

# PREDICTING TCP/IP NETWORK TRAFFIC USING TIME SERIES FORECASTING

FINAL PRESENTATION

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Thomas Mauerhofer, and Matthias Wölbitsch

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**goal: forecast TCP/IP traffic**

- real-time and short-time

**data set**

- network traffic of three months

**approaches**

- classical time series prediction methods
- artificial neural networks

## data set preparation

- split into training, and test set
- important to start at the beginning of a period

## optimize parameter

- with grid search
- from 0.01 to 1.0 with 0.05 steps

## parameter

- influence of the previous element
- weight for the level ( $\alpha$ )
- weight for the trend ( $\beta$ )
- weight for the seasonality ( $\gamma$ )

# RESULTS

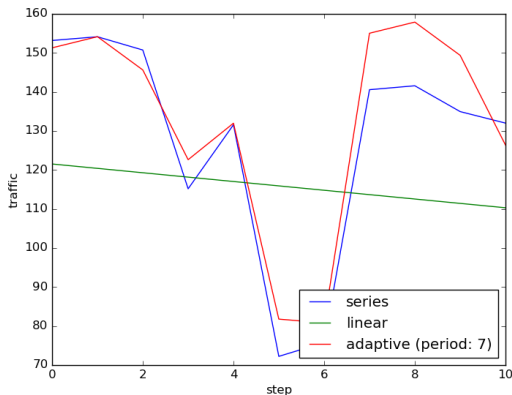
## forecast daily traffic

linear:

- $\alpha$ : 0.1
- $\beta$ : 0.2
- mase: 1.1768

addaptive:

- $\alpha$ : 0.0
- $\beta$ : 0.0
- $\gamma$ : 1.0
- period: 7
- mase: 0.3168



# RESULTS

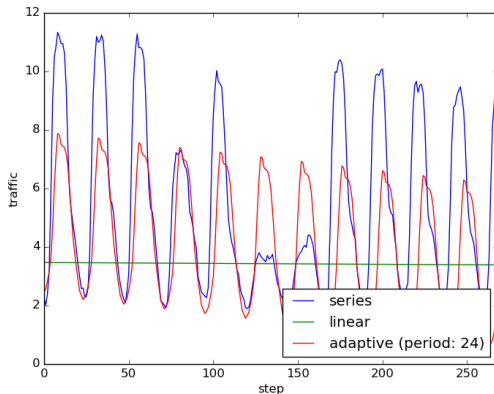
## forecast hourly traffic

linear:

- $\alpha$ : 0.15
- $\beta$ : 0.1
- mase: 4.3413

addaptive:

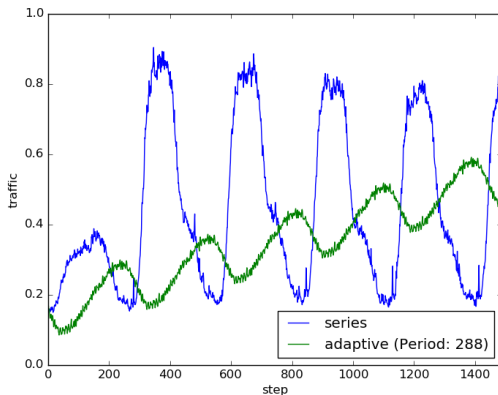
- $\alpha$ : 0.7
- $\beta$ : 0.0
- $\gamma$ : 1.0
- period: 24
- mase: 3.2444



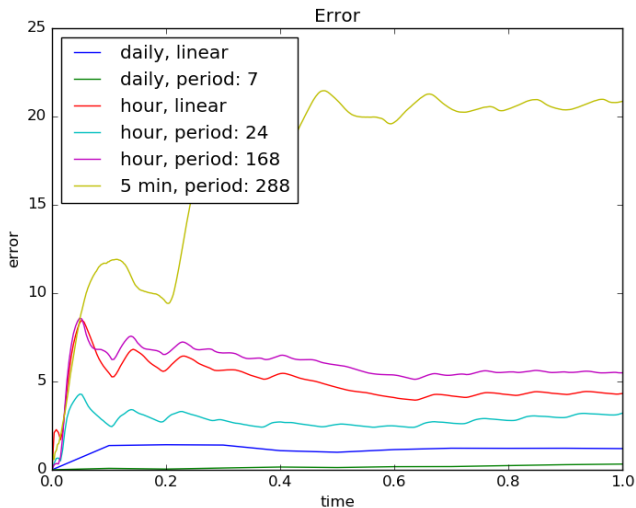
## forecast 5 minute traffic

addaptive:

- $\alpha$ : 0.9
- $\beta$ : 0.65
- $\gamma$ : 0.9
- period: 288
- mase: 20.8613



## Error-development over the complete test set



## data set preparation

- generate sequences using sliding window
- split into training, validation, and test set

## neural network library

- keras
- theano

## hyper parameter search

- sliding window, number of neurons, number of layers,...
- hyperopt library
- tree-structured parzen estimator



# RESULTS

## MLP

- $N = 25$
- $W = \{1, 2, 4, 8\} \cup \{287, 288, 289\}$

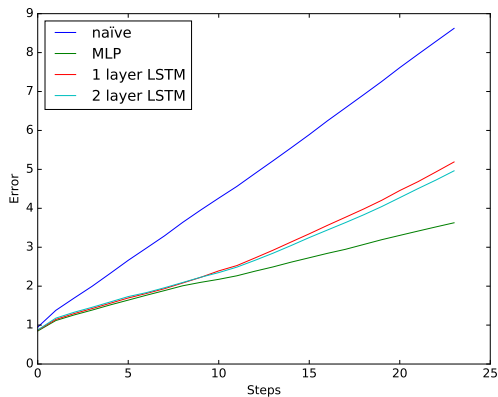
## 1 layer LSTM

- $N = 19$
- $W = \{1, 2, \dots, 19\}$

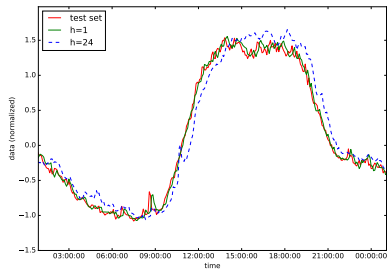
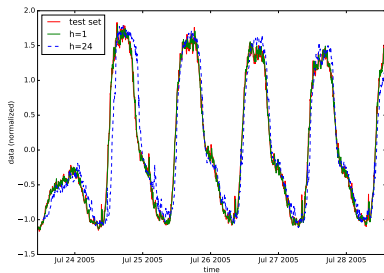
## 2 layer LSTM

- $N_1 = 13$
- $N_2 = 5$
- $W = \{1, 2, \dots, 14\}$

## forecast error for different horizons



forecasting examples with  $h = 1$  and  $h = 24$  using MLP



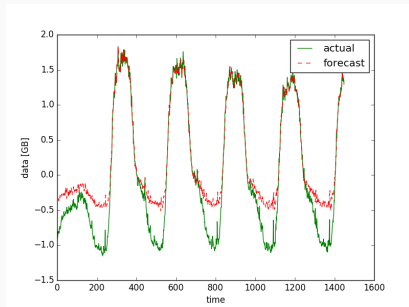
# CONCLUSION

## forecast horizon

- one step ahead forecasting
- direct vs. iterative forecasting

## training loss function

- MSLE
- penalizes underestimates
- numerical issues



## LSTM issues

- high expectations
- too few training samples
- slow

## neural networks and time series

- used often for forecasting
- numerous different approaches
- problem solved?

two very different methods

- neural networks: black box
- Holt-Winters: time series engineering

→ hard to compare

QUESTIONS?