

DATA DRIVEN ADAPTIVE TRAFFIC SIMULATION OF AN EXPRESSWAY

Abhinav Sunderrajan, Vaisagh Viswanathan, Wentong
Cai and Alois Knoll

Presented by: Wentong Cai

1. Background and Motivation.
2. Symbiotic Traffic Simulation Framework.
3. Case Study.
4. Results.
5. Conclusions and Future Work.

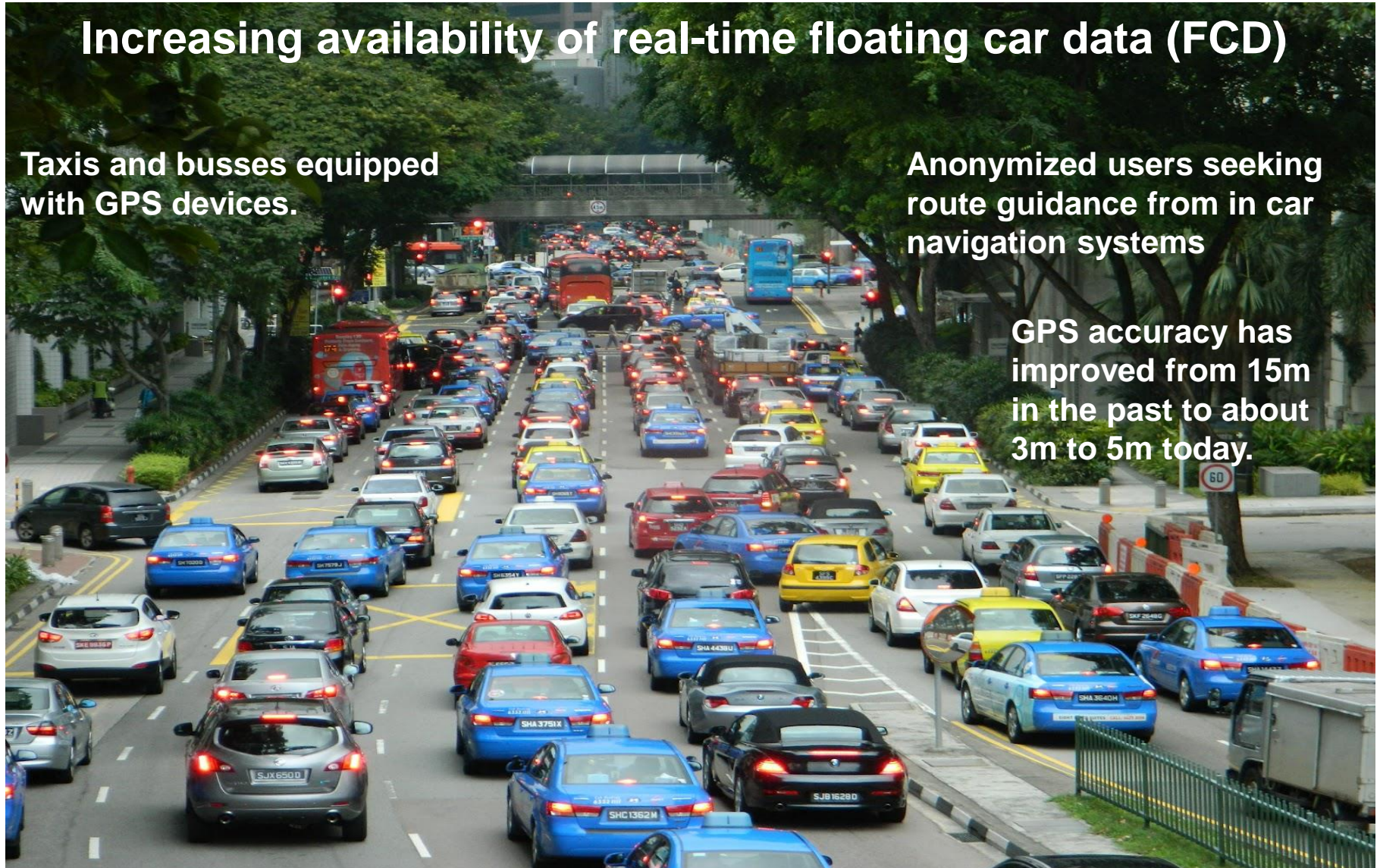
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Increasing availability of real-time floating car data (FCD)

Taxis and busses equipped with GPS devices.

Anonymized users seeking route guidance from in car navigation systems

GPS accuracy has improved from 15m in the past to about 3m to 5m today.



Next generation ERP
could be Global
navigation satellite
system based.

Thus most vehicles will
have GPS receivers
emitting data such as
location and speed at
fixed time intervals.





Can we leverage this real-time data stream for data-driven adaptive traffic flow control and optimization?

- In this talk, we present a “*Symbiotic Traffic Simulation Framework*” (STSF).
- *Symbiotic Simulation* is a special class of dynamic data driven adaptive simulation involving a mutually beneficial relationship between the *physical system* and *simulation systems*.
- The STSF receives continuous inputs from the physical system, i.e., the road network (and the vehicles driving on it) to initialize predictive faster than real time simulations.
- Based on the results of the predictive simulations, a recommendation is sent back to control the road network for optimizing traffic flow.

- The physical system is emulated using a high-fidelity *agent based microscopic traffic simulation* incorporating acceleration and lane change models.
- The prediction & control system receives traffic state inputs from the physical system uses a *macroscopic traffic flow model* to employ a simulation based optimization strategy.
- Finally the recommendations of this predictive & control system are given back to the physical system (microscopic simulation) and evaluated for efficacy.

1. Background and Motivation.

2. Symbiotic Traffic Simulation Framework.

- ❖ Overview
- ❖ Physical System
- ❖ Prediction system
- ❖ Optimization Module

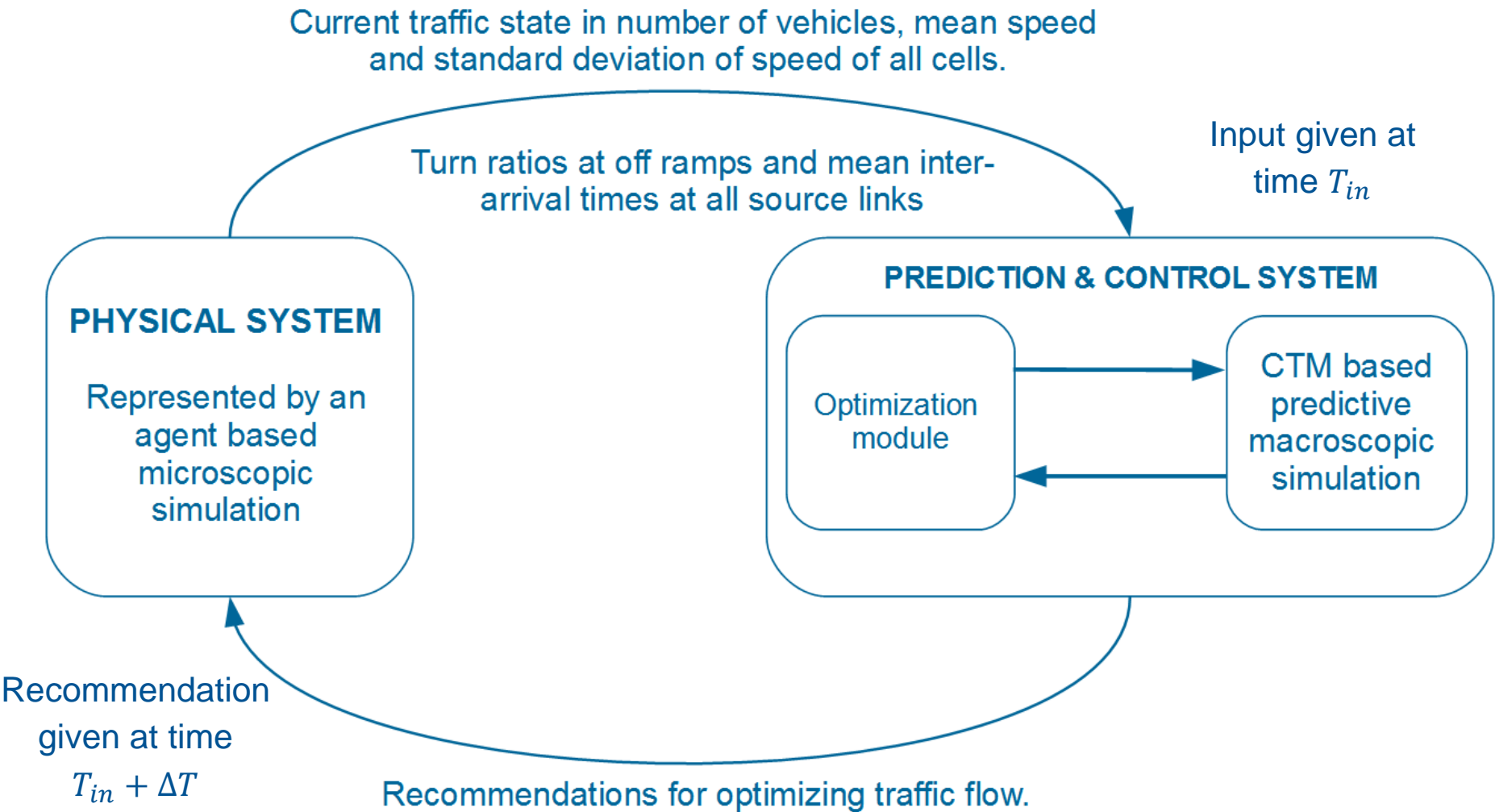
3. Case Study.

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- For a reasonable representation of the physical world, we employ an agent-based microscopic traffic simulation.
- The floating car data (FCD) provided by the microscopic simulation was used to initialize the state of the predictive Cell Transmission Model (CTM)* based macroscopic simulation.
- The predictive component works hand in hand with the optimization module to give recommendations to the physical system to optimize traffic flow after evaluating several candidate solutions.
- In this paper we optimize the traffic flow by simulating a real world expressway employing ramp-metering as the control action.

* The cell transmission model: A dynamic representation of highway traffic consistent with the hydrodynamic theory". Transportation Research Part B: Methodological 28 (4): 269–287





We employ agent-based microscopic traffic simulations to model the physical system with high fidelity.

- The agent-based simulations were employed owing to the enormous amount of resources required to implement the recommendations of the predictive simulation on a real world road network.
- Microscopic models describe traffic from the perspective of individual driver-vehicle units (DVUs).
- These high fidelity simulations help capture the heterogeneities in traffic in terms of different vehicles classes (e.g. cars, trucks) and driver behaviors (e.g. intelligent, aggressive).
- The movement of DVUs are characterized by acceleration models for longitudinal motion and lane change models for lateral movement along the road.

- The microscopic simulation is based on the SEMSim platform.
- SEMSim uses the Intelligent Driver Model (IDM) as the acceleration model for moving the agent/vehicle forward every time-step of the simulation.
- The IDM is an accident free model which ensures that a vehicle attains the desired velocity at free flow and maintains the safe bumper to bumper distance to the leading vehicle.

- ❖ Zehe, D., Knoll, A., Cai, W. and Aydt, H., 2015. SEMSim Cloud Service: Large-scale urban systems simulation in the cloud. *Simulation Modelling Practice and Theory*, 58, pp.157-171.
- ❖ Treiber, M. and Kesting, A., 2010. An open-source microscopic traffic simulator. *Intelligent Transportation Systems Magazine, IEEE*, 2(3), pp.6-13.

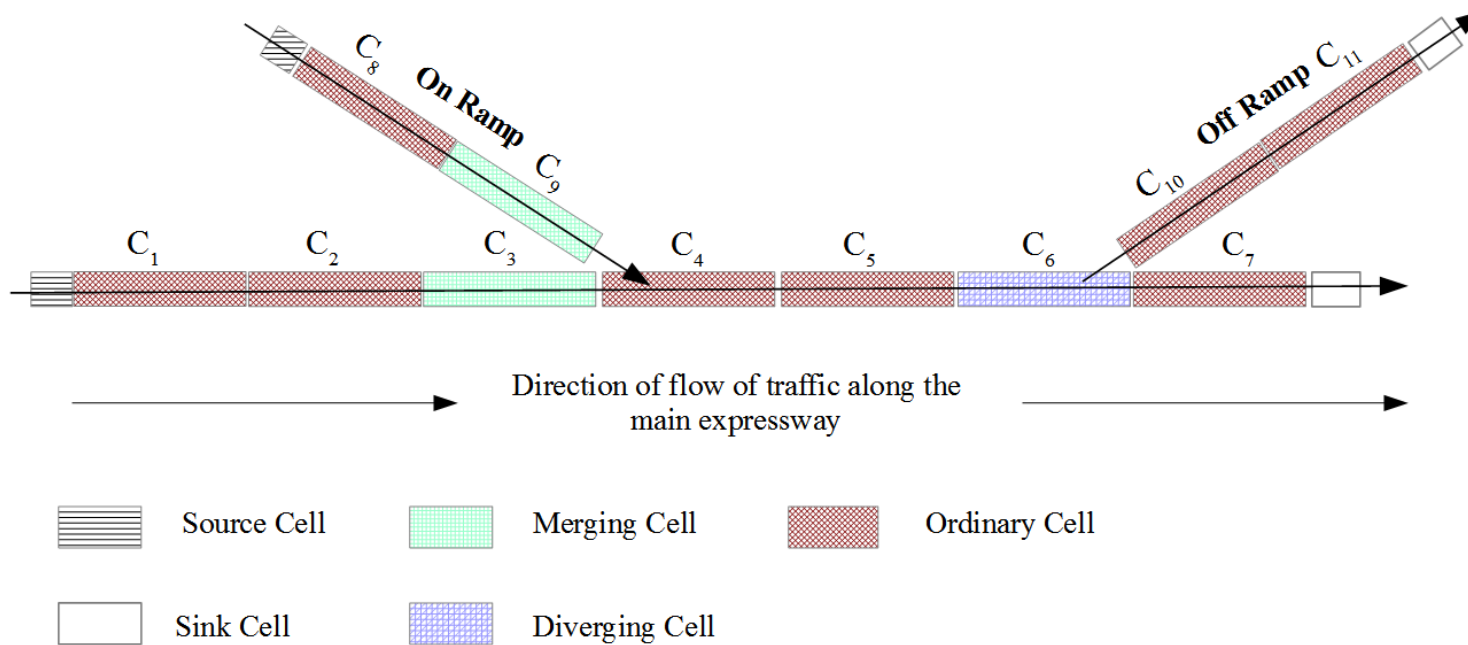
- SEMSim uses MOBIL as the lane change model.
- MOBIL ensures that the resultant accelerations and decelerations for a vehicle and its followers in the old and new lanes does not exceed a safe threshold.
- A lane change is done only if a vehicle gains speed without violating the safety and inconvenience (to the old and new followers) criteria.

Kesting, Arne, Martin Treiber, and Dirk Helbing. "General lane-changing model MOBIL for car-following models." *Transportation Research Record: Journal of the Transportation Research Board* (2007).

- The traffic simulation takes as input a road network consisting of links and the lanes constituting each link.
- Road links that do not have a preceding link are considered *sources* and those without a subsequent link are *sinks*.
- Traffic thus flows from the sources to the sinks.
- Vehicles are created at each source as a Poisson process with a constant mean inter-arrival time (IAT).
- The route taken by each DVU is determined based on static turn ratios specified at each intersection.

- The predictive, faster than real time macroscopic simulation is based on the stochastic variant of the Cell transmission model (Boel and Mihaylova 2006) and METANET (Kotsialos et al. 2002).
- The primary reason for employing a macroscopic simulation for the predictive component, is computational efficiency. A gradient-based optimization strategy involves assessing the fitness of several candidate solutions in parallel.
- Refer to Sunderrajan et al * for greater details on the algorithm and equations governing the model of the predictive simulation.

*Sunderrajan, A., V. Viswanathan, W. Cai, and A. Knoll. 2016a. "SYMBIOTIC TRAFFIC SIMULATIONFRAMEWORK". Technical report, TUM CREATE. doi:<http://dx.doi.org/10.13140/RG.2.1.2429.9128>



- The cell network, \mathcal{C} is comprised of n cells. At each time step k , $k = 0, 1, \dots, K$ (where K is the time horizon) the state of all cells are updated.
- The state of a cell $c_i \in \mathcal{C}$ at each time step k is determined by the concept of sending $S_i(k)$ and receiving potentials $R_i(k)$. $S_i(k)$ and $R_i(k)$ represent the number of vehicles cell c_i can send and receive at time-step k .

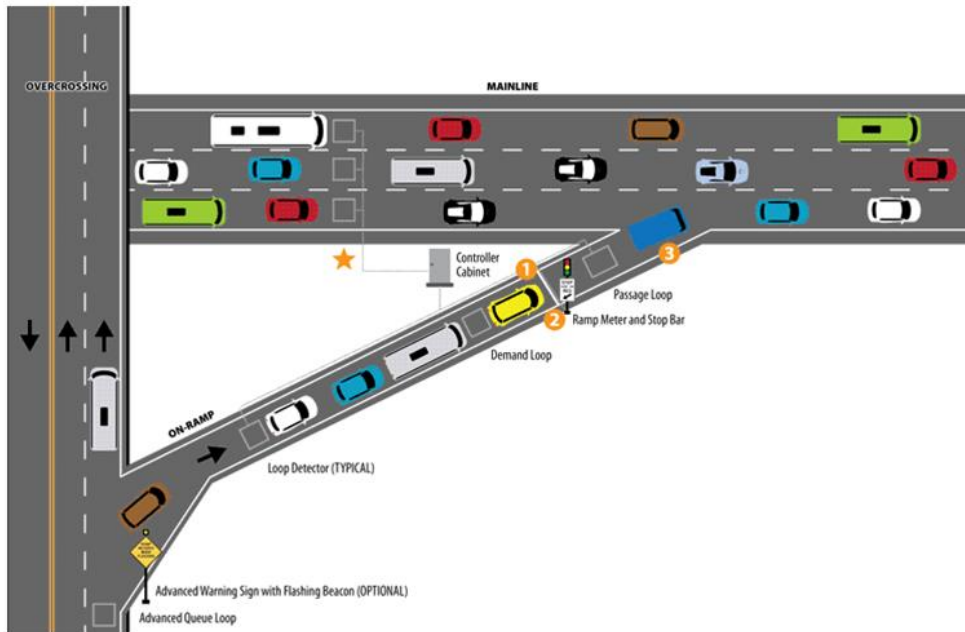
- We implemented a simulated annealing approach* for the optimization module to determine the best control strategy for the case study discussed next.
- The simulated annealing algorithm seeks to identify the best control-action by minimizing the value returned by the function $fitness(S)$ (S represents the candidate solution) running the CTM based predictive simulation over a time horizon K .
- Specifically, the fitness function returns the value N_{total} which is the total number of vehicles in the system over the predictive time horizon K .

$$N_{total} = T \sum_{k=0}^{k=K} N(k)$$

*Weise, T. 2009. "Global optimization algorithms-theory and application"

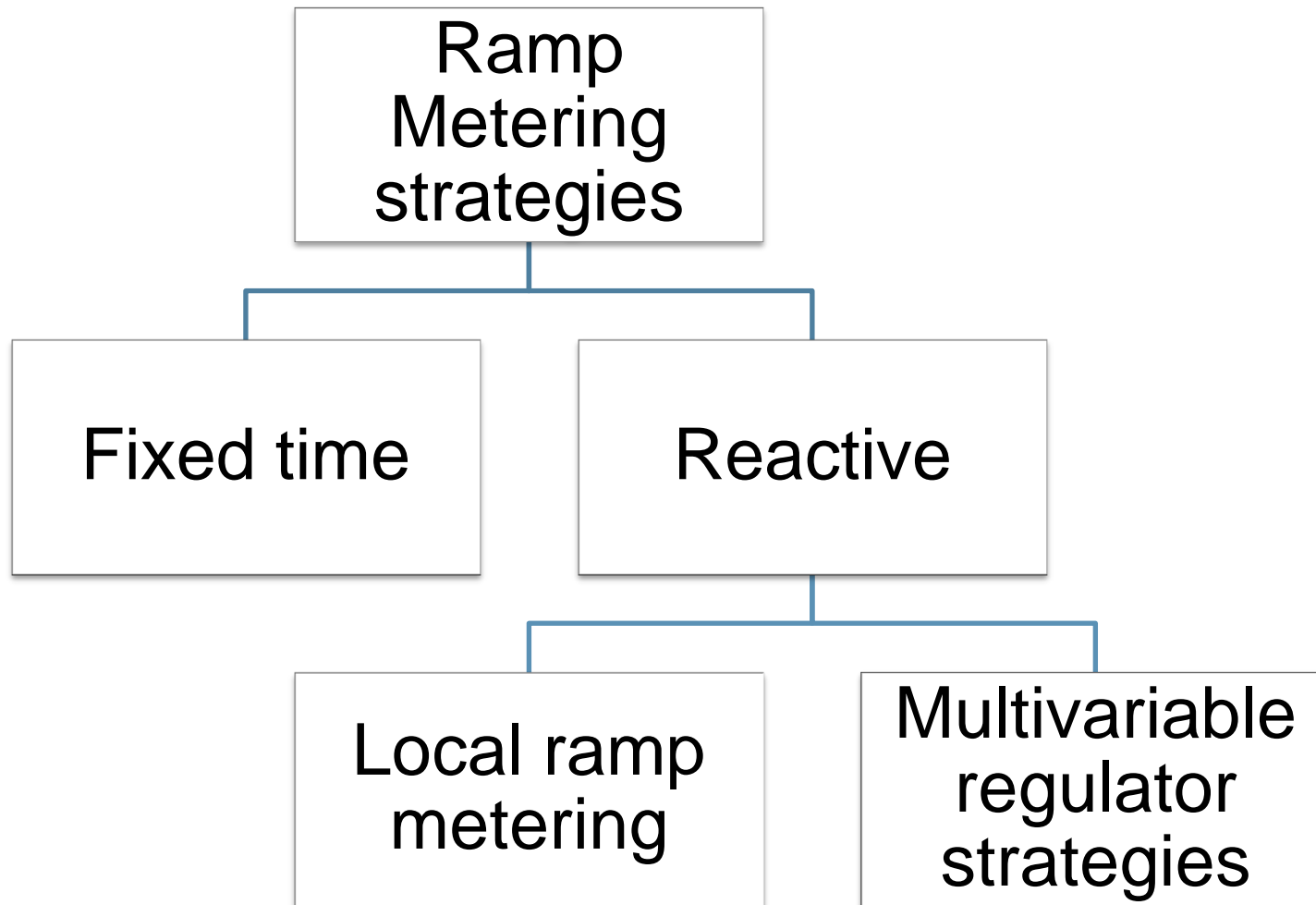
1. Background and Motivation.
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- 3. Case Study.**
 - ❖ **Ramp Metering**
 - ❖ **Simulated Physical Environment**
 - ❖ **Traffic Scenario**
4. Results.
5. Conclusions and Future Work.

Ramp Metering



TUMCREATE

- Ramp meters are traffic signals placed at the intersection of on-ramps and expressways.
- Ramp meters regulate the flow of vehicles along the ramps so as to minimize the turbulence caused due to merging vehicles disrupting the mainline flow.
- Care must be taken to ensure that the queue of the vehicles waiting along the on-ramps does not spill into the preceding urban street network.



A review on several implemented and proposed ramp-metering strategies can be found in Bogenberger, K., and A. D. May. 1999. "Advanced coordinated traffic responsive ramp metering strategies". California Partners for Advanced Transit and Highways (PATH).

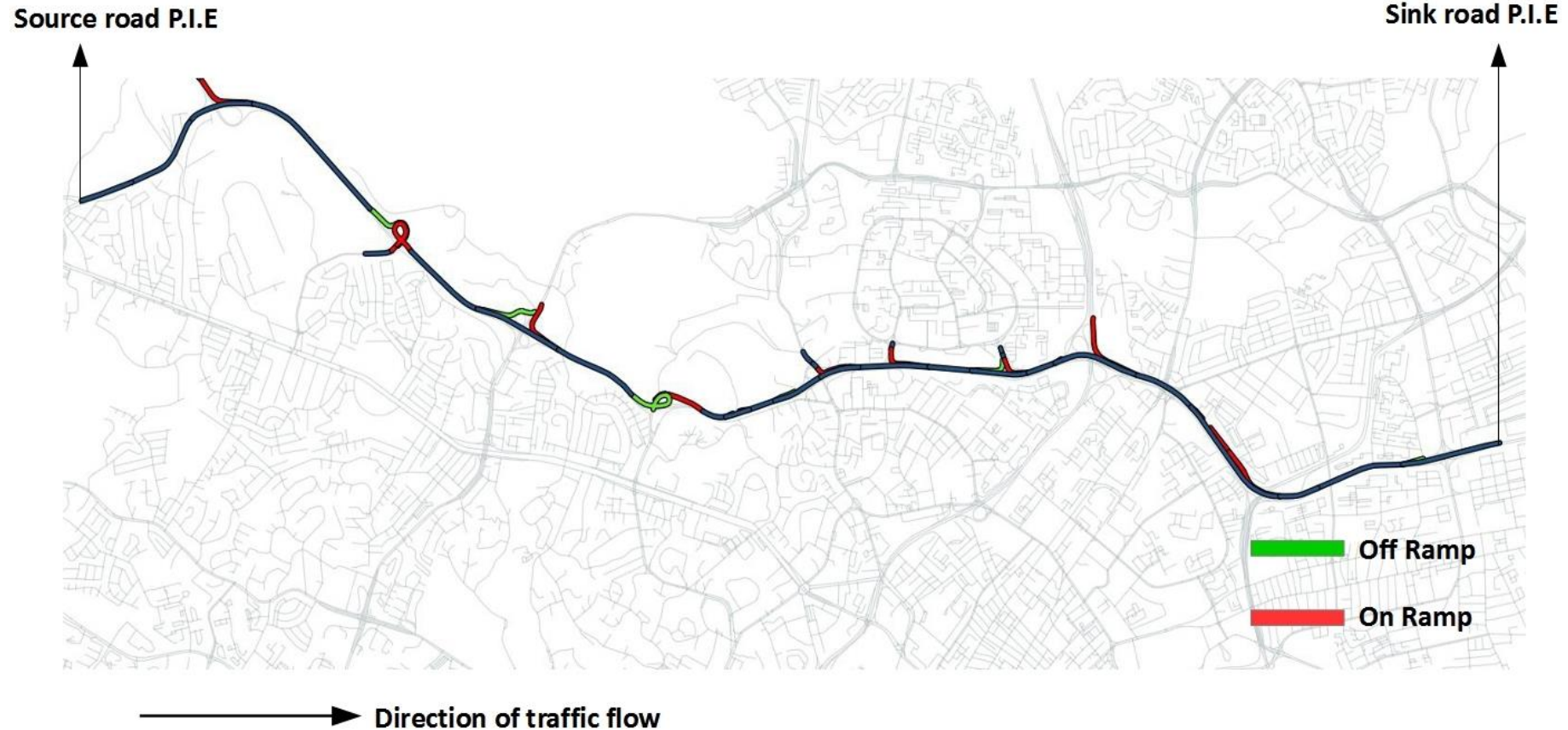
- We employ simulated annealing to develop a system wide ramp metering strategy by regulating the flow on all on-ramps simultaneously.
- The ramp controller determines the maximum allowable queue-threshold q^{th}_r for a ramp $r \in R^{on}_{ramps}$ (R^{on}_{ramps} represents the set of controllable on ramps.) before turning phase of the signal to green from red.

The queue threshold q_r^{th} is given by

$$q_r^{th} = \frac{N_r(k)}{N_r^{max}(k)} \Bigg| q_r^{th} \in [0.0, 1.0]$$

- $N_r(k)$ represents the number on the on-ramp r at time step k .
- $N_r^{max}(k)$ represents the maximum number of vehicles that can be accommodated on r at time step k .

Concretely the task is to find the ideal value of q_r^{th} for all controllable on-ramps so as to minimize N_{total} .



13 km stretch of P.I.E (Pan Island Expressway) in central Singapore with all on ramps and off ramps.

- The on ramps and the first P.I.E link are sources for the vehicles/agents entering the simulation.
- The off ramps and the last link on P.I.E are sinks through which the vehicles exit the simulation.
- The turn ratios for all off-ramps is kept constant at 0.25. This implies that 25% of all vehicles exit at a given off-ramp while the remaining 75% of the vehicles continue to travel on the main expressway.
- The number of lanes in the simulated stretch of the expressway varies between 3 and 6.

DISTANCE (m)	RAMP TYPE	ϵ_s (sec)
0.0	First P.I.E link	1.0
583.98	On-Ramp	2.0
2489.87	On-Ramp	3.6
4071.9	On-Ramp	3.6
5531.18	On-Ramp	3.6
5965.29	On-Ramp	3.6

DISTANCE (m)	RAMP TYPE	ϵ_s (sec)
7025.15	On-Ramp	2.0
7658.4	On-Ramp	2.0
8554.28	On-Ramp	3.6
9591.84	On-Ramp	3.6
11286.2	On-Ramp	3.6
11637.04	On-Ramp	3.6

- The traffic state of the expressway at the end of a time horizon is determined by the inter-arrival time ϵ_s for all source links and cells.
- Notice that the flow of vehicles into the expressway along all on-ramps are significantly less (1000 vehicles/hour) except for the ones at 583 m, 7025 m and 7658 m.
- The system thus needs to find an optimal ramp metering strategy which balances the flow along all on-ramps so as to minimize the surge of vehicles along the three ramps with relatively higher inflow of vehicles.

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 - ❖ **Calibration of predictive Simulation**
 - ❖ **Experimental setup**
 - ❖ **Efficacy of Ramp Metering**
 - ❖ **Computational efficiency**
5. Conclusions and Future Work.

- To ensure that the state predicted by the macroscopic simulation accurately represents the state of the physical system, the model parameters (of the predictive macroscopic simulation) have to be calibrated.
- The parameters which will have significant impact in terms of bridging the difference in the state of the physical system and that of the predictive simulation at the end of a given time horizon.

Note that calibration of the predictive simulation with respect to the physical system is an offline one step process.

PHYSICAL SYSTEM (SEMSim, microscopic traffic simulation)

- Initialize the turn ratios for all off-ramp expressway intersections.
- Initialize the mean inter-arrival rates for all sources.

- Run SEMSim for a time horizon of $K = 2000$ seconds.
- At the end of time horizon, determine the number of vehicles in each cell corresponding to the number predictive simulation.
- $N_i^{SEMSim}(K)$ represents the number of vehicles in each cell i at the end of simulation at time-step K .

CTM based predictive simulation

- Initialize the turn ratios for all off-ramp expressway intersections.
- Initialize the mean inter-arrival rates for all sources.
(same values as the physical system represented by SEMSim)

For j in 1:1600

- Initialize random simulation parameters.
- At the end of time horizon determine the number of vehicles in each cell.
- Compute $N_{ji}^{CTM}(K)$ representing the number of vehicles in each cell i at the end of predictive simulation in iteration j .
- Compute $\sum (N_{ji}^{CTM}(K) - N_i^{SEMSim}(K))^2$.

Determine CTM model parameters so as to minimize

$\sum_j \sum_i (N_{ji}^{CTM}(K) - N_i^{SEMSim}(K))^2$ using linear regression.

Constraints

- The minimum phase time for both the red and green phases are 12 seconds.
- The signal at an on-ramp can be continuously red only for a maximum of 120 seconds. After a 120 second red phase, there is a mandatory green phase for 24 seconds.

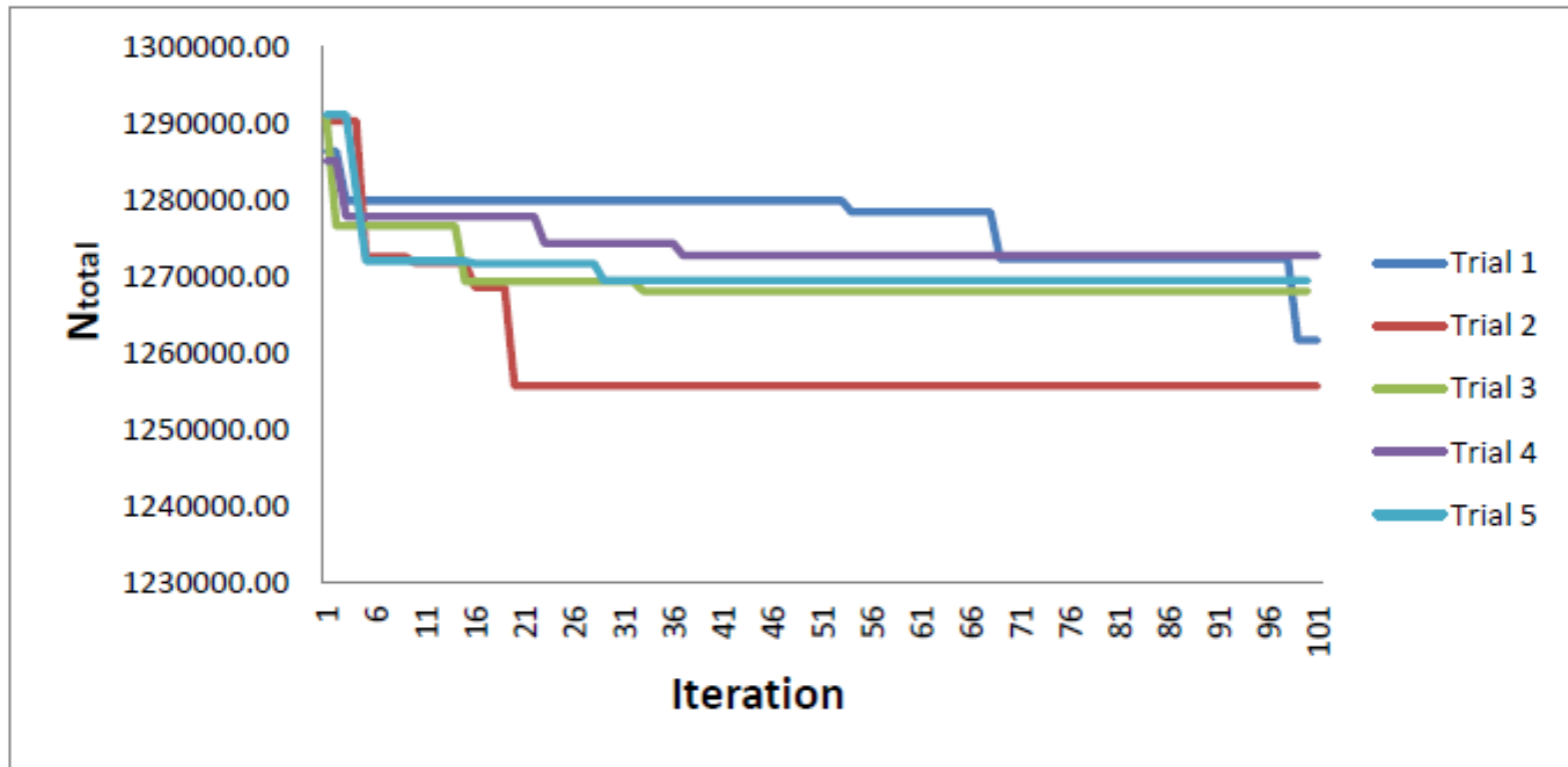
The fitness function of a solution fitness(S) has been modified as follows

$$N_{total} = \frac{1}{5} \sum_{i=1}^{i=5} N_{total}^i + \alpha \sum_{r \in R} q_r^{th}$$

The mean value of N_{total} is computed by averaging over 5 different runs of CTM (using different seeds) for a single queue-threshold configuration to account for model stochasticity.

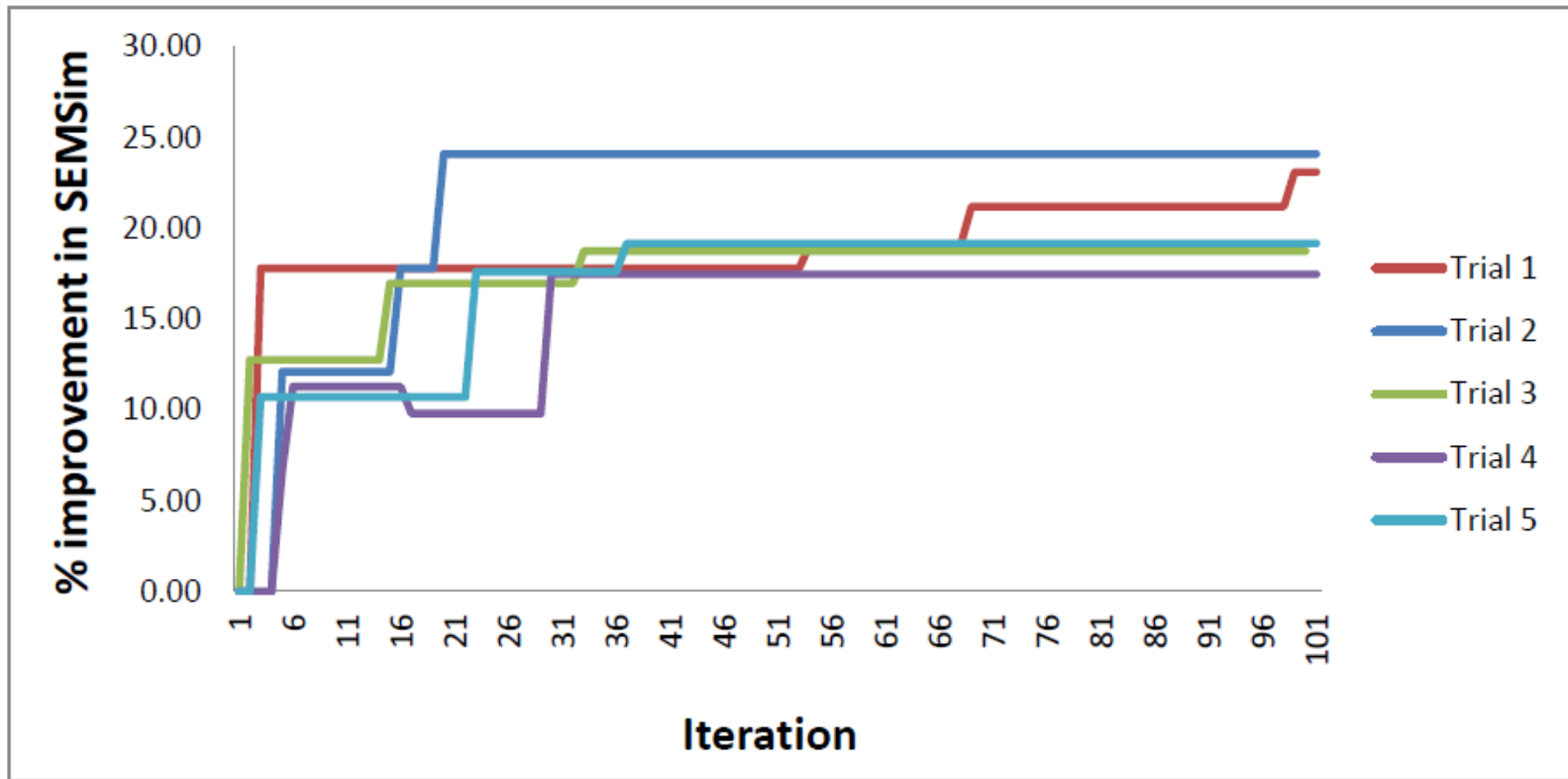
$$N_{total} = \frac{1}{5} \sum_{i=1}^{i=5} N_{total}^i + \alpha \sum_{r \in R} q_r^{th}$$

- The penalty factor α in the above fitness function for solution S penalizes the imposition of queuing control at an on ramp r .
- α prevents the algorithm from increasing q_r^{th} for minimal benefits in optimizing traffic flow.
- The penalty factor thus reduces the variance in the results obtained for the queue-threshold configuration.



Decrease in N_{total} for predictive simulated annealing algorithm.

Each trial represents different mean inter-arrival rates at the source links



Percentage improvement in N_{total} when recommendations are given to SEMSim

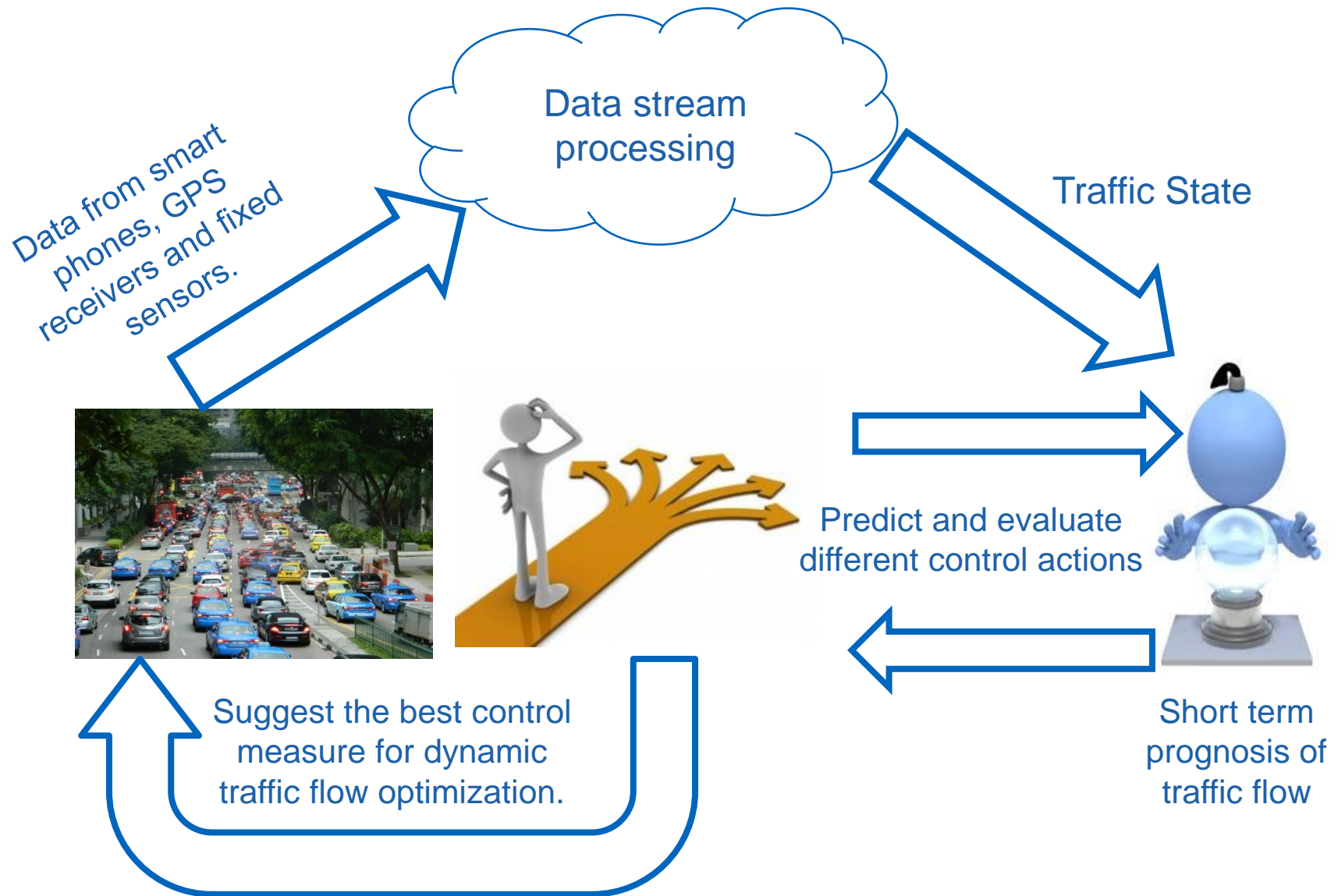
Each trial represents different mean inter-arrival rates at the source links

Computational Efficiency of Predictive Simulation

- A data driven adaptive simulation and prediction framework for traffic systems should work under reasonable time constraints for giving back recommendations to optimize traffic flow.
- The computational time for the CTM based simulation (used for determining N_{total}) over a time horizon of 1800 seconds is around 75 milliseconds. The CTM simulation was coded in Java SE 7 and measured in a 2.5 GHz Intel i5 system running on Windows 7.
- The entire run of the simulated annealing algorithm over 100 iterations took around 42 seconds to complete, thus satisfying the soft real time constraints for a symbiotic traffic simulation.

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- In this work we have shown that data driven predictive simulations can be beneficial towards optimizing traffic flow.
- The prediction and optimization system should receive fairly accurate and continuous information on the current traffic state. This information is used for initialization, calibration and steering of the predictive simulations.
- The simulation model and optimization strategy used in the prediction & control system can be varied depending upon accuracy, efficacy and computational time constraints.



Symbiotic Traffic Simulation Platform

- Symbiotic traffic simulations also offer exciting opportunities to implement and optimize several techniques for traffic flow optimization such as adaptive speed limits and dynamic routing.
- Mobile applications and in car navigation systems provide a great means to disperse information to the traffic participants while the control system receives user anonymized data about vehicle speed, location and even origin-destination flows.
- This form of a symbiotic simulation based traffic prediction and optimization framework directed towards receiving user anonymized data from individual drivers and providing them personalized updates is an interesting area for future research.

Thanks for you attention!
Any Questions or suggestions?