

# Deep Learning for Traffic and Air Pollution Prediction

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[www.scads.de](http://www.scads.de)



## Goals

- Predict air pollution
- Investigate the influence of car traffic

## Problems

- Are the sensor data reliable?
- Can we build a successful model of the complex system?
- **How can we get traffic data in a similar resolution, time and location?**

# How to get traffic data?

## Public webcam



Alles im Fluss? Webcam hat Bau und Verkehr im ...  
dewest.de



Placenza: Verkehr A01 - KM 61,0 - Placenz...  
webcamgalore.de



Webcam zeigt den Verkehr am Grenzübergang  
heimatzeitung.de



Verkehrs-Webcams | Brennerautobahn A22 ...  
webcam-suedtirol.com



Röszke - Horgos Grenzübergang Verkehr Liv...  
elivewebcams.com



Webcams A23 und aktueller Verkehr auf der S...  
vienna.at



Webcam-Quelle: Webcam Gotthard-Tunnel W...  
webcam-4insiders.com



Porotto: Verkehr A13 - KM 40,0 - Ferrara Nord ...  
webcamgalore.de



Verkehr Aktuell - hamburg.de



Webcams in Köln | köln.de



A9 München - Nürnberg



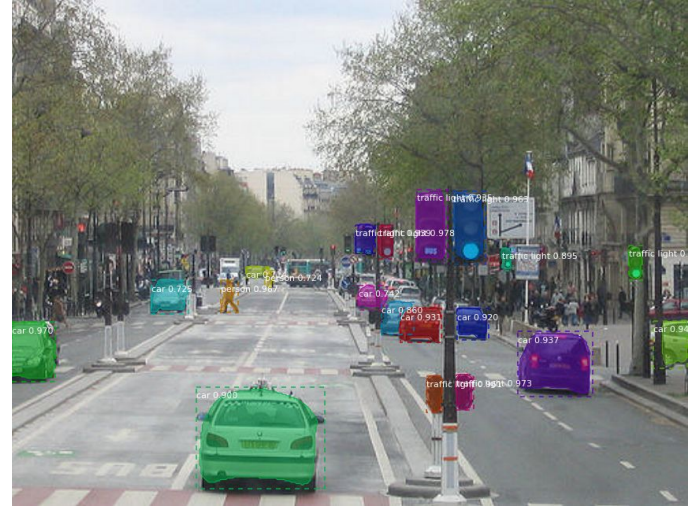
Stadt Heidelberg - Stadtblatt Online



Webcams in Karlsruhe - aktuelle Bilde...

+

## Deep Learning for object recognition<sup>1</sup>



= ?

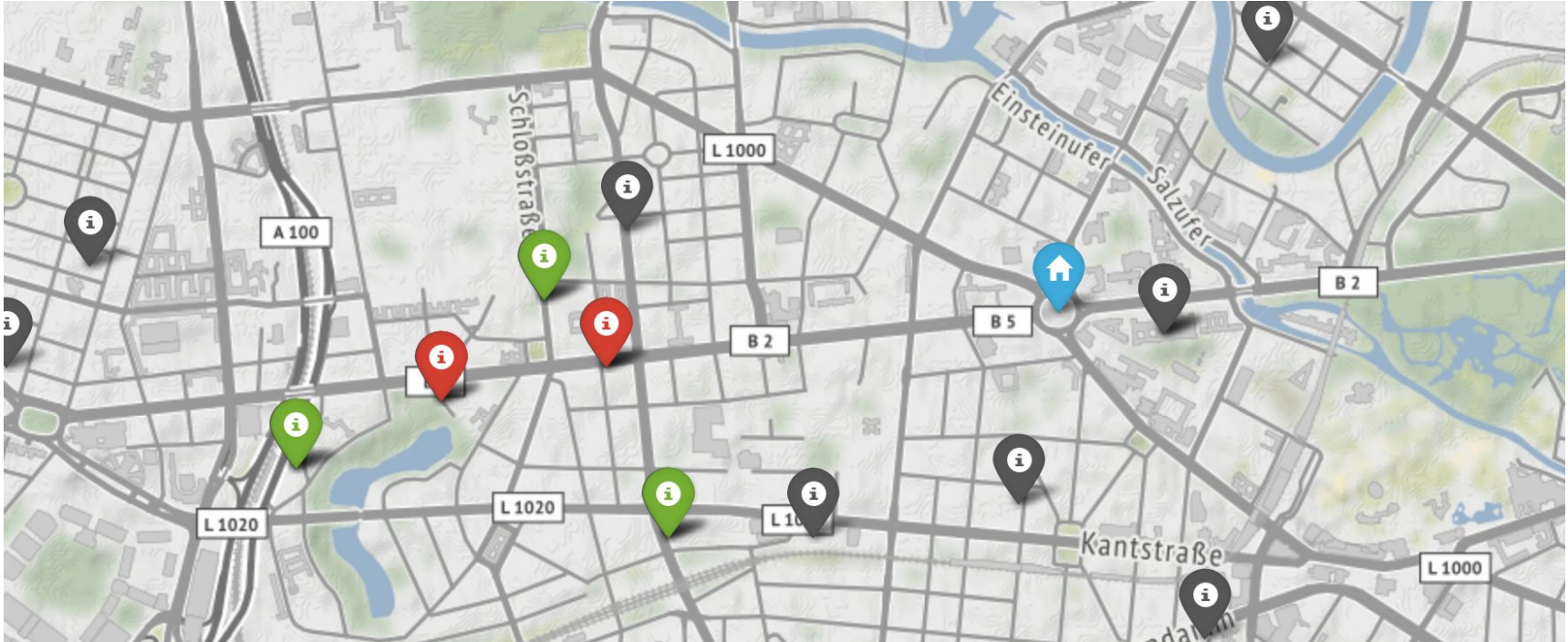
<sup>1</sup>[https://github.com/matterport/Mask\\_RCNN](https://github.com/matterport/Mask_RCNN)

# Steps

- Finding the Right Spot
- Preprocessing Sensor Data
- Building a Prediction Model
- Traffic Data
- Transfer Learning
- Results and Outlook

# Finding the right spot

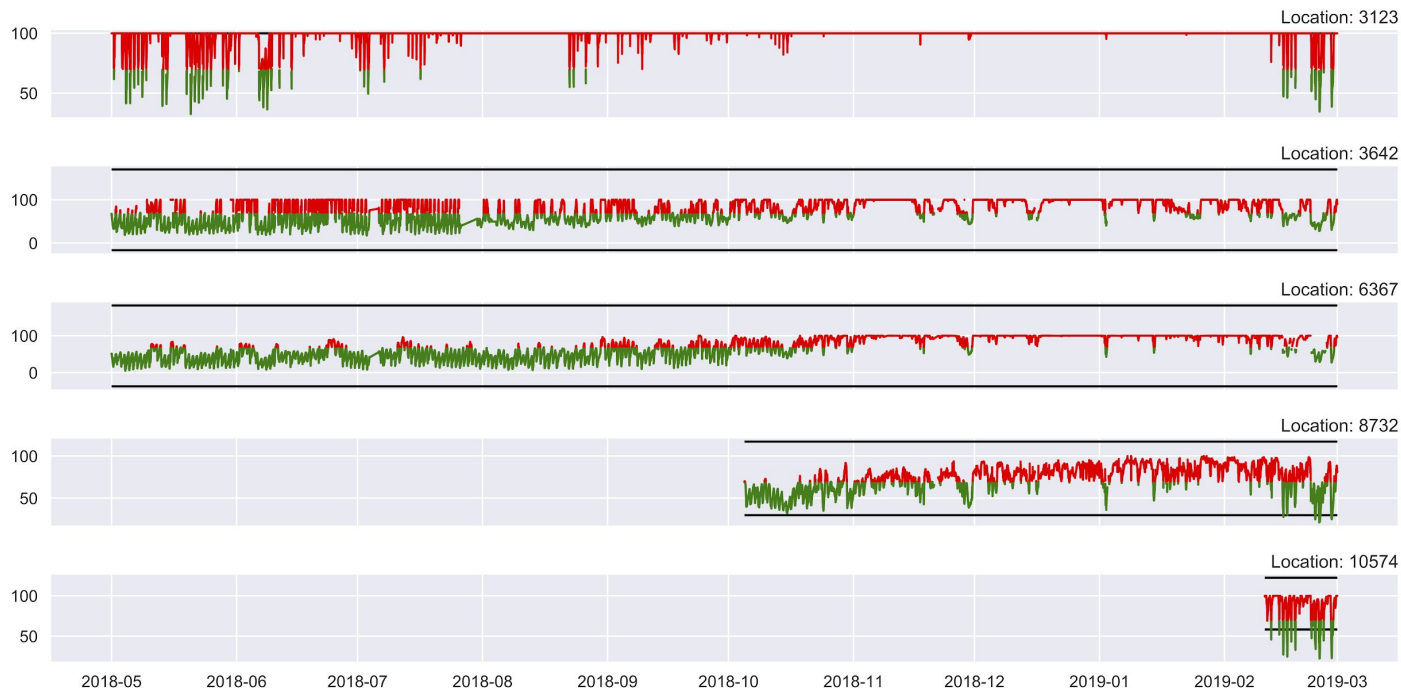
## Finding the right spot



blue: camera, red: too many invalid data points, green: chosen

# Data Preprocessing

# Too much humidity



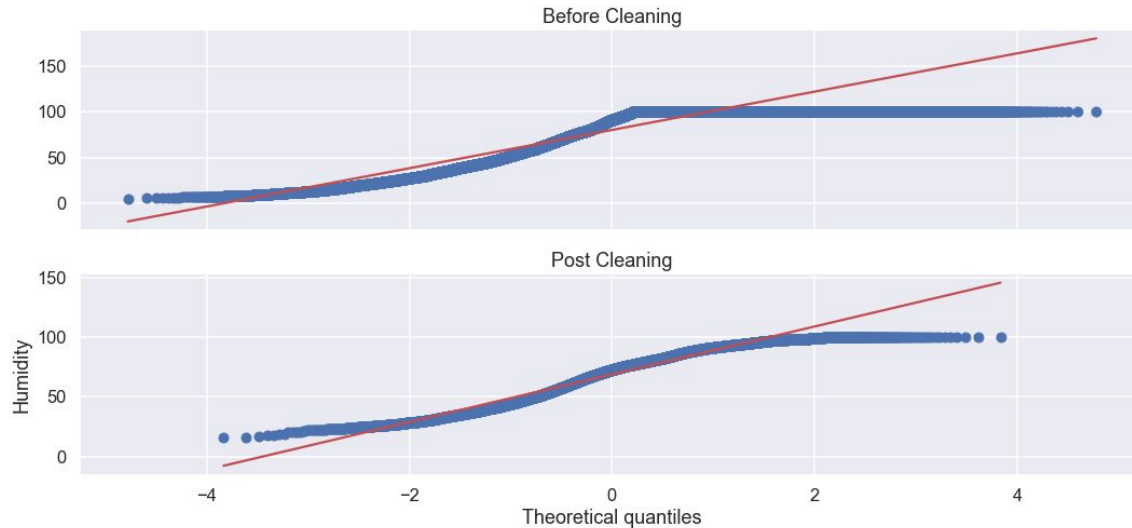
Red: Not usable data points for PM measurements

Discarded locations 3123, 10574



# Data Preprocessing Steps

- remove measurements where humidity > 70%
- outliers removal
- average over all (3) locations



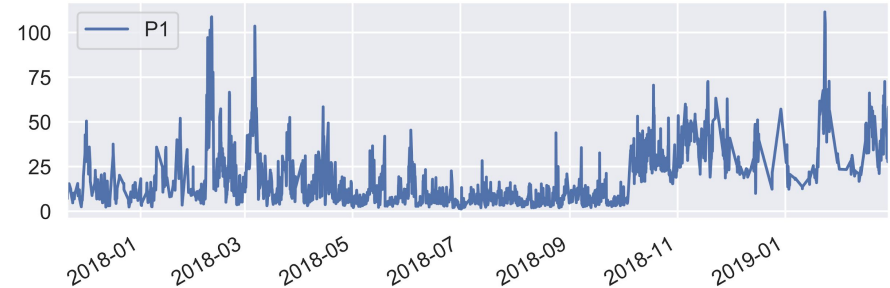
# Time series prediction model

# Model: learning, baseline, results

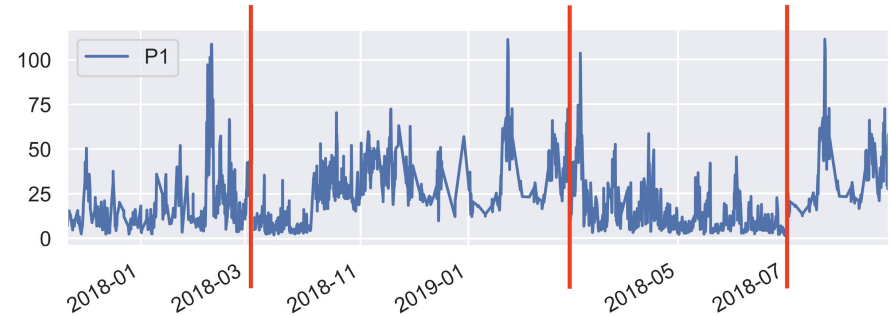
- because of possible time dependency  
→ LSTM net usage

## Models:

1. Baseline/Persistence Model  
( $PM10_t = PM10_{t-1}$ )
2. TS Model (Stateful LSTM)
3. Normal Model



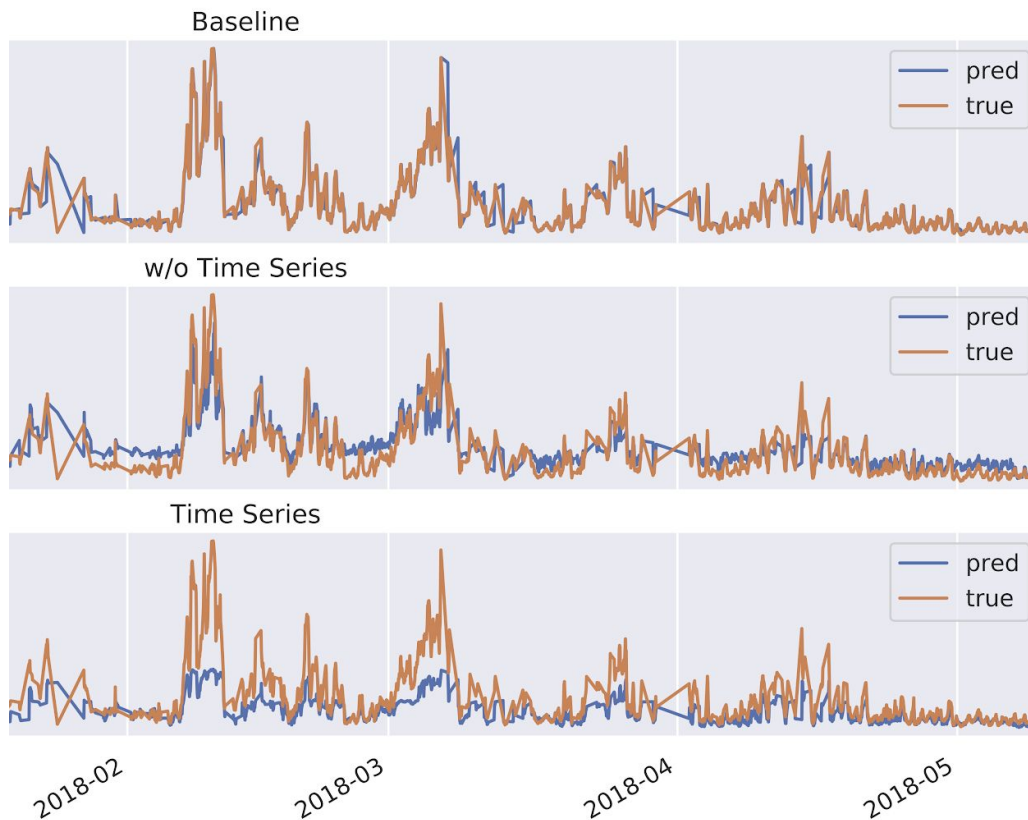
Input as time series for Model 2



Shuffled and broken time series for Model 3

# Results: Models vs. Baseline

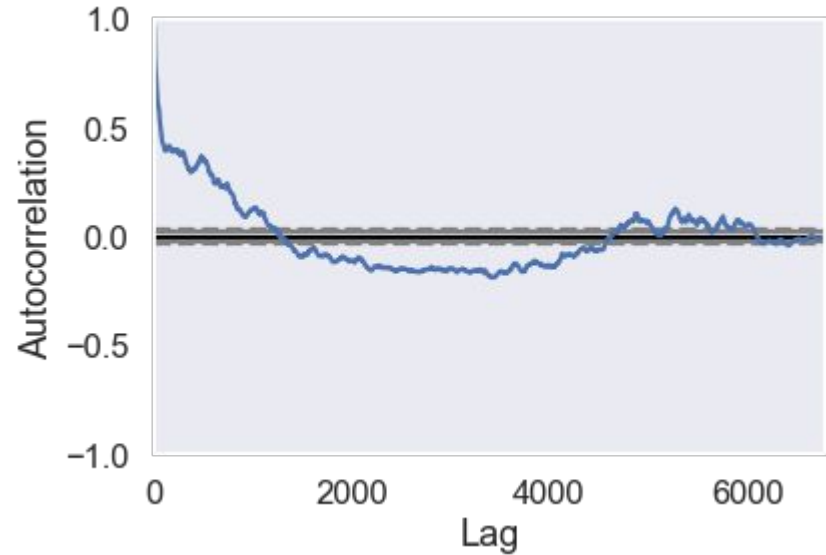
	Persistence Model	Normal Model	TS Model
MSE	<b>0.00447</b>	<b>0.01781</b>	<b>0.02953</b>
std	<b>0.00236</b>	<b>0.00967</b>	<b>0.01947</b>



## Data: do we have a Random Walk?

- Persistence Model is still the best
- Temporal dependence from used features is strong

→ add more features to Model

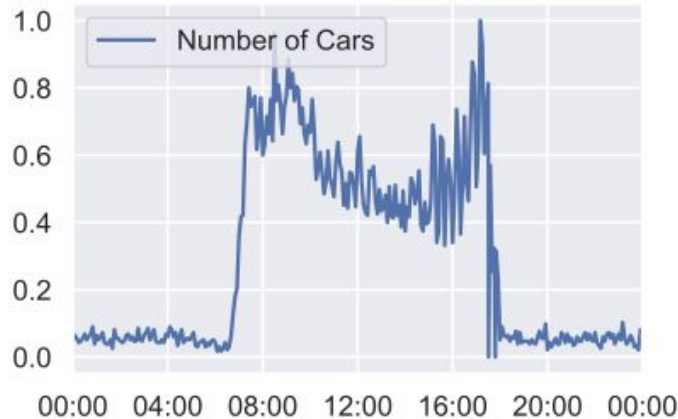


**PM10 Autocorrelation**

# Working with webcam data

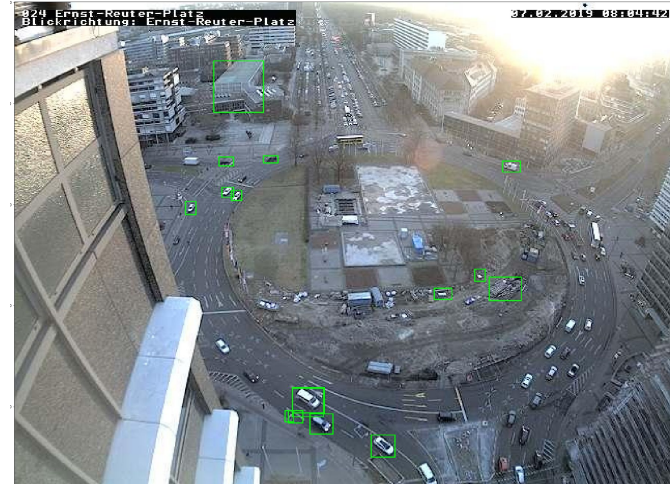
# Webcam Data

- Store an image every 5 minutes
- Process with pre-trained neural network (Mask R-CNN / COCO)
- Result: car timeseries for february 2019



# Webcam Problems

- Only short timespan
- Pre-trained model not optimized on perspective and night
- No exact numbers, only magnitude



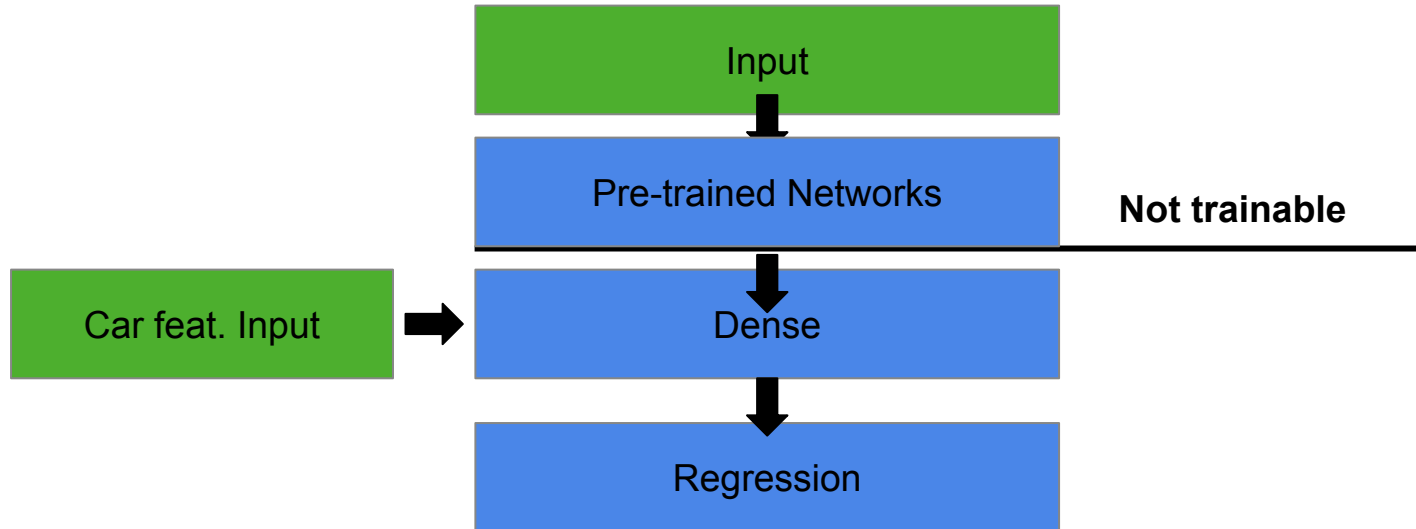


Combining both data sources

# Transfer Learning

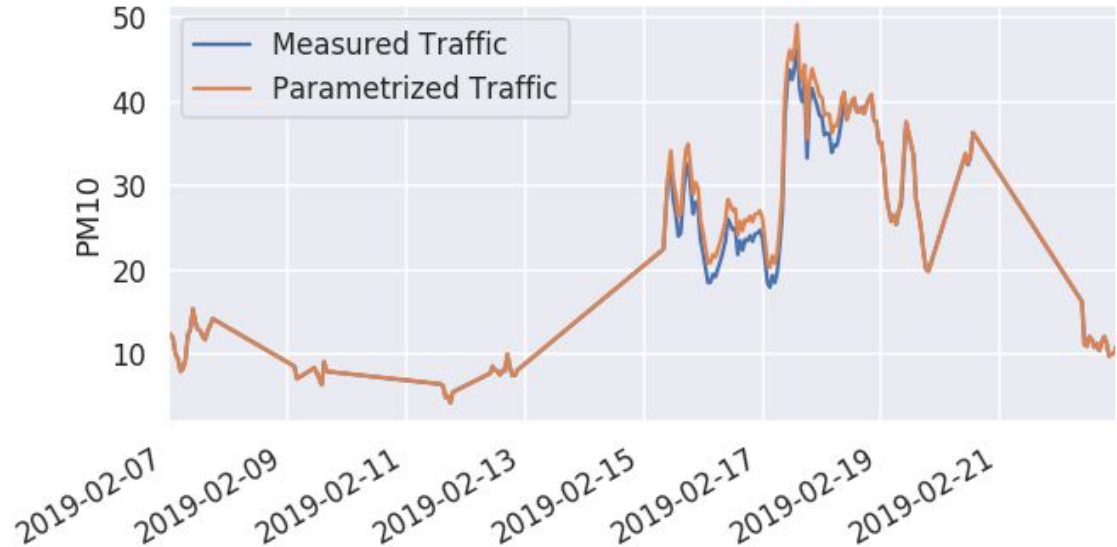
# Transfer Learning

- Car Dataset is very small compared to other data
- Approach: train model first and use extracted features with car data



# The Influence of Traffic

New models traffic  
can now be  
**parametrized  
manually** to **simulate  
future traffic**  
behaviour



Blue: PM measures for actual traffic  
Orange: PM measures for simulated traffic  
(higher traffic higher traffic density)

# Conclusion and Outlook

## Conclusion and Outlook

- Pollution time series can not easily be forecasted from the given data
- Traffic Data can be obtained from public sources
- To investigate the relationship between particulate matter pollution and traffic, different features and more data are needed

# On Github

<https://github.com/GeorgesAlkhourigolddust>

Thank you for your attention.

Questions?