An aerial photograph of a city intersection with several roads. Overlaid on the road surface are yellow, concentric, triangular lines that form a grid-like pattern, likely representing a traffic control or sensor network. Various vehicles, including cars, buses, and a truck, are visible on the roads. The text is overlaid on the top half of the image.

A two-level urban traffic control for autonomous vehicles to improve network-wide performance

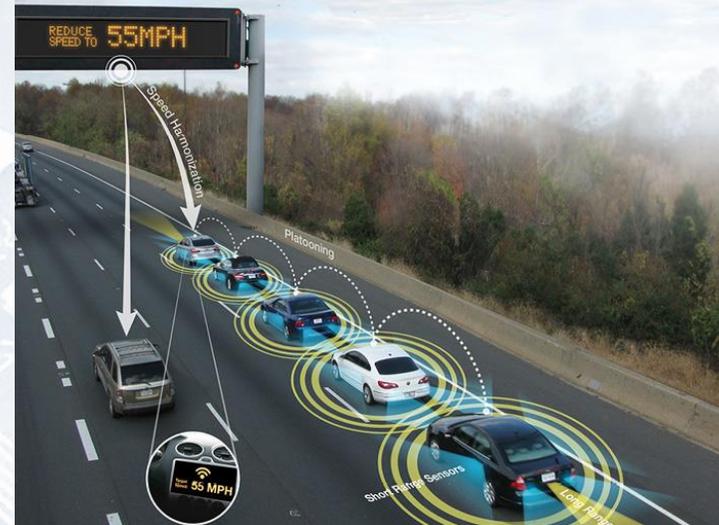
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^b Institute for Intelligent Systems Research and Innovation, Deakin University, Australia

Motivation:

- Manual versus autonomous vehicles
- Also on traffic network level



Autonomous vehicles at junctions

→ autonomous intersection without physical traffic lights

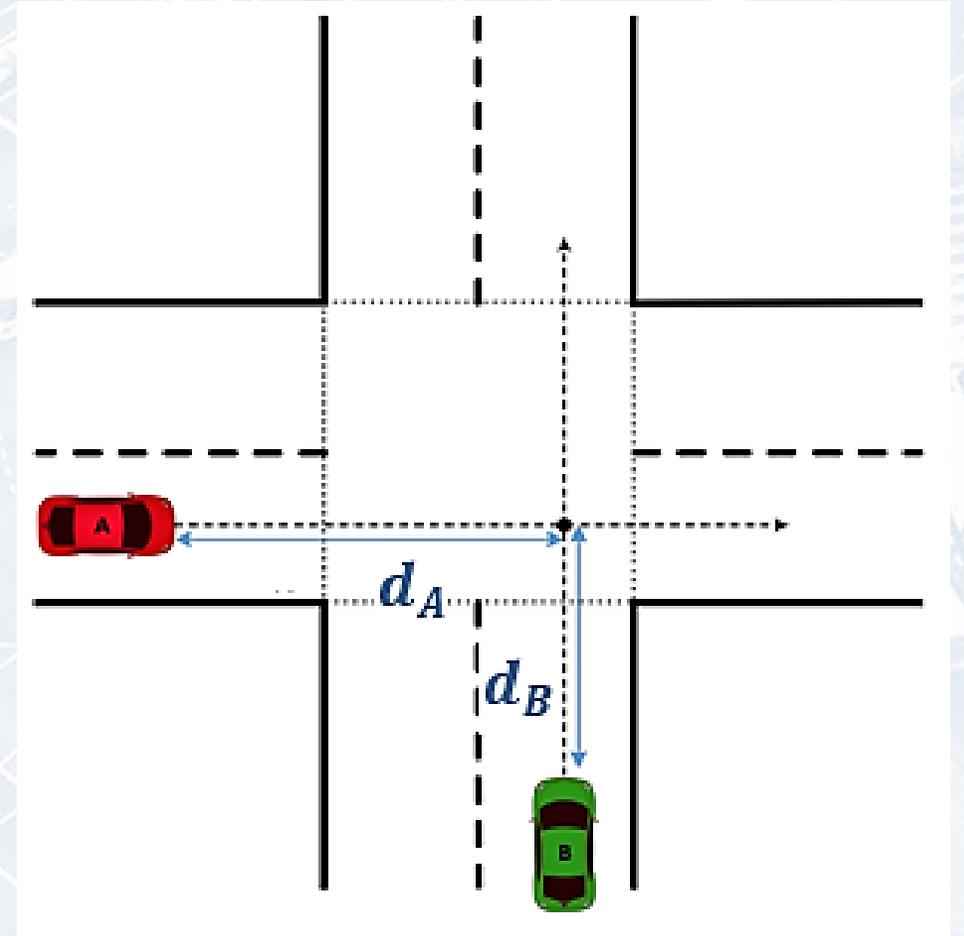


Junction traffic model

$$\vec{S}(k+1) = \vec{S}(k) + \vec{d}(\vec{S}(k)) \cdot v(k) \cdot T$$

To avoid collision, condition $|d_A - d_B| > d_{min}$ must be valid at all times.

This constraint is considered later in the control design.



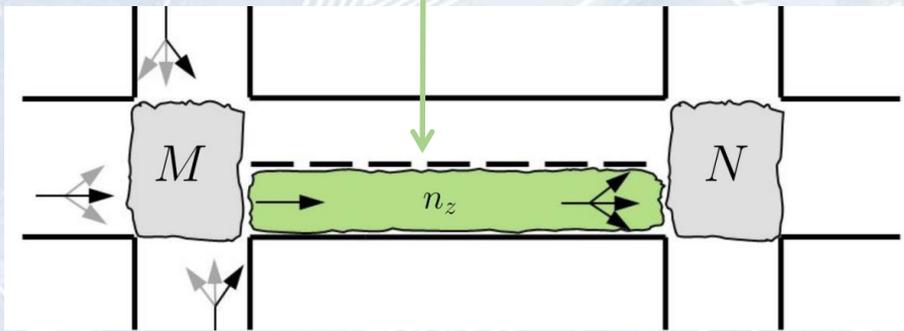
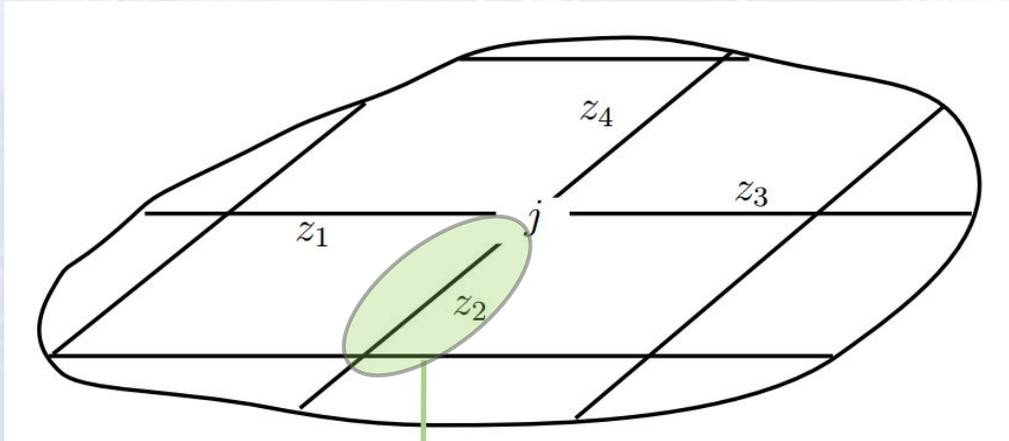
Emission model was also considered in the control design

- Traffic emission mainly consists of CO , NO_x , HC , and CO_2 .
- For microscopic (vehicle based) emission the COPERT IV model was adopted:

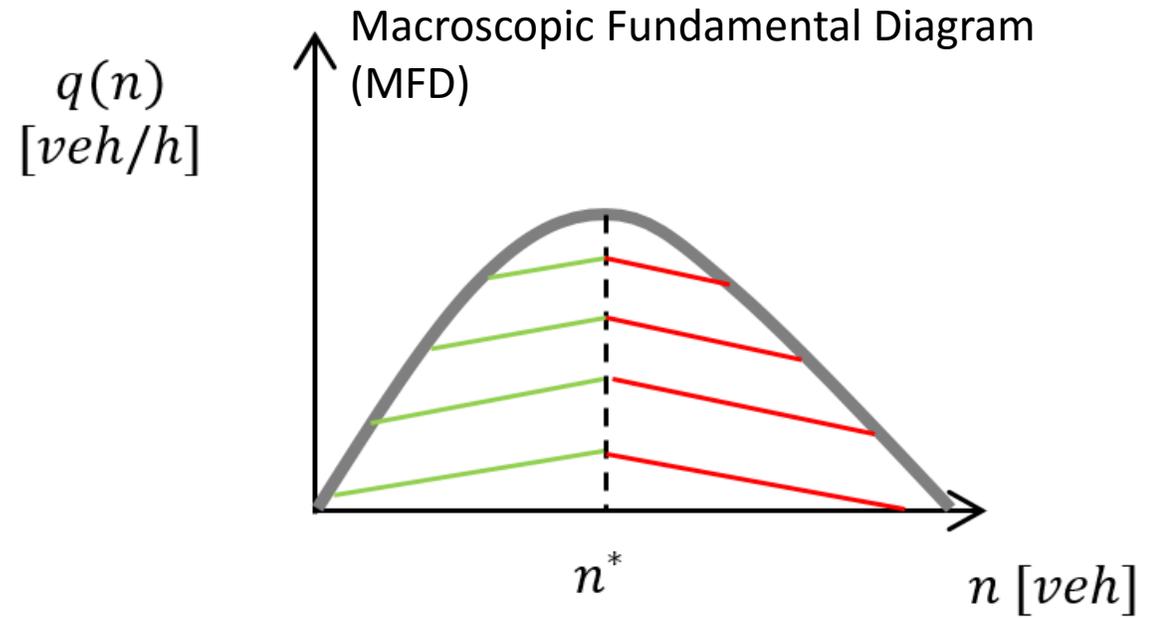
$$ef(v) = \alpha_2^p v^2 + \alpha_1^p v + \alpha_0^p,$$

where α_i^p denotes the emission parameters for pollutant p

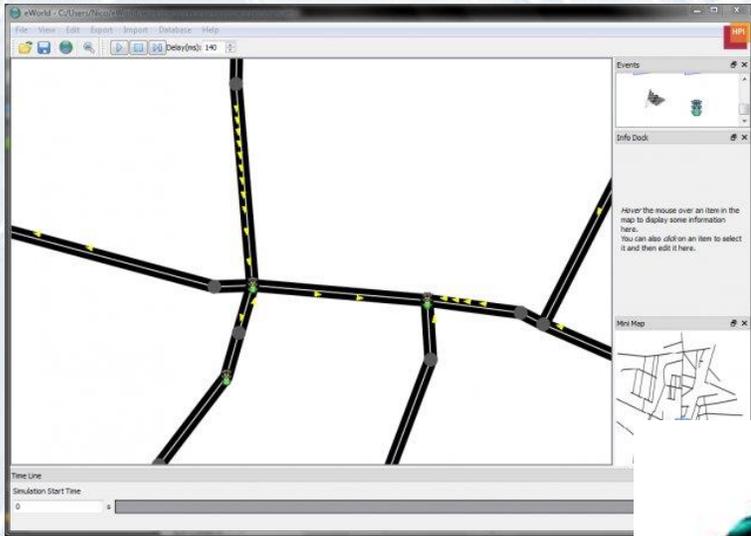
Network traffic model



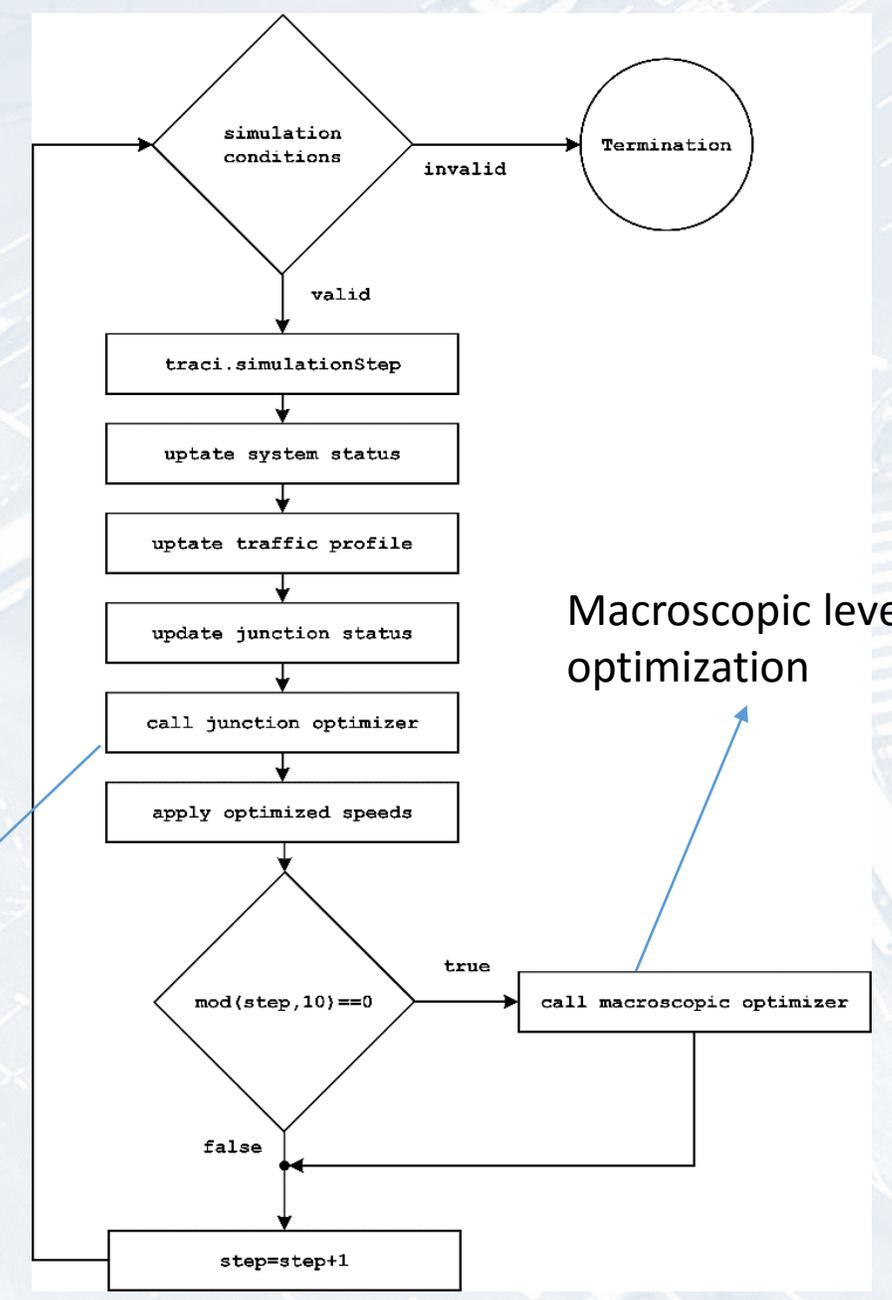
$$n_z(k+1) = n_z(k) + T \left[\sum_{w,z} \alpha_{w,z} q_{w,z}(k) - q_z(k) \right]$$



Two level optimization using SUMO traffic simulator and MATLAB



Microscopic level optimization



Low level control in order to avoid collision

→ Nonlinear Model Predictive Control (MPC)

Highest mobility
Lowest emission

$$\begin{aligned} & \min_{u(k+l-1)} J(k), \\ & \text{s. t.} \quad u(k+l-1) \in \mathbb{U}, \\ & \quad \quad x(k+l) \in \mathbb{X}, \\ & \quad \quad l = 1, 2, \dots, K \end{aligned}$$

Constrained optimization

$$J(k) = \sum_{l=1}^K \sum_{i=1}^{N_{vehicles}} [\alpha[1 + p_i]v_i^2(k+l) + \beta[v_i(k+l) - v_i(k+l+1)]^2 + \gamma ef_i^2(v_i(k+l))],$$

$$\begin{aligned} D &= [d_{il}]_{N_{vehicles} \times K}, \\ \min(D) &> d_{\min} \end{aligned}$$

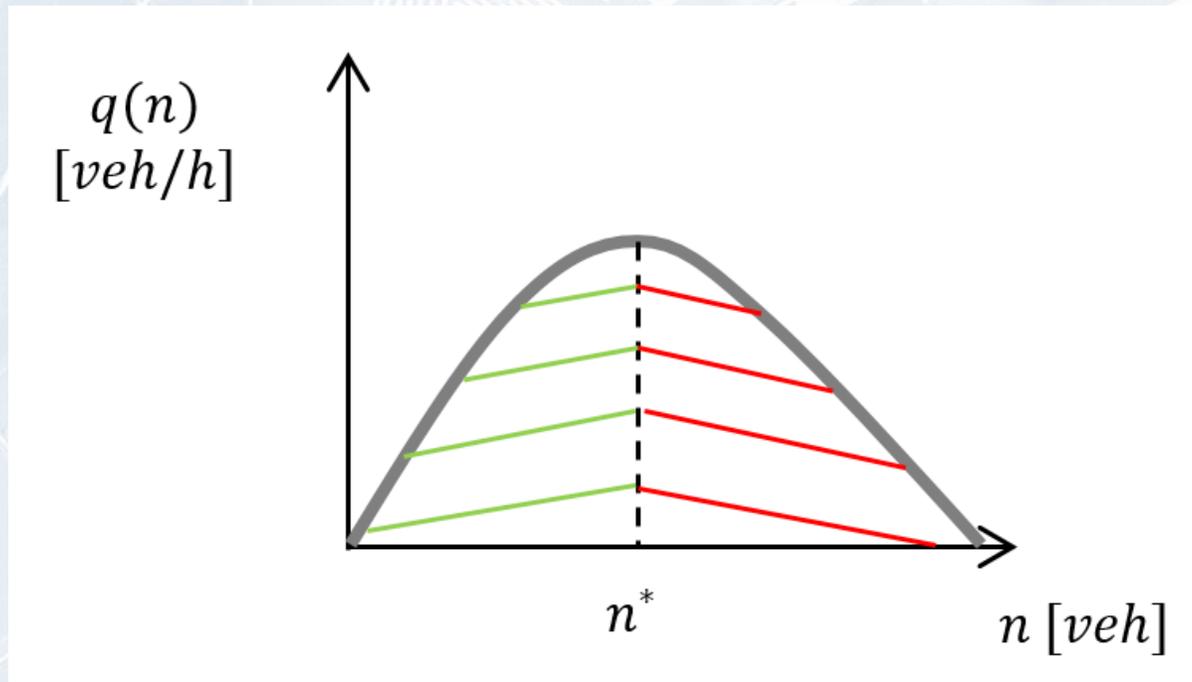
Avoid collision

Acceptable distance

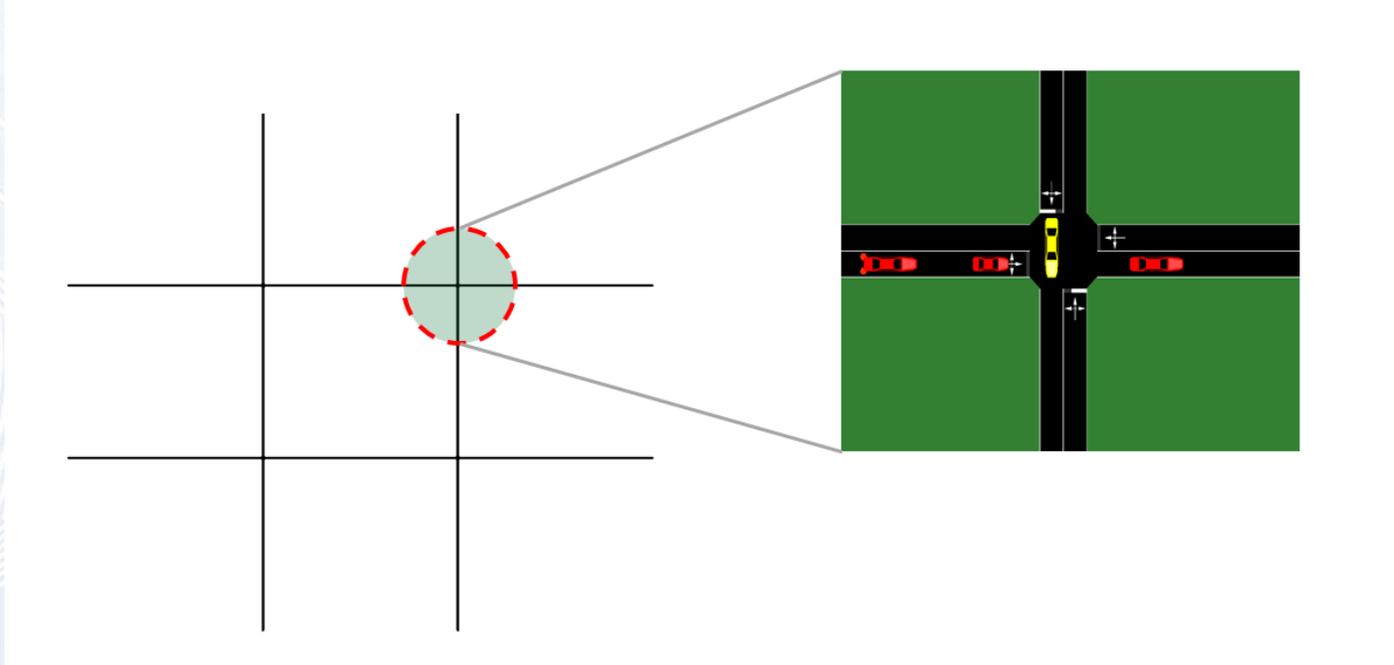
Priority parameter

Macroscopic level control

Priority parameter is calculated based on the MFD model:
$$p_z = \begin{cases} 0, & n_z \leq n_z^* \\ \frac{n_z - n_z^*}{n_z^*}, & n_z > n_z^* \end{cases}$$



Simulations



MATLAB+SUMO (TraCI) <http://www.dlr.de/>

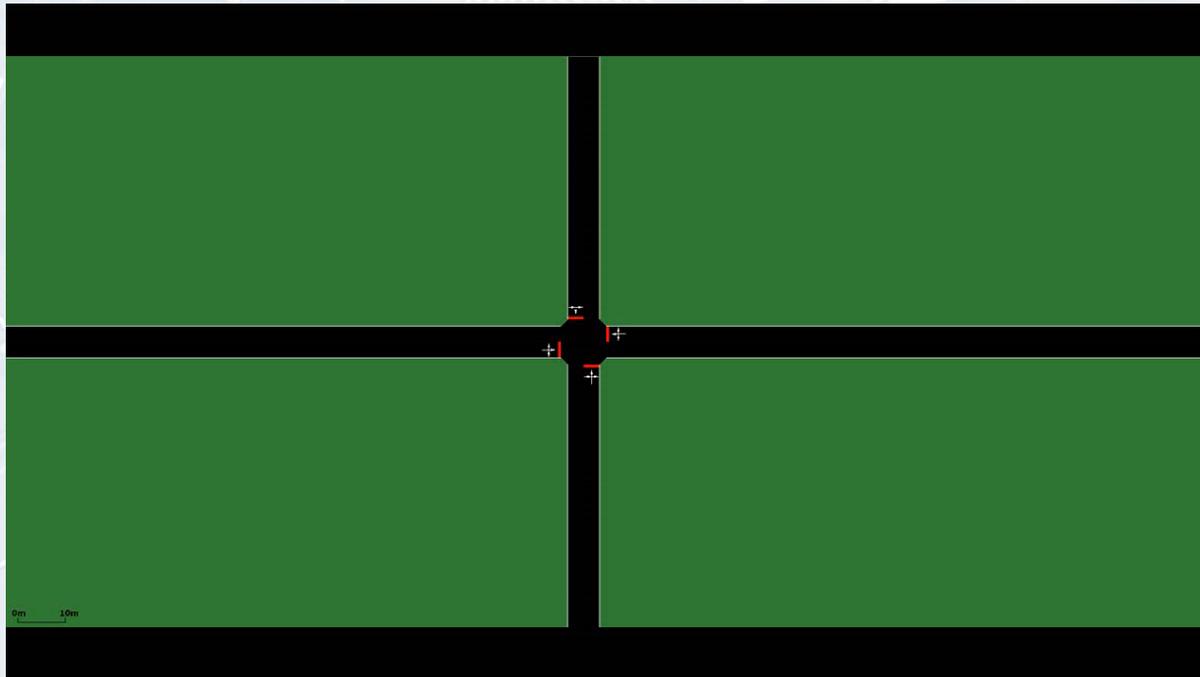
Test network with 4 intersections

Prediction and control horizons = 20 seconds

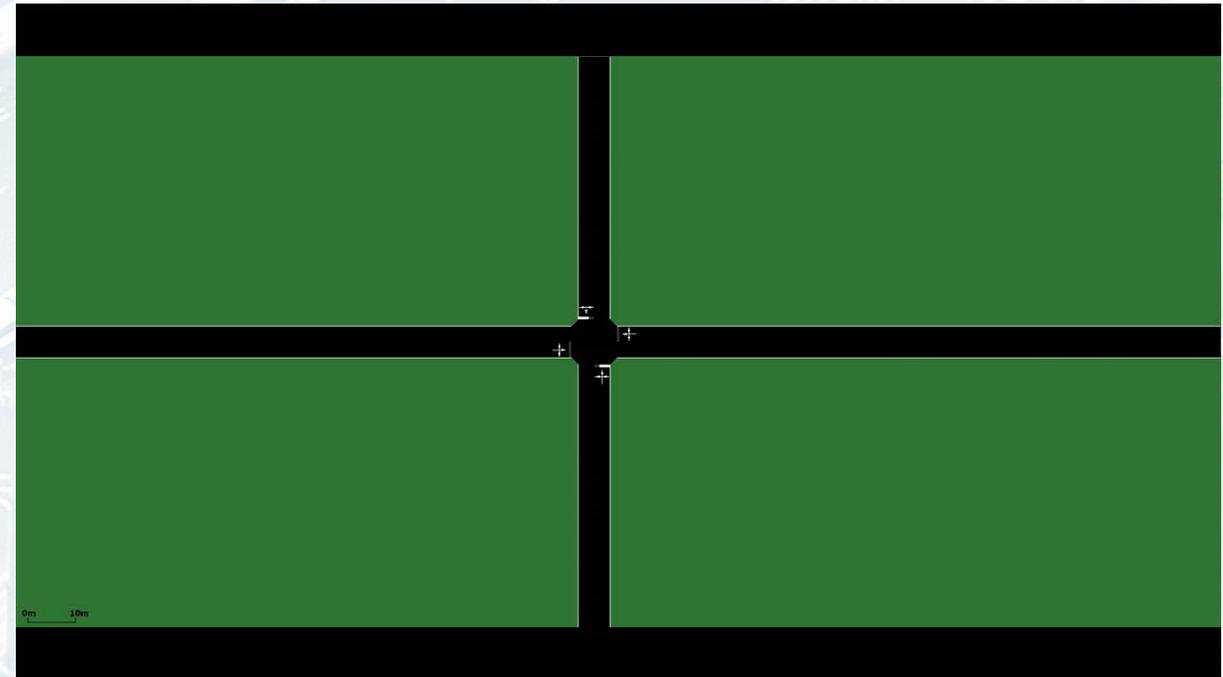
MPC optimizer: nonlinear (fmincon)

Visual results

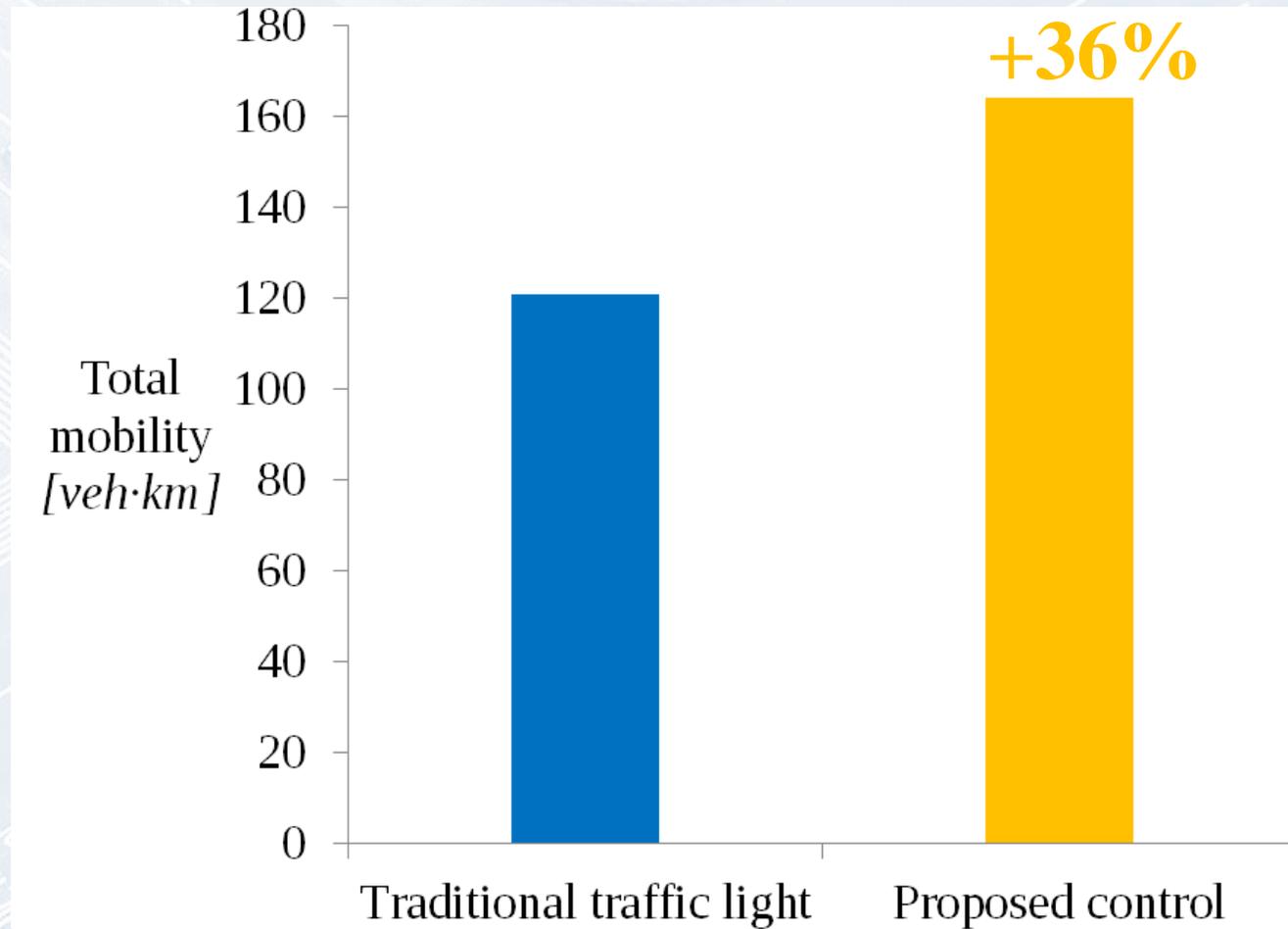
Actuated
(time gap based actuated control)



Autonomous intersection control

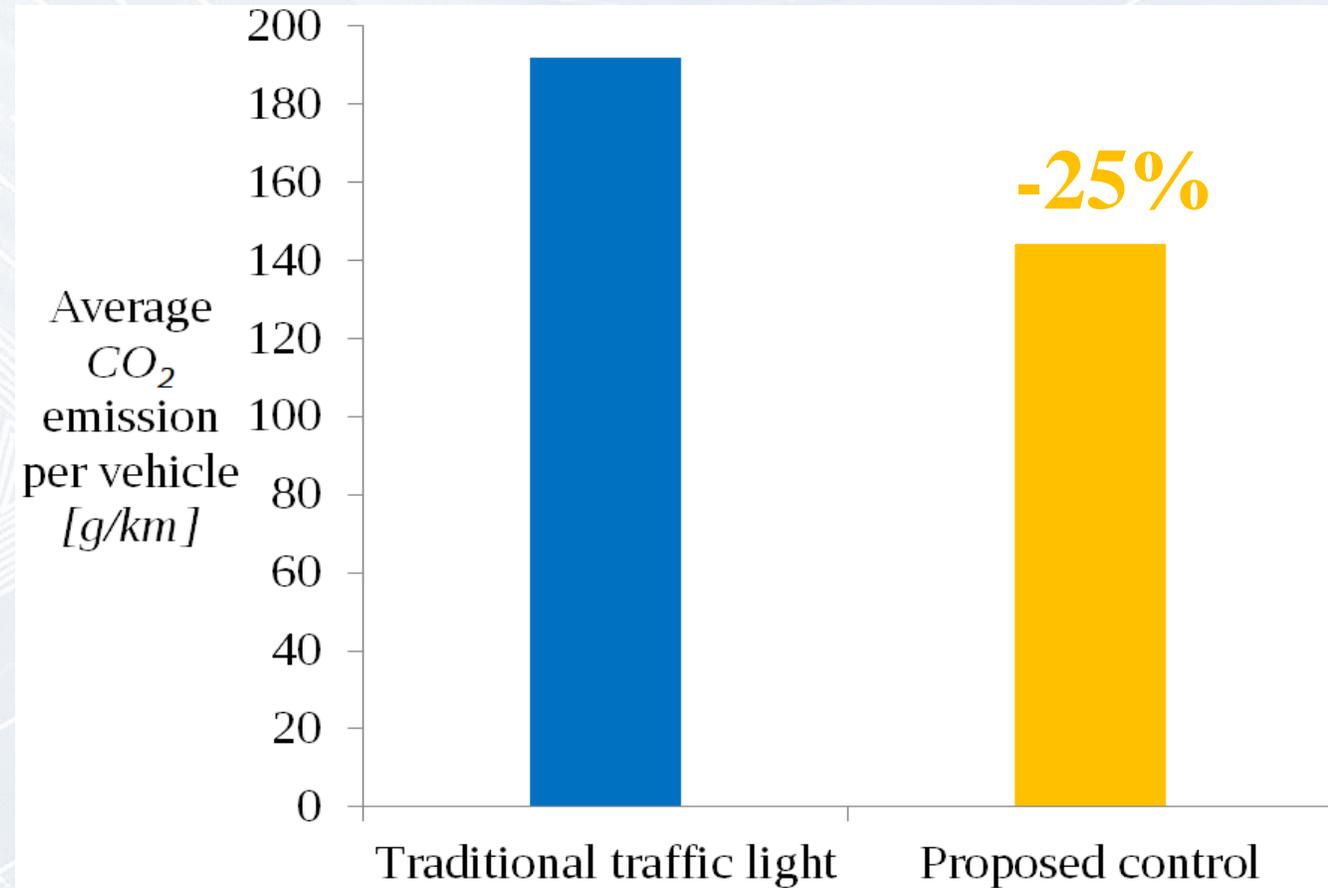


Results



The comparison of network mobility between the traditional and proposed methods.

Results



The comparison of CO_2 emission (based on HBEFA v3.1 data) between the traditional and proposed methods.

Conclusion and future work

- The performance of proposed control was justified:
 - High performance
 - Higher mobility
 - Lower emission
- Problems to overcome within the control design:
 - Disturbance can be present in the system, e.g. pedestrian crossing
 - Solution for the transition period (when traditional and autonomous cars are running together)

Thank you for your attention!!

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