# ~ Seasonal time series ~

Forecast time series with multiple seasonalities

Neue Fische - Capstone project Sandra Groß

### Structur

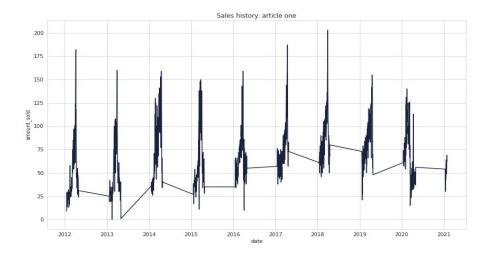
- > Introduction
- > EDA
- Models and forecast
- Conclusion & future work

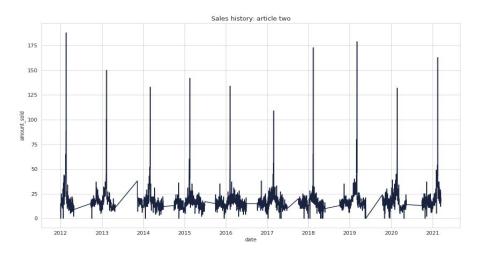


### Introduction

- Daily sales data
- 3 seasonal bakery products
  - A1: Osterpinze, A2 & A3: Berliner
- 1 Bakery branch in Wien

- Goal: Sales prediction for 2021
- No exogenous data

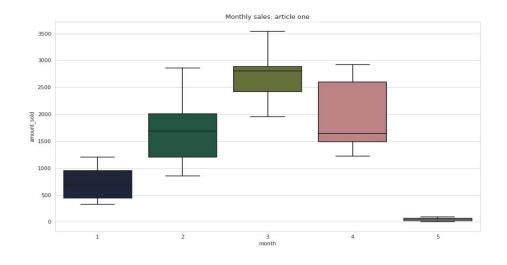


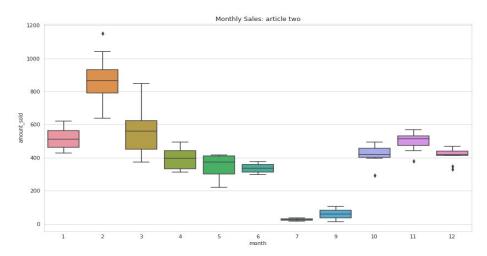


### Course of the time series

- A1: one season
  - beginning of the year
  - Seasonal frequency: 365

- > A2 & A3: two seasons
  - o beginning and end of the year
  - Seasonal frequency: 365

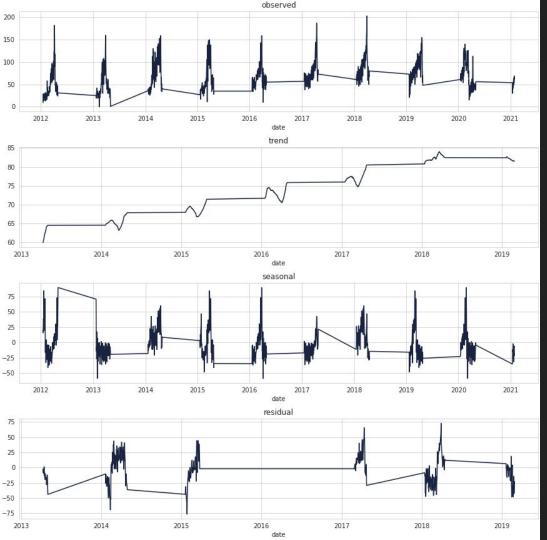




## Monthly sales

- A1: January April
  - Peak around easter
  - Seasonal frequency: 5 & 12

- A2 & A3: Oktober March/April
  - Peaks February & November
  - Seasonal frequency: 12



# Simple decompose

- requency = 365
- correlated Residuen
- underlying seasonalities
- Periodogram
  - f = 5,7, 12, 18, 43.5, 53, 77, 100, 365

#### Characteristics ts

Article one:

Multiple seasonalities

Gaps

Article two & three:

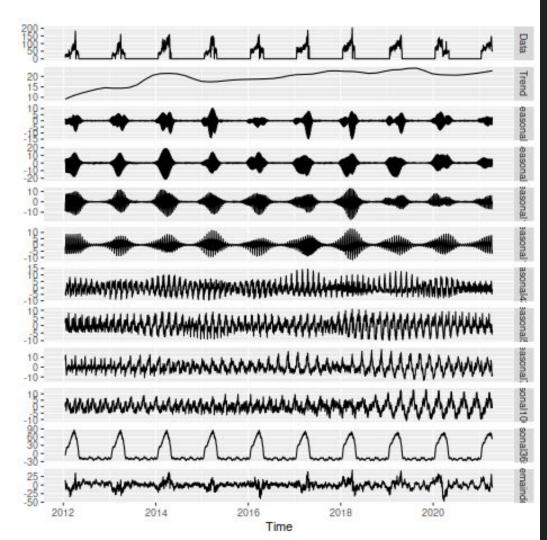
Article two & three:

Univariate

- Difference stationarity
- Normally distributed

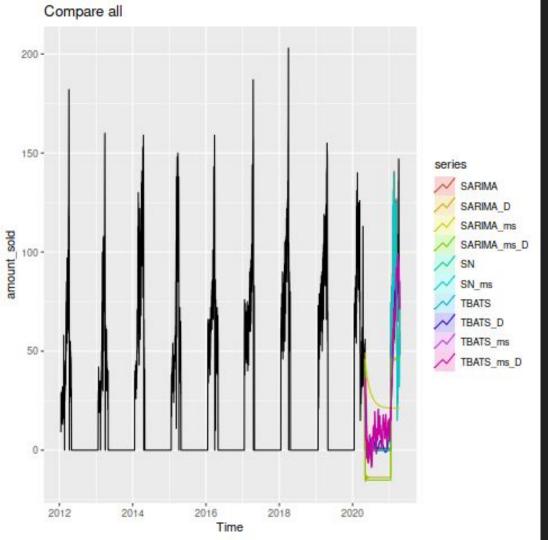
- Stationarity
- right skewed

correlated residuen



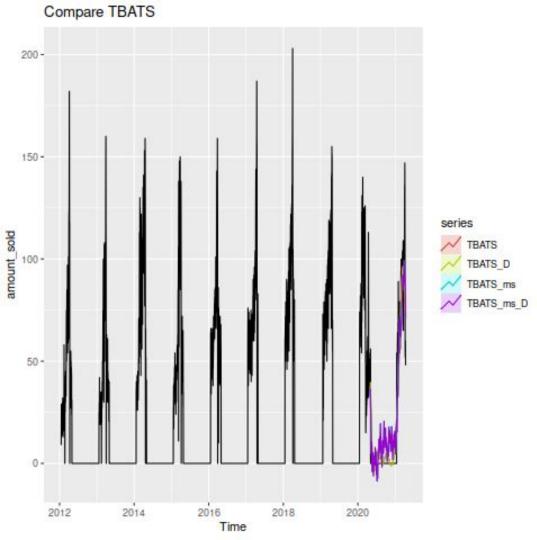
# Decompose multiple seasonalities

- Testet:
  - Only one frequency: 365
    - ts objects
  - Multiple seasonal periods
    - msts objects
  - Sale time dummy variable



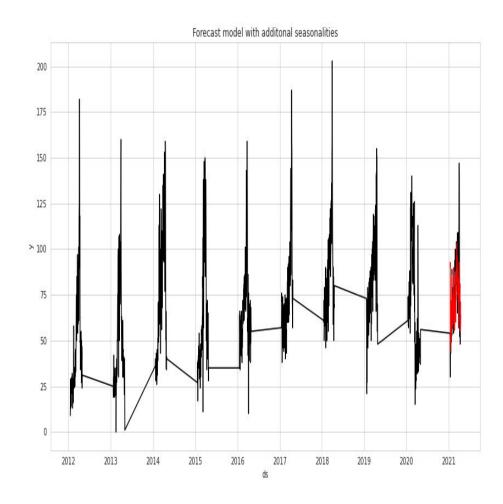
# Comparison all R models

- Metric: RMSE
- → ARIMA + ms → bottom
- Seasonal naiv → medium
  - No effect ms
- > TBATS plain → best



# TBATS comparison

- ➤ Best: pure TBATS
  - o RMSE: 11.61
- Dummies no influence
- ms marginally worse



## **fbProphet**

- 'Out of the box' forecasting
  - RMSE: 22.70
- Adding seasonalities
- Adding holidays Austria
  - RMSE: 23.17

### Conclusion & future work

- ms can be incorporated
  - o In classic ML models
  - In fbProphet
- Dummies as regressors is not sufficient

- Try further models
- Repeat for the other articles
- Hyperparameter tuning

### Thank you very much for your attention

