

# Forecasting number of natural gas consumers and their total consumption with R

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# Motivation

- Total consumption and price per unit are essential inputs for forecasting the revenues from delivered energy.
- Energy retail prices can differ for various tariffs.
- Larger customers get usually lower price and vice versa.
- The tariffs can be assigned automatically based on historical consumption.

# Goal of our research

We wanted to develop a prediction model with the following properties:

- 1 we forecast customer counts and their total consumption within each tariff class
- 2 tariffs are assigned to customers based on their consumption level
- 3 forecasts are based on regular invoicing data
- 4 forecast is conditioned by a long-term normal temperature
- 5 model should be implemented in a user-friendly way and run on standard PC

# Challenges

Factors that make reaching the goal difficult

- Forecast variables are not independent:
  - ① total consumptions (naturally) depends on the customer count
  - ② customers can switch between tariffs (as a result of consumption level variability)

**Result:** covariance structure should be considered

- Invoicing periods differ between various customers

**Result:** data need to be transformed

# Package structure

- 1 Data preprocessing – conversion of invoicing data to input data
- 2 Parameter estimation
- 3 Forecasting

# Standardized load profiles

- Model for disaggregation of consumption
- Makes daily consumptions from time aggregates, e.g. annual
- GAM with temperature and calendar as explanatory variables
- Brabec et. al (2015). *Statistical Models for Disaggregation and Reaggregation of Natural Gas Consumption Data*. Journal of Applied Statistics 42(5)

# Prediction model construction

## Basic ideas I

- 1 Two-level model – customer counts forecast (incl. tariff switches) as the first level, consumption totals forecast as the second
- 2 Transition from the forecast time series to a new one – time series of tariff assignment for a particular customer
- 3 Forecasting based on Markov property

$$\hat{p}_{t+1} = p_t P_t$$

# Prediction model construction

## Basic ideas II

- 1 Probability can be estimated using relative frequency

$$\hat{p}_{ct} = N_{ct}/N_{\bullet t}$$

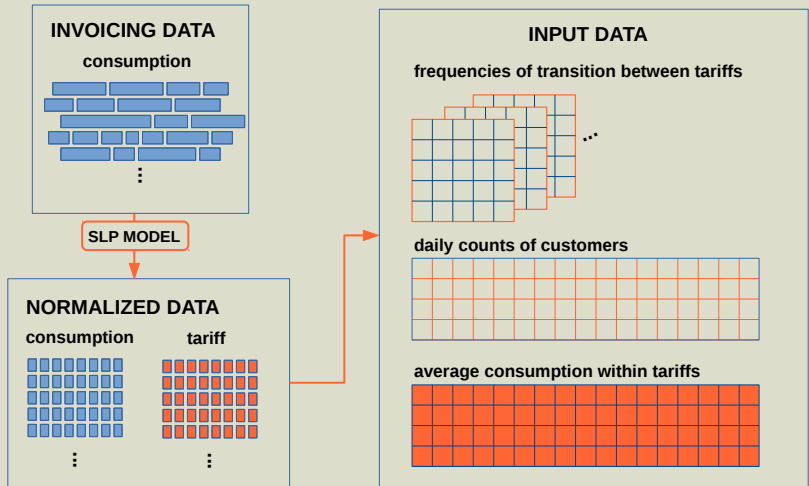
we can invert the procedure and work with

$$\hat{N}_{ct} = p_{ct}N_{\bullet t}$$

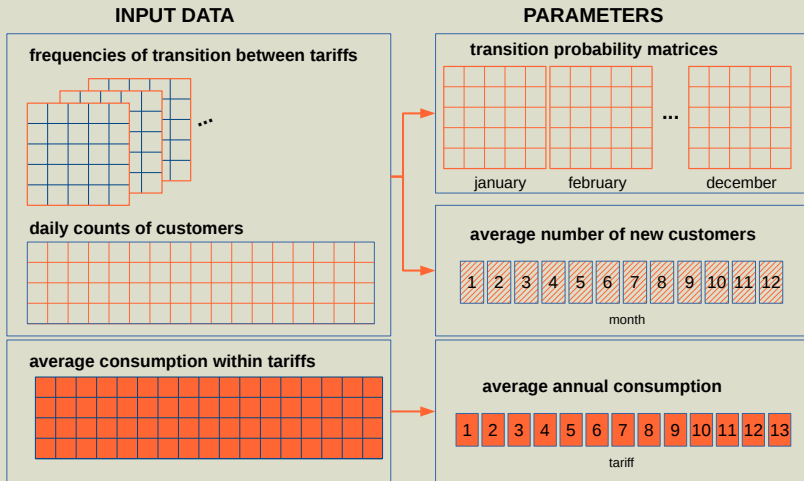
- 2 Number of new customers forecast as a separate module
- 3 Customer counts forecasts are multiplied by average consumption forecasts



# Data preprocessing

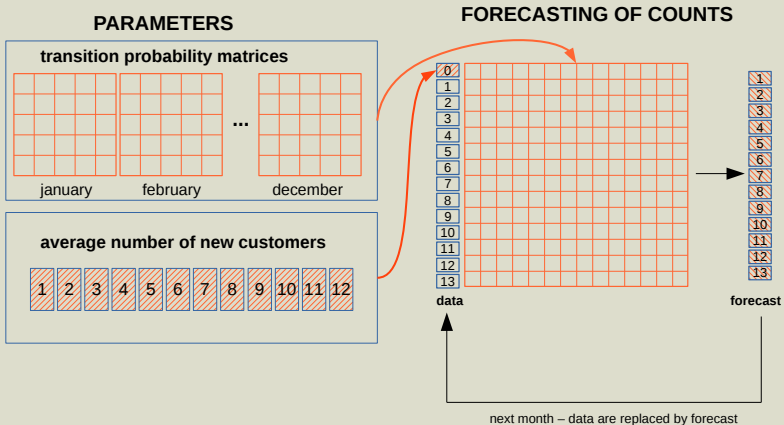


# Estimation of parameters



# Forecasting

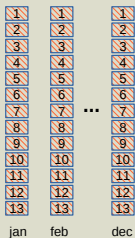
## Level 1



# Forecasting

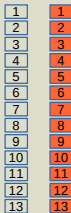
## Level 2

monthly counts forecasts



averageing

annual counts forecast



product

tariff by tariff

annual consumption forecast



### PARAMETERS

average annual consumption within tariffs



Thank you for your attention.