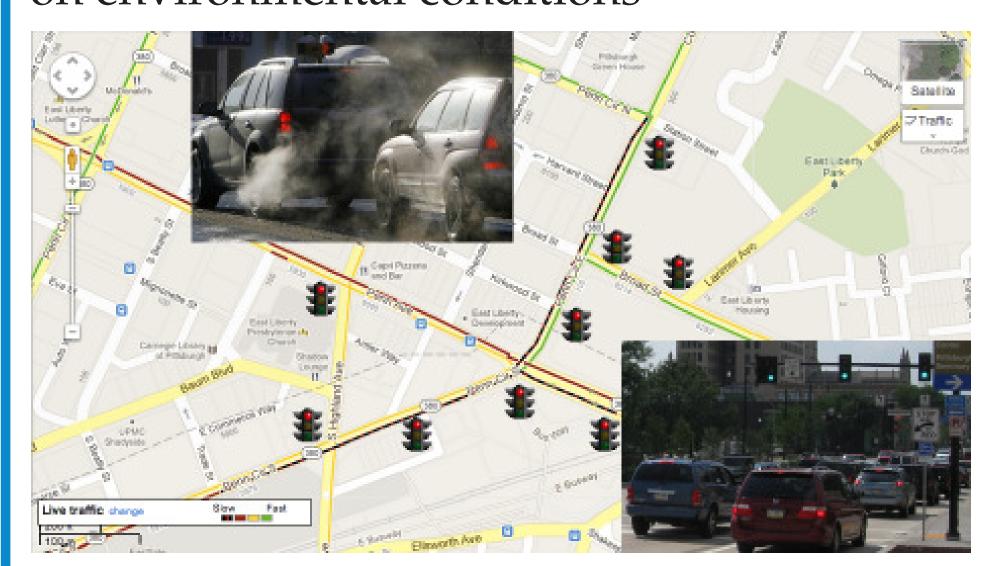
Coordinated Look-Ahead Scheduling for Real-Time Traffic Signal Control

Xiao-Feng Xie, Stephen F. Smith, Gregory J. Barlow {xfxie,sfs,gjb}@cs.cmu.edu The Robotics Institute, Carnegie Mellon University



PROBLEM

• Traffic congestion is a practical problem resulting in substantial delays and extra fuel costs for drivers, and has negative impacts on environmental conditions



- For urban road networks, better traffic flow requires better traffic signal control, and realtime, adaptive strategies offer the biggest payoff
- Real-time decisions: traffic light cycles through a sequence of phases I, each phase i has a variable duration that can range between a minimum and a maximum
- Local observation: inflows of vehicles in the prediction horizon (H), the current phase index and duration of traffic light, and the current decision time

CHALLENGE

Goal: Scalable network-wide optimization

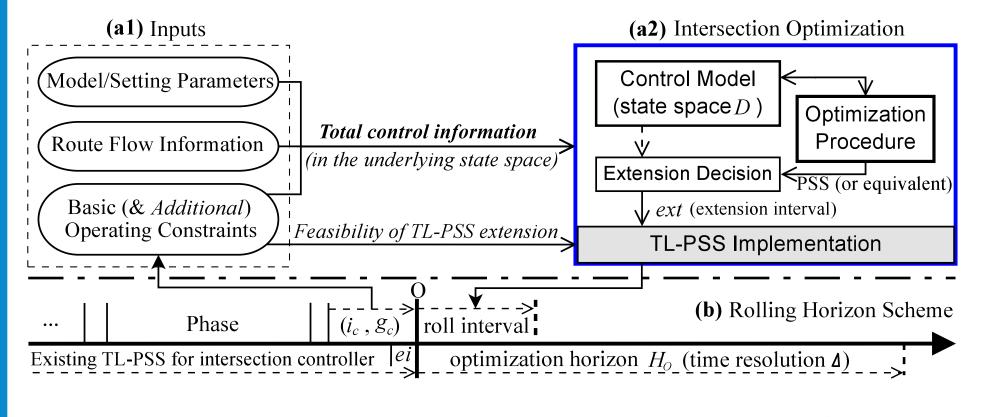
- Intersection level: the number of joint signal control sequences and local observations is huge in the prediction horizon
- Network level: effective coordination for handling non-local impacts between tightlycoupled intersections in a complex network

CONTRIBUTIONS

- Real-time traffic signal control based on coordinated look-ahead scheduling
- Each intersection is locally controlled by an agent using a schedule-driven intersection control strategy (SchIC) [3]. At each decision point, each agent constructs a schedule that optimizes movement of the observable traffic through its intersection, and uses this schedule to determine the best control action to take
- For strengthening its local view, each agent queries the scheduled outflows from its upstream neighbors to obtain an optimistic observation, which is capable of incorporating non-local impacts from indirect neighbors
- * Summary: Multi-agent coordination = lookahead scheduling + coordination mechanism(s)

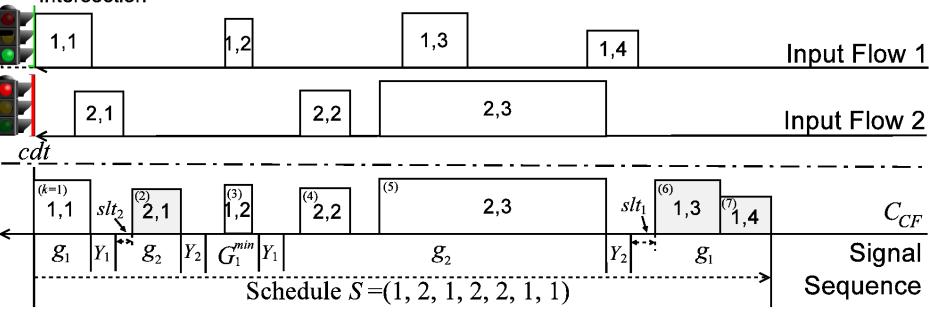
INTERSECTION CONTROL

• Look-ahead scheduling in a rolling horizon



• Aggregate non-uniform flow into jobs

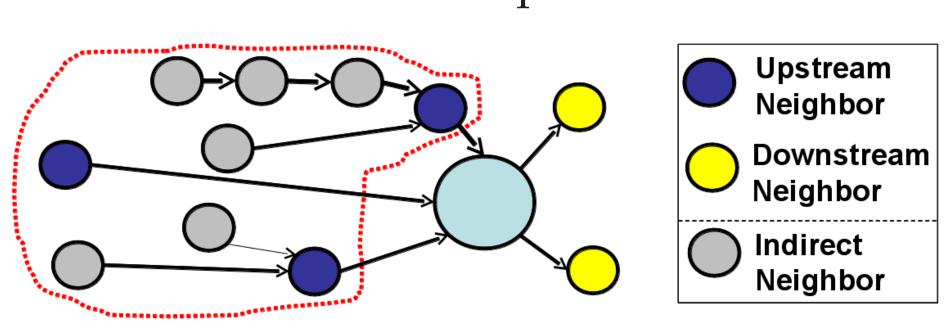




- Construct a schedule that optimizes movement of the currently approaching traffic in the local observation
- Worst-case complexity: $|I|^2 \cdot \prod_{i=1}^{|I|} (|C_{IF,i}| + 1)$ state updates, where |I| is the number of phases/inflows, and $|C_{IF,i}|$ is the number of jobs on inflow i
- Performs 2-4 orders of magnitude faster than COP [2]

COODINATION IN NETWORK

Consider non-local impacts in the network



Intuition: Predicted inflows $(IFs) \xrightarrow{schedule}$ Control flow $(CF) \Rightarrow$ Planned outflows (OFs) = Predicted nonlocal inflows for downstream neighbors

- Use an optimistic coordination protocol
- Decentralized / Scalable: each agent only communicates with its direct neighbors, but can incorporate non-local impacts from direct and indirect upstream neighbors
- Optimistic: each agent tries to follow its own schedule, minor schedule changes in neighbors can be absorbed

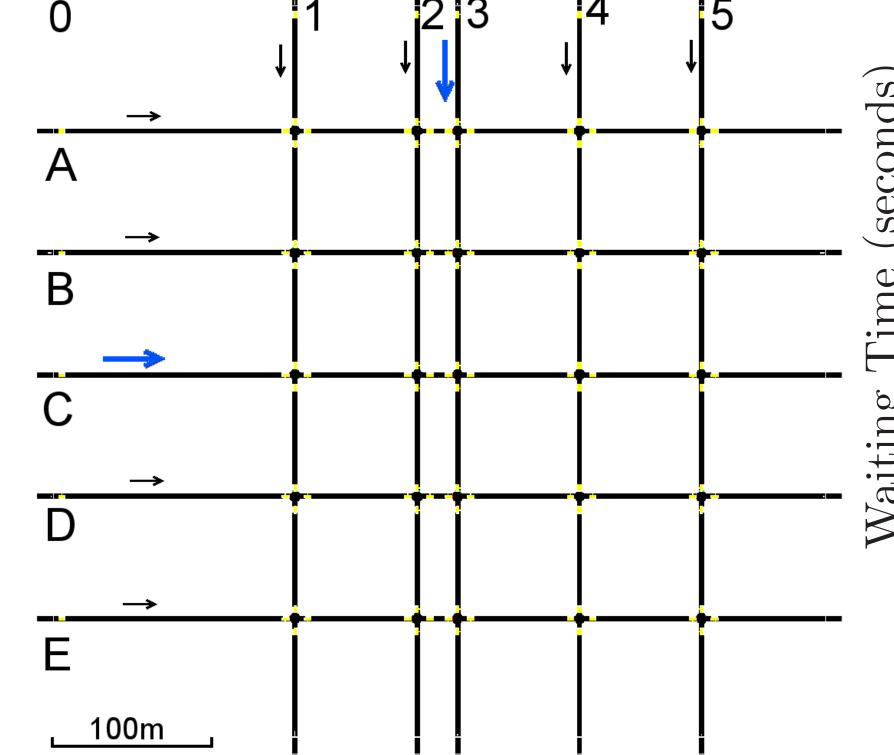
Algorithm 1 Obtain an optimistic non-local observation

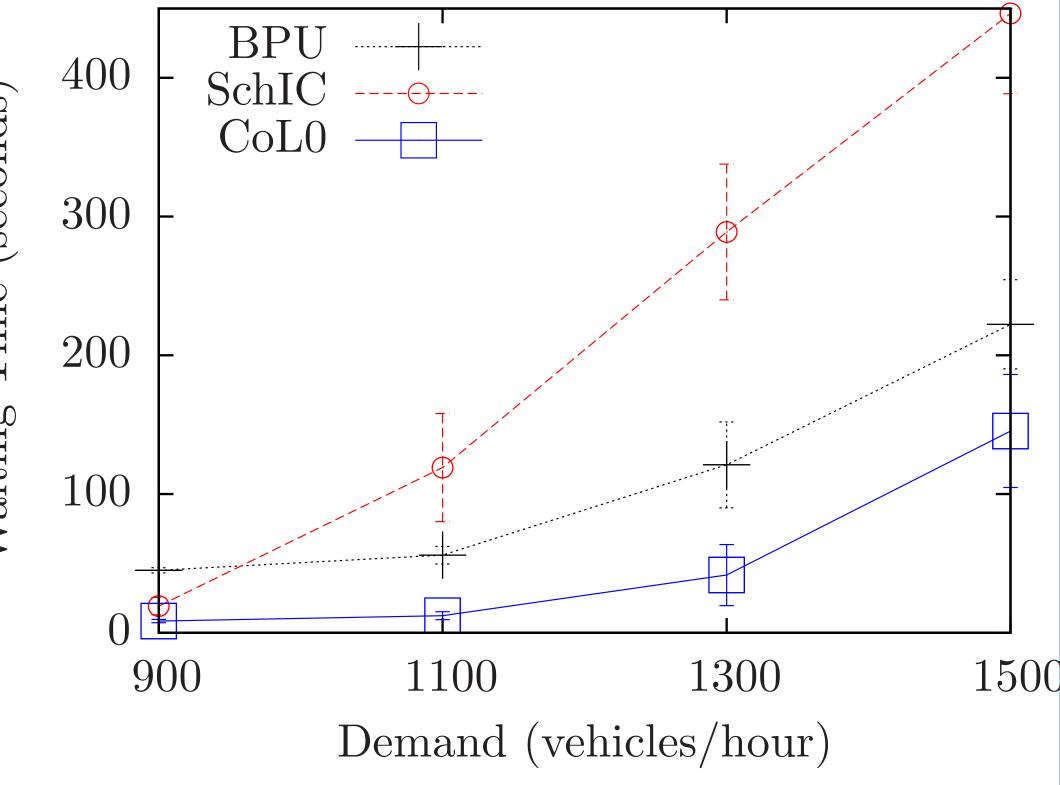
- 1: m = GetEntryRoadByPhase(i)
- $\{\text{For each phase } i\}$ 2: UpAgent = GetUpstreamAgent(m)
- 3: Request C_{OF} from UpAgent using (cdt, m, H_{ext})
- 4: $Shift(C_{O_{\xi}})$, the travel time on m)
- **Algorithm 2** Return C_{OF} for a message (cdt, n, H_{ext})
- 1: $(C_{OF}, S_{OF}) = (C_{CF}^*, S^*) \cap [cdt, cdt + H_{ext}]$
- for $k = |C_{OF}|$ to 1 do

5: Append C_{OF} into $C_{IF,i}$

- $|c_{OF,k}| = |c_{OF,k}| \cdot tp(s_{OF,k}, n)$ {turning proportion}
- 4: end for

PERFORMANCE EVALUATION





Simulation Settings

- Scenario: grid network with tightly-coupled intersections (with 2.5 or 7.5-second travel time on one edge)
- Dynamic demands on the bottleneck intersection C3

Control Strategies

- BPU: Balanced phase utilization [1] (offset calculation)
- SchIC: Schedule-driven intersection control [3]
- CoL0: SchIC + Optimistic non-local observation
- CoL0 produced lower waiting times than both other strategies. Comparison to SchIC demonstrates the added benefit of optimistic non-local observation. Furthermore, CoL0 outperforms BPU without requiring explicit offset calculation; coordination between neighbors is instead accomplished implicitly by looking ahead to upstream output flows.

REFERENCES

- [1] G. J. Barlow. Improving Memory for Optimization and Learning in Dynamic Environments. PhD Thesis, 2011.
- [2] S. Sen and K. L. Head. Controlled optimization of phases at an intersection. Transport. Sci., 31, 1997.
- [3] X.-F. Xie, S. F. Smith, L. Lu, G. J. Barlow. Scheduledriven intersection control. Transport. Res. C, 2012.

ACKNOWLEDGEMENTS

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ONGOING WORK AND FUTURE DIRECTIONS



- Pilot test: Scalable urban traffic coordinator
 - Currently testing approach on a 9-intersection 2-way road network in the East Liberty area of Pittsburgh, PA, USA
 - Real-world challenges: uncertainty, robustness to failures
- URL: https://pilot.surtrac.net (available soon)
- Advanced coordination mechanisms
- Pricing mechanisms to dampen any disruptive changes on schedules made by upstream agents (intersections)
- Negotiation mechanisms to reach for an equilibrium in an over-saturated traffic sub-network
- Dynamic learning of edge weights for critical flows