

# Predicting Time-to-Green for Fully-actuated Signal Control Systems with Deep Learning Models

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29 March 2022, Lausanne



# MOTIVATION



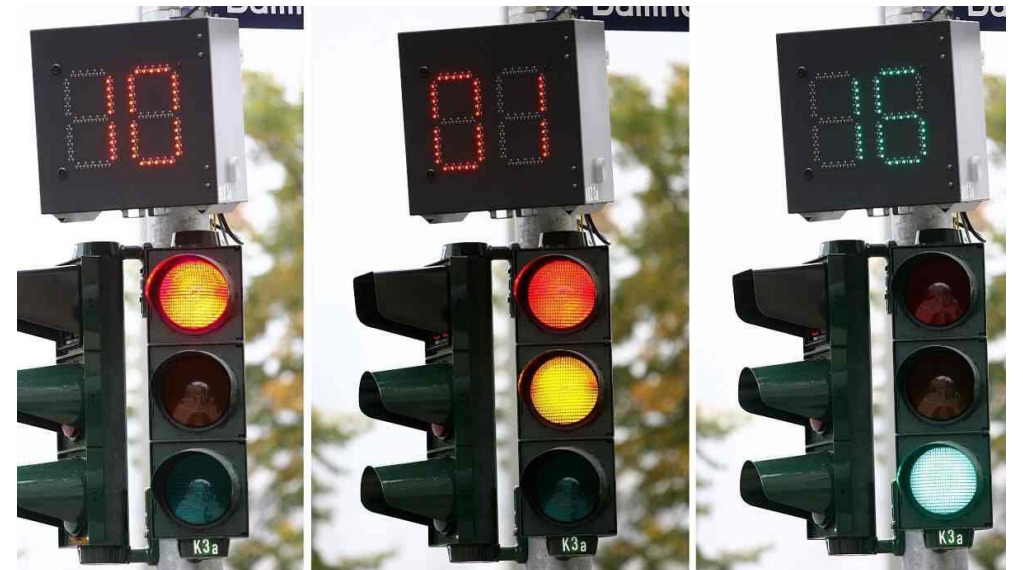
# Motivation – Time-2-Green countdowns

- Red and green phases dependent on traffic demand and public transportation

## Oft unregelmässig lange Rotphasen

In der Schweiz hatte es 2015 einen Vorstoss auf Bundesebene für «Countdown-Ampeln» landesweit gegeben. Die Forderung der Berner SP-Nationalrätin Margret Kiener Nellen wurde vom Bundesrat abgewiesen. Die Landesregierung begründete ihre Haltung unter anderem damit, dass sich Rot- und Grünphasen von Ampeln oft nach dem Verkehrsaufkommen sowie dem öffentlichen Nahverkehr richten und somit die Rotphasen unterschiedlich lang sein würden.

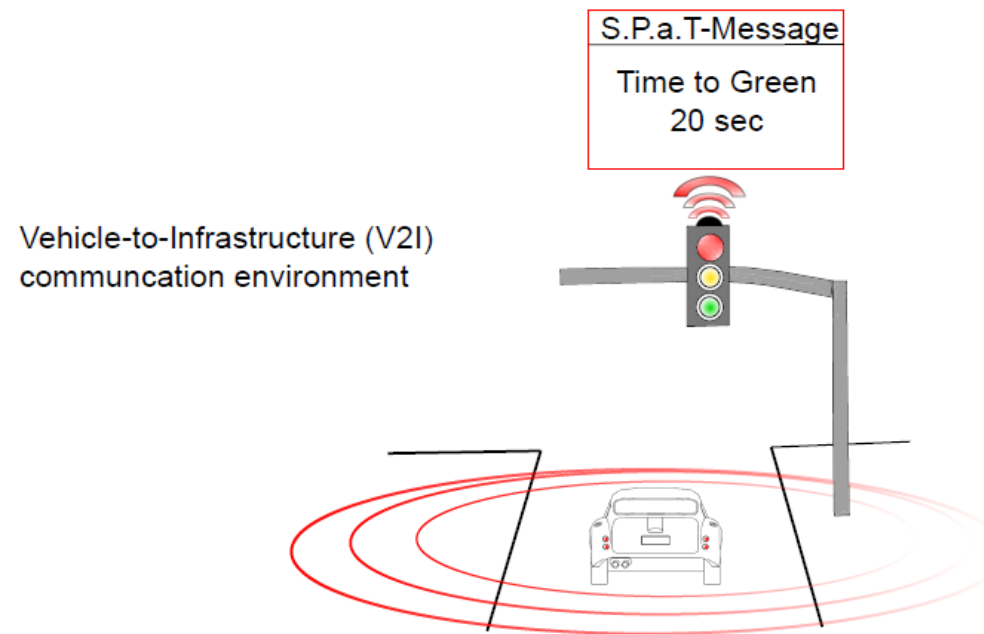
Source: nau.ch, 2020.



Source: rp-online.de, 2019.

# Motivation – Speed advisory systems

- Enhancement of Signal Phase and Timing (SPaT) messages
  - Beneficial for **speed advisory systems**
  - Efficient and environmental friendly motion planning (**homogeneous** speed profiles)
- Requirement - Robust prediction of Time to Green (T2G)



# PROBLEM DEFINITION

# State-of-the-art signal control systems

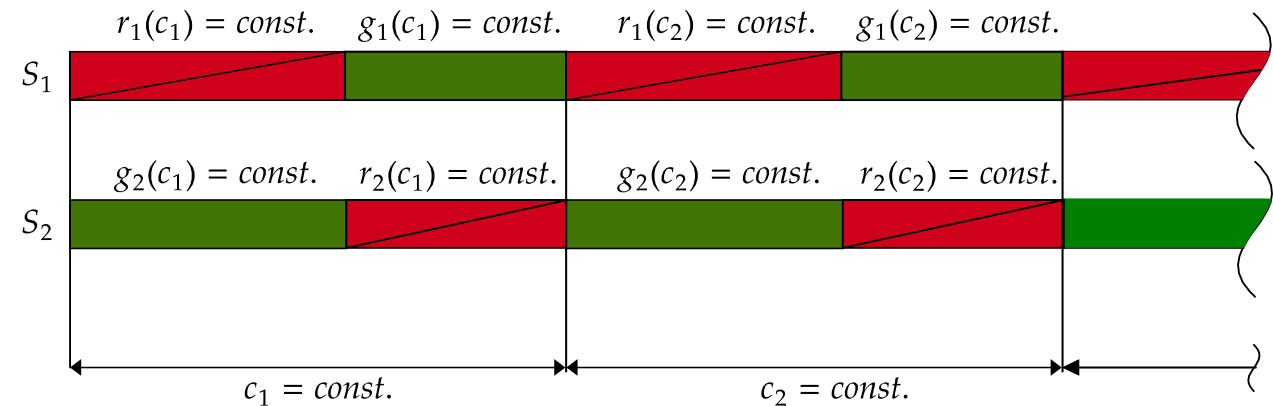
- Types of signal control systems
  - Non-actuated
  - Semi-actuated
  - Fully-actuated
- None to fully flexible systems for control according to traffic dynamics



# Non-actuated signal control systems

- Non-actuated signal control system
  - Green time  $g_i(c_i)$ , red time  $r_i(c_i)$ , cycle-time  $c_i$  of signal  $i$  are **constant**
  - **No reaction** to traffic dynamics / public transportation

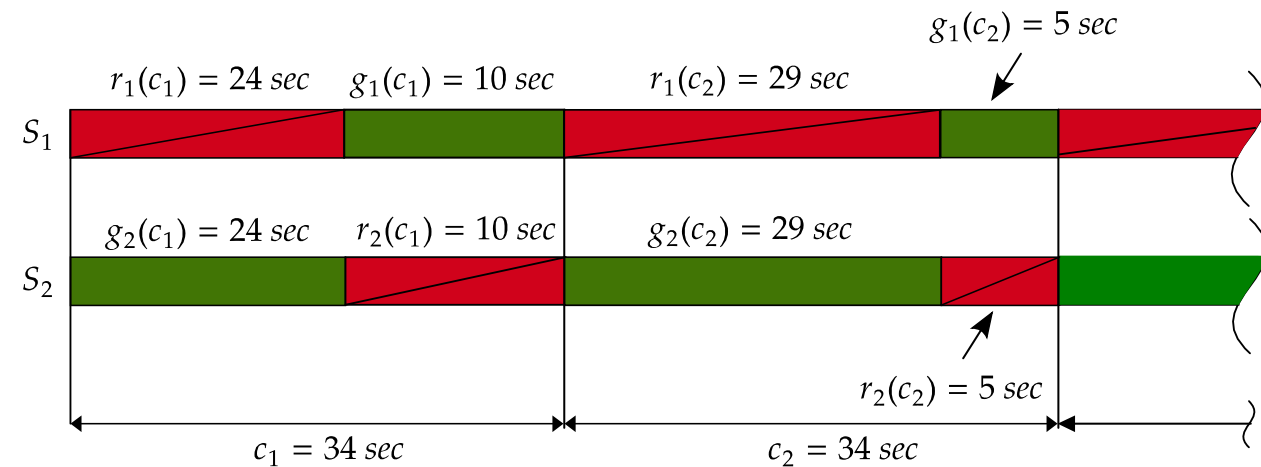
System	$r_i(c_i)$	$g_i(c_i)$	$c_i$
None	const.	const.	const.
Semi	nonconst.	nonconst.	const.
Fully	nonconst.	nonconst.	nonconst.



# Semi-actuated signal control systems

- Semi-actuated signal control system
  - Green time  $g_i(c_i)$ , red time  $r_i(c_i)$  of signal  $i$  are **non-constant**
  - Extension of green-time (e.g., priority for public transportation)
- T2G prediction with constraint that **cycle durations are fixed**

System	$r_i(c_i)$	$g_i(c_i)$	$c_i$
None	const.	const.	const.
Semi	nonconst.	nonconst.	const.
Fully	nonconst.	nonconst.	nonconst.

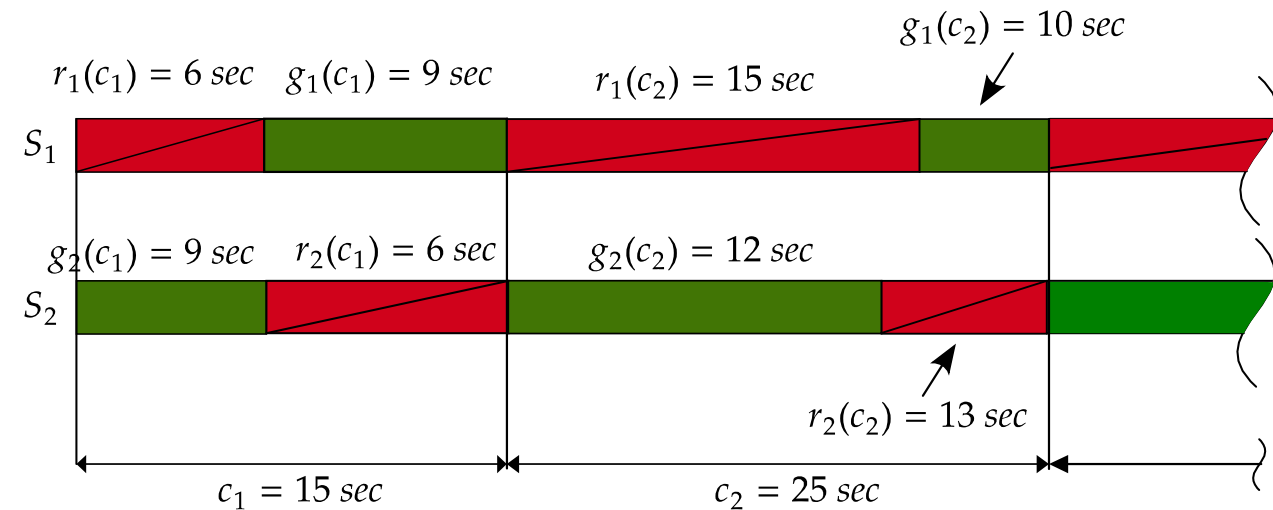




# Fully-actuated signal control systems

- Full-actuated signal control system
  - Green time  $g_i(c_i)$ , red time  $r_i(c_i)$ , cycle-time  $c_i$  of signal  $i$  are **non-constant**
  - No variables have a fixed time quantity
- T2G prediction without pre-defined constraints

System	$r_i(c_i)$	$g_i(c_i)$	$c_i$
None	const.	const.	const.
Semi	nonconst.	nonconst.	const.
Fully	nonconst.	nonconst.	nonconst.



# Problem definition

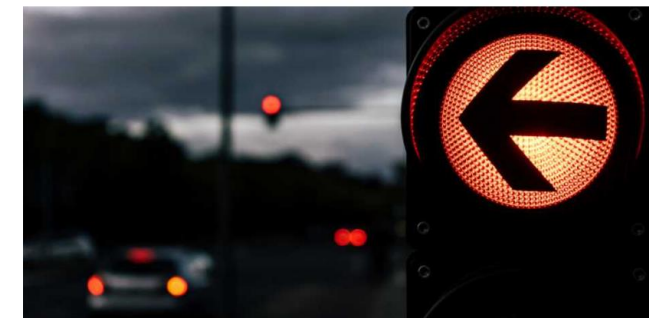
- **Continuous development** of traffic signal control systems
  - Flexible systems through sensor technology (detectors, Bluetooth, thermal cameras, etc.)
  - Optimization methods (VS-PLUS, Self-control, etc.)
- Cycle times, green or red times **not constant**
- Prediction model must capture **the target's variance**

## Pilot study in the city of Lucerne: Self-control light-signal systems

16.12.2020

By: SVT, City of Lucerne

Rising requirements for mobility and the environment demand new ideas, especially in regard to efficiency and sustainability.



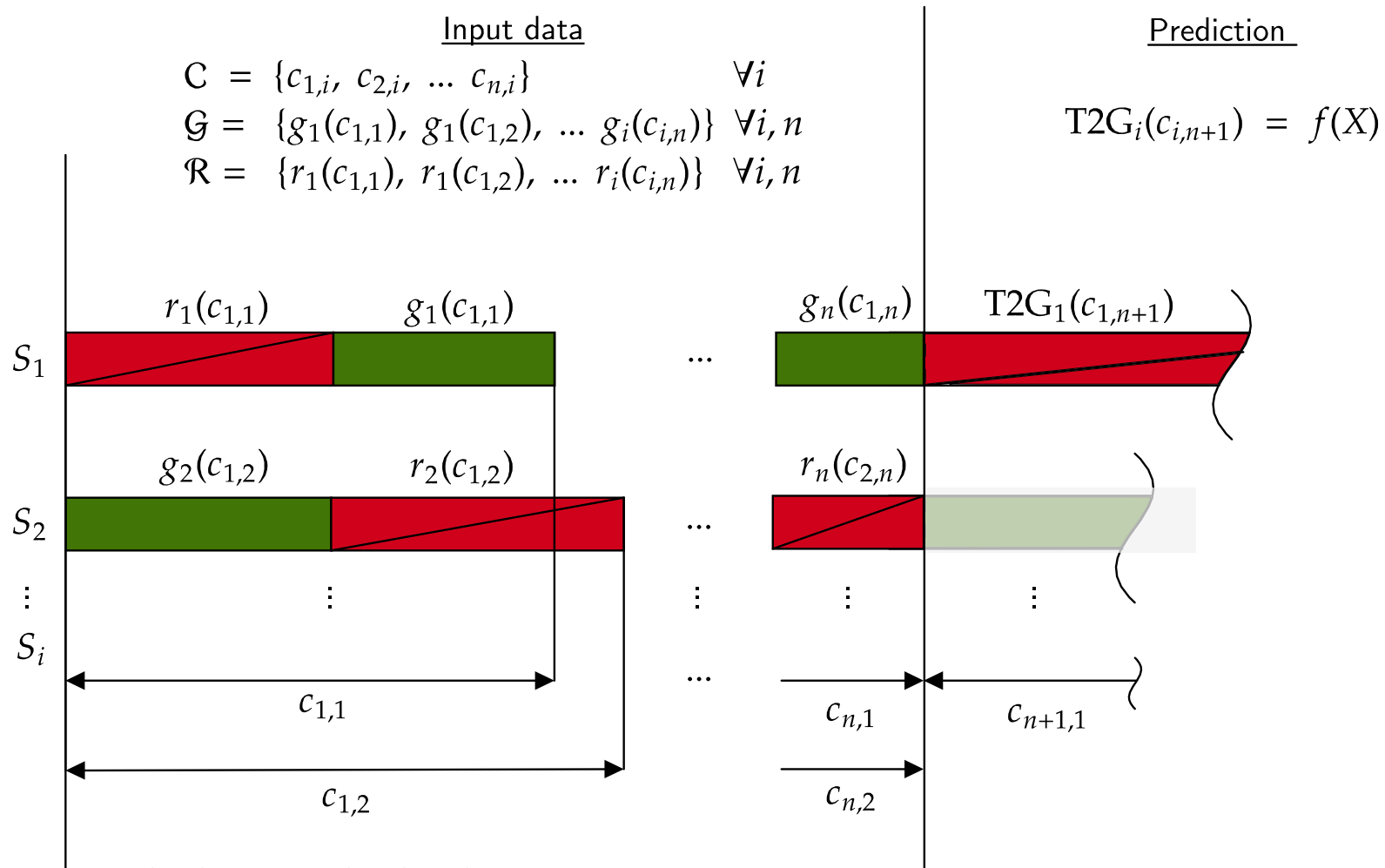
Traffic light (CC0 1.0 / S. Sakharovskiy via Unsplash)

The city of Lucerne has analyzed the improvement of current traffic flow using existing infrastructure. In a pilot study a brand new approach for light-signal systems has been tested. [Read on](#)

IVT, ETH, 2020.

# Problem definition

- Prediction of T2G with  $f(X)$ , where  $X$  contains all concatenated inputs



# Problem definition

- Previous research is based on
  - Vehicle trajectories (demanding data requirements)
  - Considering **only traffic signal data**
  - **Fixed** cycle times (semi-actuated)
  - **No consideration of public transportation** dynamics

Prediction of Time-2-Green (T2G) for fully-actuated signal control system by utilizing detector and traffic signal data.



# STUDY AREA AND DATA SET

# Study area

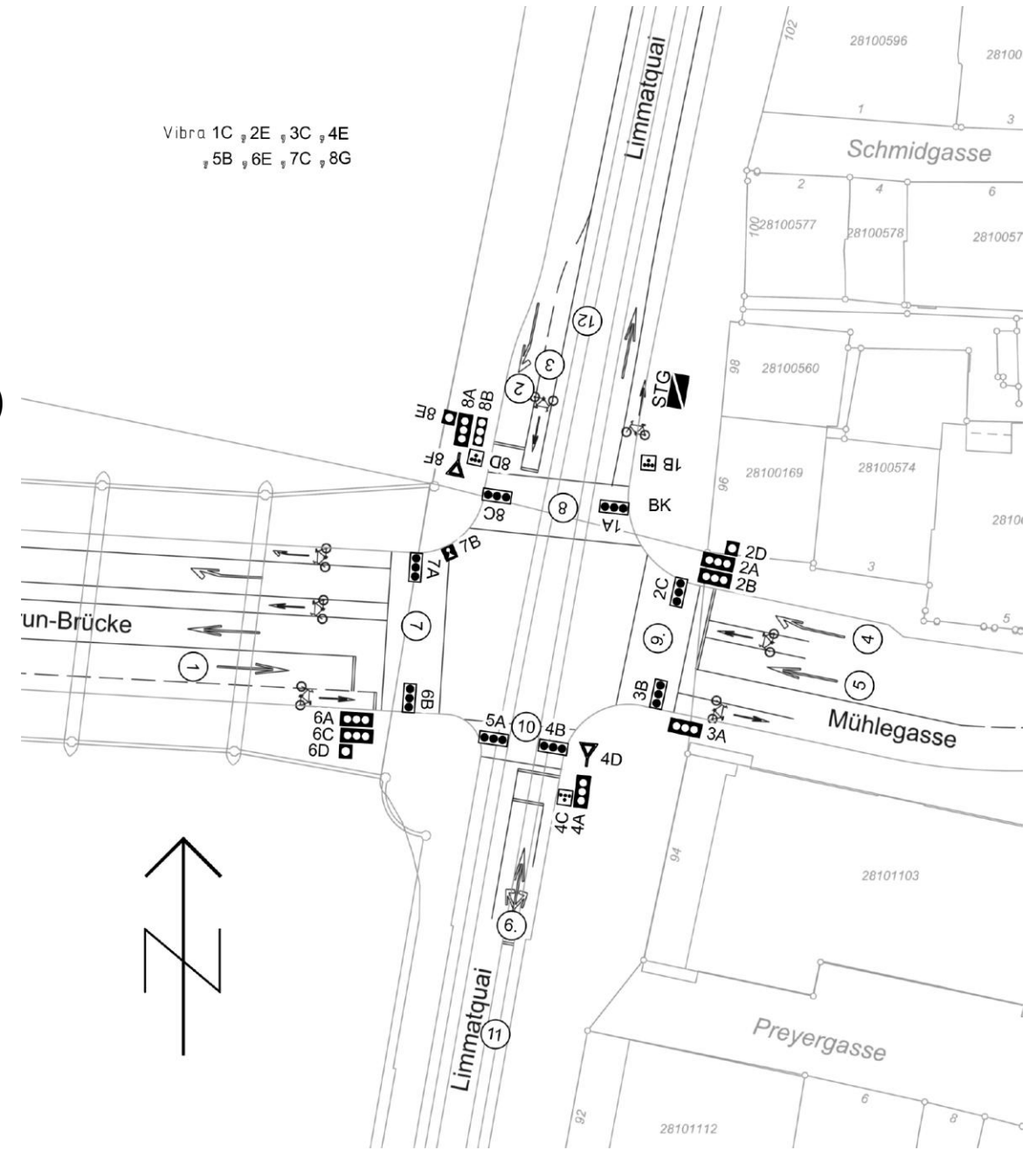
- Intersection in the center of Zurich, CH
- Traffic modes: Individual, public transportation, pedestrians and cyclists
- Fully-actuated signal control system



Source: OpenStreetMap (2022)

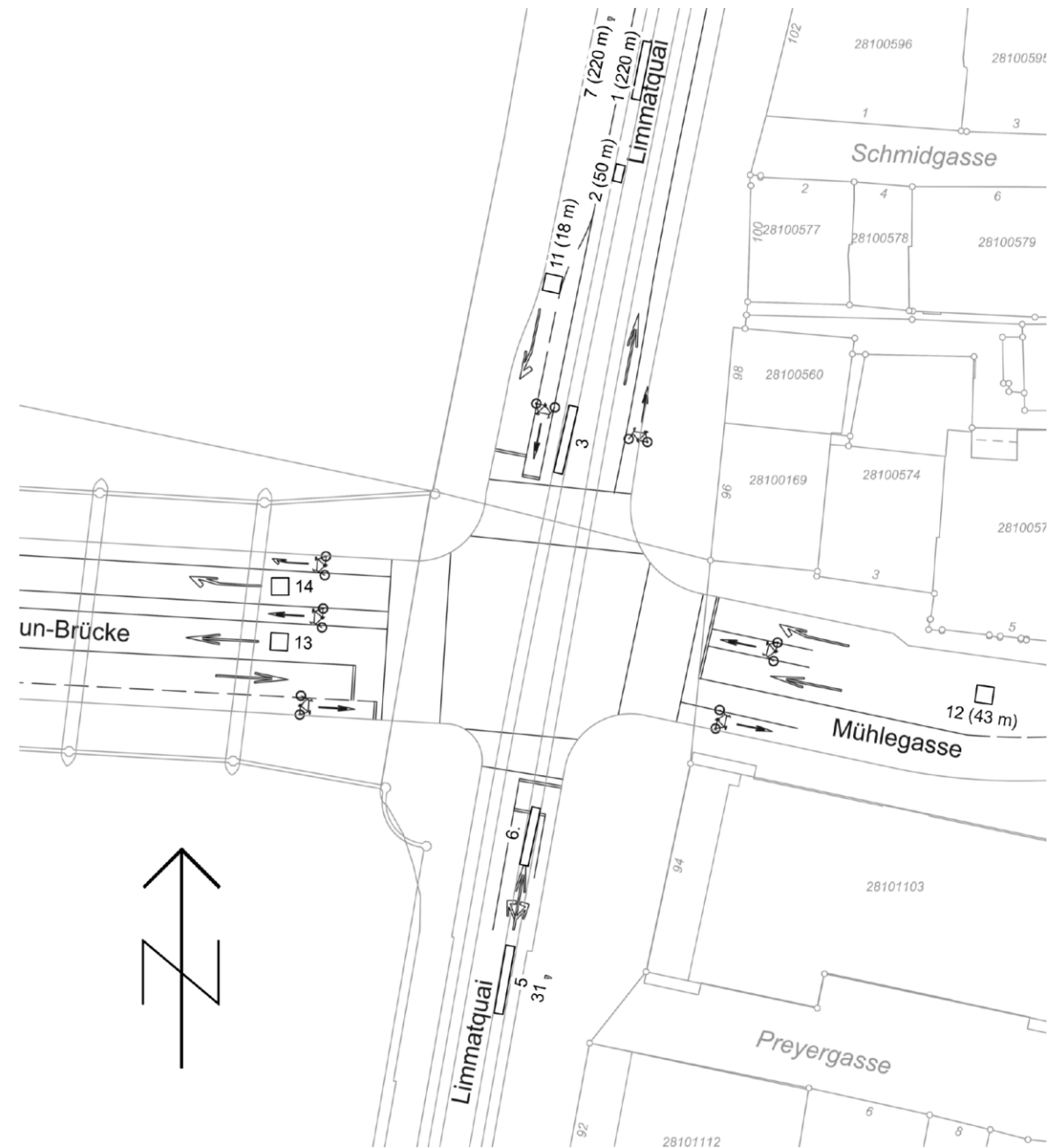
# Study area

- All traffic flows controlled by traffic control system
- 12 traffic lights employed (3 signals are for tram line)
- Pedestrian flows are co-regulated with individual transportation



# Study area

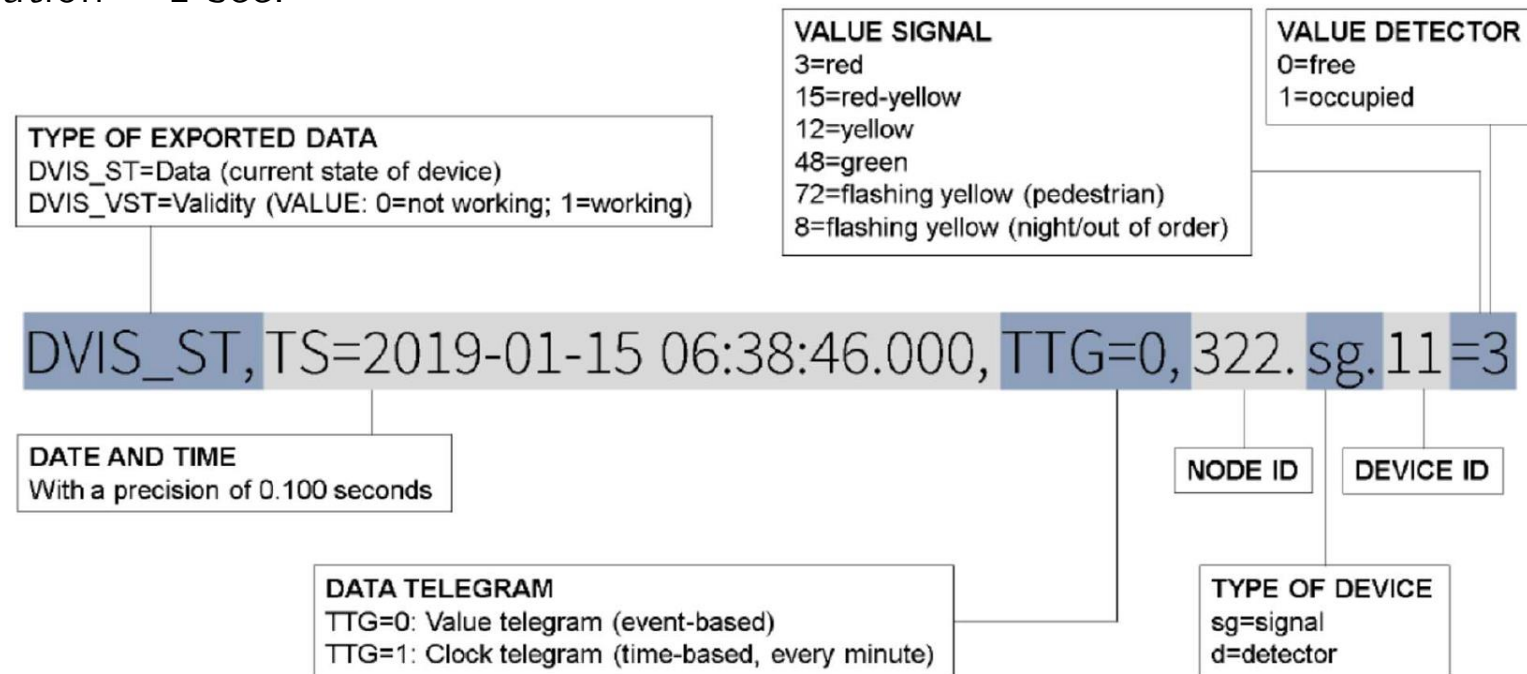
- 12 inductive loop detectors installed
- Every arriving vehicle is detected
- Dedicated loop detectors for trams
- No separate detection of bicycle traffic





# Data format and processing

- Data processing of event-based telegrams including detectors and traffic signals
- Transformation into a data set as input for machine learning
  - Time series for each device
  - Data resolution = 1 sec.



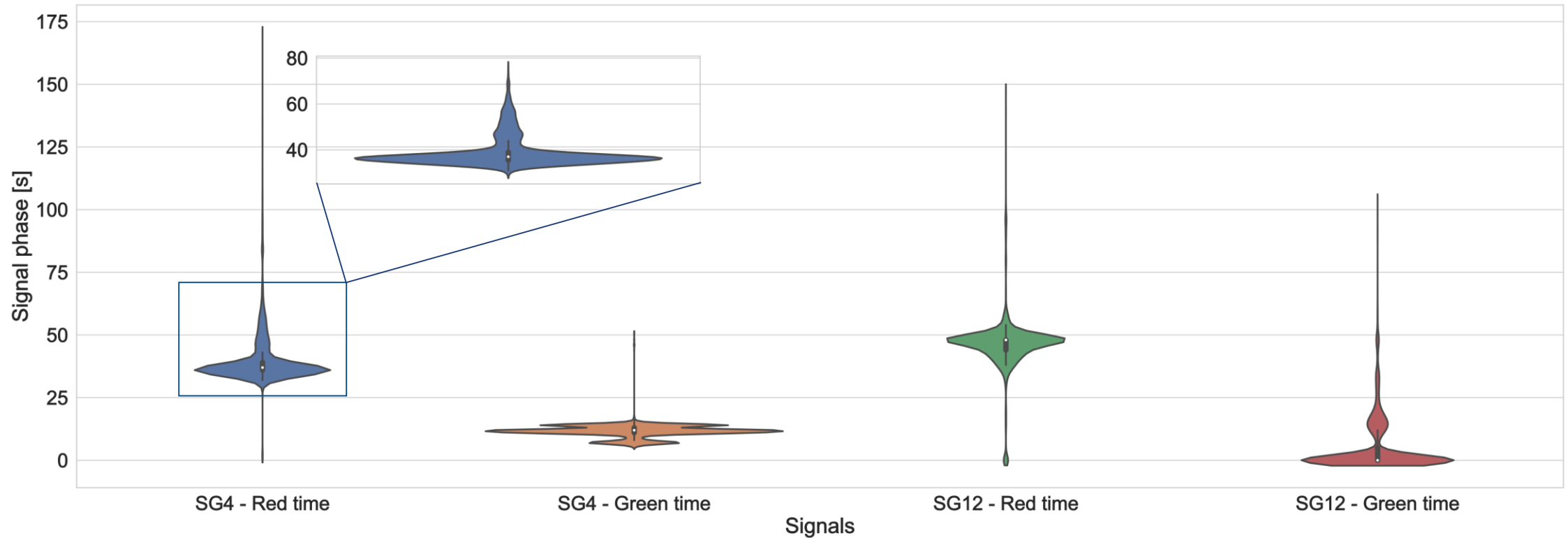
# Feature Engineering

- Aggregated data set (cycles) of traffic signals and detectors
- Two weeks of data (Monday – Friday, 7:00 and 20:00)
- Computation of feature set (R=red time, G=green time):

Feature	Variable
Red and green time [s]	$r_i(c_{n,i}), g_i(c_{n,i})$
Traffic flow at red and green time [veh/phase]	$q_{i,R}(\cdot), q_{i,G}(\cdot)$
Detector occupancy for red and green time [detections/phase]	$o_{i,R}(\cdot), o_{i,G}(\cdot)$
Congestion indicator [-]	$u_i(\cdot)$
Queue indicator [-]	$v_i(\cdot)$

# Feature Engineering

- Long tailed features due to extreme events (high traffic demand, public transportation priority, etc.)
- Other features show similar characteristics

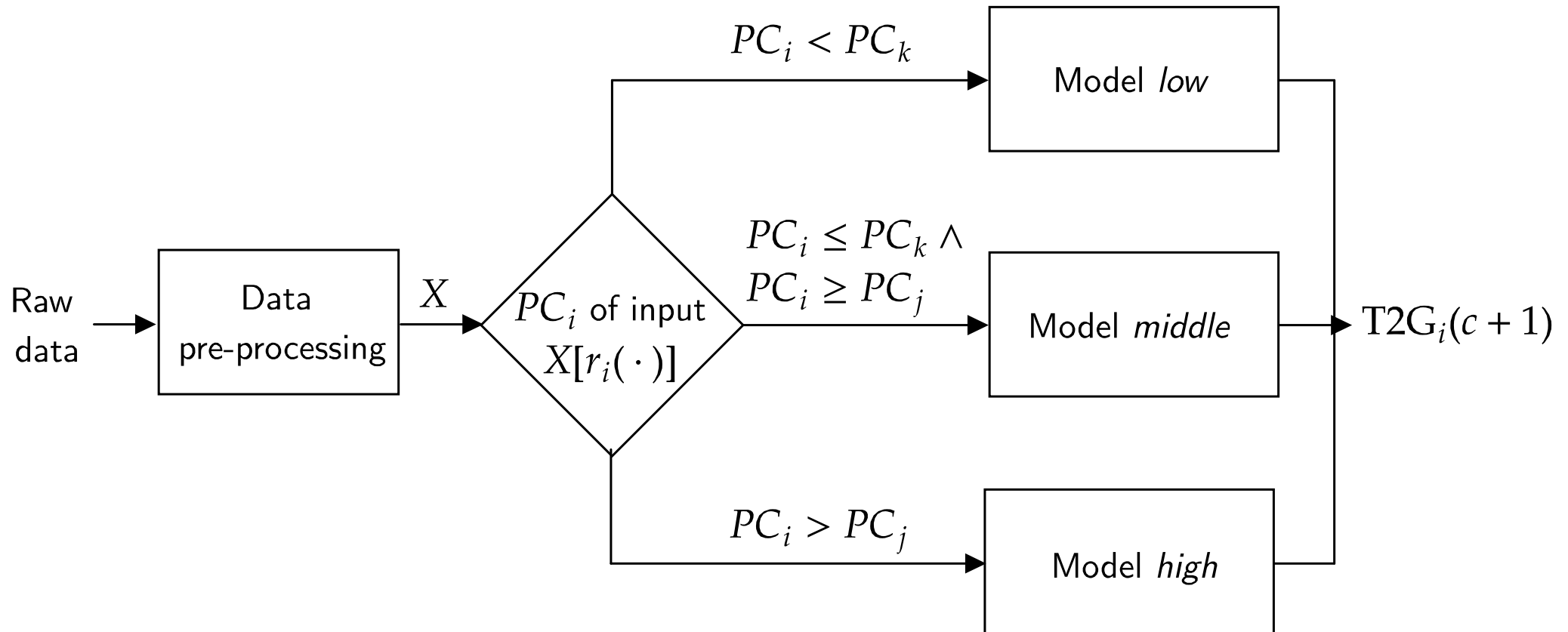


# METHODOLOGY T2G-PREDICTION



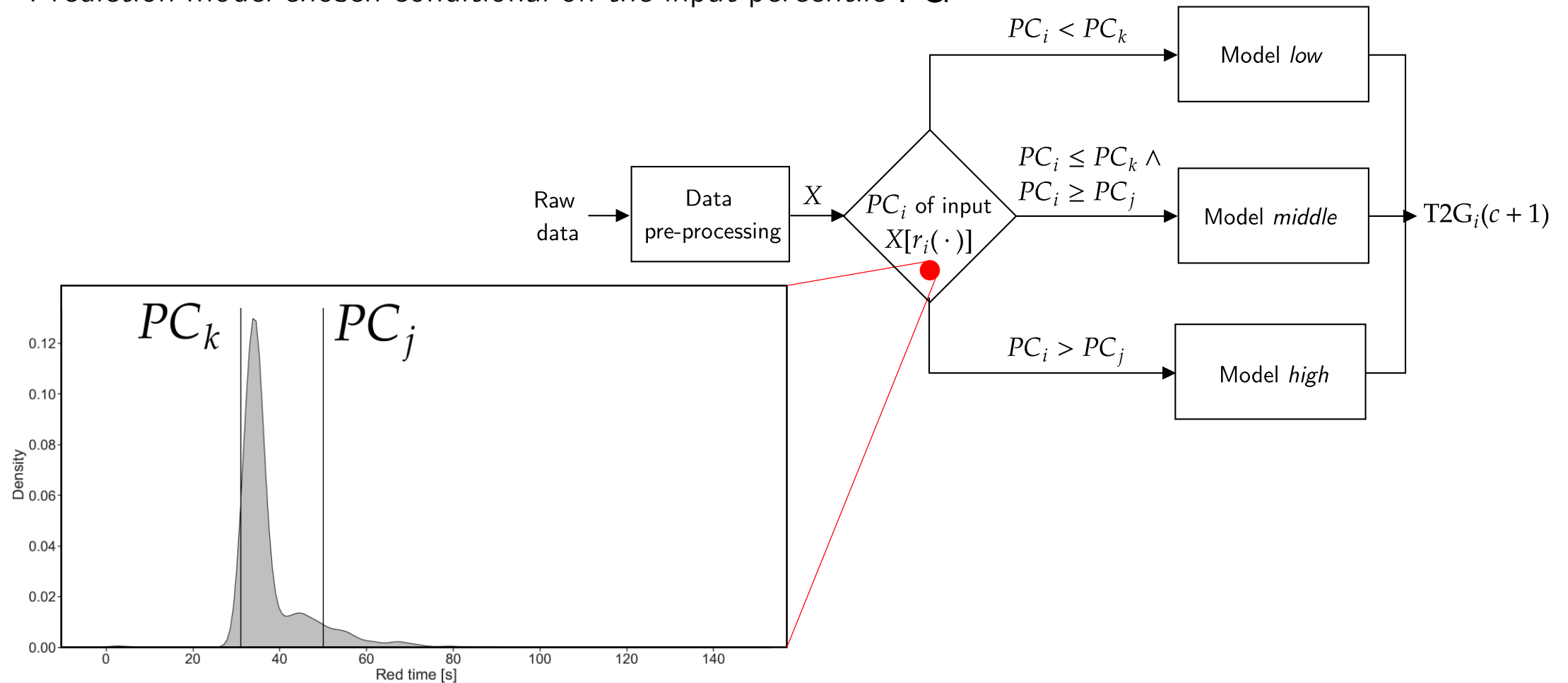
# Framework methodology

- Prediction model chosen conditional on the input percentile  $PC_i$



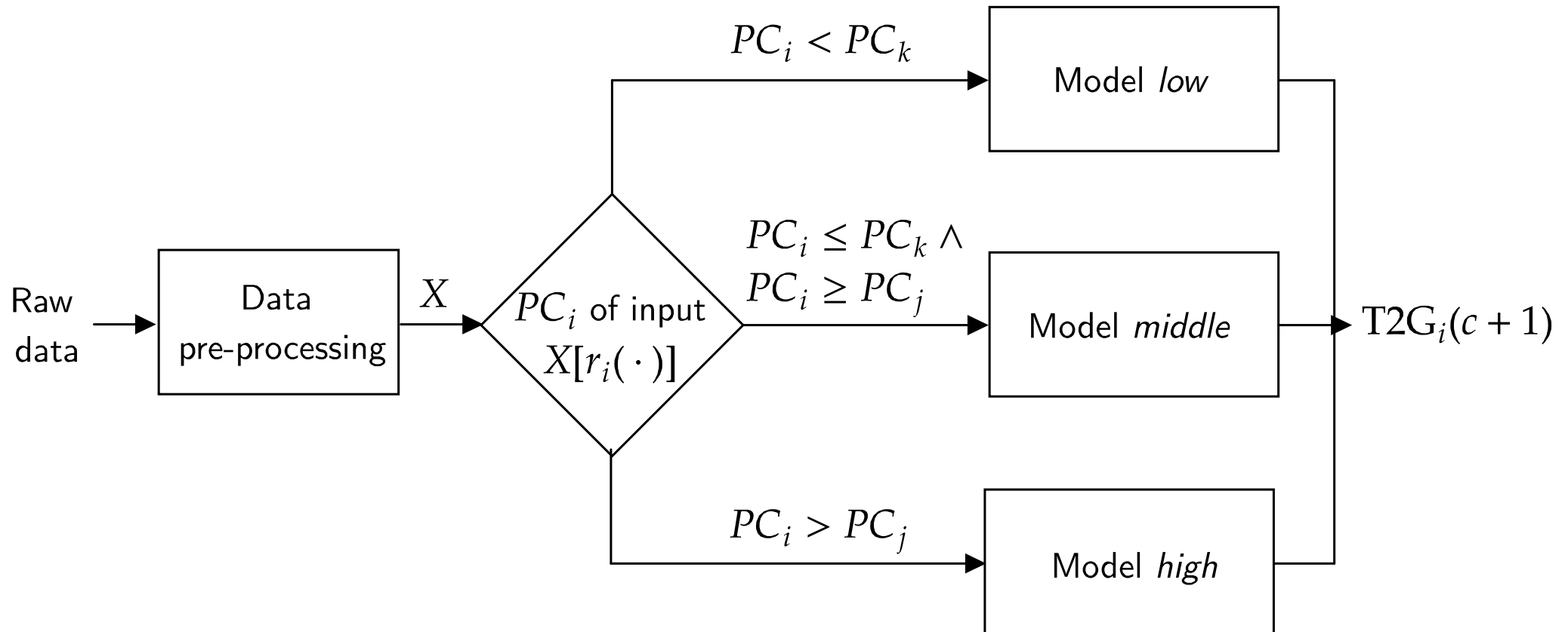
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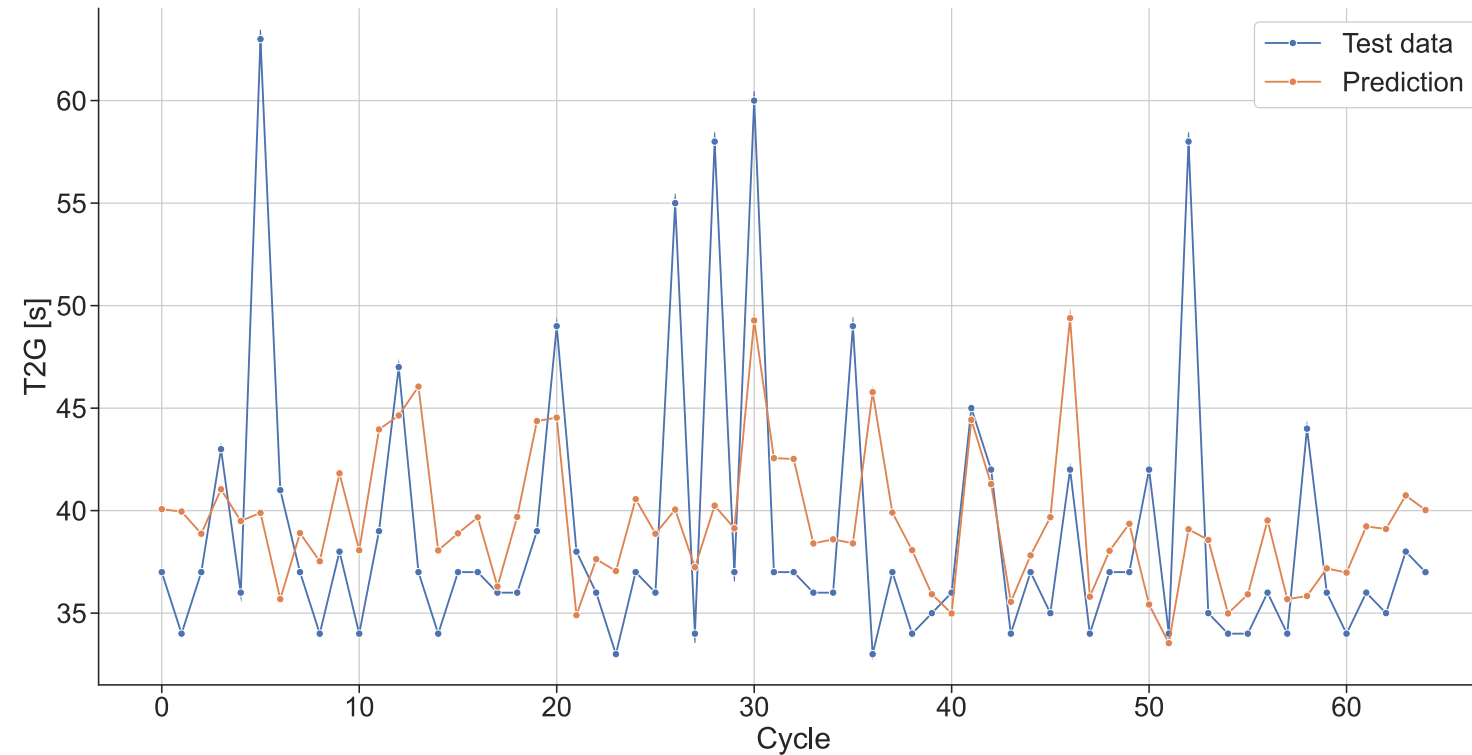
# Model selection and performance metrics

- Model selection
  - Multiple Linear regression (MLR)
  - Random forest (RF)
  - Random forest **with distribution split** (RFDS)
- Performance metrics
  - Mean Absolute Error
  - **Exact hit (EH)**      – T2G is predicted with an error of 0 sec.
  - **Near misses (NM)**      – T2G is predicted with an error of < 1 sec.

# RESULTS

# Results

## Multiple Linear Regression (MLR)



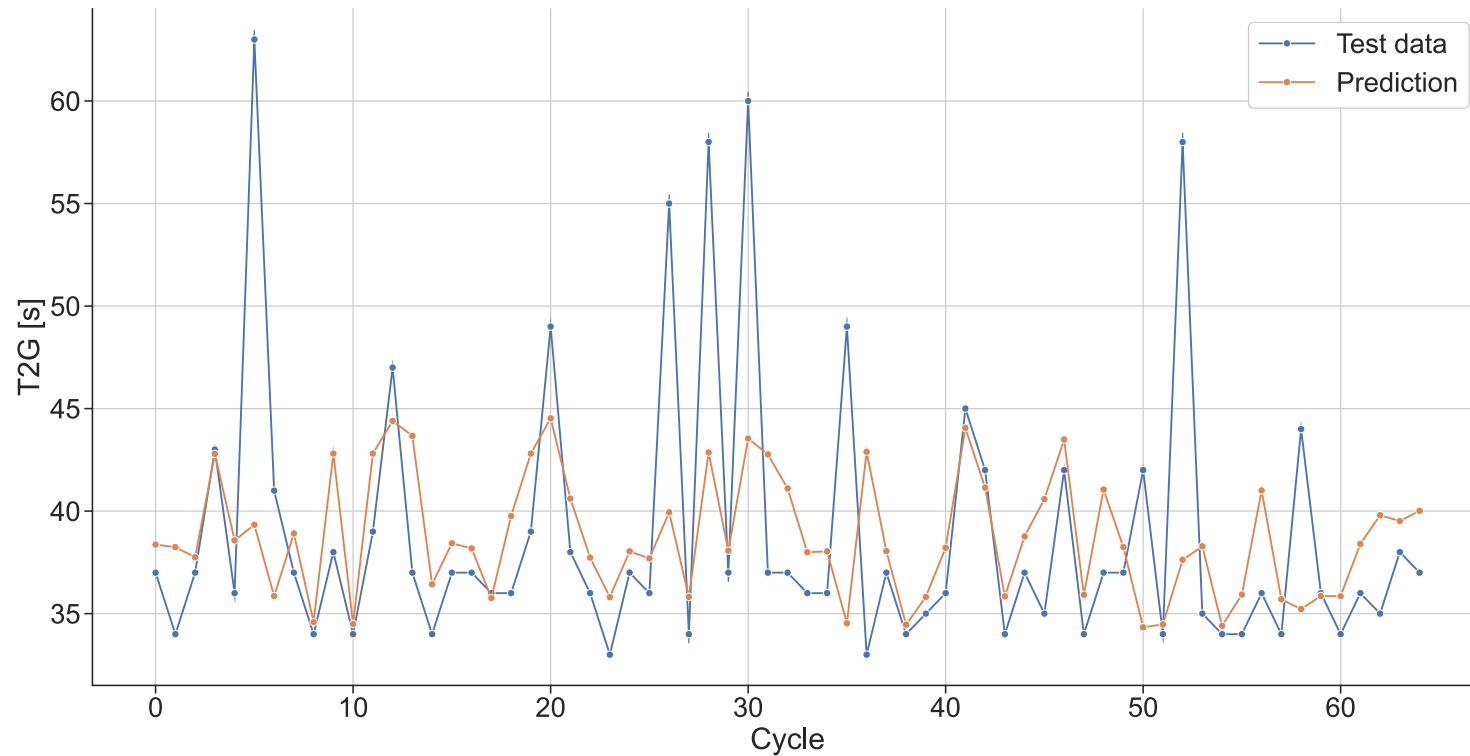
$i$	MAE	MLR EH[%]	NM[%]
1	4.75	5.07	16.94
2	4.23	13.29	42.60
3	4.63	11.70	35.42
<b>4</b>	<b>4.81</b>	<b>4.57</b>	<b>17.87</b>
5	4.90	4.15	16.50
...	...	...	...
12	63.12	0.88	1.32

Signals  $i = \{12\}$  are only utilized for public transport.



# Results

## Random Forest (RF)

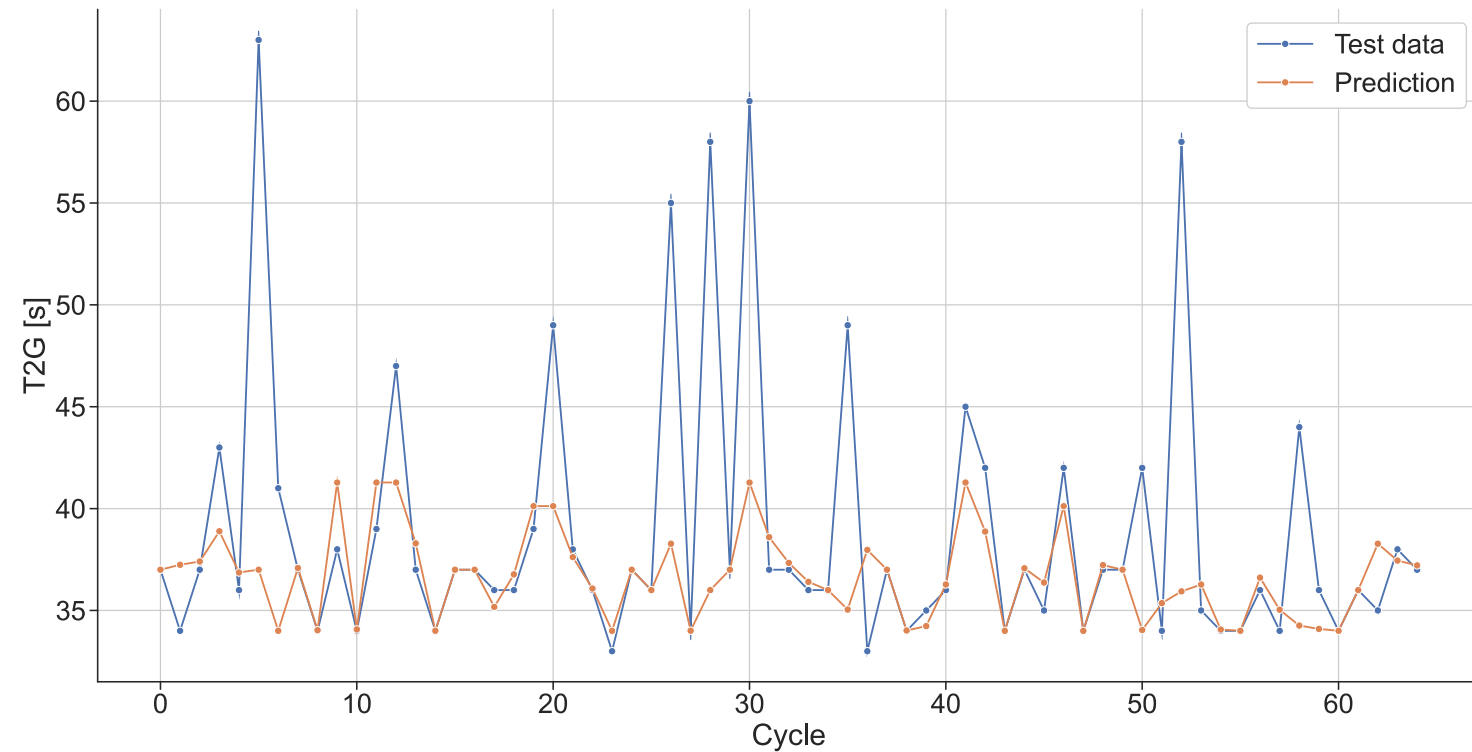


$i$	MAE	RF EH[%]	NM[%]
1	3.66	49.29	63.59
2	3.78	44.32	58.42
3	3.39	52.57	66.63
<b>4</b>	<b>4.54</b>	<b>6.60</b>	<b>26.80</b>
5	3.76	51.06	66.97
...	...	...	...
12	57.892	0.88	1.76

Signals  $i = \{12\}$  are only utilized for public transport.

# Results

## Random Forest with distribution split (RFDS)

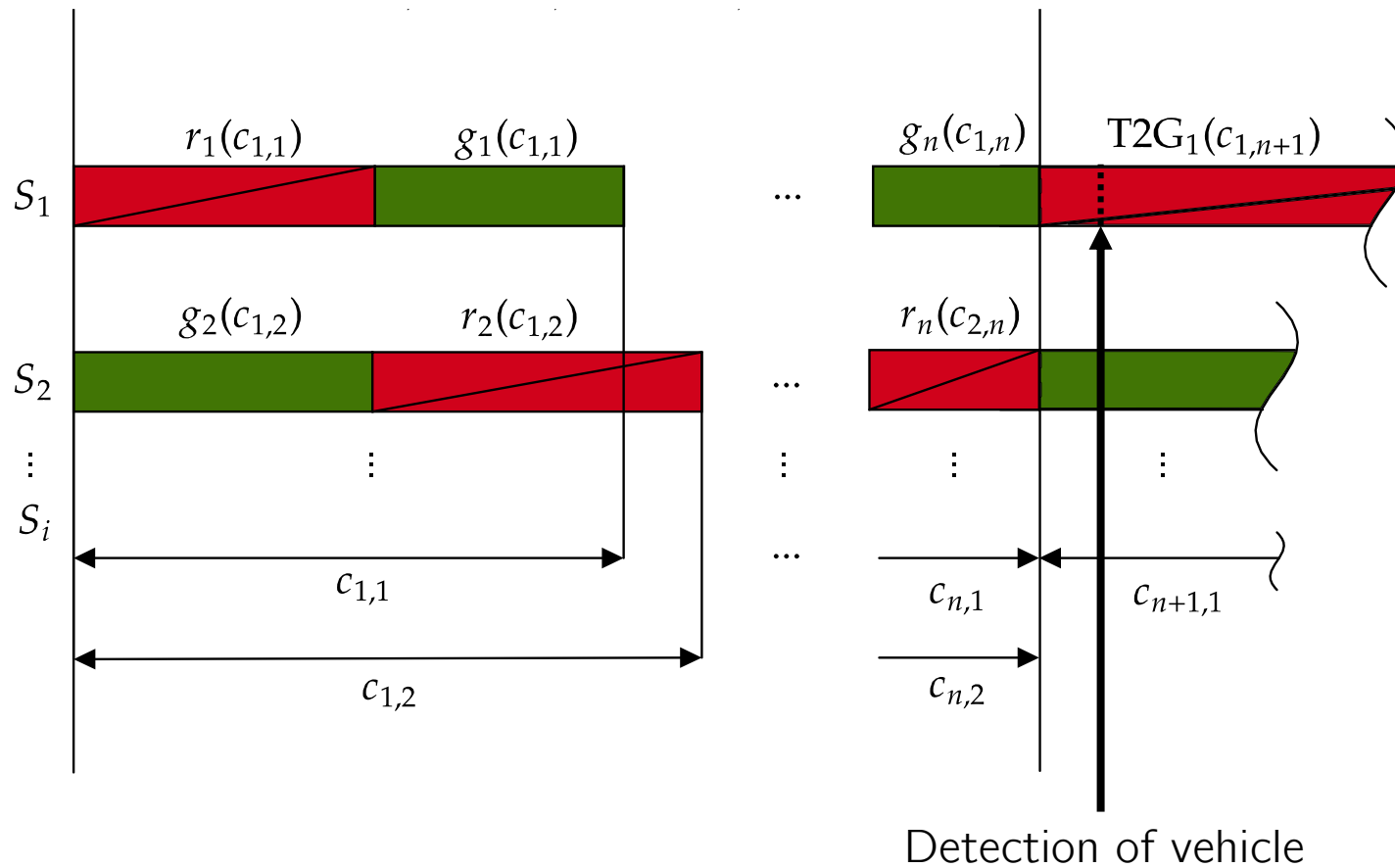


$i$	MAE	RFDS	
		EH[%]	NM[%]
1	3.44	50.00	66.02
2	3.50	44.02	61.56
3	3.44	50.92	64.99
<b>4</b>	<b>3.79</b>	<b>47.01</b>	<b>62.82</b>
5	4.02	47.01	62.82
...	...	...	...
12	58.32	0.88	4.41

Signals  $i = \{12\}$  are only utilized for public transport.

# Limitation of framework

- Model can not detect red time extensions that occur after the prediction
- Inferring this dynamic behavior can improve model robustness



# CONCLUSION AND FUTURE WORK

# Conclusion and future work

- Conclusion
  - T2G prediction framework for e.g., **SPaT message enhancement**
  - **Capturing of non-linear relationship** between traffic signal and loop detector data
- Future work
  - **Feature to model detections** occurring **after prediction**
  - Comparison to other machine learning candidates, e.g., XGBoost, LSTM
  - Test framework on **multiple** intersections (various characteristics)

Thank you very much for your attention!

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