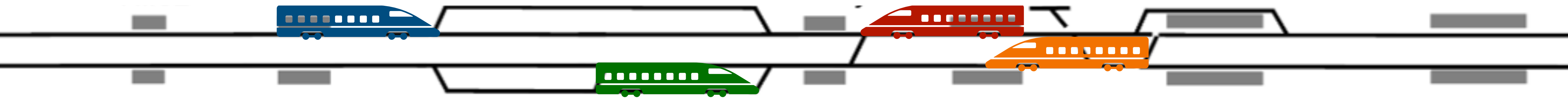
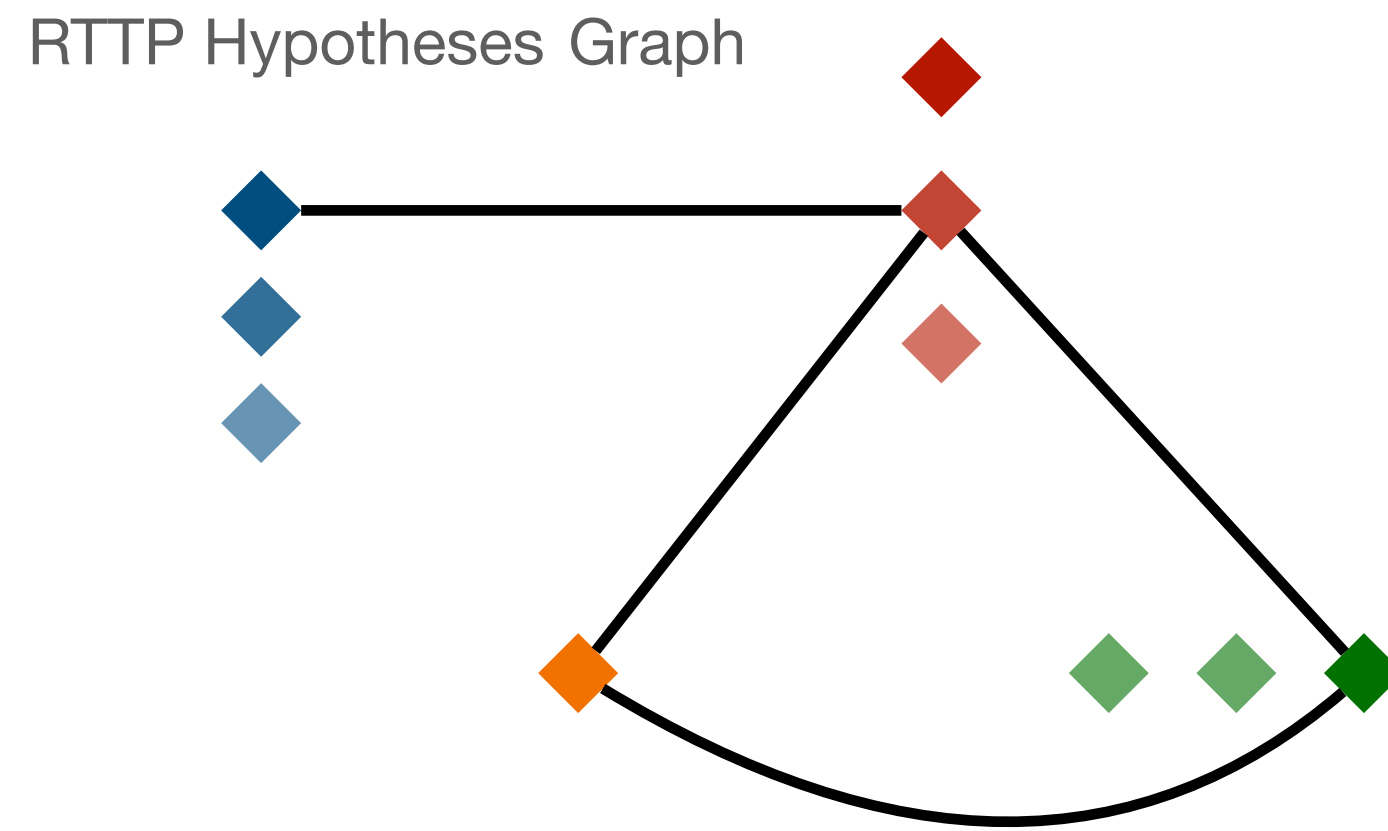
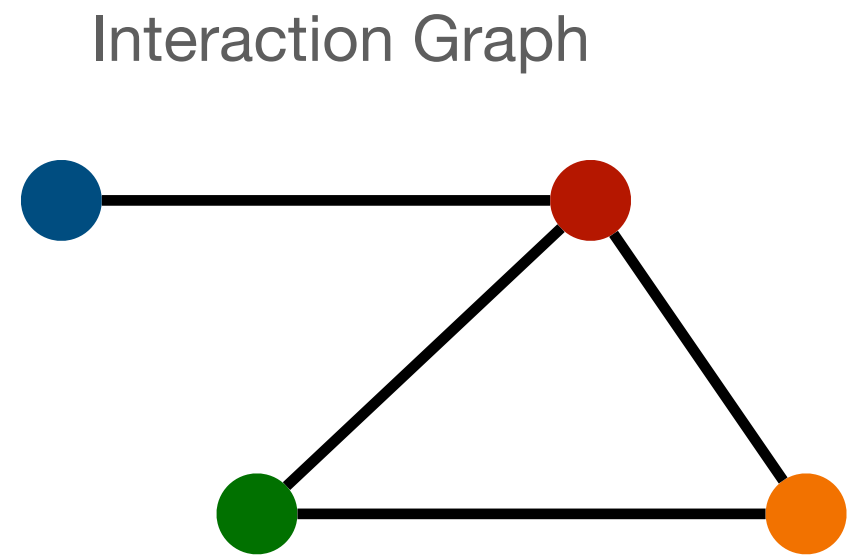
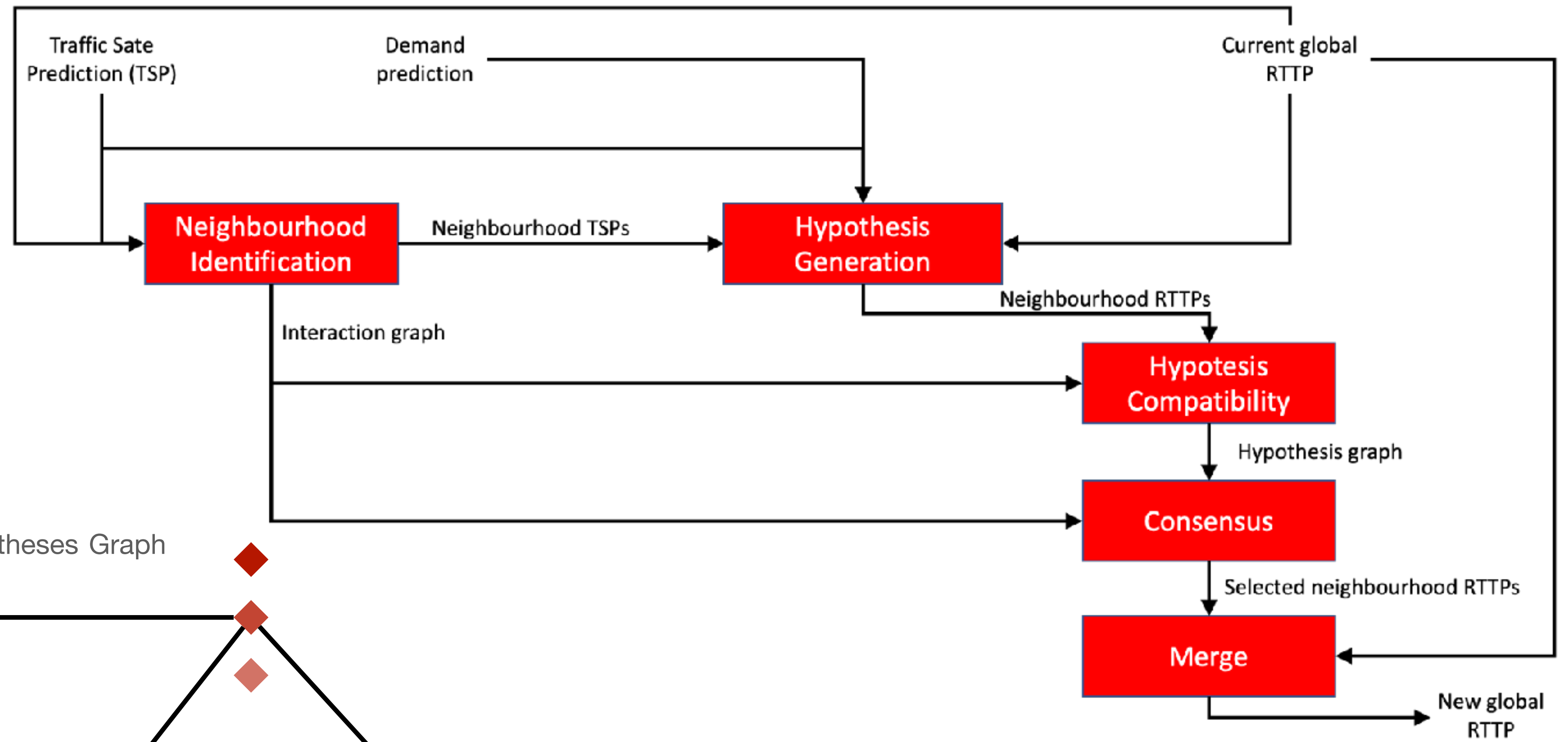
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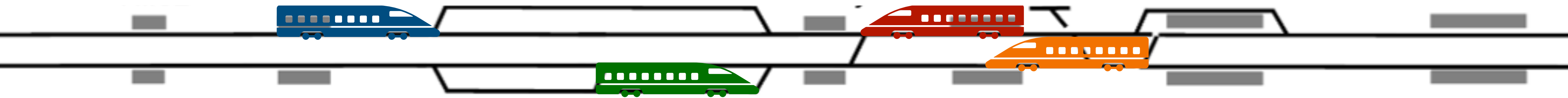
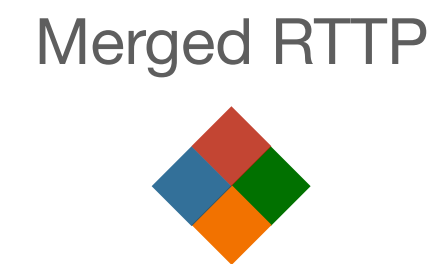
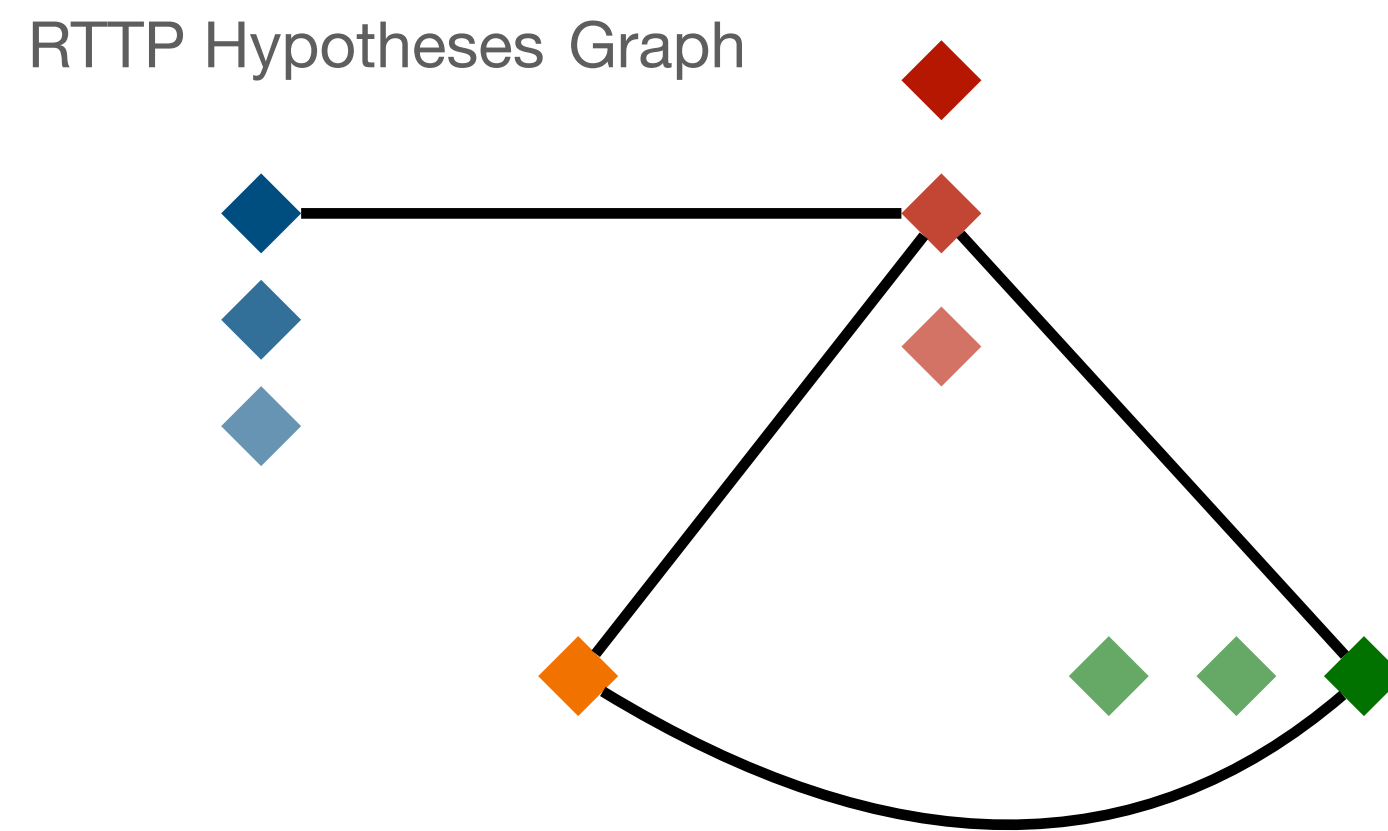
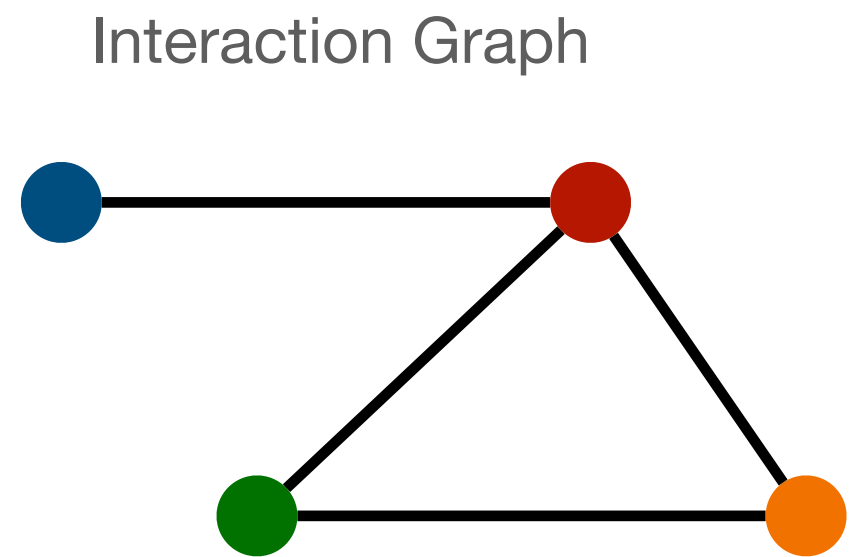
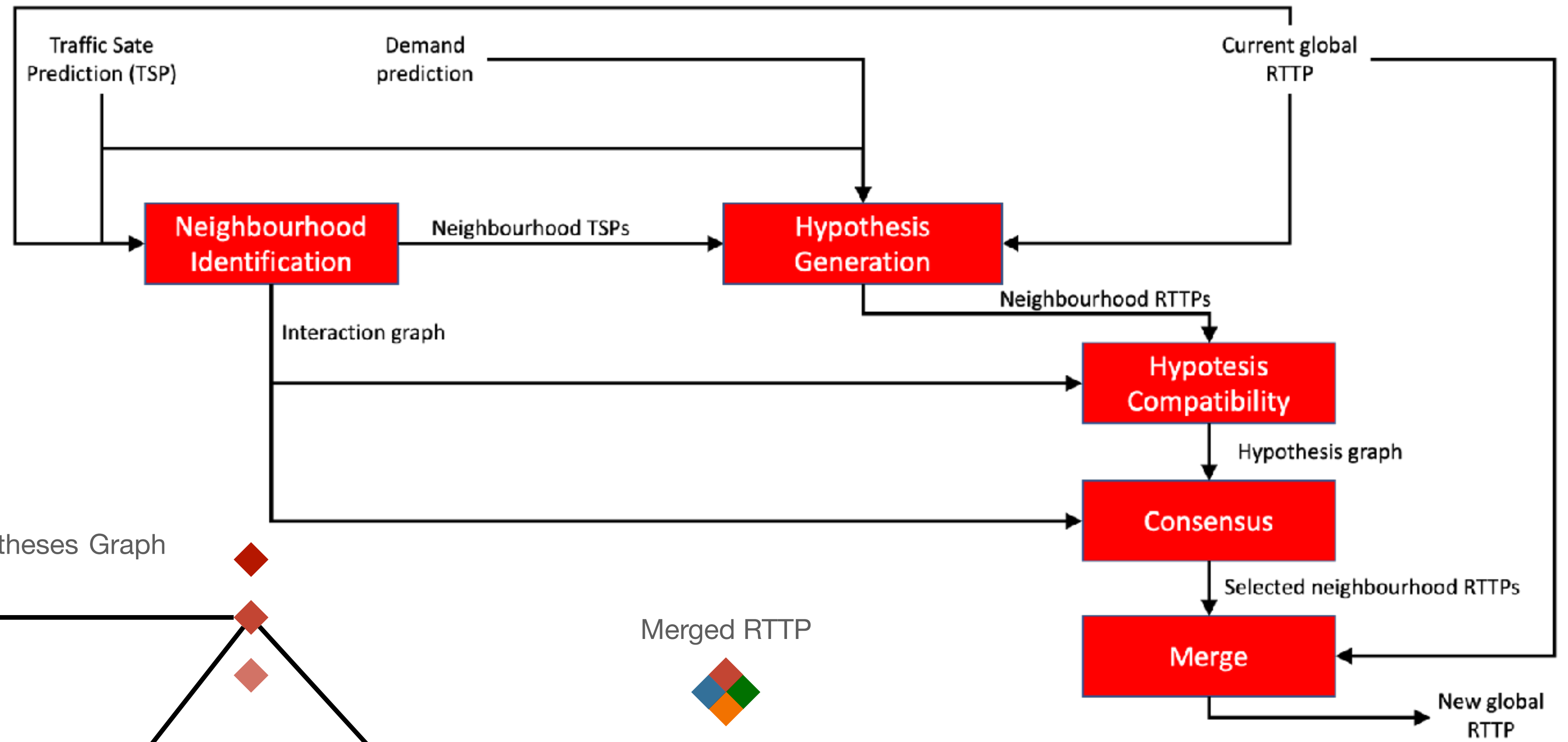


RTP Hypotheses





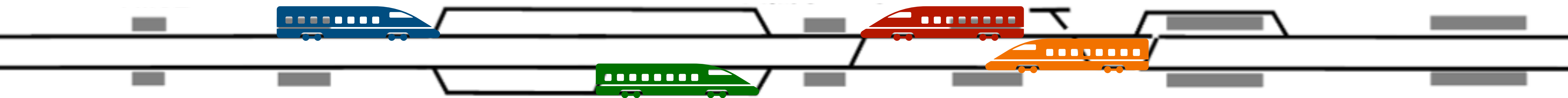




Neighbourhood Selection

detecting interactions among trains

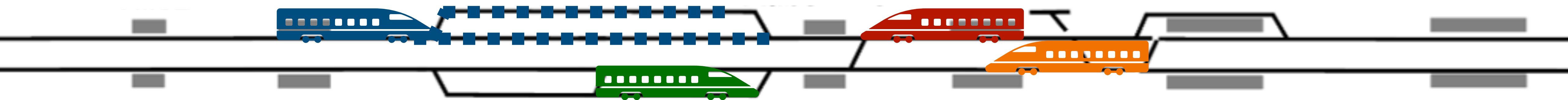
- Two trains interact if they may use the same track within a given time horizon
 - Consider all trains that will be in the area within the time horizon
 - Consider all possible routes of these trains, limited to the time horizon
 - Check for trains that may pass through the same track
- Choosing the best time horizon is crucial



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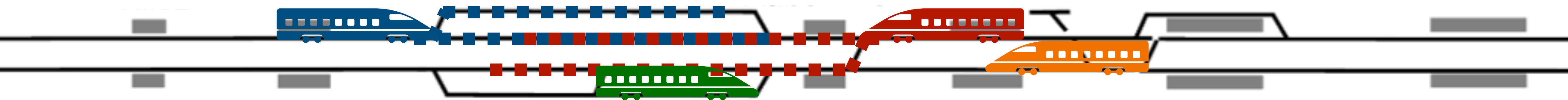
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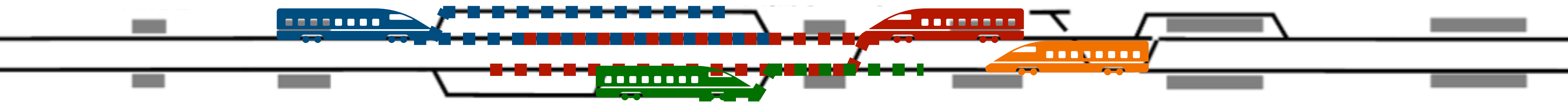
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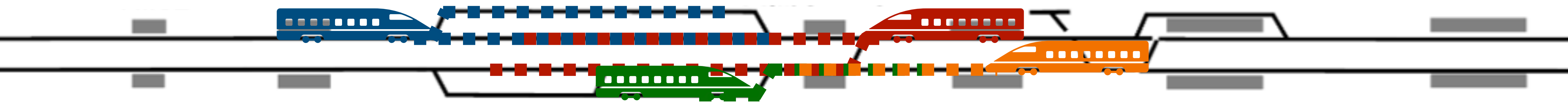
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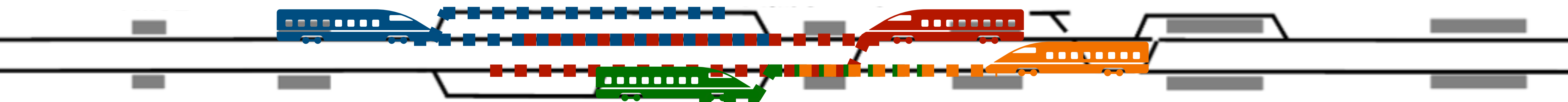
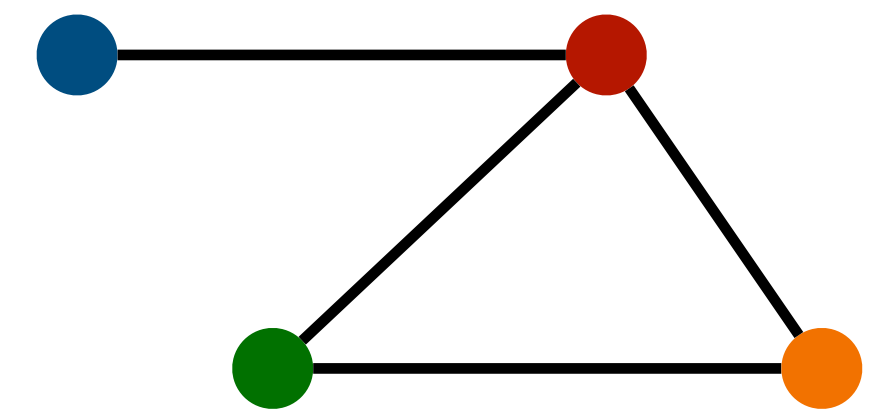


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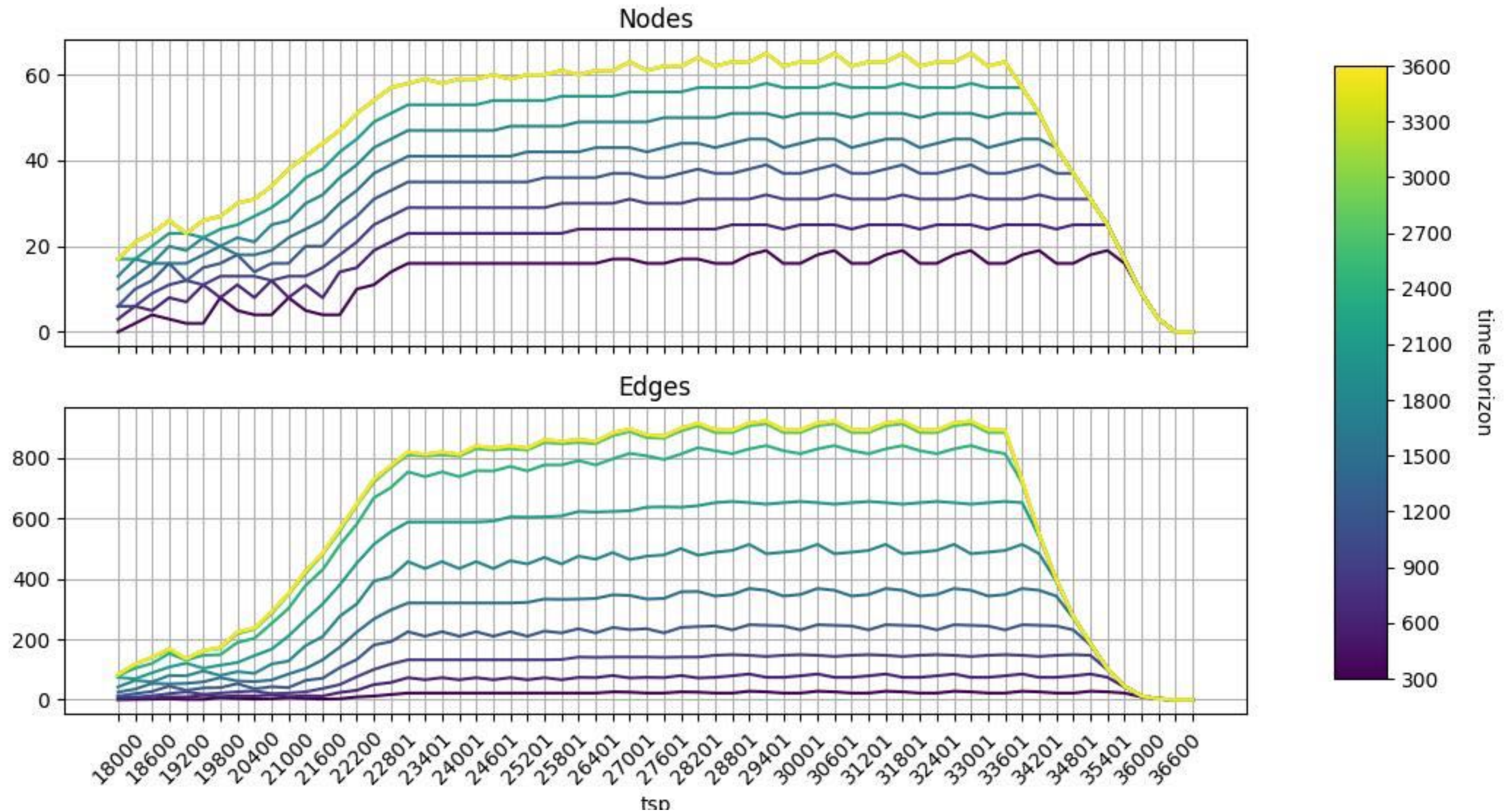
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Interaction Graph



Neighbourhood Selection

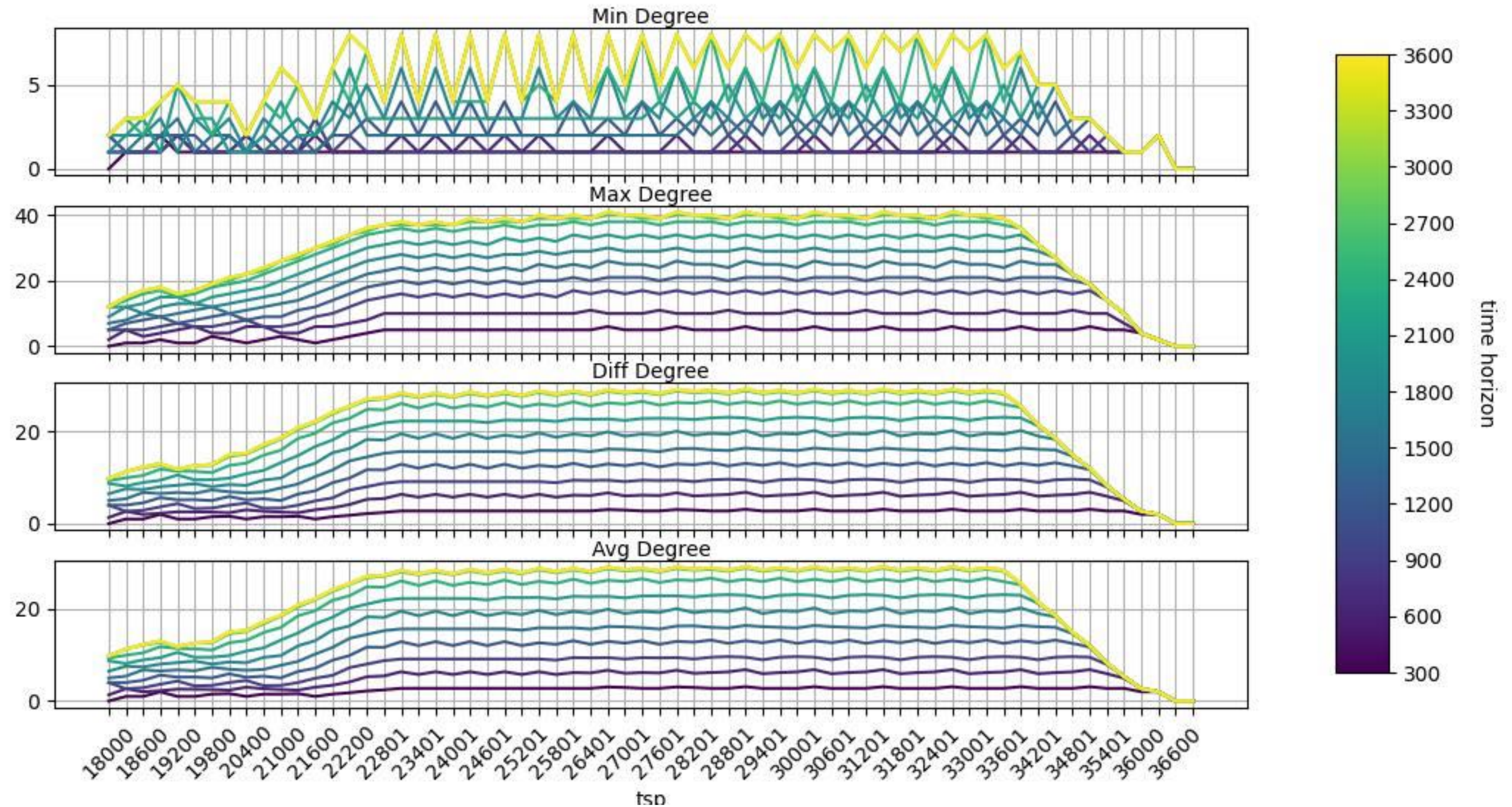
Effects of varying time horizon





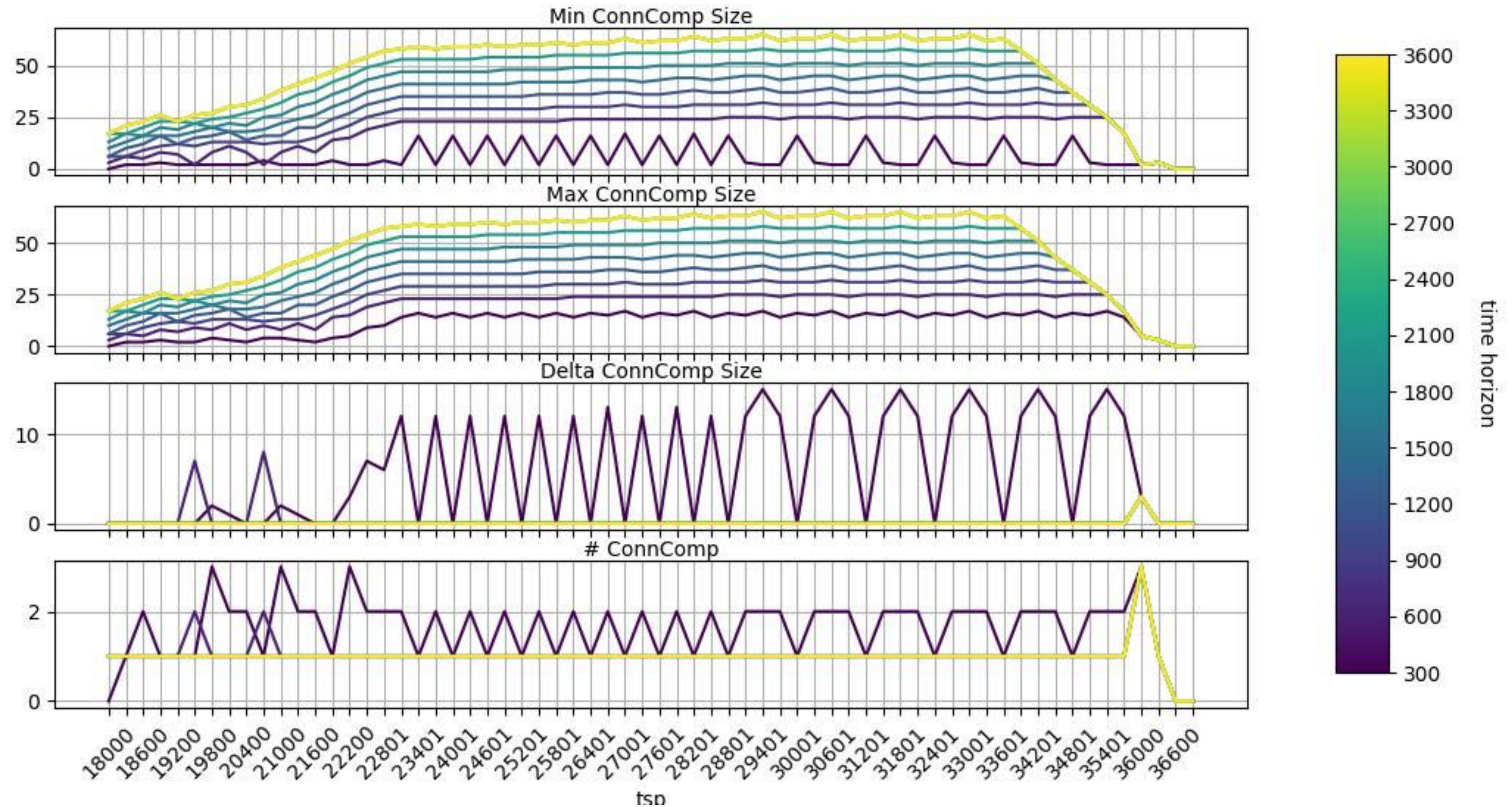
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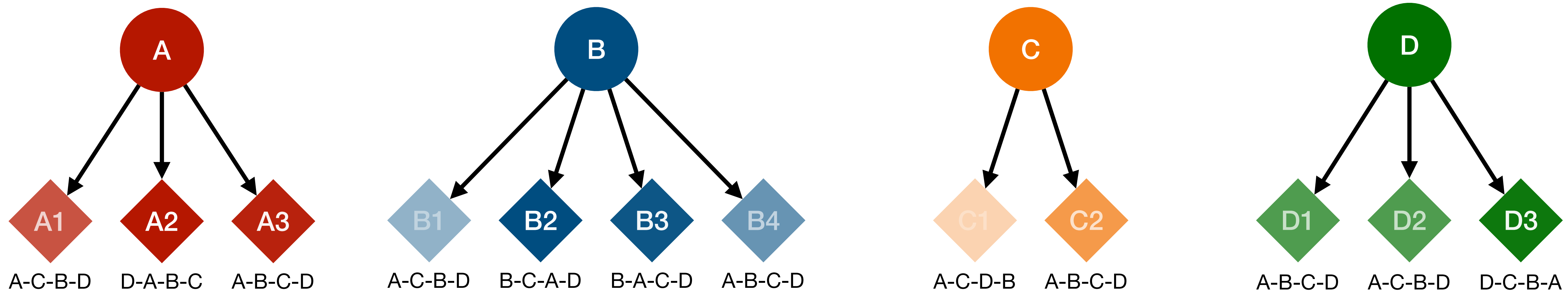
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Hypothesis Generation

producing alternative solutions

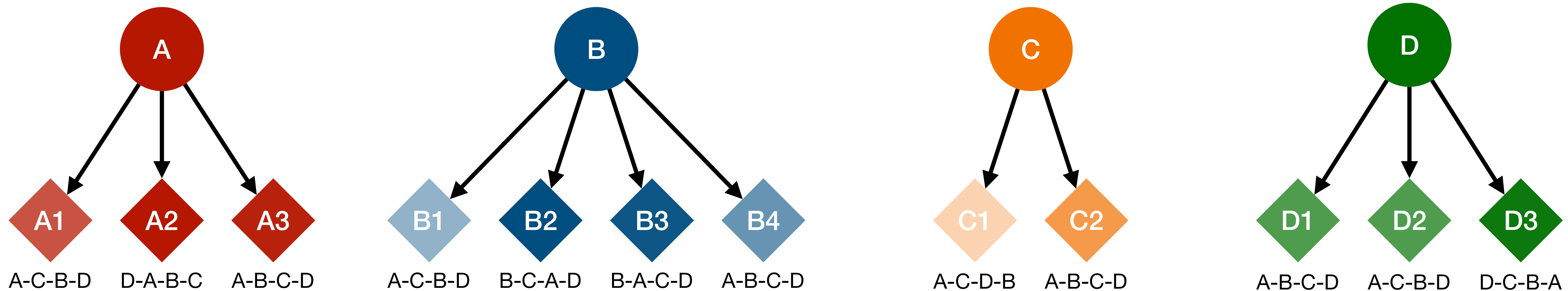
- Each train independently solves a local traffic management problem exploiting a custom version of RECIFE-MILP
 - The focal train may weight differently from other trains in the objective function
 - Passenger demand is considered at this stage
 - Only trains belonging to the neighbourhood are optimised



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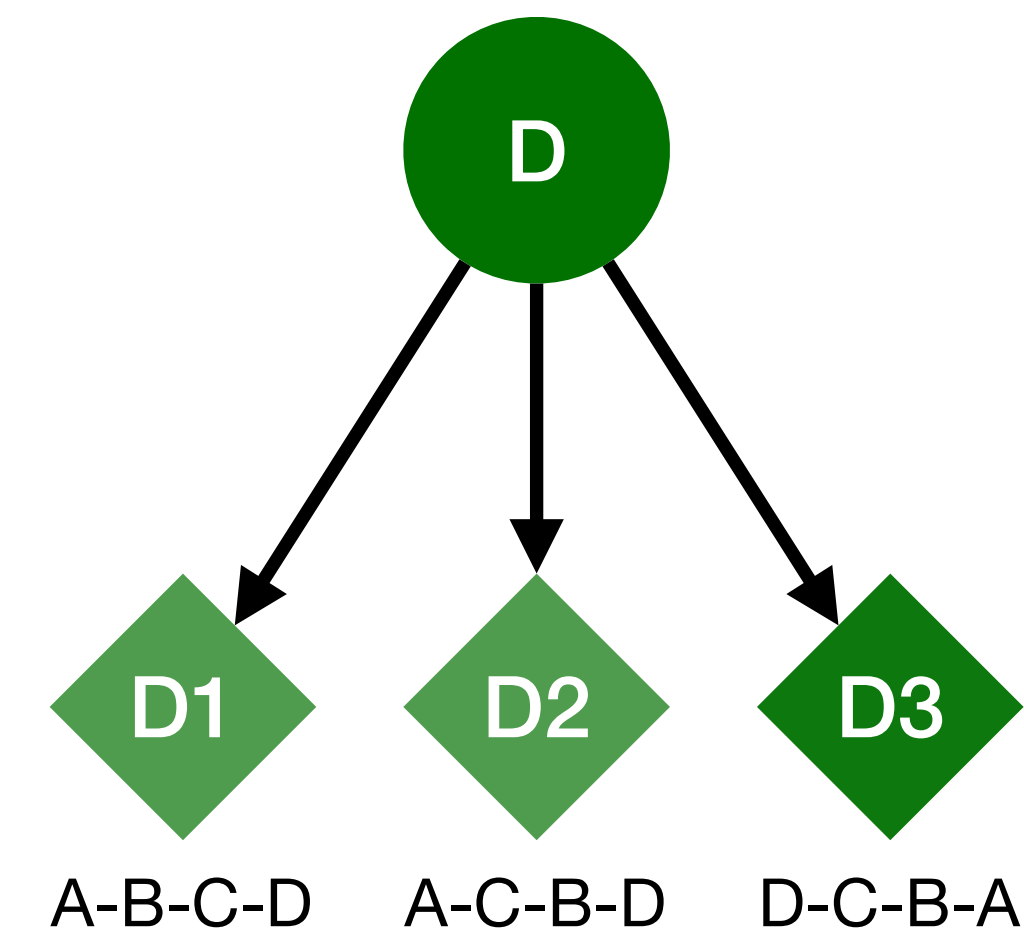
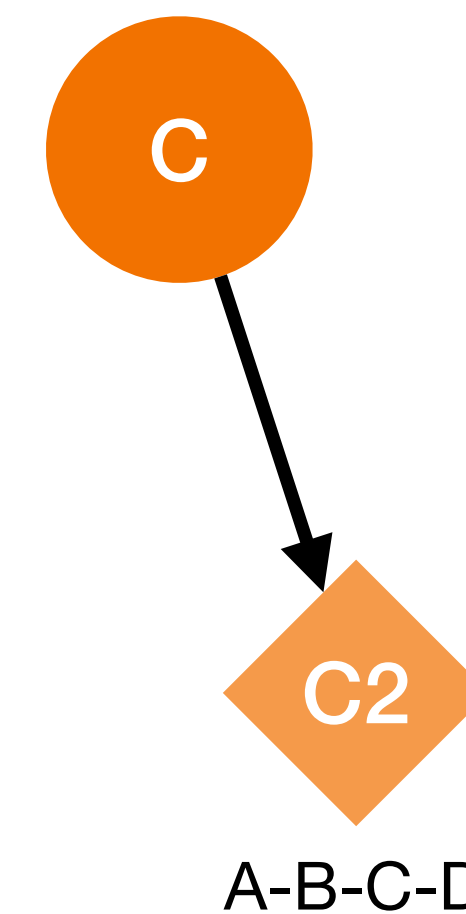
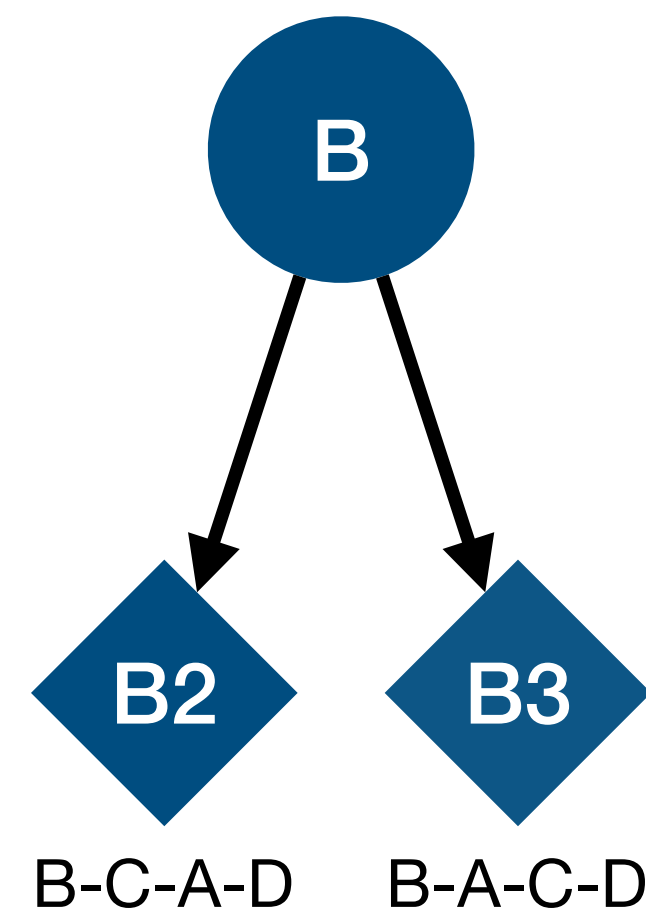
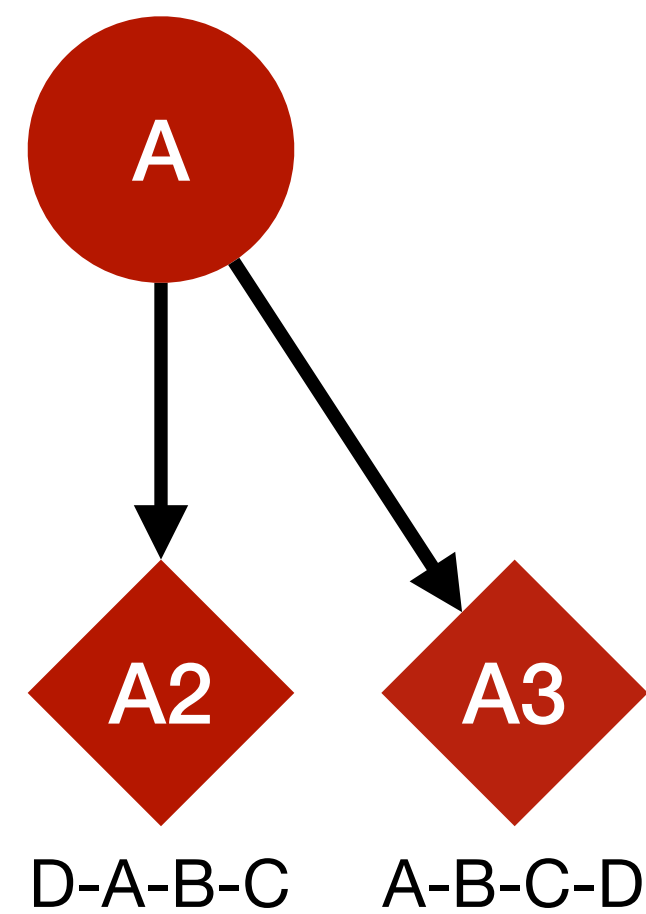
- One or more different solution hypotheses are produced
 - Retain only solutions within a certain margin from the optimal one
 - Retain only a maximum number of solutions
- Individual hypotheses can be shared within a neighbourhood



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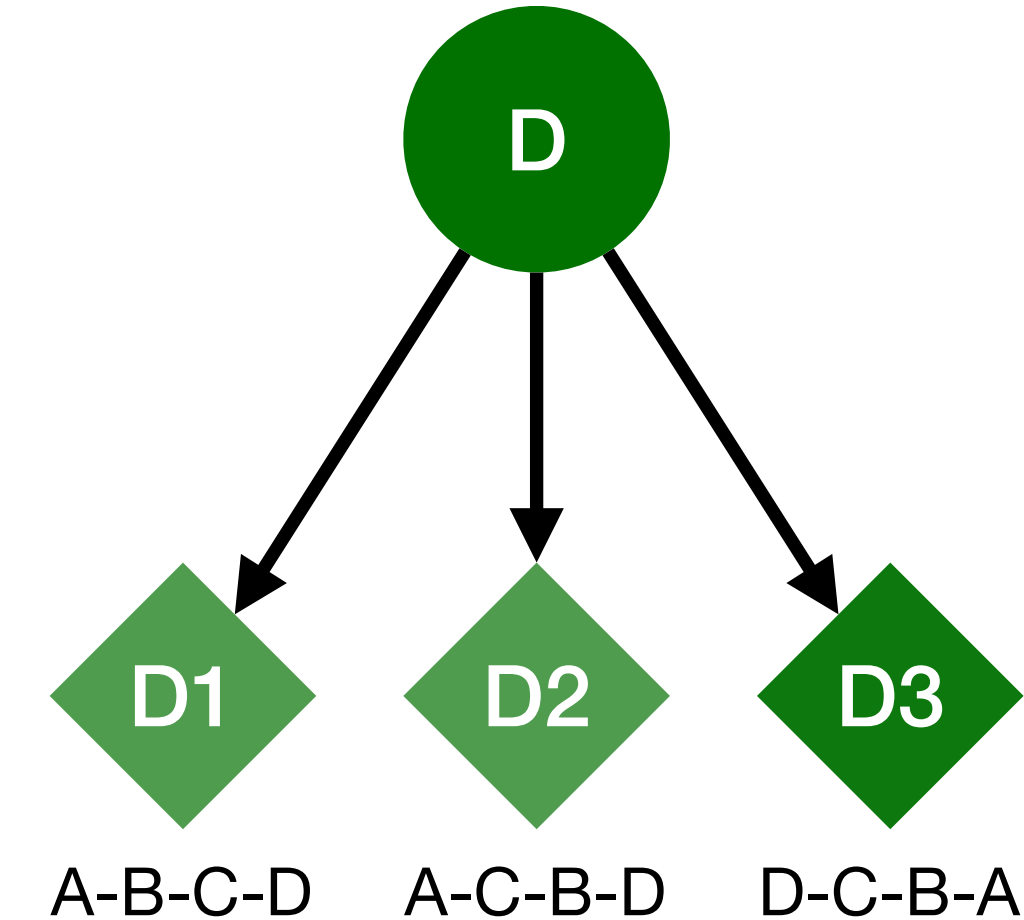
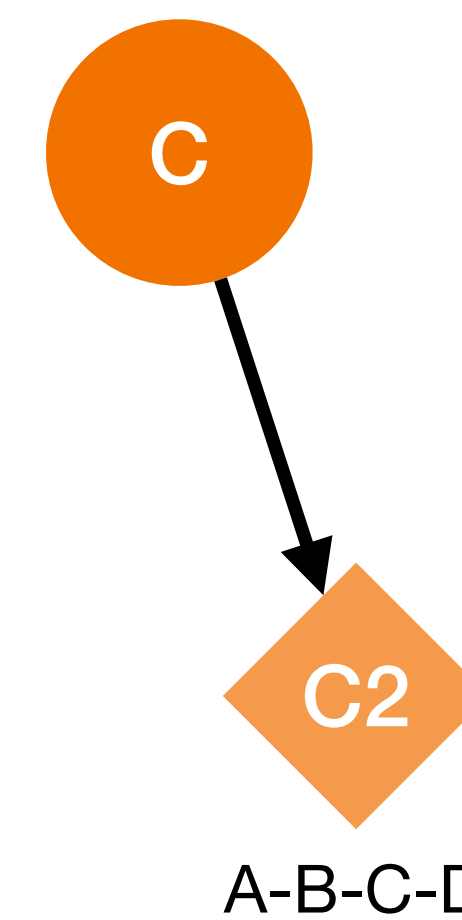
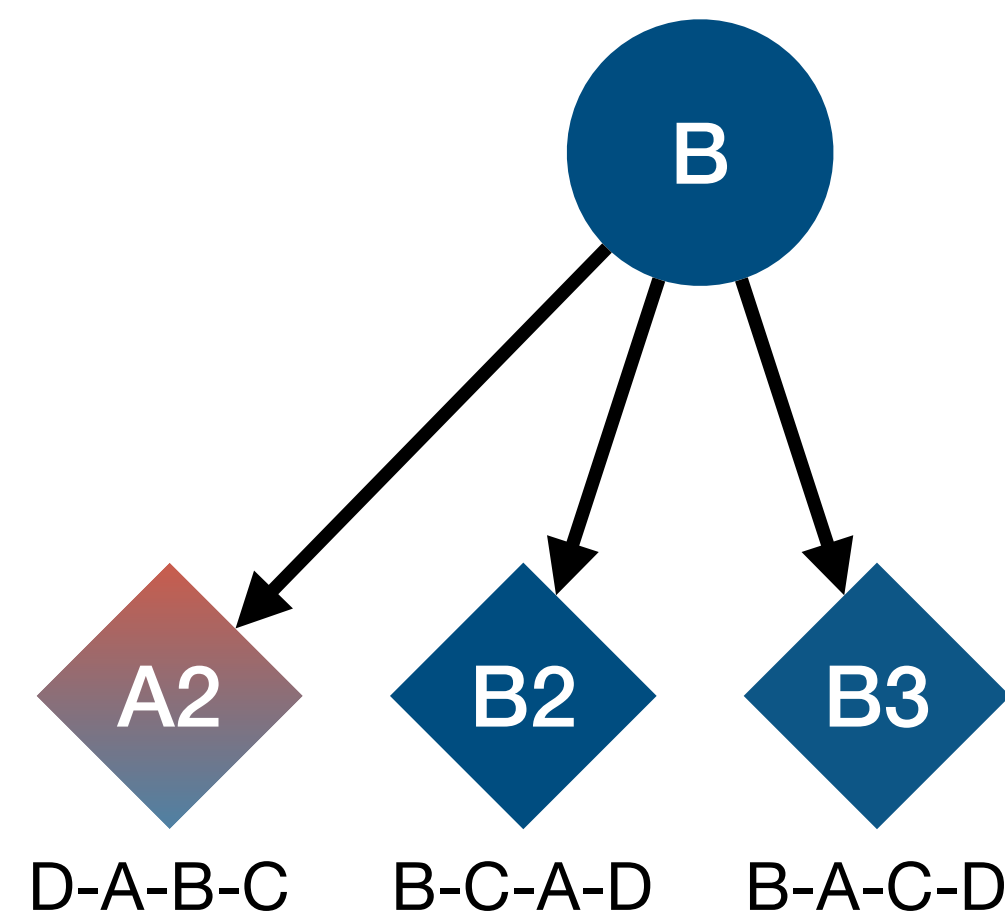
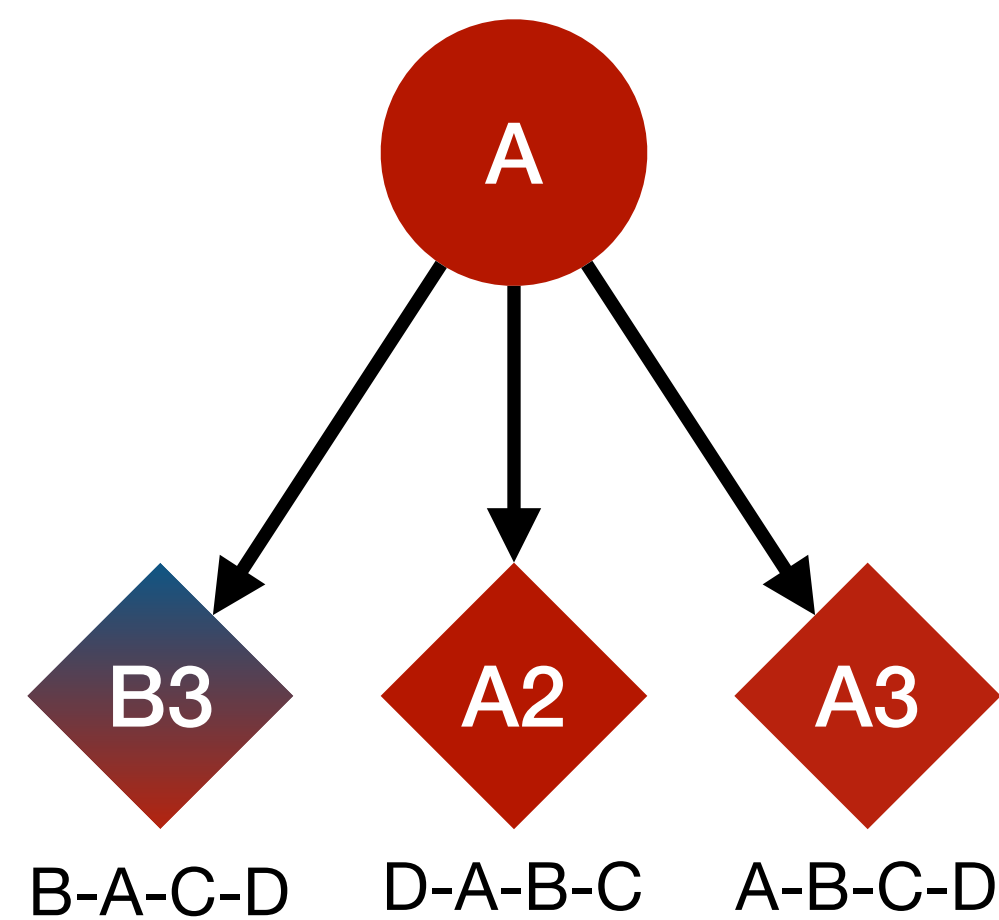
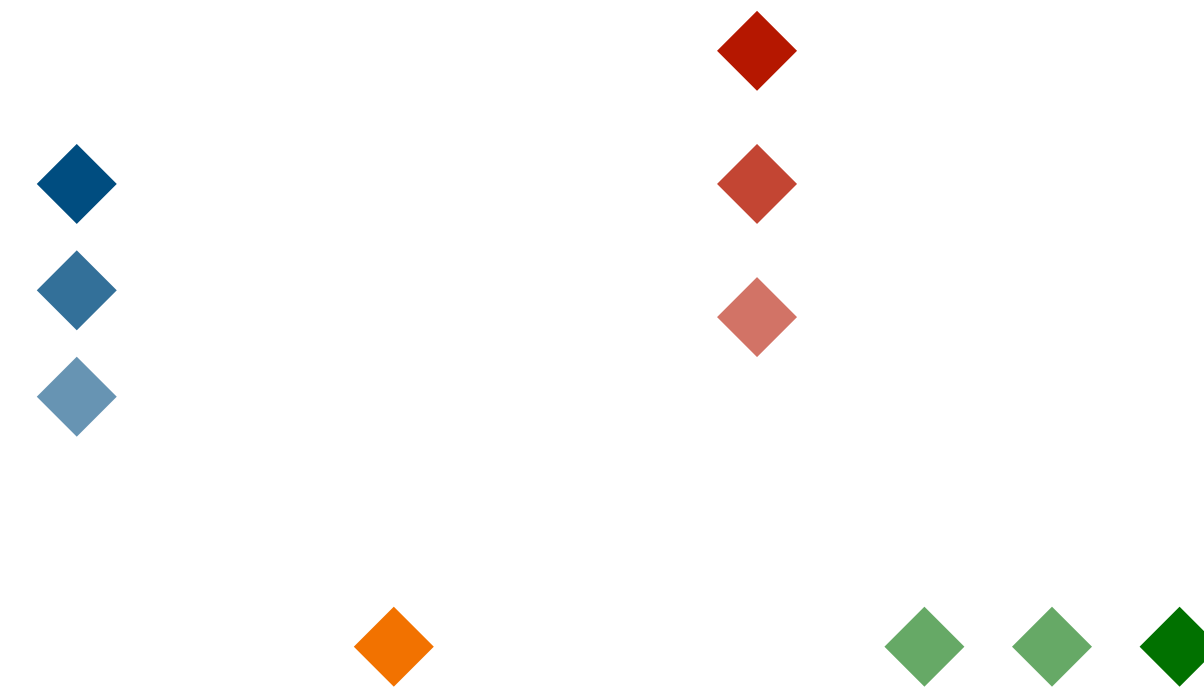
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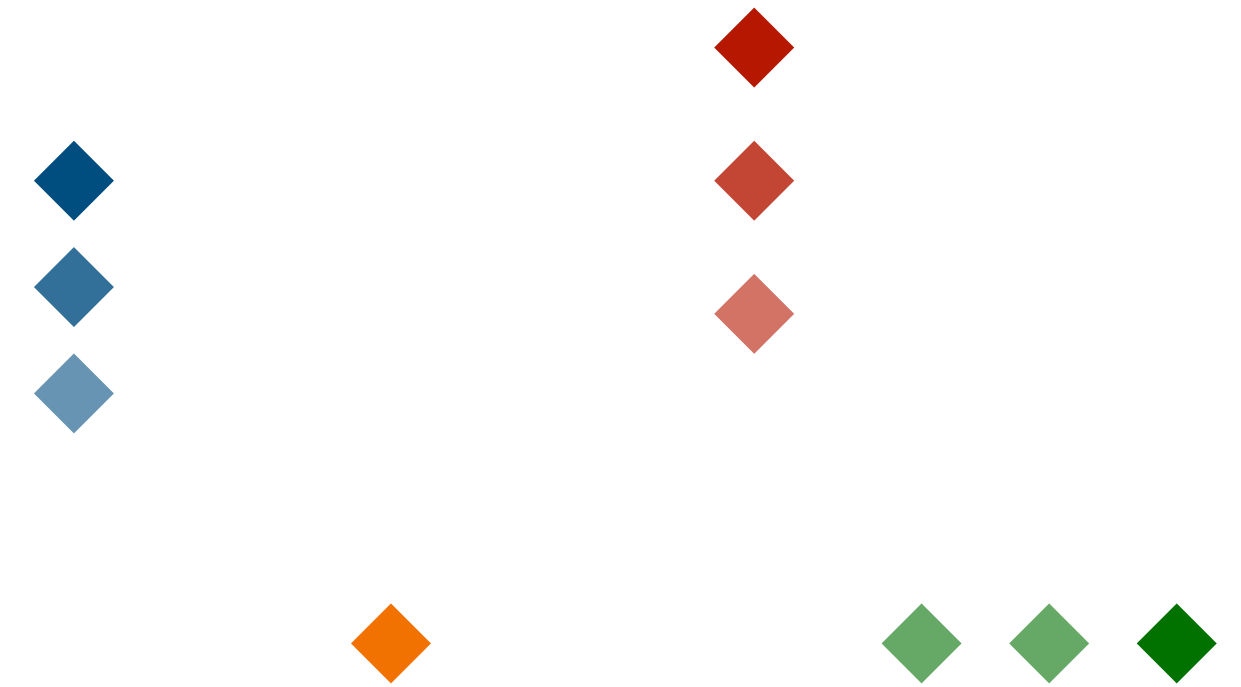
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Hypothesis Compatibility

can different get merged without conflicts?

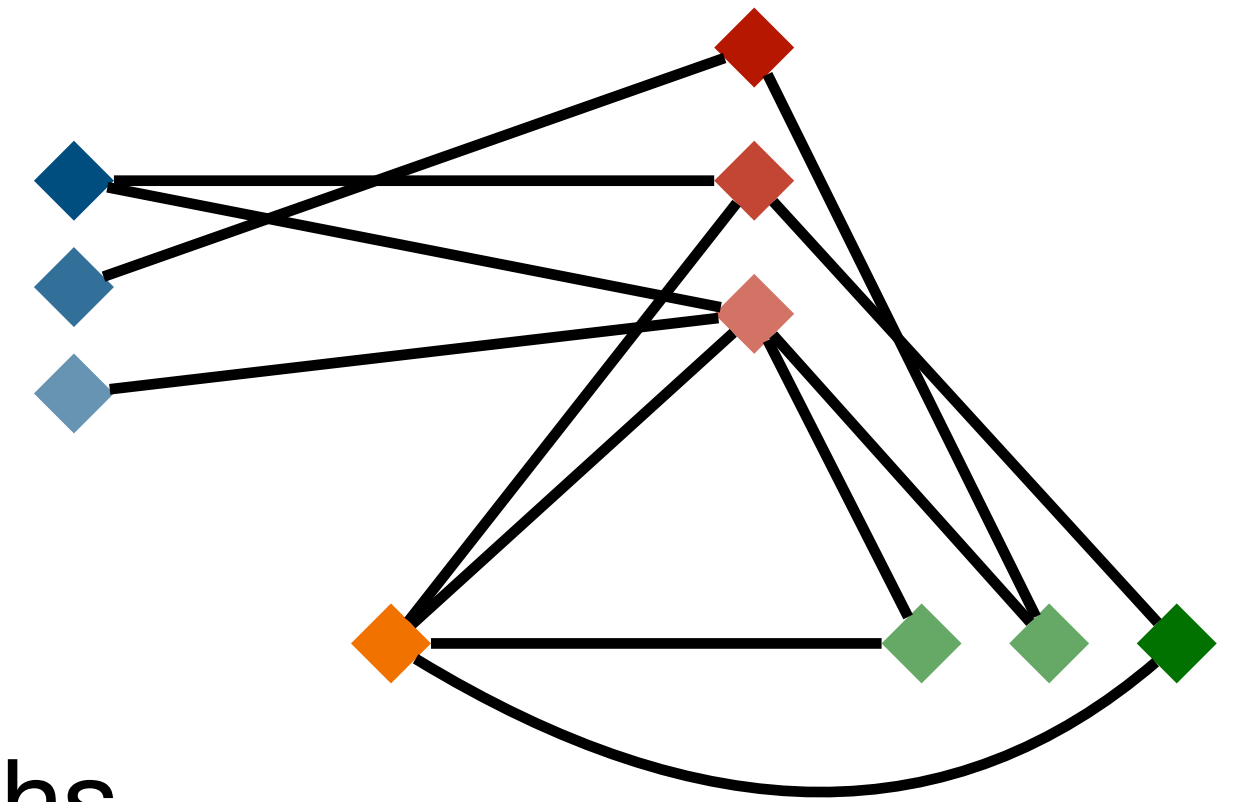
- Pairwise evaluation of hypothesis of different trains
- Compatibility strength: how strict is the evaluation?
 - **Strong compatibility:** require that all trains have compatible paths
 - **Weak compatibility:** require that focal trains have compatible paths
- Compatibility outcome: how to create the hypothesis graph?
 - **Binary compatibility:** graph edges exists or not
 - **Continuous compatibility:** weighted edges in the graph
- Current solution: **weak** and **binary**



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Consensus

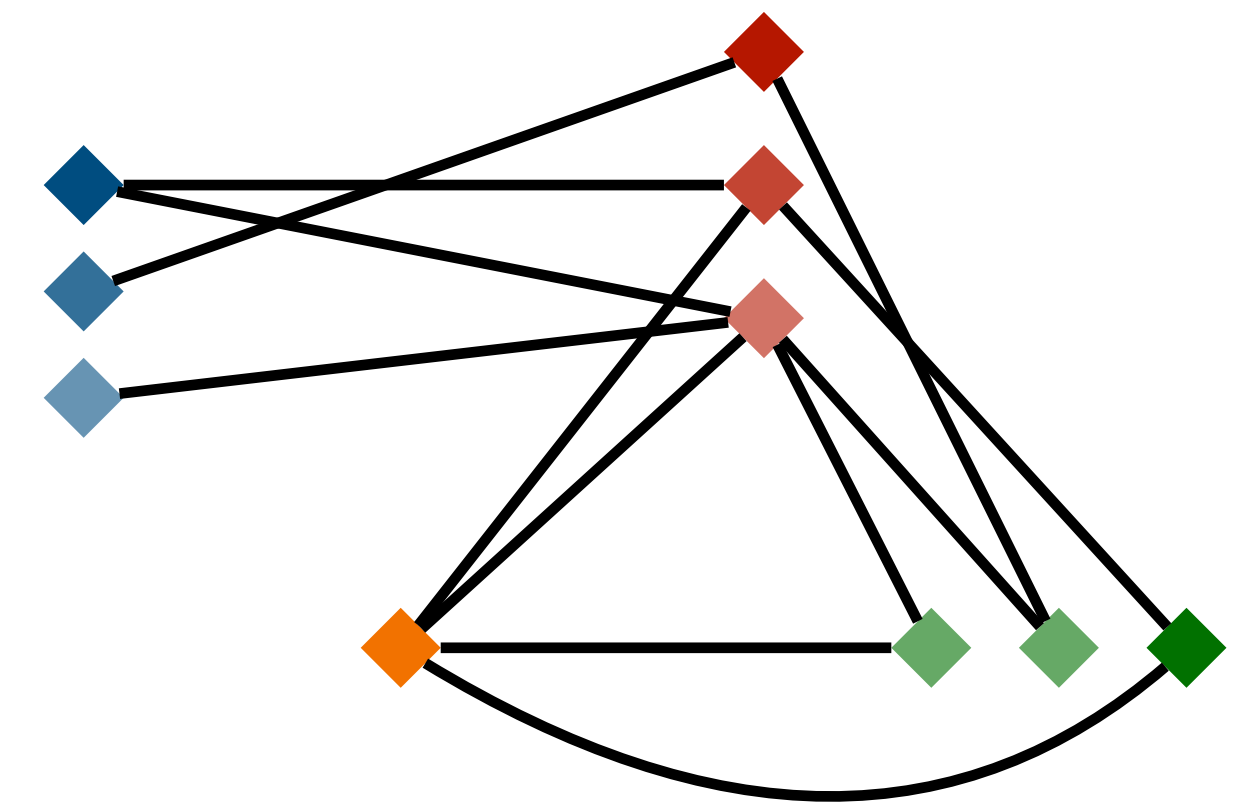
how to select the best set of hypotheses?

- Given knowledge about hypothesis compatibility, find a global solution
 - each train selects an hypothesis compatible with the one of the neighbours
 - each train selects an hypothesis to optimise the objective function
- A solution is a subgraph of the hypothesis graph
- SORTEDMOBILITY:
Decentralised consensus process based on voter models

Consensus

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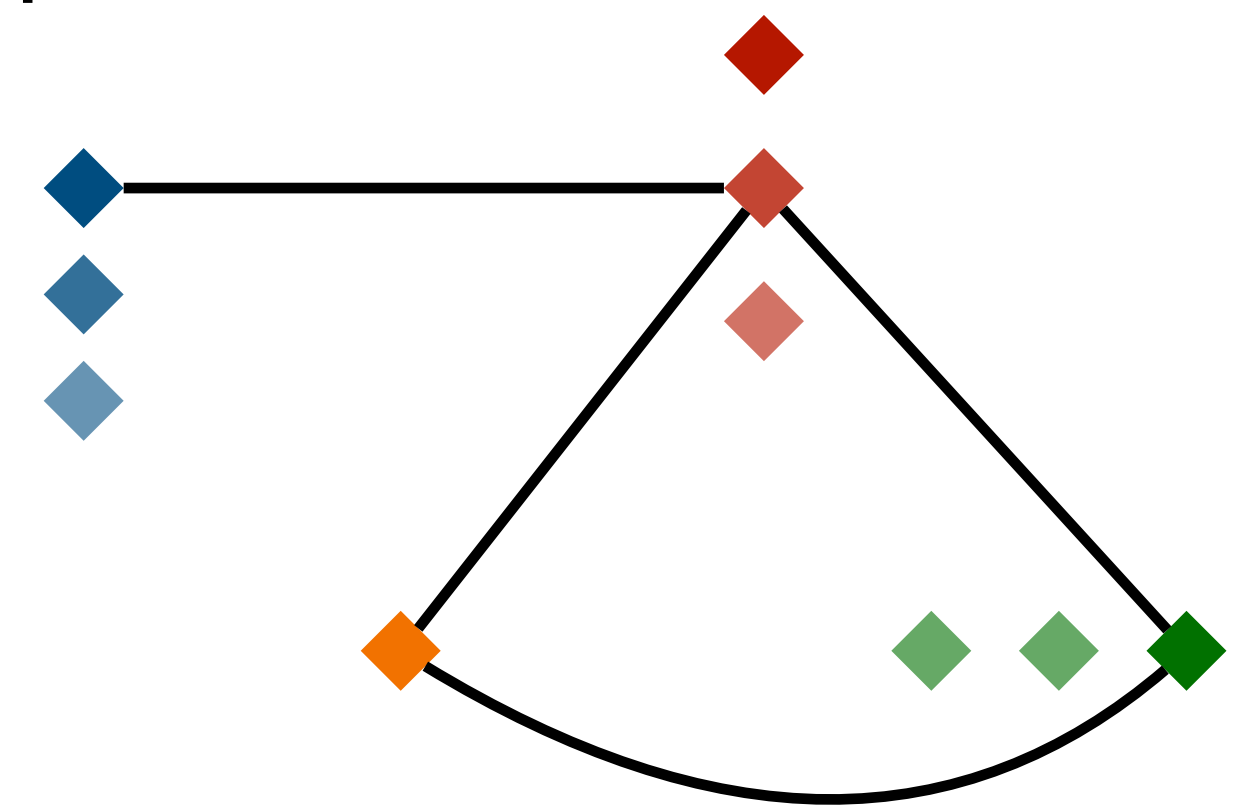
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Consensus

iterative stochastic decentralised algorithm

- Train t selects hypothesis $h_t \in H_t$ at start, select its best hypothesis h_t^*
- Train t selects a subset \mathcal{N}'_t of k neighbours ($|\mathcal{N}'| = k$)
- Train t ranks its hypotheses in H_t for compatibility with \mathcal{N}'
- If h_t is a top-ranked hypothesis, keep it
- Otherwise, chose a top-ranked hypothesis proportionally to its utility $u(h_t)$



Consensus

results on abstract problem instances

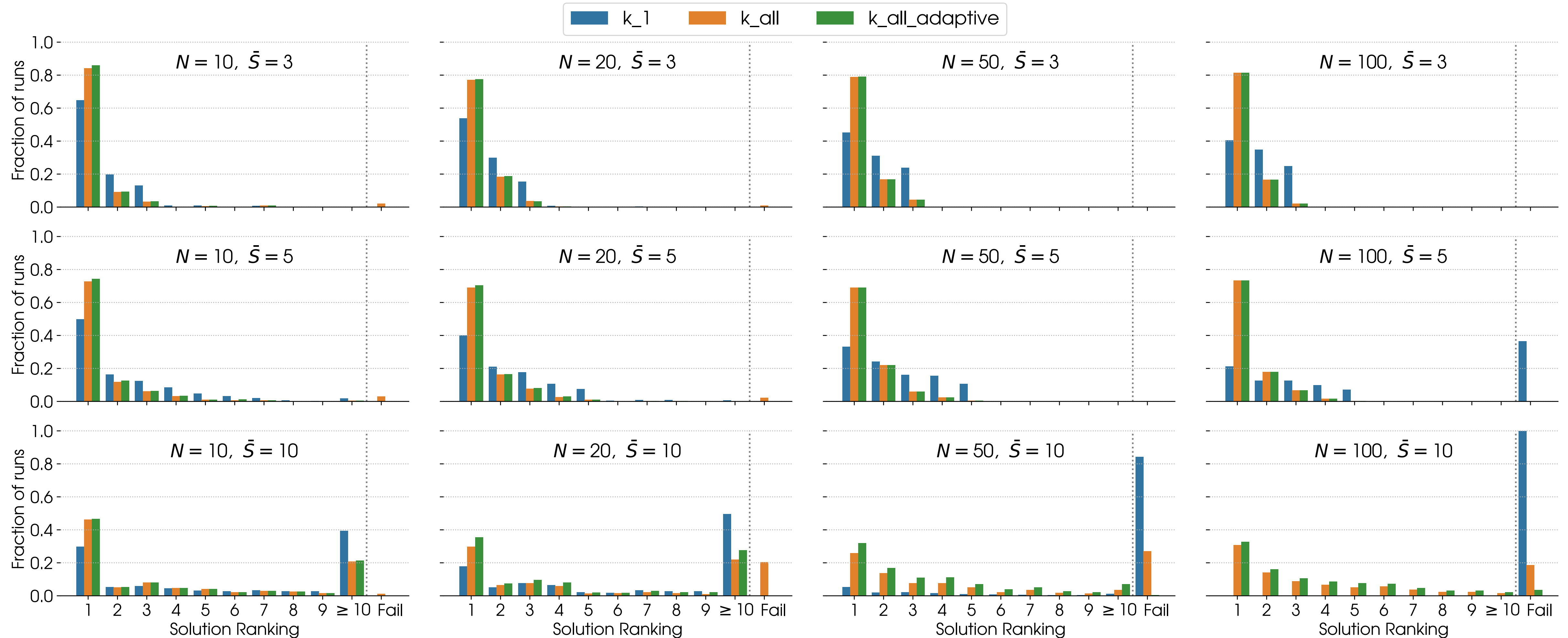
- We test three different implementations
 - $k = 1$: only choose a single neighbour at a time
 - $k = \infty$: always choose all neighbours
 - k adaptive: start with ∞ and slowly decrease to 1
- Convergence when a global solution is found (absorbing state)
- Goals:
 1. Select the best global solution
 2. Minimise the time to convergence





Consensus

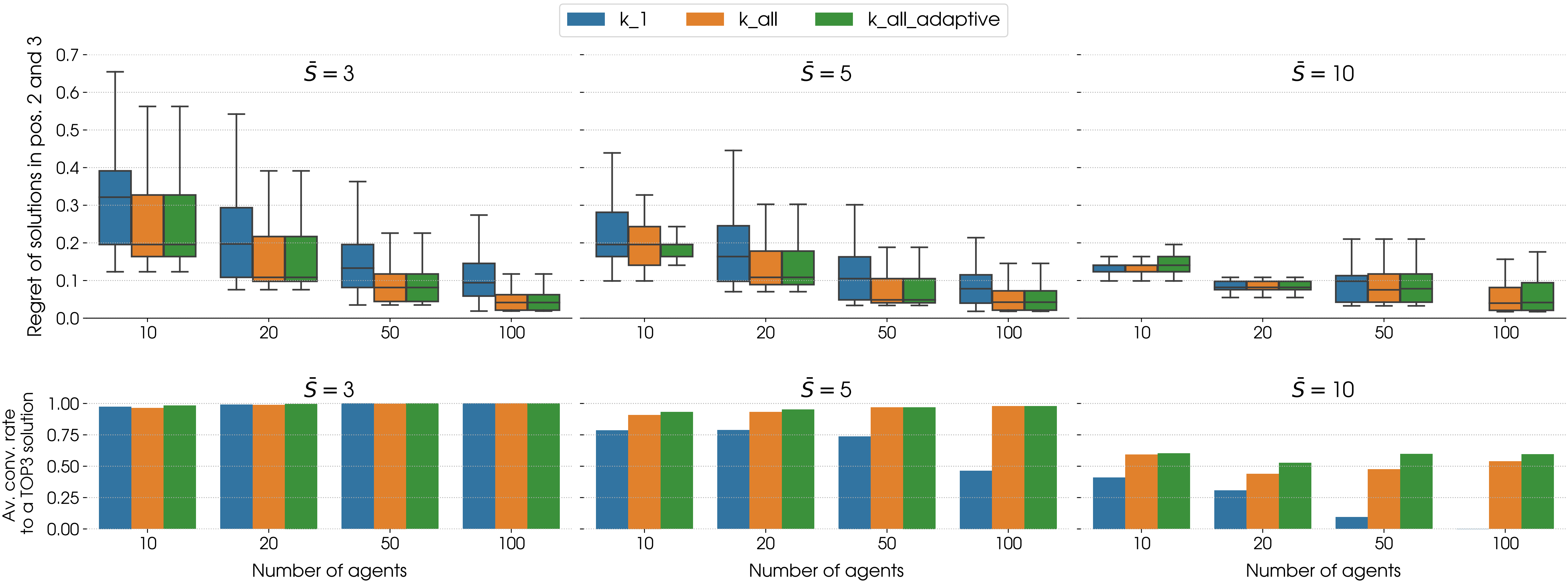
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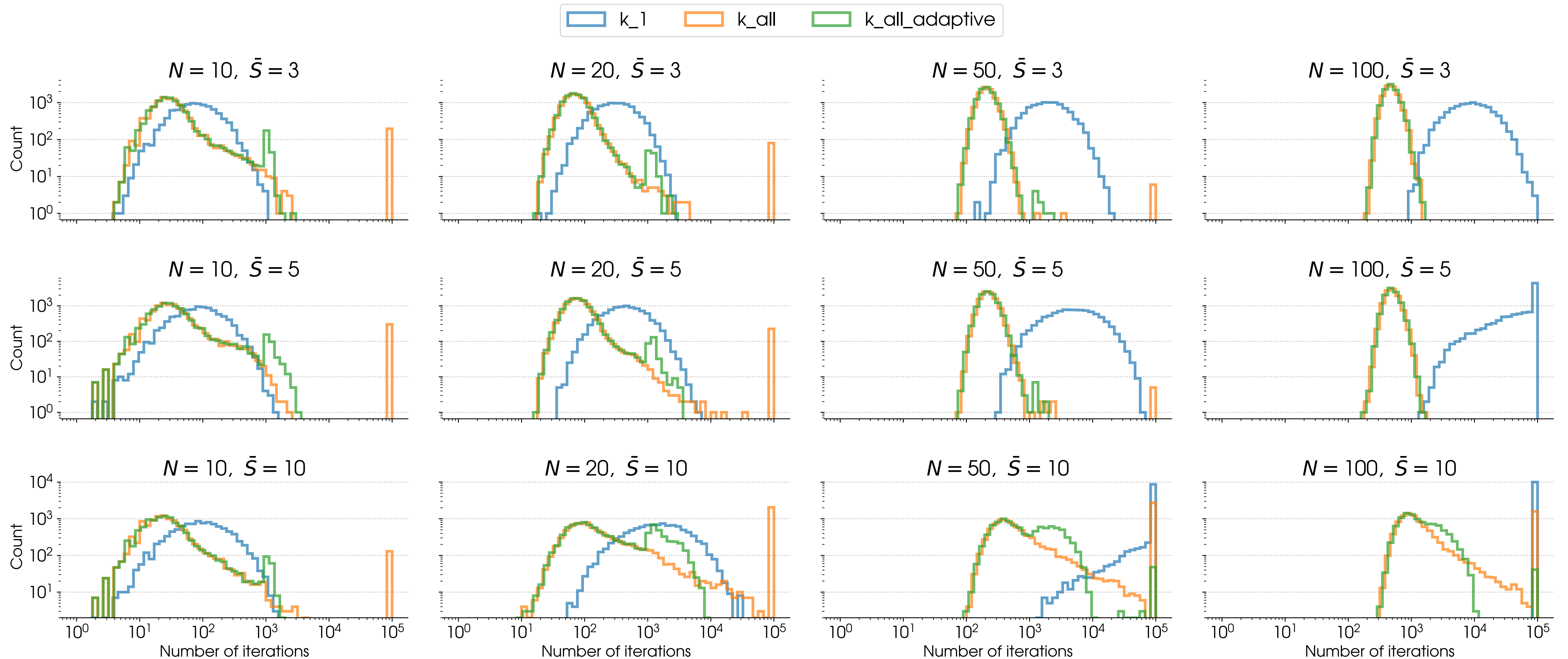
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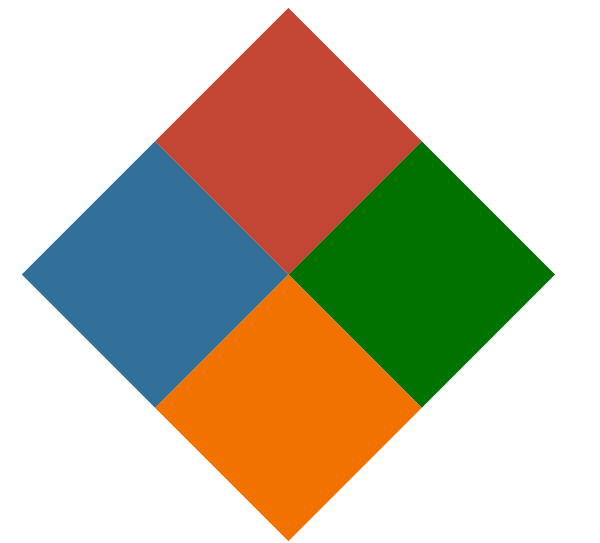
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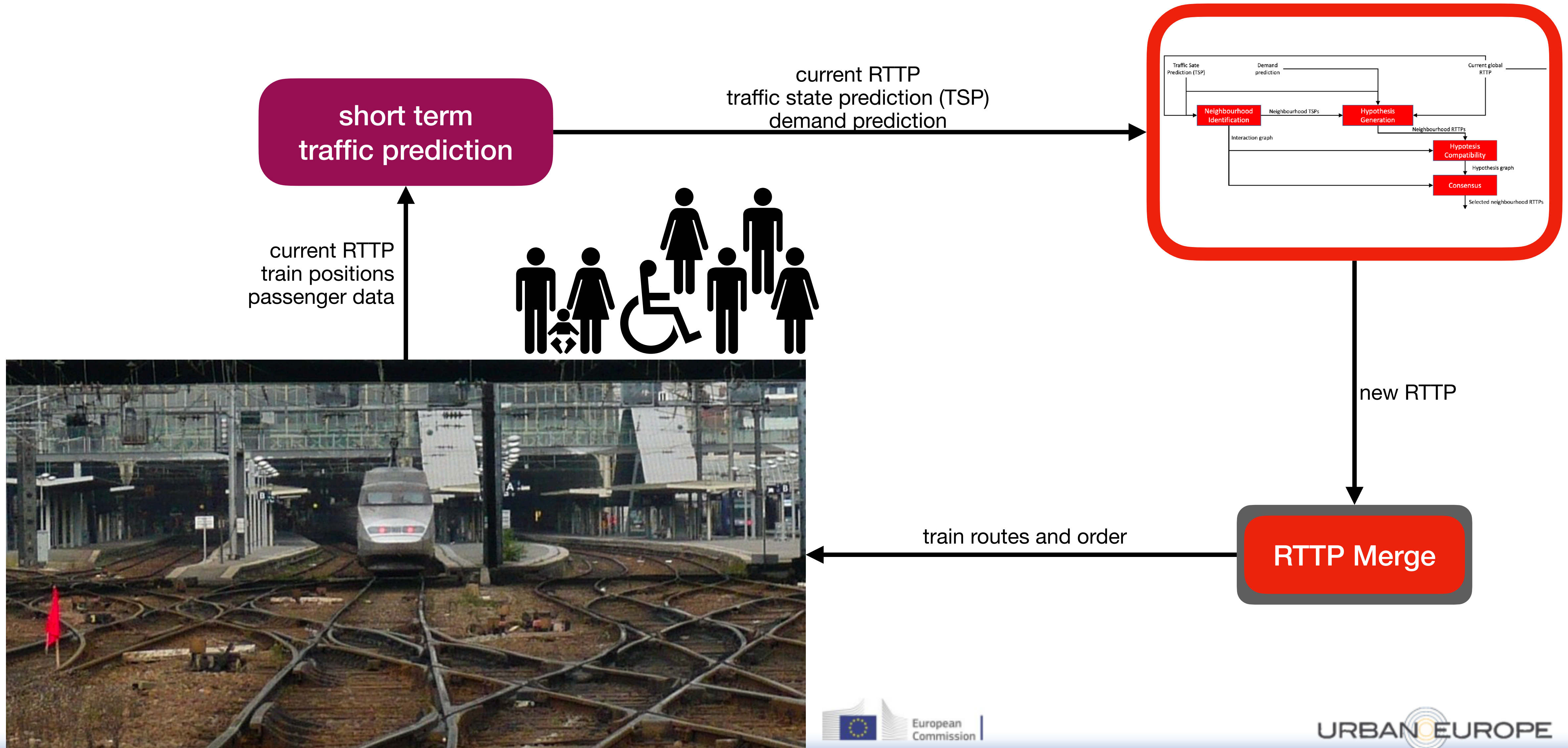
Merge

how to generate a global RTTP

- After consensus, each train proposes an hypothesis (RTTP)
- All hypotheses are merged to create a new, well-formed, RTTP
- The merge process is centralised at the train control center
 - The path of each train t is extracted from the selected hypothesis h_t
 - All previous paths are replaced by the new one into the RTTP
 - If consensus is not achieved (or partially achieved), the previous paths remain valid
- In the unlikely case that the merge process produces incompatibilities in the long term, these are repaired centrally



Closed-Loop Traffic Management



Thanks for your attention!



This project is supported by the European Commission and funded under the Horizon 2020 ERA-NET Cofund scheme under grant agreement N° 875022