

An Optimal Control Framework for Reliable & Sustainable Bus Operation

GSC'25

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Agenda



Motivation



Methodology



Case Studies



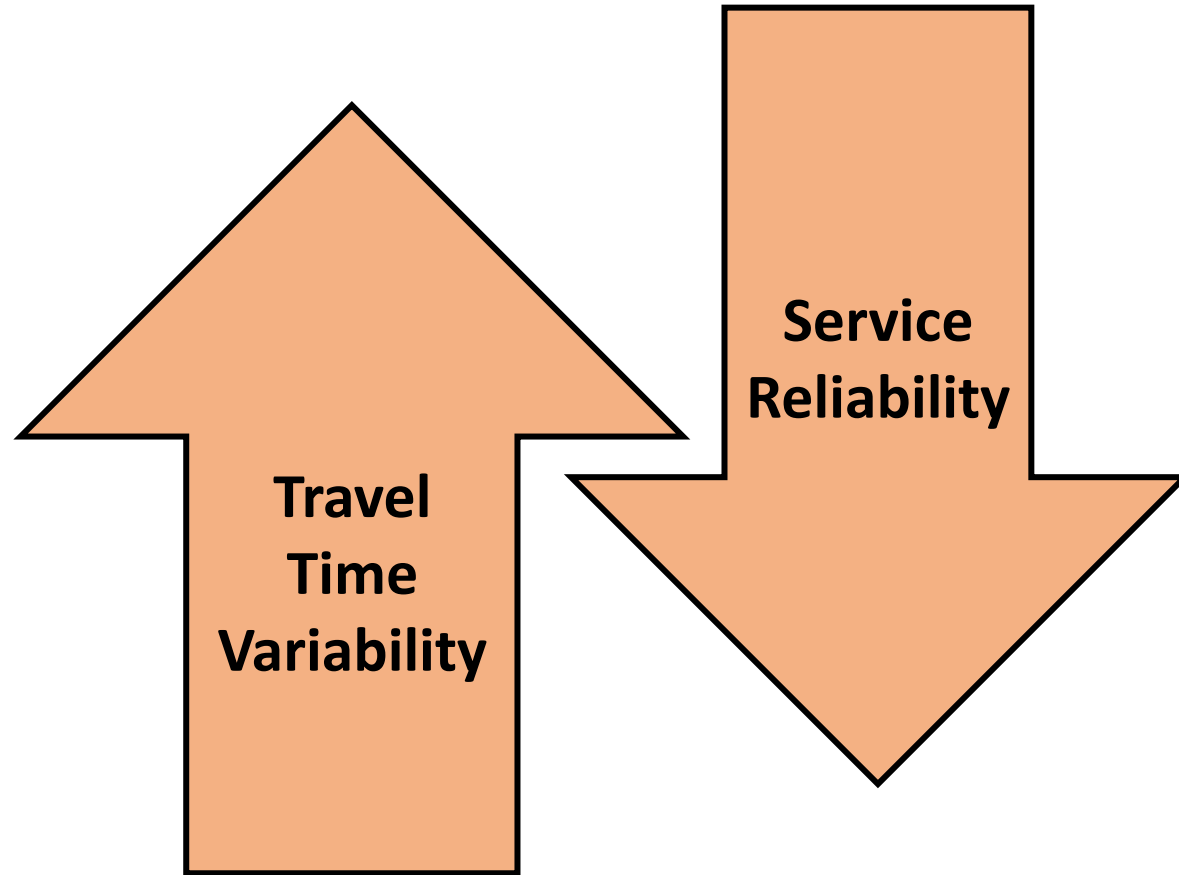
Results



Conclusion Remarks

Motivation

An optimal control framework for reliable and sustainable public transit operation



By Felix O - 55 Buses, Clapton Pond Uploaded by oxyman, CC BY-SA 2.0,
<https://commons.wikimedia.org/w/index.php?curid=10935479>

Motivation

An optimal control framework for reliable and sustainable public transit operation



To provide an efficient and sustainable service



To deliver service as planned

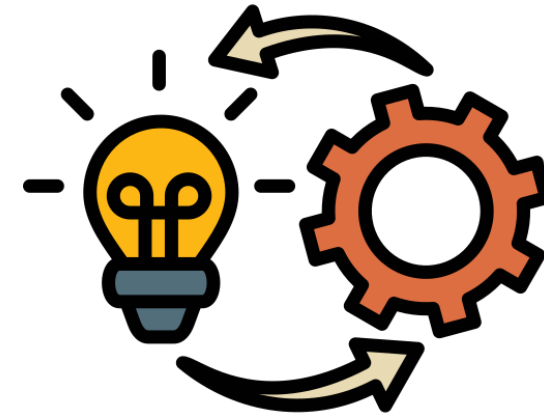
Motivation

An optimal control framework for reliable and sustainable public transit operation



State of the art

The two objectives have been studied separately

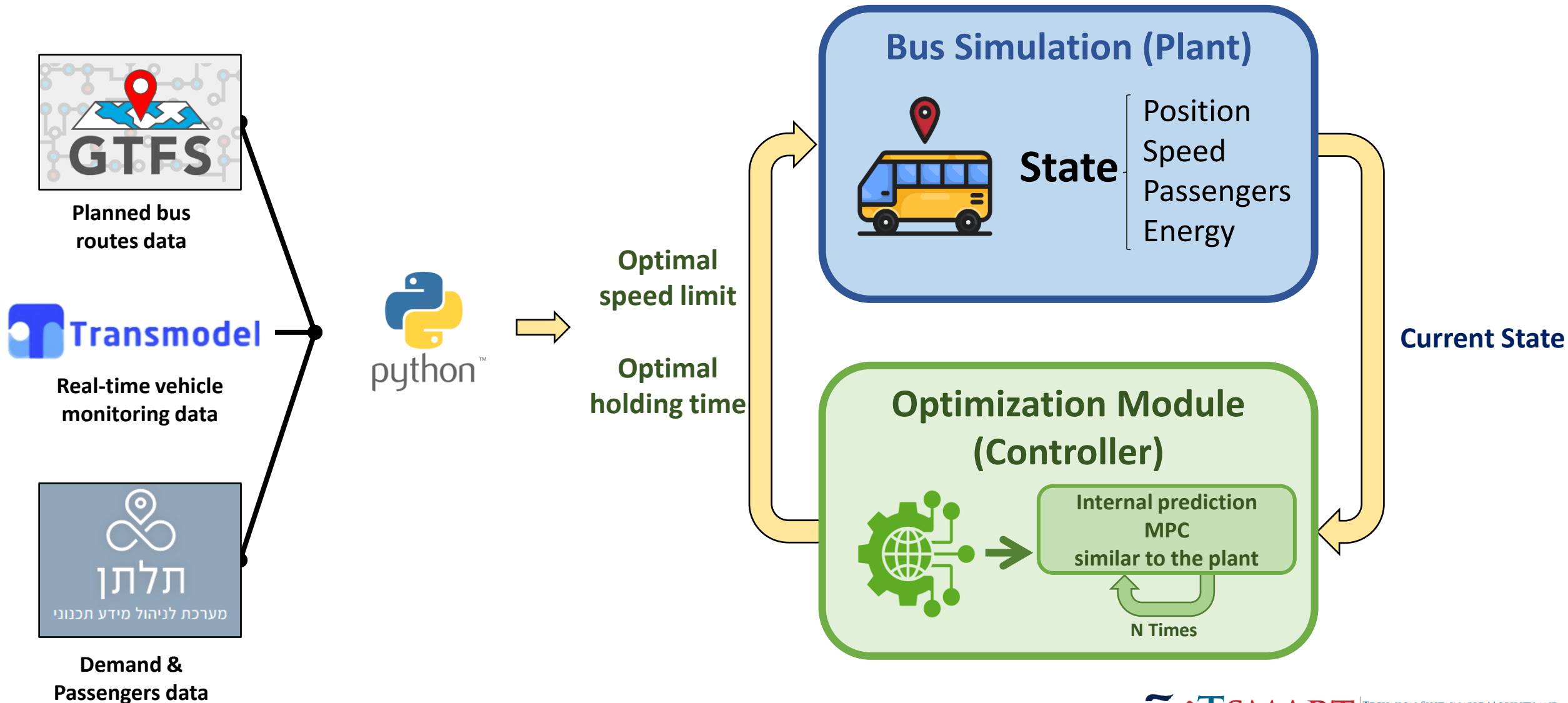


State of the Practice

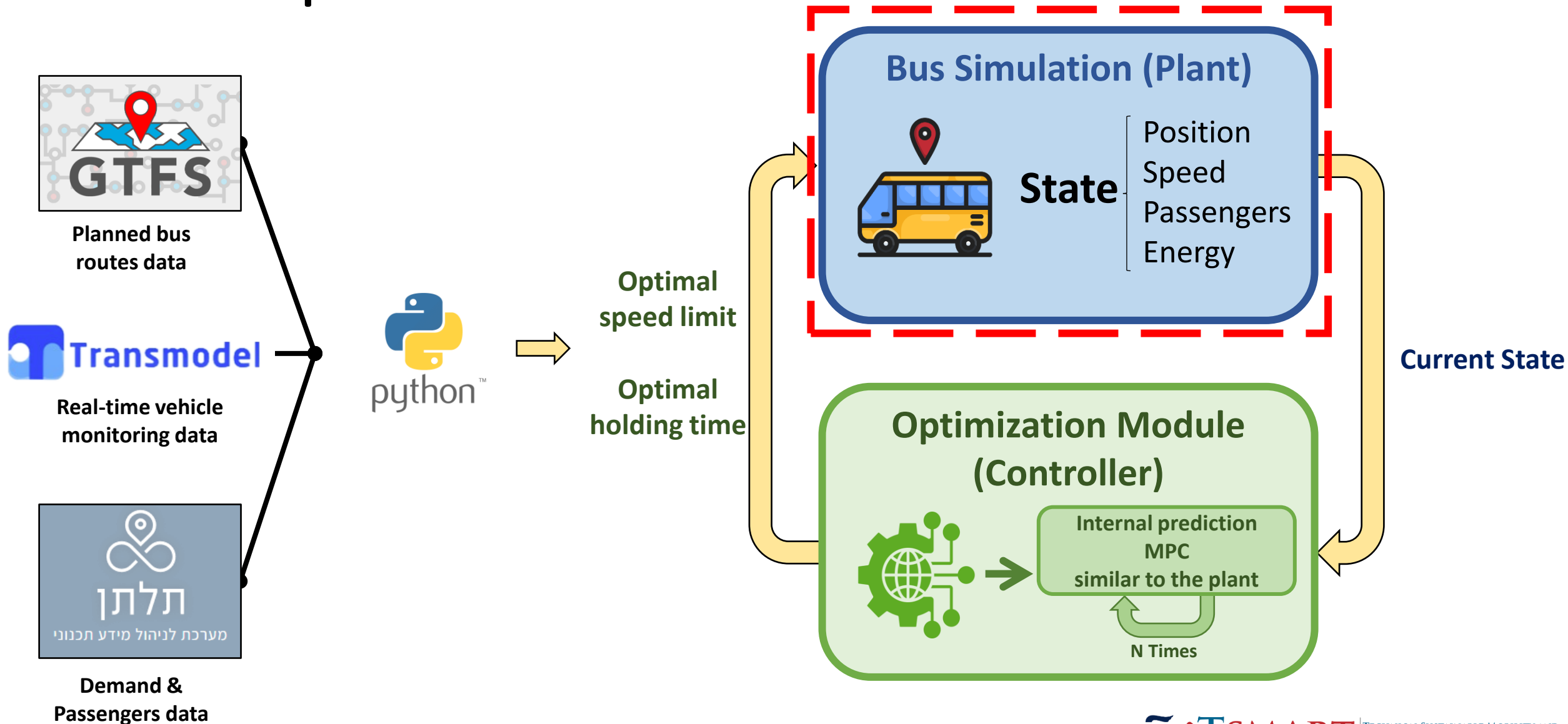
Initial approaches have seen limited implementation globally

Methodology

Developed Simulation Framework

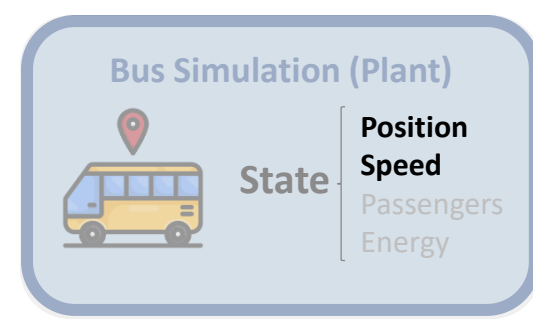


Developed Simulation Framework



Bus Transport System Modeling

Simulating Traffic Conditions



Deviations from historical average*

$$\sigma_{traffic}^2 = 5 \text{ m}^2/\text{s}^2$$



Traffic state is changed every 10 seconds

$$Traffic = \max(0, \min(SL, \mathcal{N}(avg. speed, \sigma_{traffic}^2)))$$



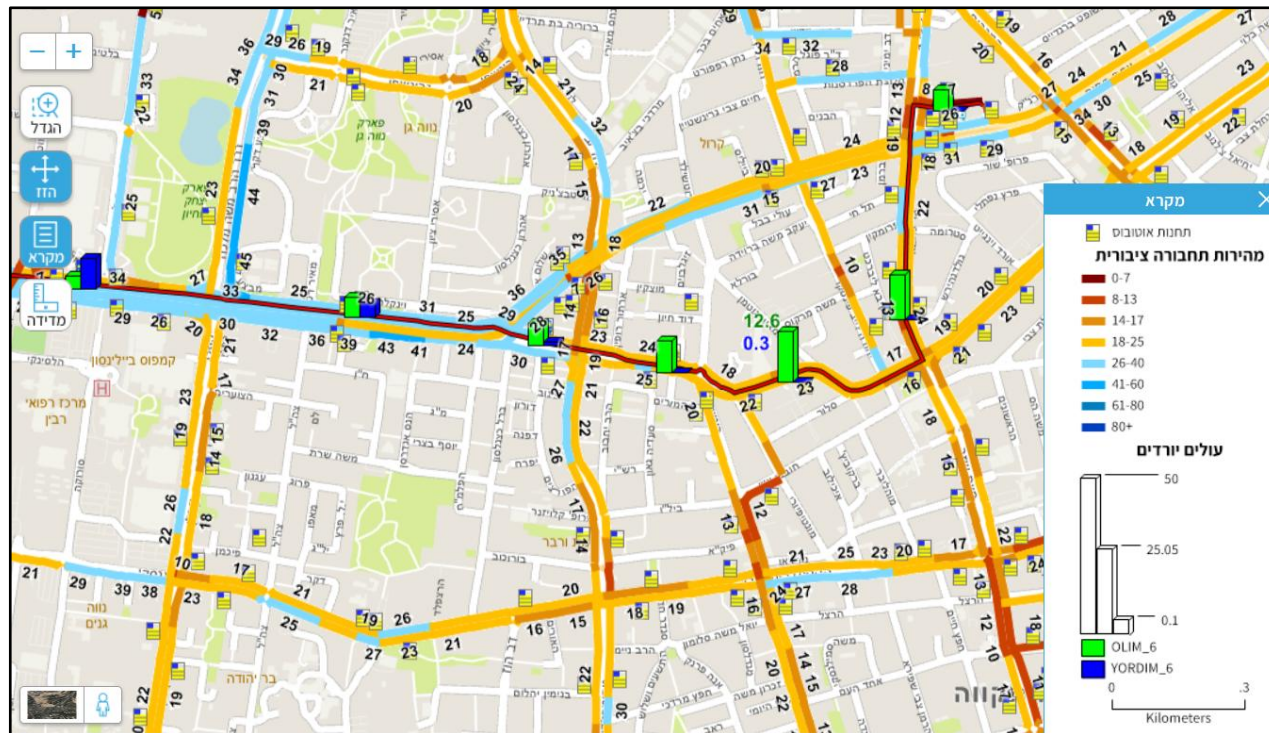
Smoothing Factor*

$$\alpha = 0.2$$



Actual bus speed

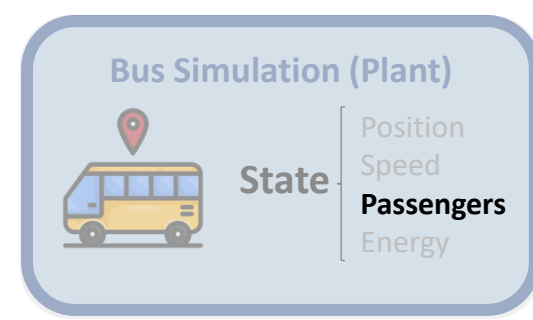
$$Speed(t) = \alpha \cdot Speed(t - 1) + [1 - \alpha] \cdot Traffic$$



<https://tiltan.ayalonhw.co.il/tiltan/App.aspx>

Bus Transport System Modeling

Simulating Passengers & Demand



Bus i passenger accumulation n at stop s :

$$n_i^s = \lambda_i^s(t_i^s - t_{i+1}^s) + (1 - \mu_i^s) * n_i^{s-1}$$



Bus i is expected to find $\lambda_i^s(t_i^s - t_{i+1}^s)$ passengers, as λ varies at each stop following demand profile

$\mu_i^s \rightarrow$ alighting proportion at stop s following demand profile

$$0 \leq n_i^s \leq n_{\max}$$

$$0 \leq \mu_i^s \leq 1$$

$$0 \leq \lambda_i^s$$

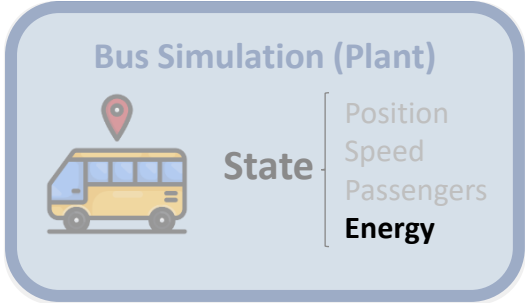
Mass of bus i :



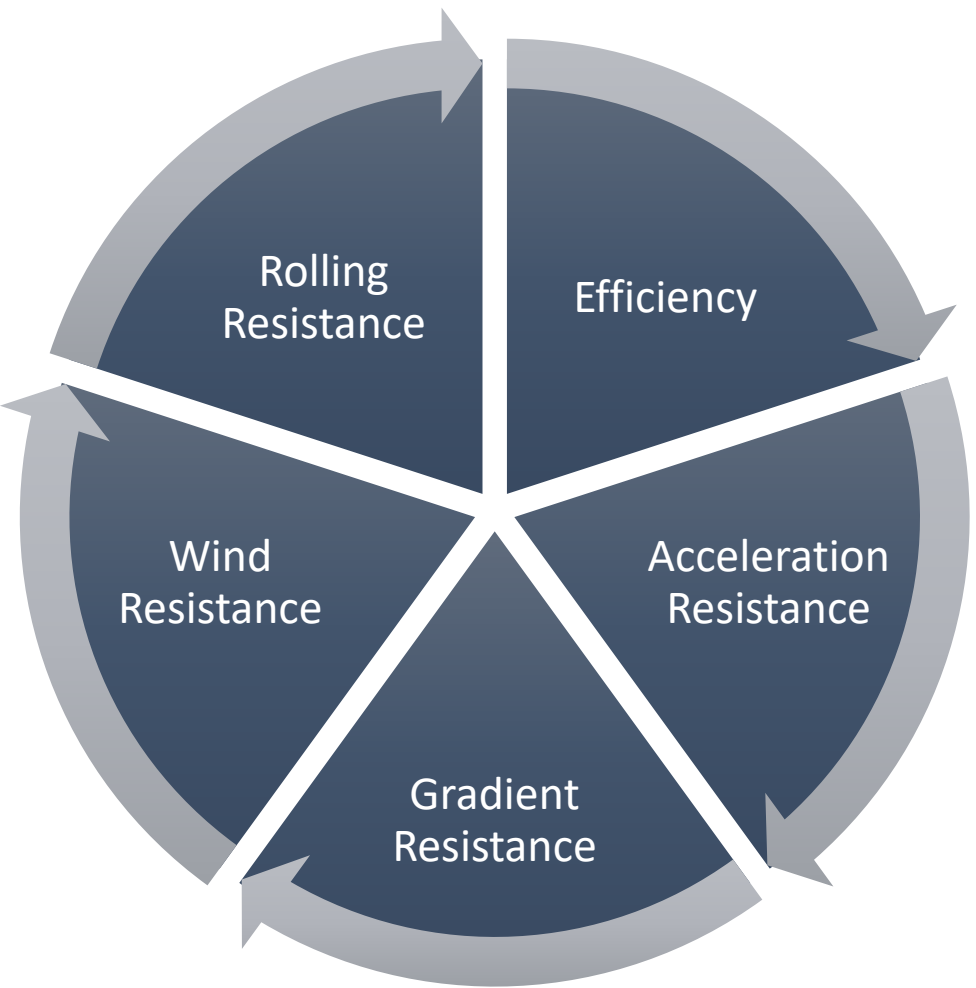
$$m_i(t) = m_{\text{emp}} + n_i(t) * m_{\text{pax}}$$

$m_{\text{pax}} \rightarrow$ An average passenger mass [kg]

$$m_{\text{emp}} \leq m_i(t) \leq m_{\text{full}}$$



Energy Model



Power $\rightarrow P = \frac{v}{\eta_t} \cdot F_t = \frac{v}{\eta_t} \cdot (F_r + F_w + F_i + F_j) + \underbrace{v(t) * m(t) * a(t) * \eta_{KERS}}_{\text{Regenerative Braking}}$

Energy $\rightarrow EC = \int_0^T P(t)dt \text{ [kWh]}$



η_t

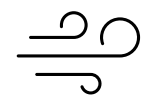
Total Efficiency



F_r

Rolling resistance

$m_i(t) \cdot g \cdot C_{rr} \cdot \cos(\alpha)$



F_w

Air resistance & drag

$\frac{1}{2} \rho \cdot A \cdot C_D \cdot v_i^2$



F_i

Gradient resistance

$m_i(t) \cdot g \cdot \tan(\alpha)$

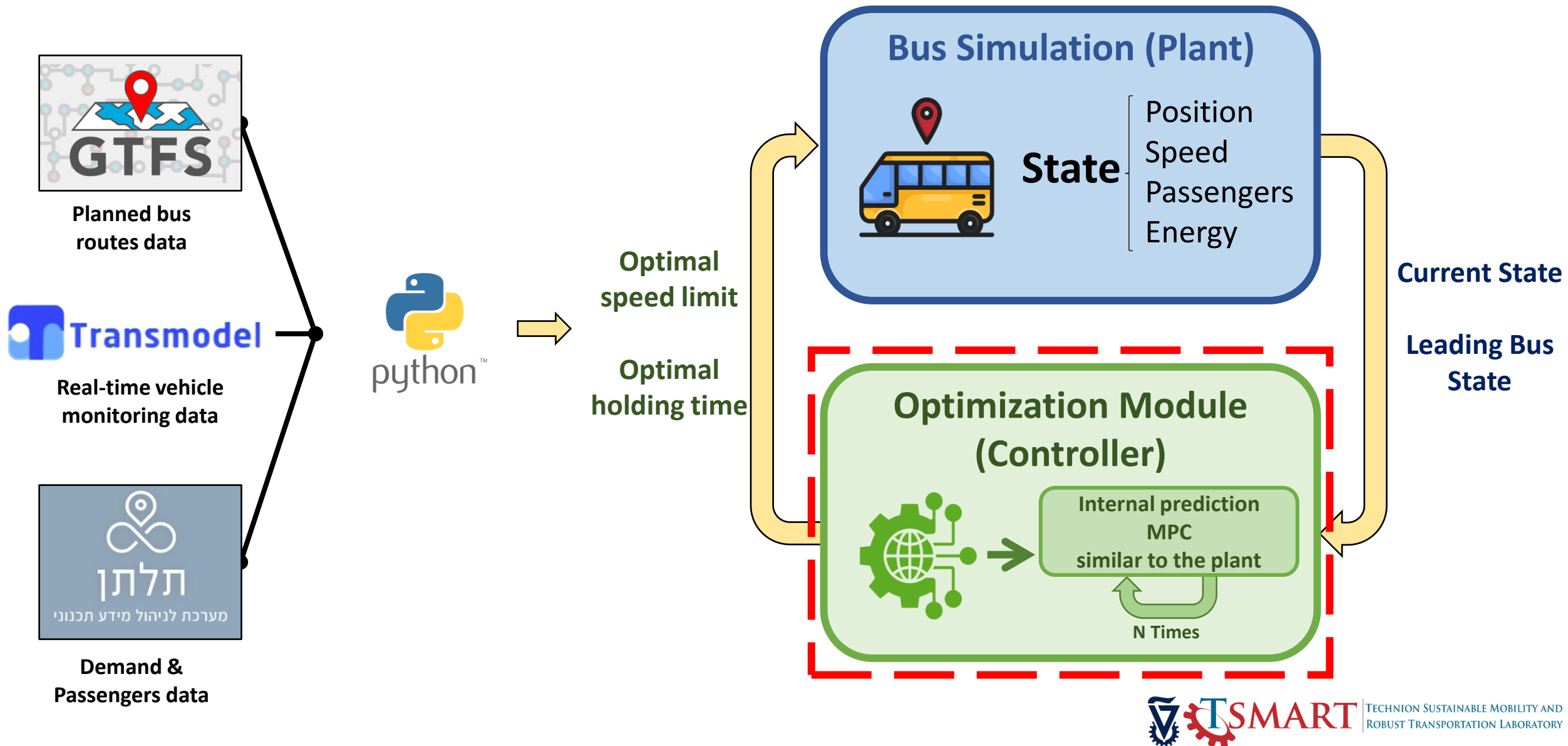


F_j

Acceleration resistance

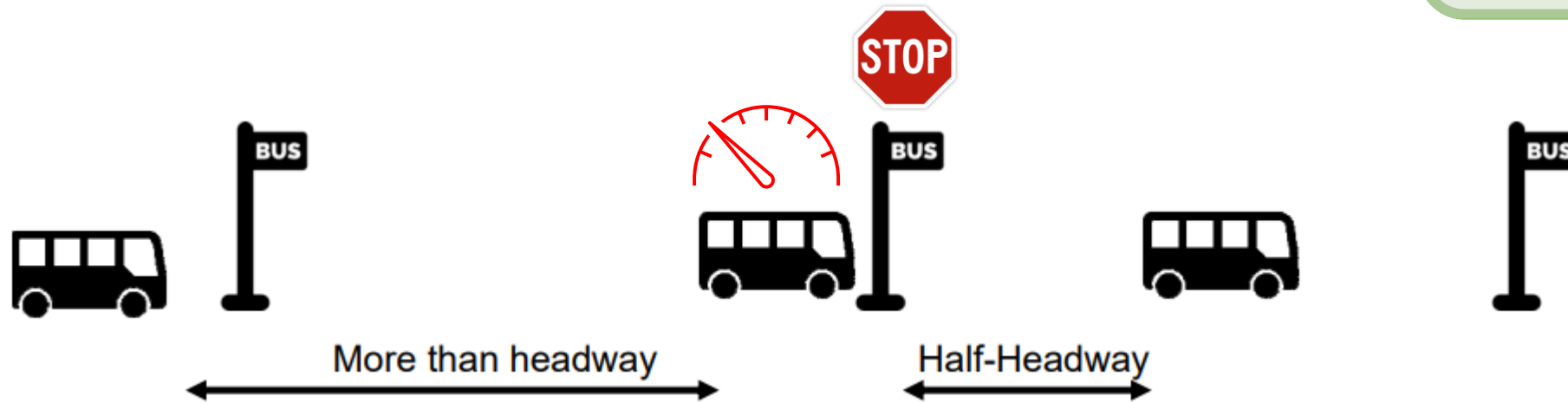
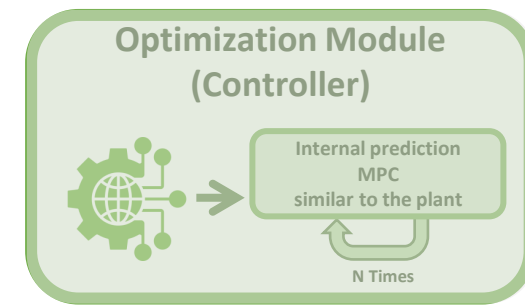
$m_i \cdot \frac{\partial v}{\partial t}$

Developed Simulation Framework



Control System

Speed Limit & Holding Adjustments



Control variables



u_h

Holding time at the current stop



u_{SL}

Speed limit for the next section

State variables

Position

Arrival times at stops

Passengers on board

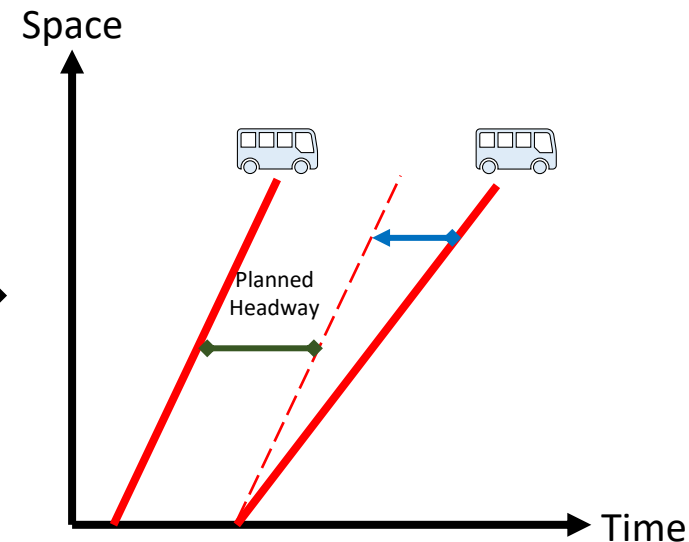
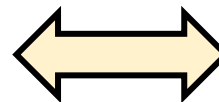
Speed

Energy consumption

Demand profiles

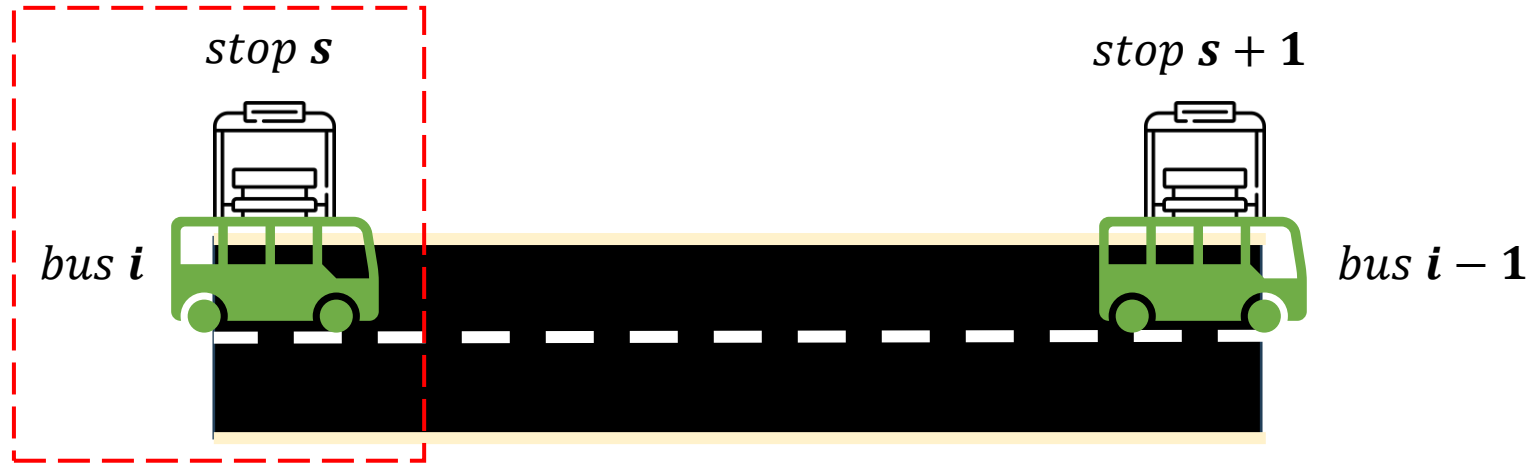
⋮

And many others



Optimal Control Framework

Event-based optimization occurs for bus i at each stop s



$$\min \sum_{i=1}^n \omega_1 \cdot \left(H_{\text{ref}} - H_{i-1,i} \right)^2 + \omega_2 \cdot EC$$

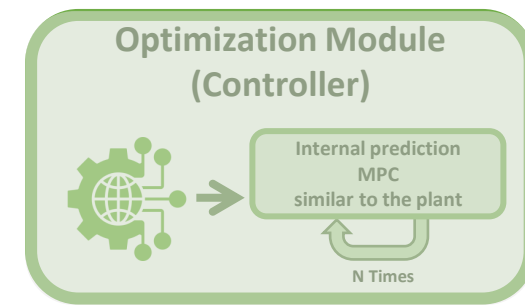
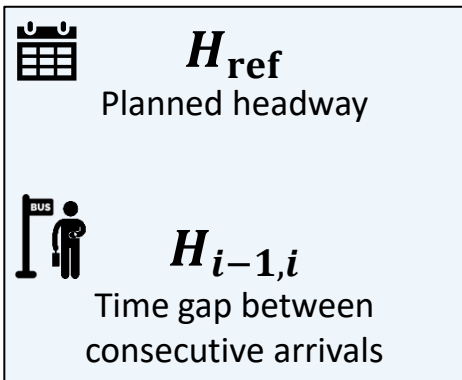
s.t.

$$0 \leq u_h(t) \leq u_{h_{\max}} \quad \text{Holding time constraint}$$

$$0 \leq u_{SL}(t) \leq SL_{s,s+1} \quad \text{Speed Limit constraint}$$

⋮

All other state constraints



Calibration

Weights are calibrated using the **Weighted Sum Approach**

Prediction

Decentralized Model Predictive Control

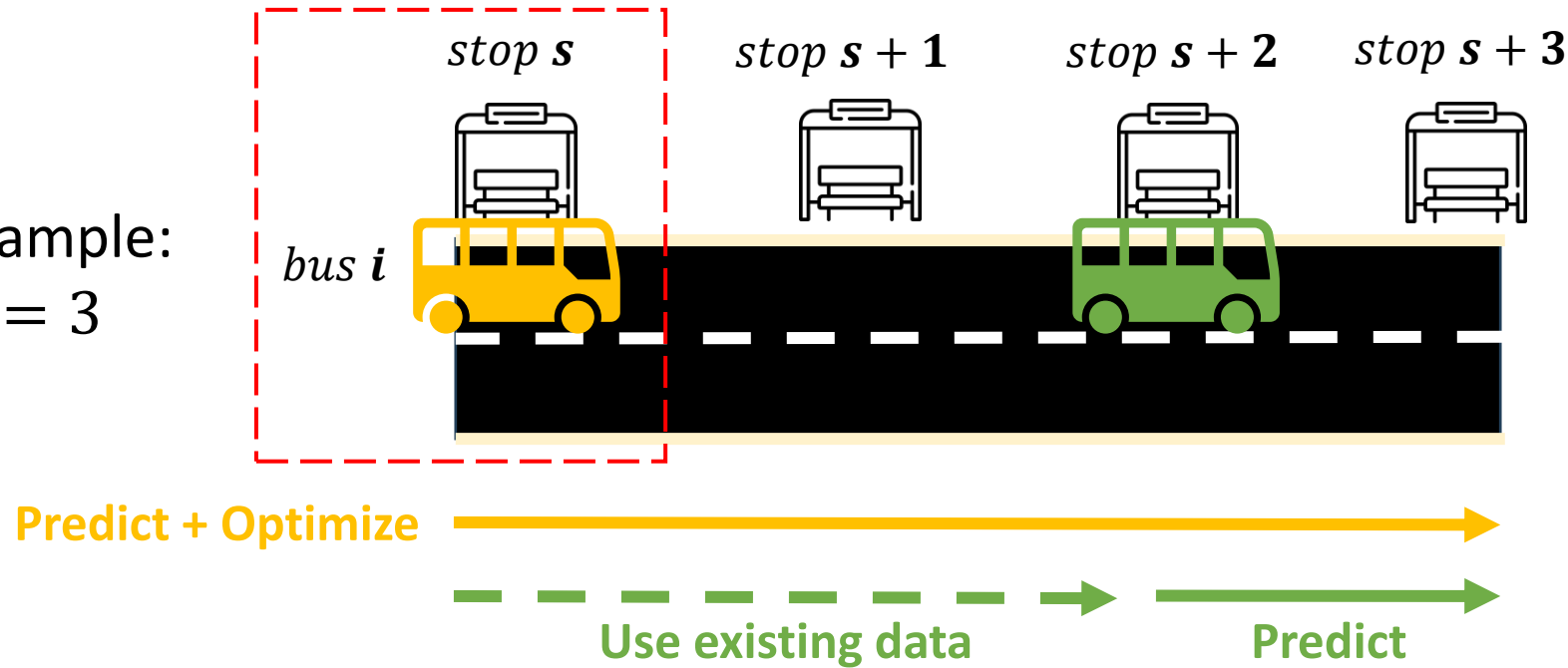
Optimization

SQP Algorithm
Sequential Quadratic Programming

Optimal Control Framework

Decentralized Model Predictive Control

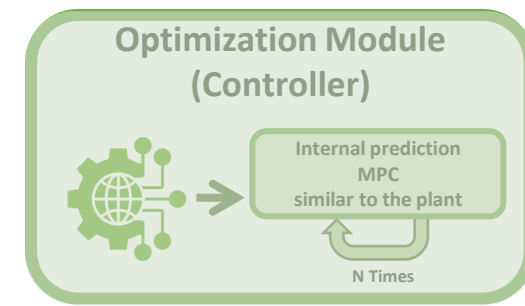
For example:
 $N_p = 3$



Chosen Horizons

Prediction horizon $N_p = 2 \text{ stops}$

Control horizon $N_c = 1 \text{ stop}$



Calibration

Weights are
calibrated using
the **Weighted
Sum Approach**

Prediction

Decentralized
**Model
Predictive
Control**

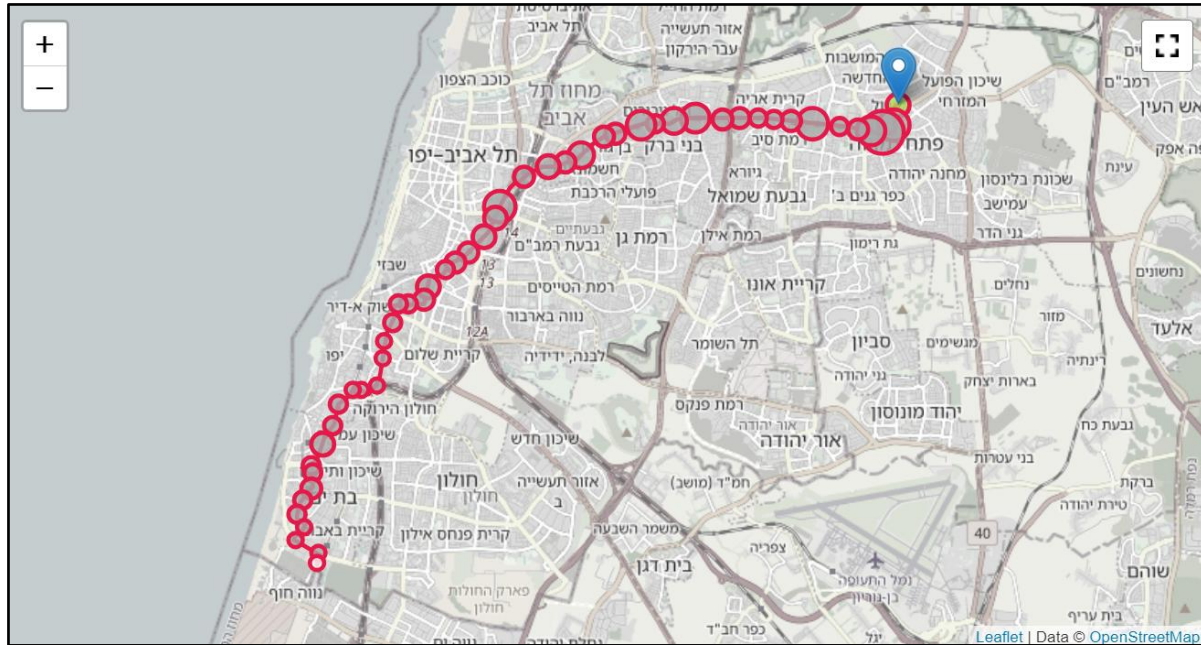
Optimization

SQP Algorithm
Sequential
Quadratic
Programming

Case Studies & Results

Line 1 (81001) – Simulation

Bus Transport System Setting



*<https://markav.net/line/81001/>

Parameter	Numerical Value	Units
Route length	22,000	$[m]$
Number of buses	6	$[-]$
Number of stops	51	$[-]$
Bus headway	10	$[min]$
Speed boundaries	$[0, 50]$	$\left[\frac{km}{hr}\right]$
Acceleration boundaries	$[-2, 1]$	$\left[\frac{m}{s^2}\right]$
Passenger accumulation boundaries	$[0, 80]$	$[pax]$

Key Metric of Stability – Cvh

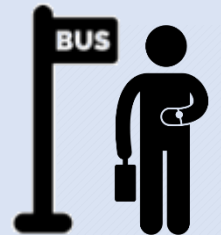
Coefficient of Variation of Headways

$$C_{vh} = \frac{\sigma_{Hd}}{H}$$

⇒ σ_{Hd} - Standard deviation of the deviations from the scheduled headway

⇒ H - Scheduled headway between two consecutive trips

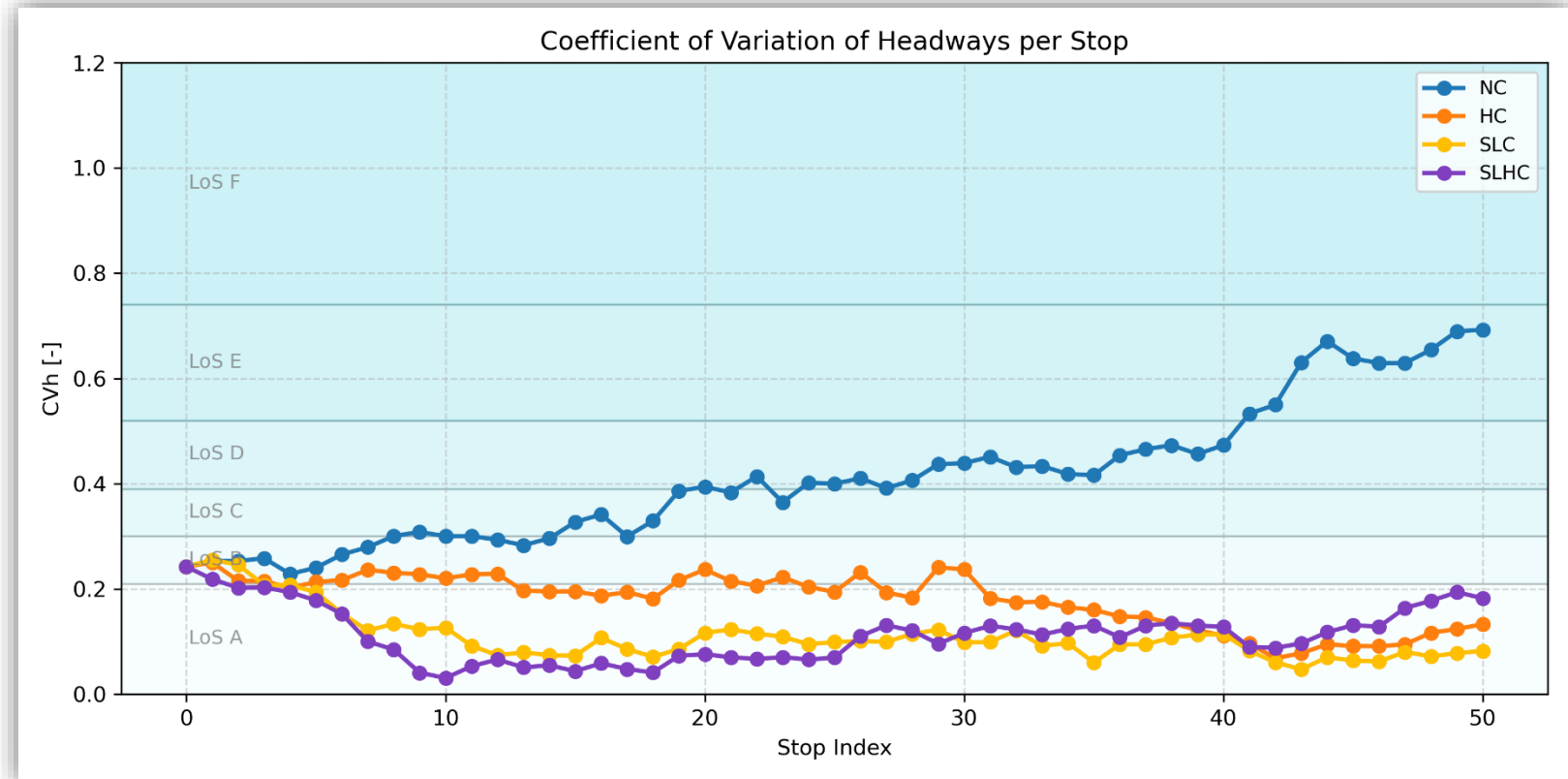
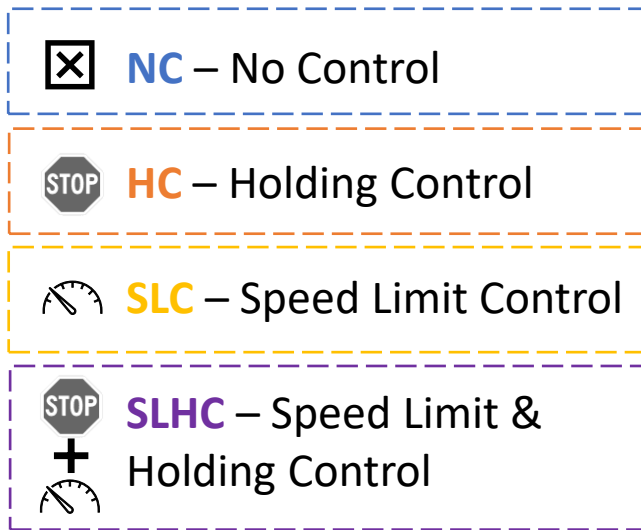
Example



1. The scheduled headway of a certain line is every 10 minutes.
2. In practice, the measured headways at a certain stop are 12, 8, 14, 6, 13, 7.
3. Therefore, the deviations from the scheduled headway are +2, -2, +4, -4, +3, -3.
4. The standard deviation of the deviations from the headway is 3.4.
5. Hence, the value of the headway variance index is 0.34 and the level of service is “C” (Moderate)*.

Results – Line 1 (81001)

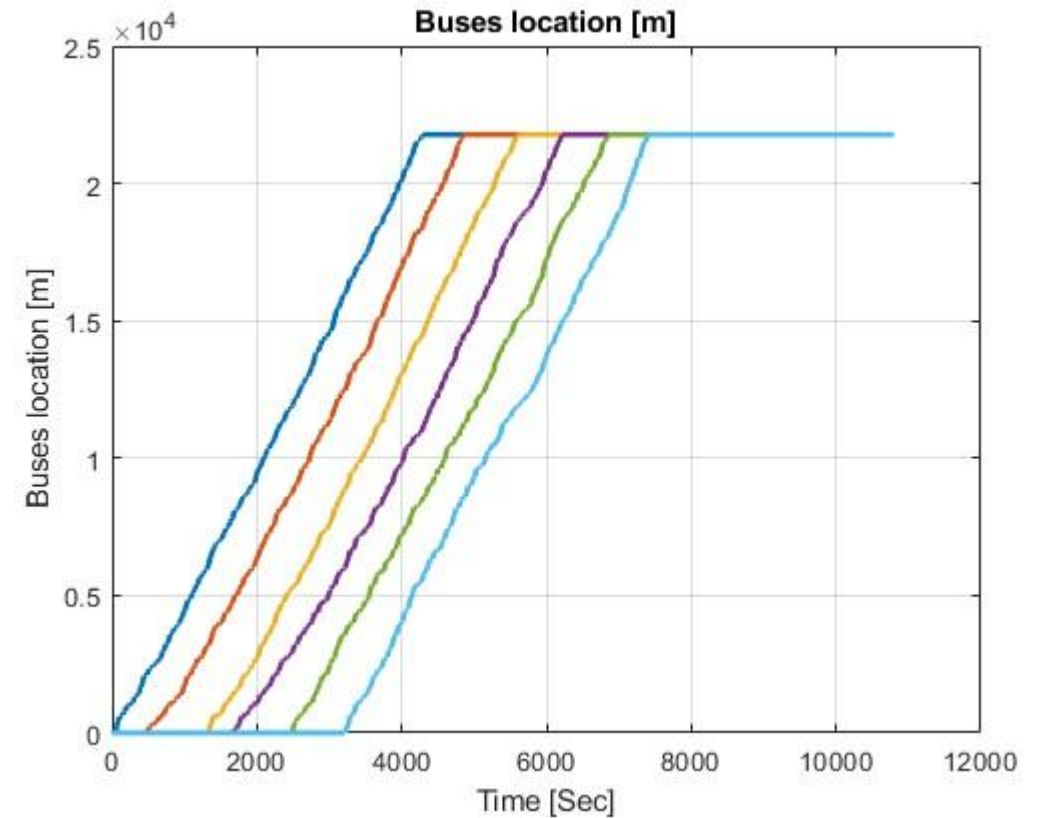
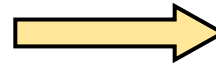
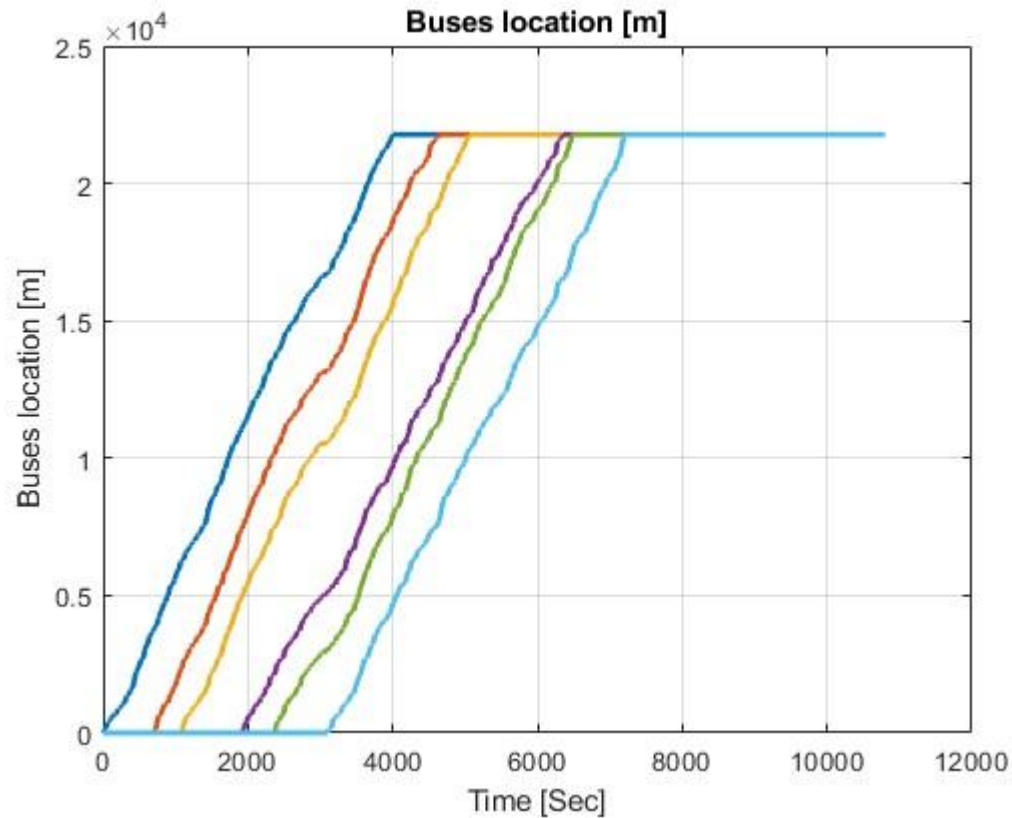
Moderate inconsistency in departures was examined



Cvh of Line 1 – 81001

Results – Line 1 (81001)

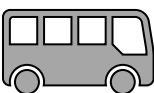
Moderate inconsistency in departures was examined



Examined Routes – Peak Hours [6-8 AM]

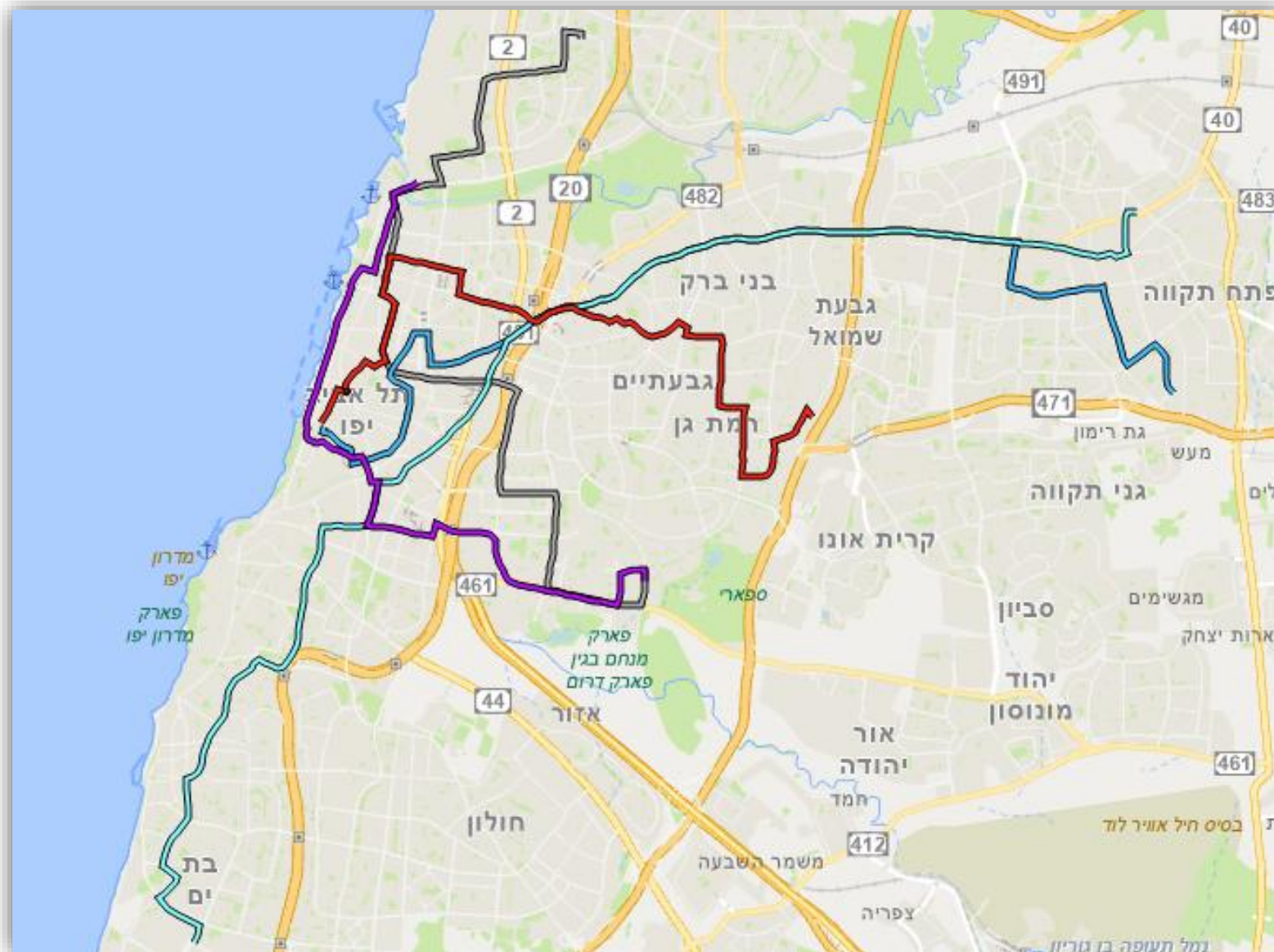
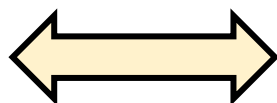
Route 1 

Route 61 

Route 8 

Route 82 

Route 4 



Examined Routes – Peak Hours [6-8 AM]

	Line no.	Frequency $\left[\frac{trips}{hr}\right]$	Route length [km]	Vehicle Type
81001 	1	6	21.8	Articulated bus
16061 	61	6	12.7	Urban bus
55008 	8	8	14.7	Urban bus
29082 	82	10	17.3	Urban bus
27004 	4	10	11.8	Urban bus

High-frequency routes with heavy demand
and significant traffic

Results

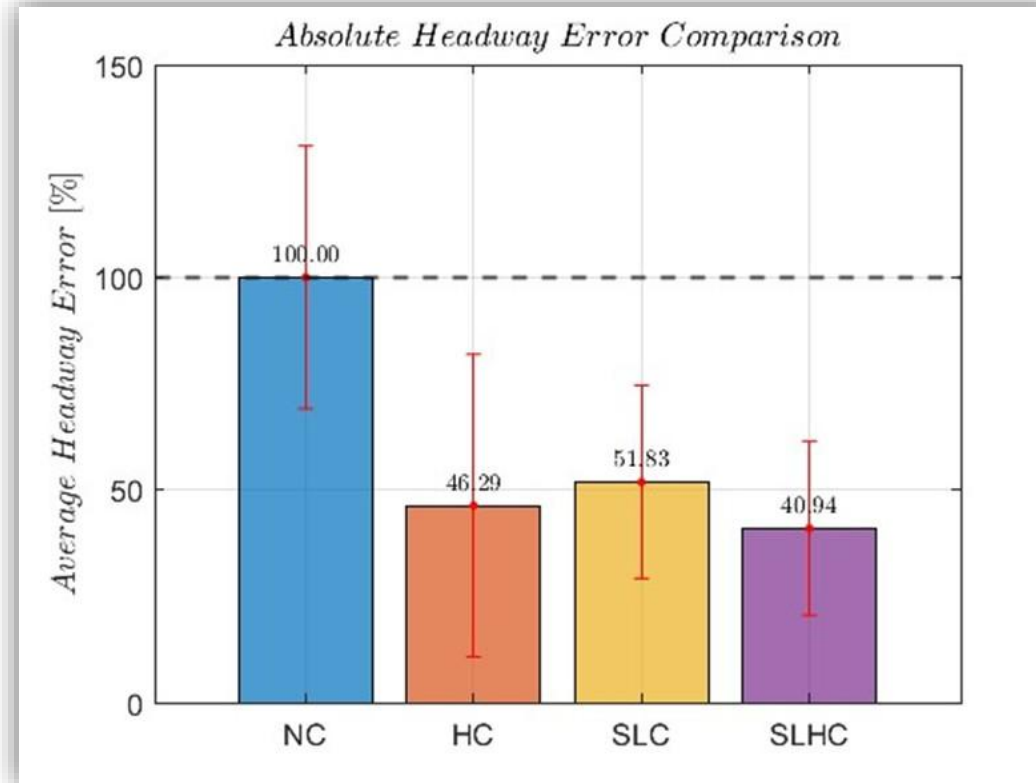
Moderate inconsistency in departures was examined

✕ **NC** – No Control

STOP **HC** – Holding Control

⌚ **SLC** – Speed Limit Control

STOP + ⌚ **SLHC** – Speed Limit & Holding Control

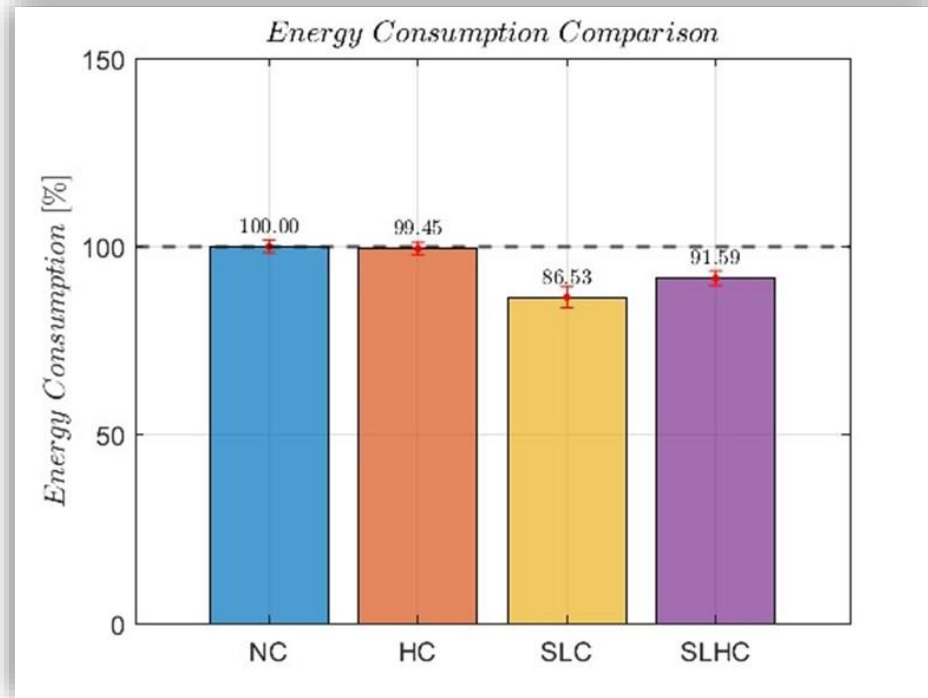


Line 1 - For Example:
Average HW error without control - **~250 seconds**

Average HW error with control - **~100 seconds**

60% reductions in deviations from reference headway

Results



More than **13% Energy savings**

$\sim 0.2 \frac{kWh}{km}$ consumption reduction $\rightarrow 0.15 \frac{NIS}{km}$ savings

X

In Israel: 300,000,000 km urban service per year*

=

$\sim 50,000,000 \left[\frac{NIS}{year} \right]$ **Potential Savings**



NC – No Control



HC – Holding Control



SLC – Speed Limit Control



SLHC – Speed Limit & Holding Control

*chrome-extension://efaidnbmnnnibpcajpcglclefindmkaj/https://www.gov.il/BlobFolder/reports/special_reports/he/%D7%A8%D7%99%D7%9B%D7%95%D7%96%D7%A0%D7%AA%D7%95%D7%A0%D7%99%D7%9D%D7%9E%D7%A8%D7%9B%D7%96%D7%99%D7%99%D7%9D%D7%91%D7%AA%D7%97%D7%91%D7%95%D7%A8%D7%94%D7%94%D7%A6%D7%99%D7%91%D7%95%D7%A8%D7%99%D7%AA-%D7%99%D7%95%D7%9C%D7%99%202021.pdf



TECHNION SUSTAINABLE MOBILITY AND ROBUST TRANSPORTATION LABORATORY

Results

✕ **NC** – No Control

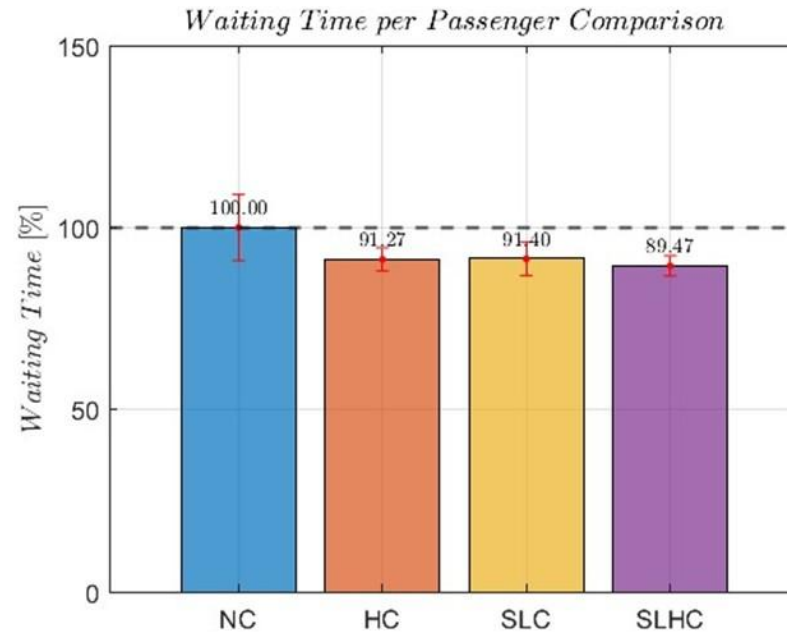
STOP **HC** – Holding Control

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STOP + ⌚ **SLHC** – Speed Limit & Holding Control

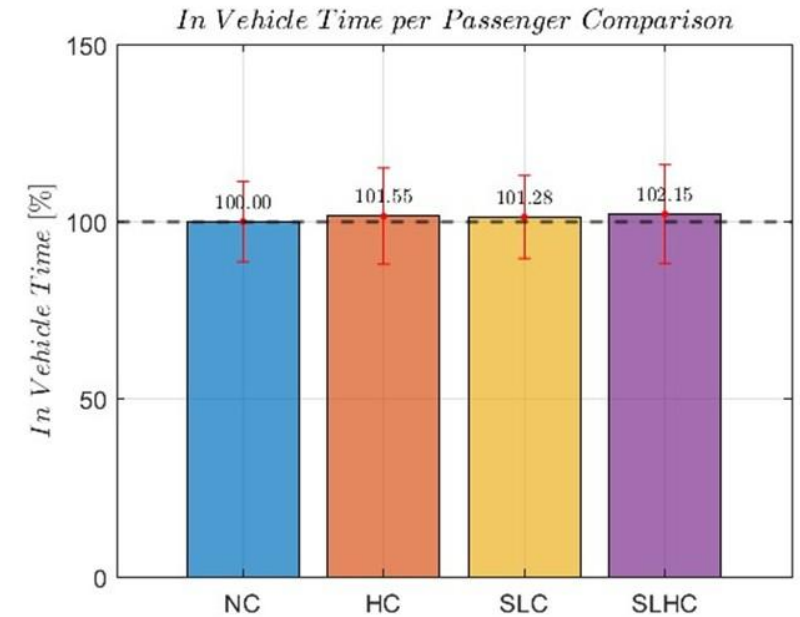
Impact on actual travel time:

Impact on passengers' perceived travel time*:



~0.5-1 min. Waiting time **reduction** per passenger

~1-2 min. Waiting time **reduction** per passenger



~0.5-1 min. In-Vehicle Time **increase** per passenger

~0.5-1 min. In-Vehicle Time **increase** per passenger

=

>

*Wardman, Mark. "Public transport values of time." *Transport policy* 11.4 (2004): 363-377.

Conclusions



Public Transport System

- Combining speed limit and holding strategies yields the most effective headway regulation
- Total travel time remains nearly unchanged, as reduced waiting offsets increased in-vehicle time



Energy Consumption

- Speed limit control alone maximizes energy efficiency



Control System

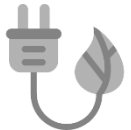
- The framework remains robust across diverse case studies and under various disturbances and uncertainties

Future Work



Public Transport System

- Exploring traffic light priority as a control strategy
- Expanding the research to different modes – Light rails & Metro
- We aim to test the methodology on a real bus line and assess its performance



Energy Consumption

- Assessment of external factors: temperature, different terrains, and vehicle types



Control System

- Evaluation and comparison of different control horizons
- Add driver's non-compliance with control inputs

Thank You!

