

Modeling and forecasting of electricity prices and demand

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1 Introduction

Struktura energetyki w Danii

The electricity market in Denmark is divided into 2 groups (DK1 and DK2). First group (DK1) consists of regions: Nordjylland, Midtjylland and Syddanmark; second group (DK2) consists of regions Sjælland and Hovedstaden with the capital Copenhagen.



Figure 1: An image of a galaxy

2 Forecasting methods

3 Data analysis

All data I used to perform forecasts I downloaded from the official webpage of Nordpool power exchange[1]. Datasets are divided into the year files and periods (hours, weeks etc.). I managed to download following datasets (valid for the day 14.05.2020):

- Consumption - hourly
- Consumption prognosis - hourly
- Wind power - hourly
- Wind power prognosis - hourly
- Elspot prices (as Price) - hourly

All of the datasets were available for years 2013-2020, except for Consumption prognosis (2015-2020). So I decided to focus on analysis only on the period 2015-2020, because 4 years time frame is still sufficient for calculations.

Units of downloaded data are following:

- Consumption and Wind Power- MWh
- Price - DKK/MWh

The files were downloaded, merged, splitted for regions DK1 and DK2, pivoted in order to have separated hours as parameters for each day and merged for all years. Example for consumption DK1 is presented below.

	date	holiday	0	1	2	...	22	23
0	2016-01-01	1	1818.0	1741.0	1660.0	...	1858.0	1713.0
1	2016-01-02	0	1615.0	1510.0	1461.0	...	2027.0	1822.0
2	2016-01-03	1	1724.0	1665.0	1671.0	...	2127.0	1998.0
3	2016-01-04	0	1844.0	1803.0	1789.0	...	2293.0	2079.0
4	2016-01-05	0	1940.0	1891.0	1952.0	...	2372.0	2193.0

Table 1: First 5 rows of merged file Consumption DK1.

Missing values

Data was very consistent and almost only single values were missing. Single missing values were replaced by average of neighbor cells and in case of missing value in neighbor cell, value was fixed manually (with file `fill_empty_cells.py`). Half of day 2018-09-18 from Wind prognosis files was filled taking closest neighbors and counting average for whole vector (with file `fill_empty_cells_wind_prognosis_DK.py`). Missing values were reduced to zero.

Dataset	DK1	DK2
Consumption	5	5
Consumption prognosis	5	5
Price	21	12
Wind power	12	6
Wind power prognosis	18	19

Table 2: Missing values in files.

3.1 Load data

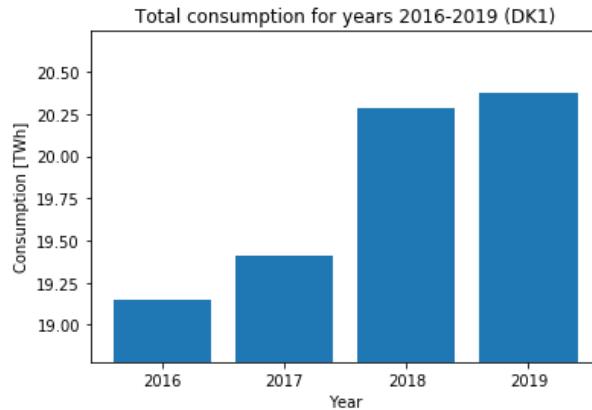


Figure 2:

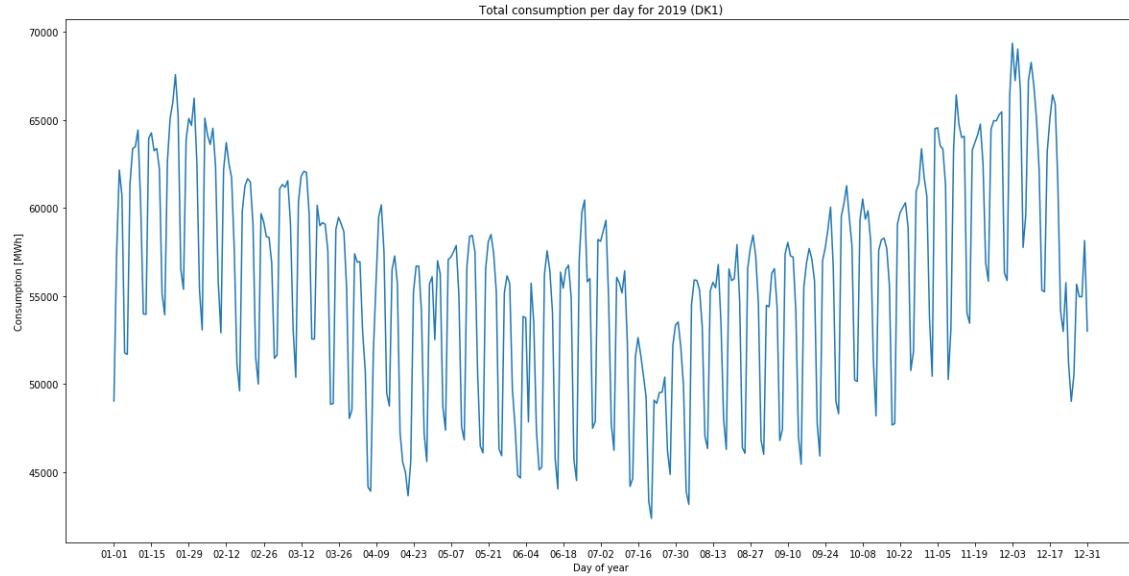


Figure 3:

Average consumption for years 2016-2020 (DK1)

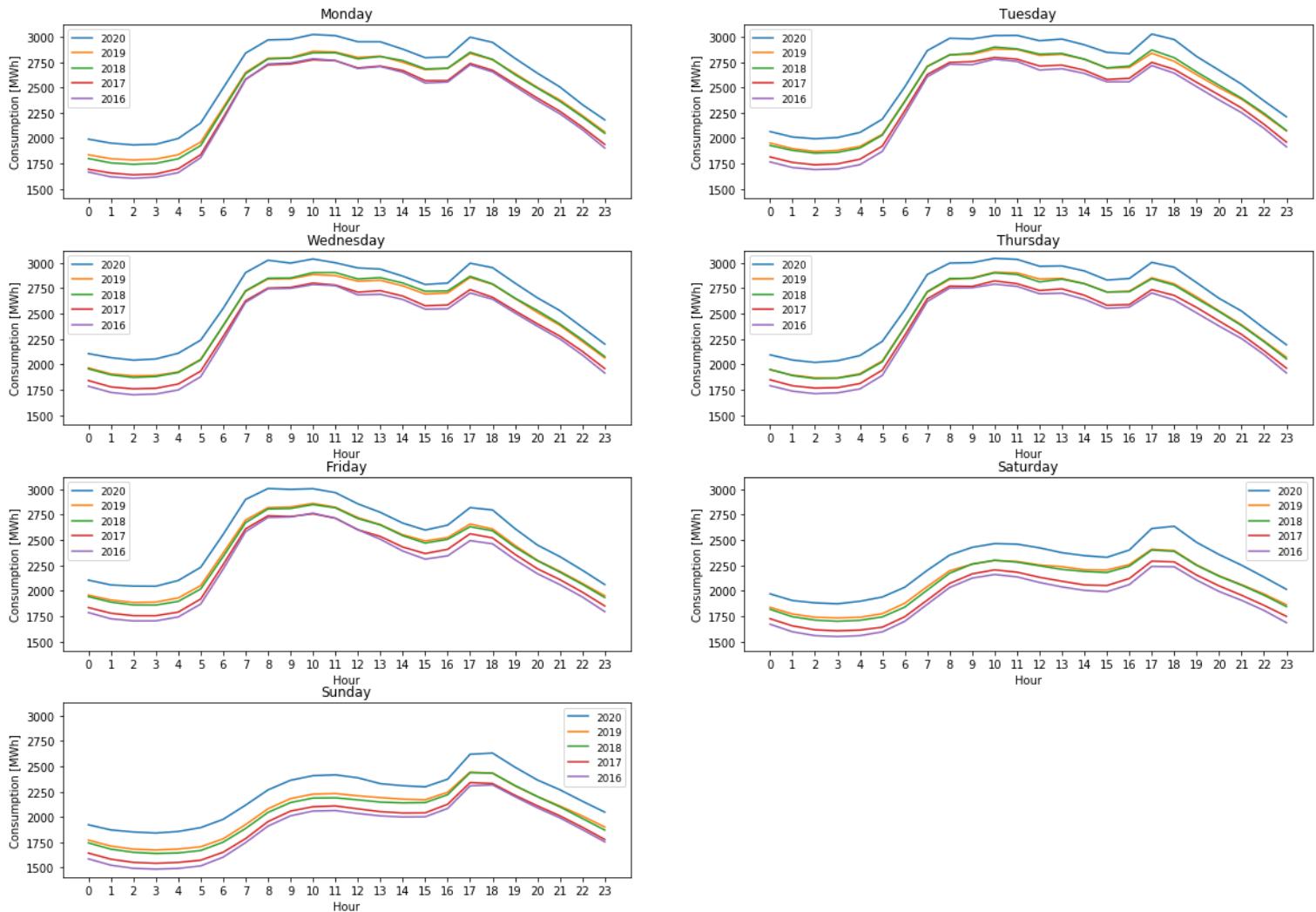


Figure 4:

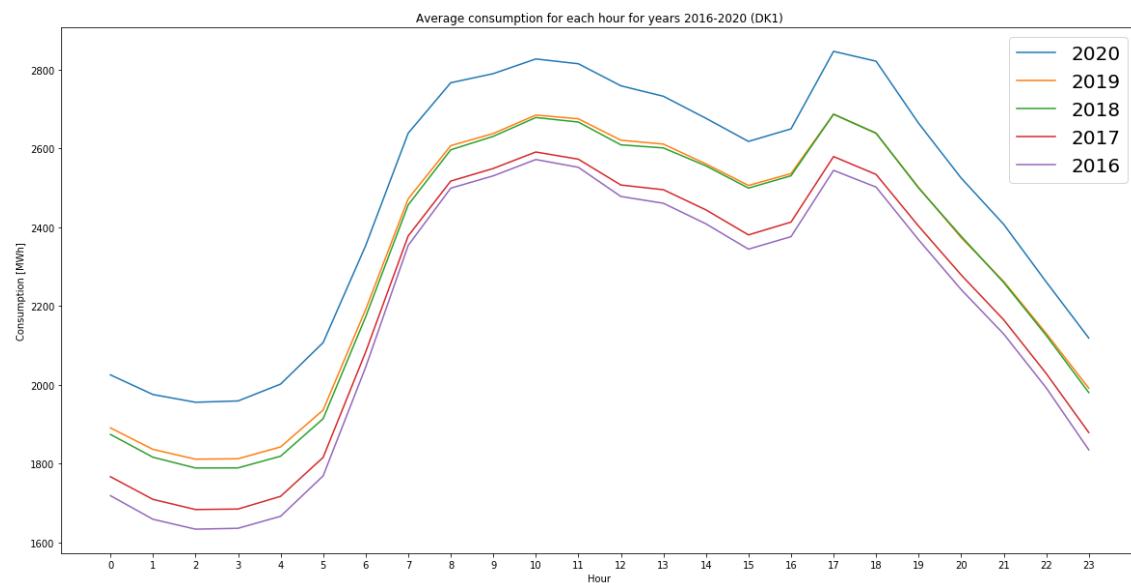


Figure 5:

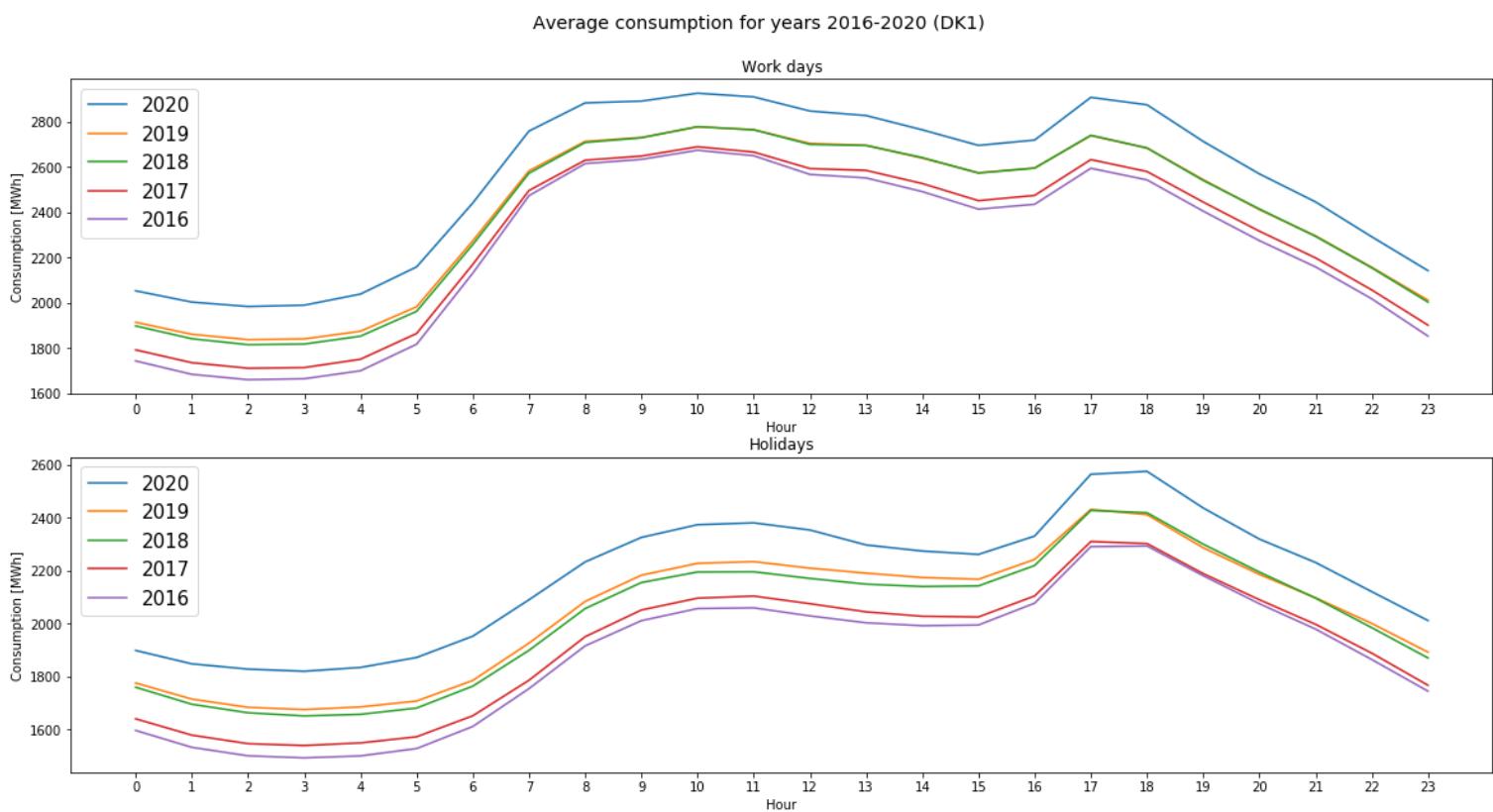


Figure 6:

Median consumption for years 2016-2020 (DK1)

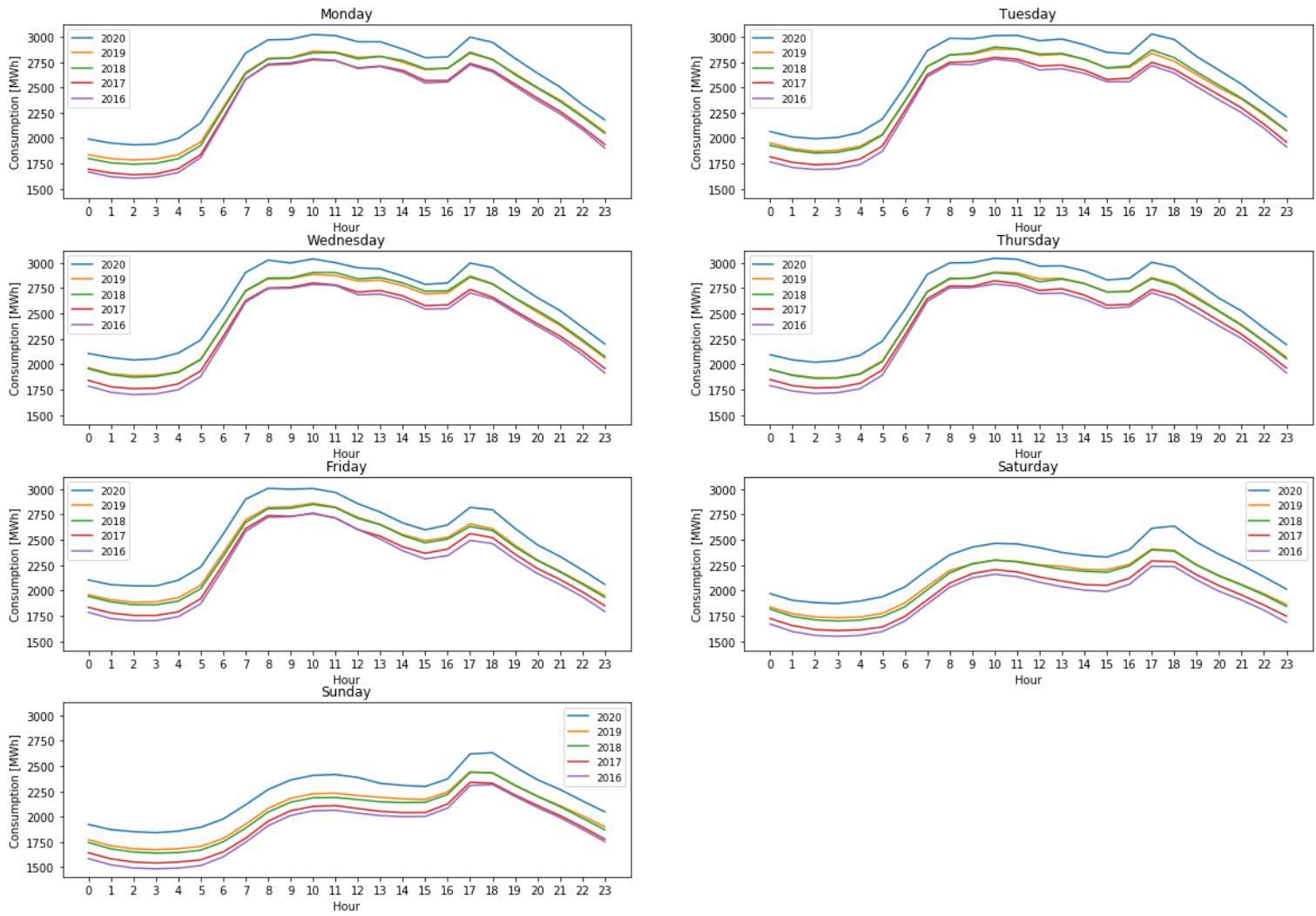


Figure 7:

Standard deviation of consumption for years 2016-2020 (DK1)

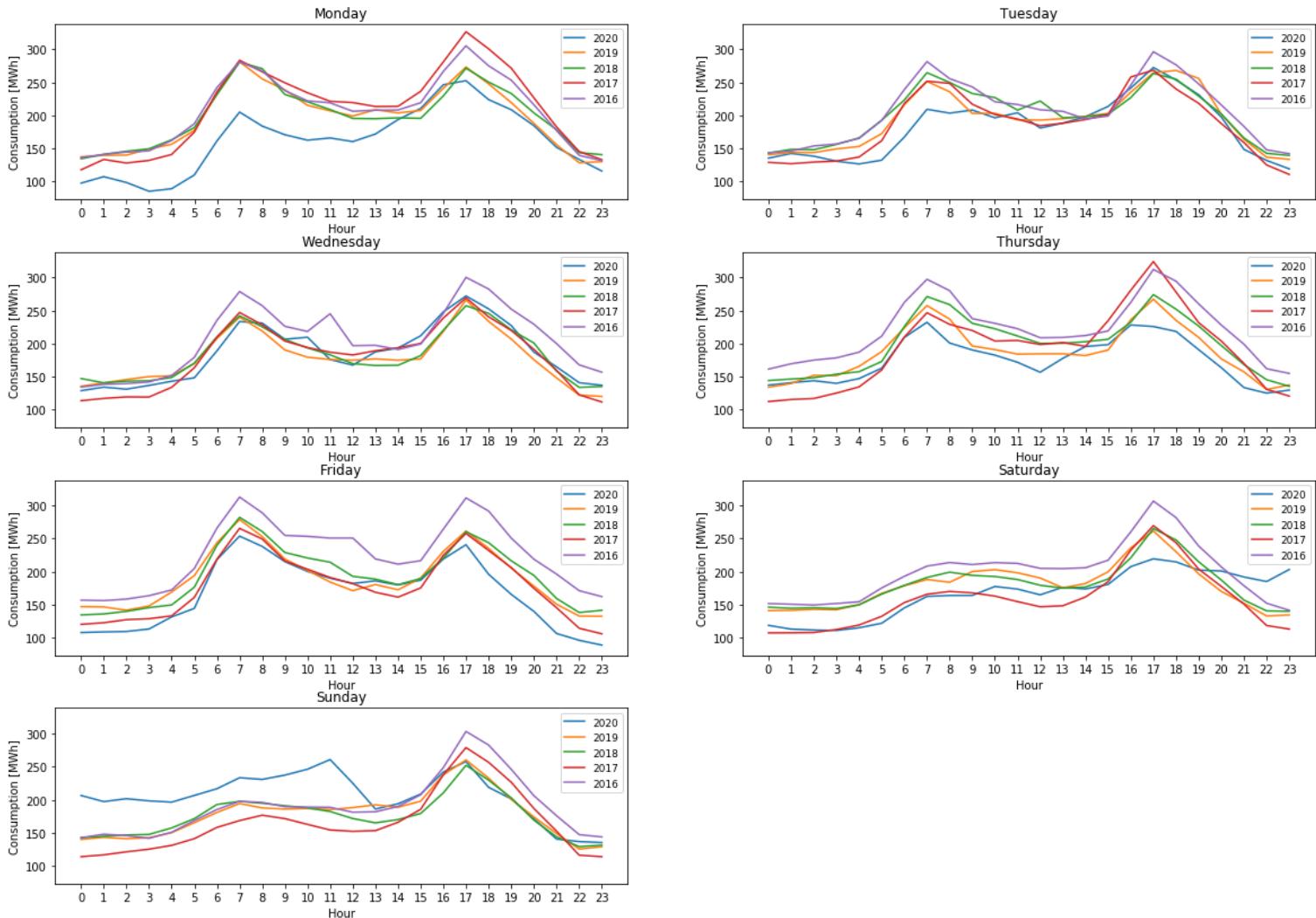


Figure 8:

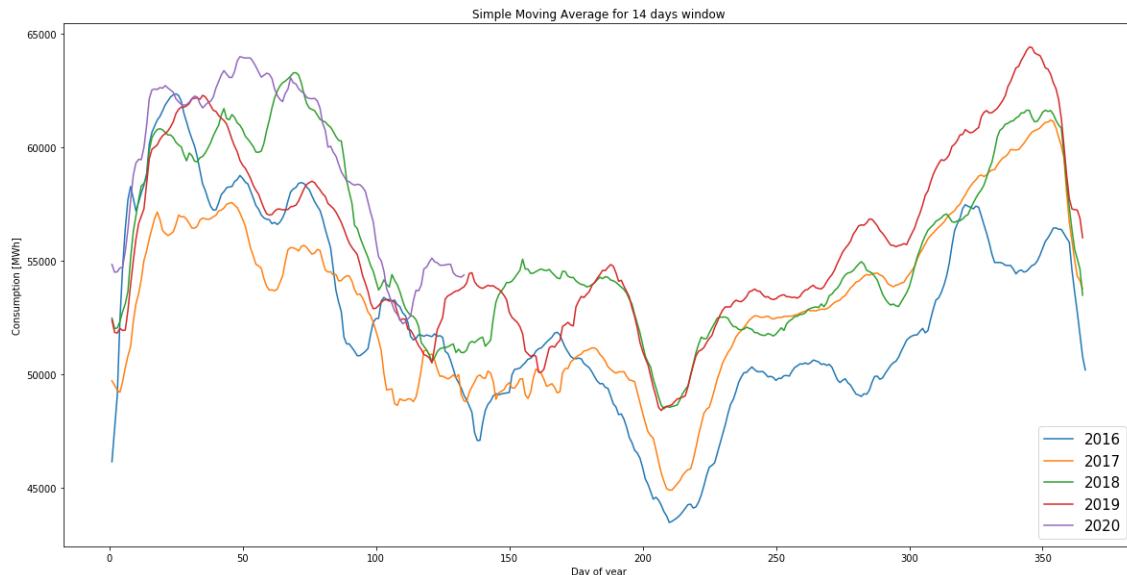


Figure 9:

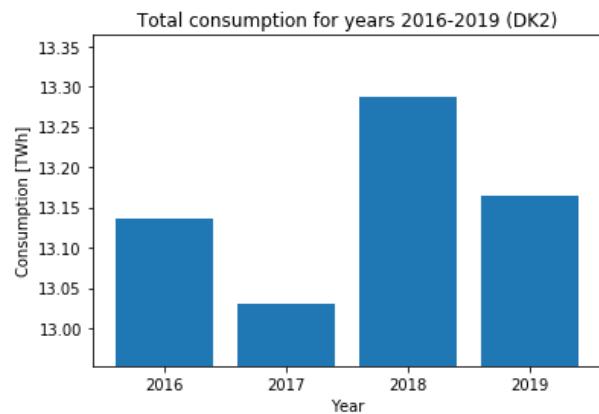


Figure 10:

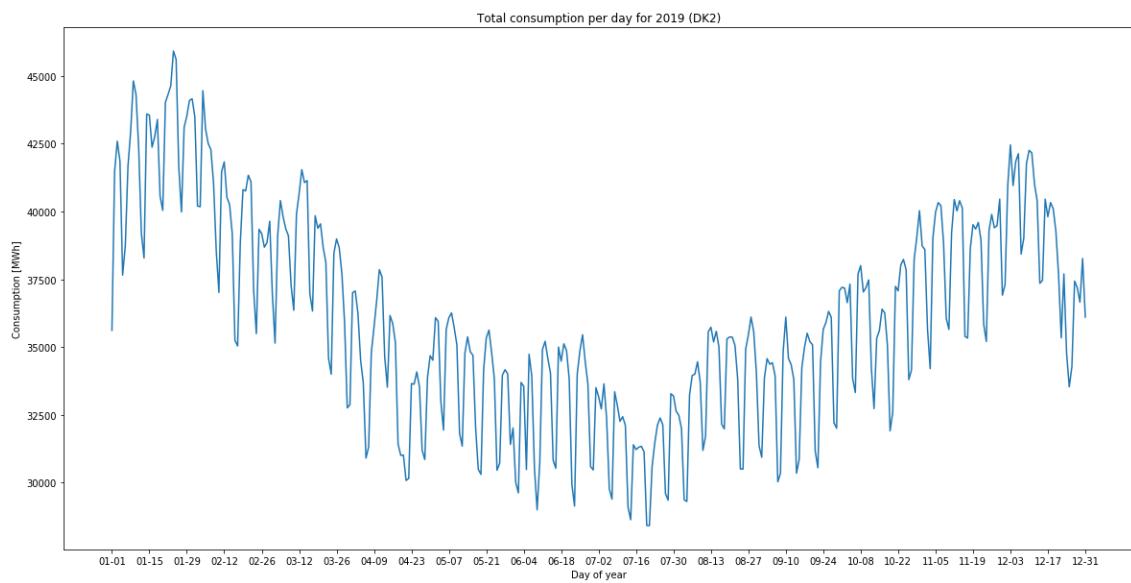


Figure 11:

Average consumption for years 2016-2020 (DK2)

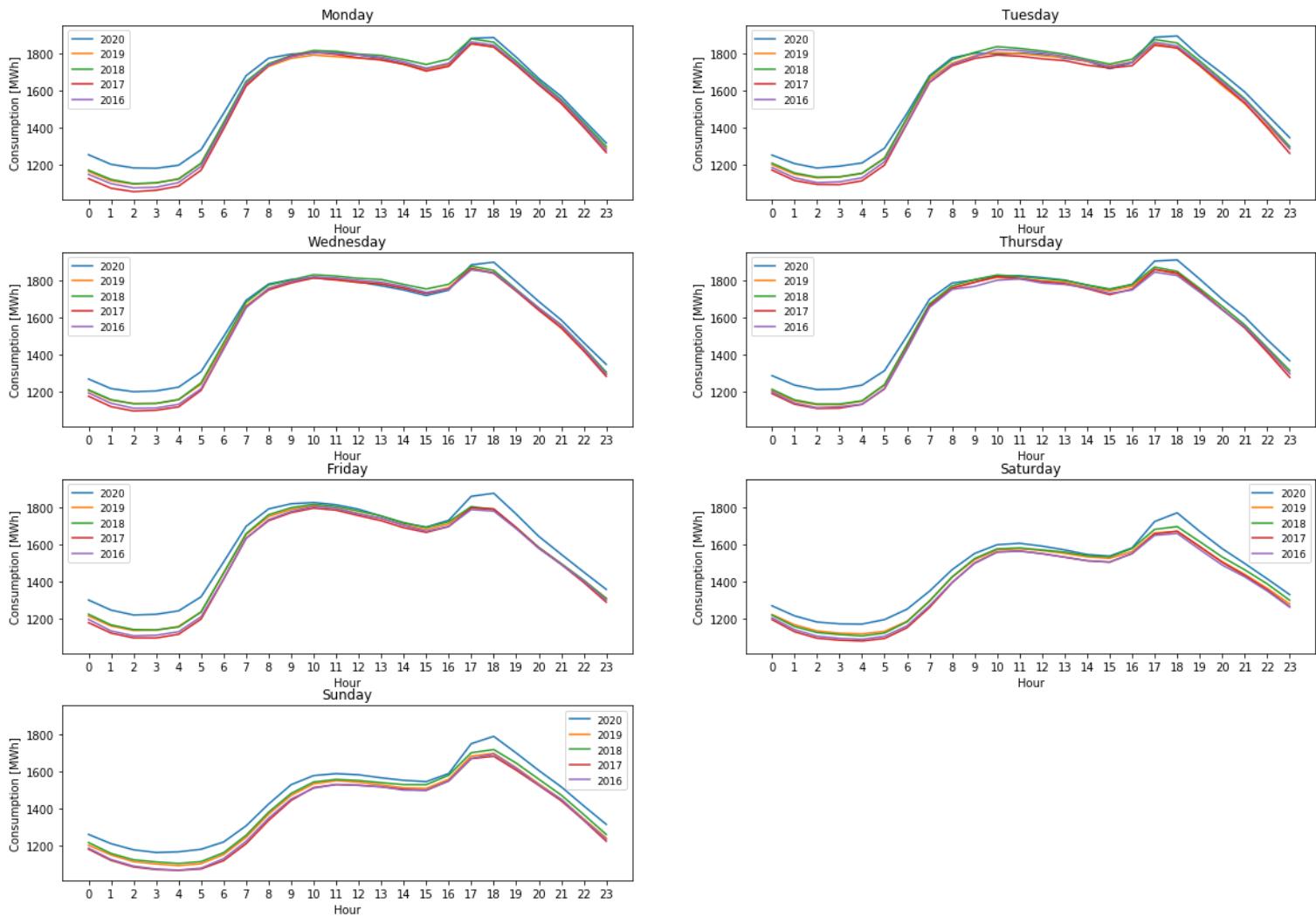


Figure 12:

Average consumption for each hour for years 2016-2020 (DK2)

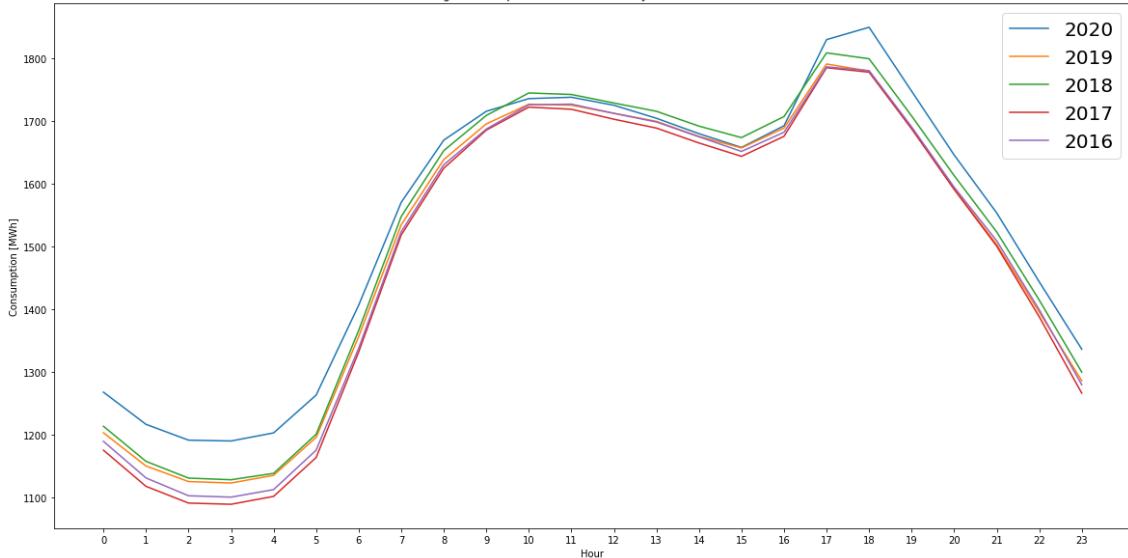


Figure 13:

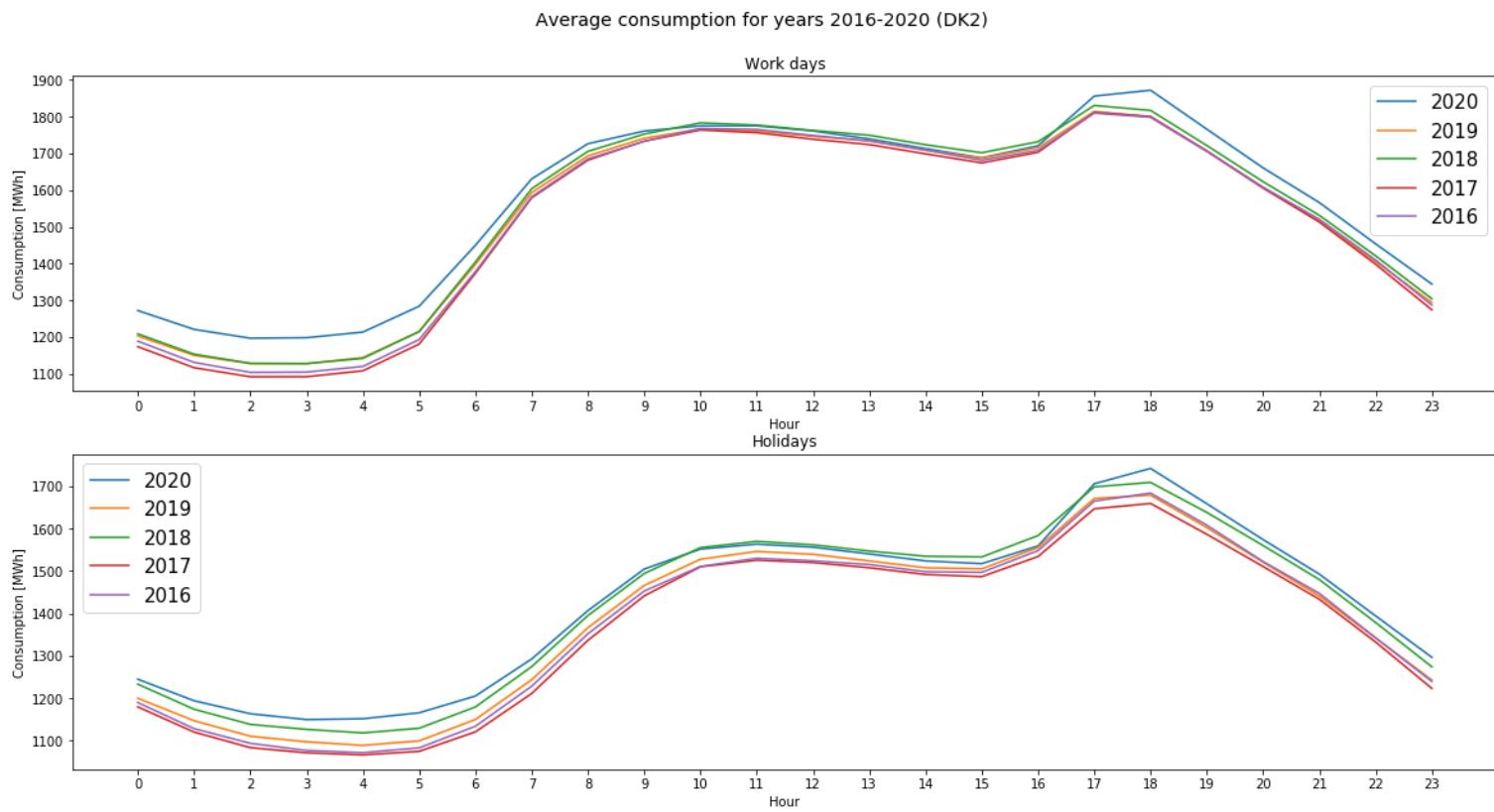


Figure 14:

Median consumption for years 2016-2020 (DK2)

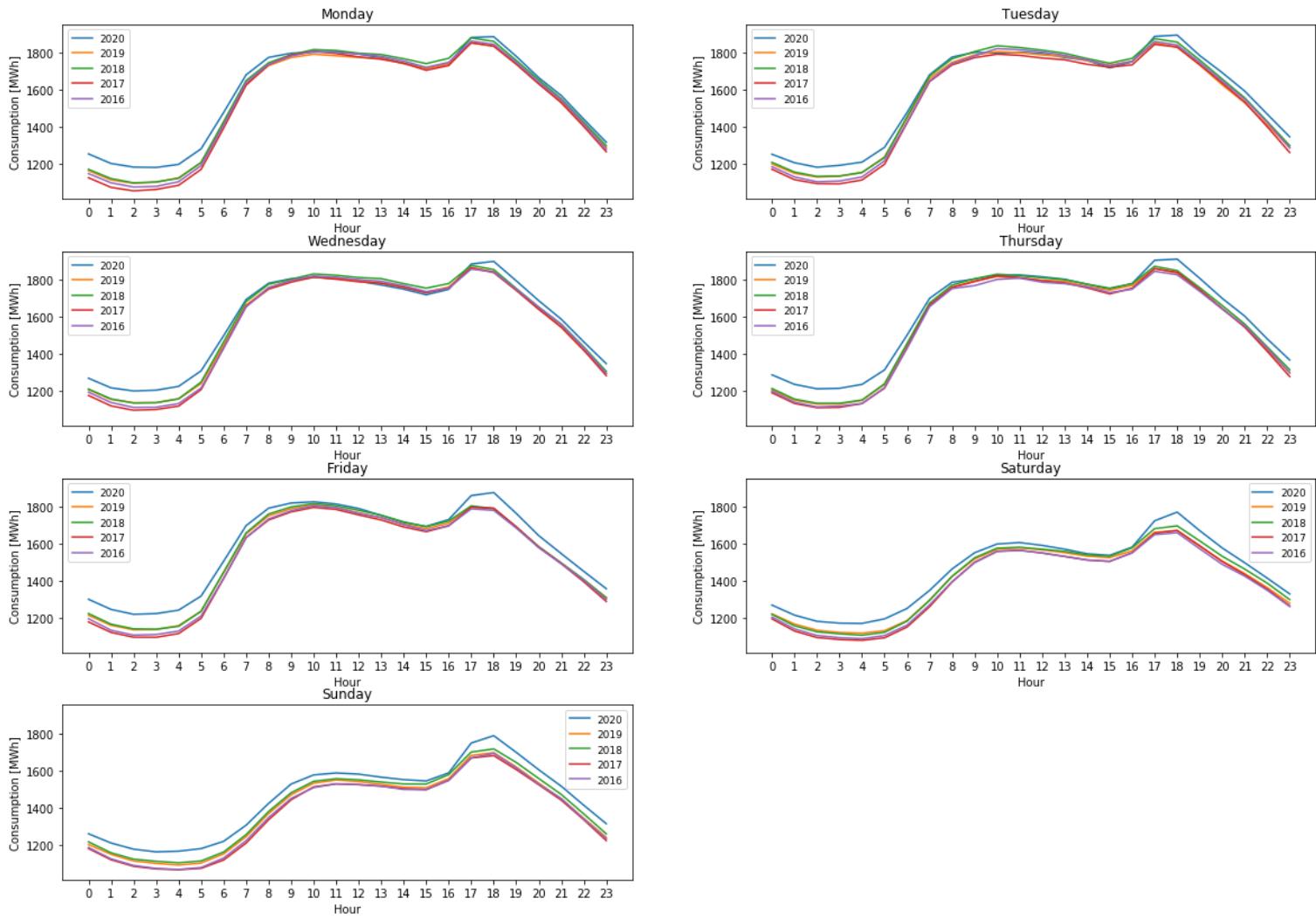


Figure 15:

Standard deviation of consumption for years 2016-2020 (DK2)

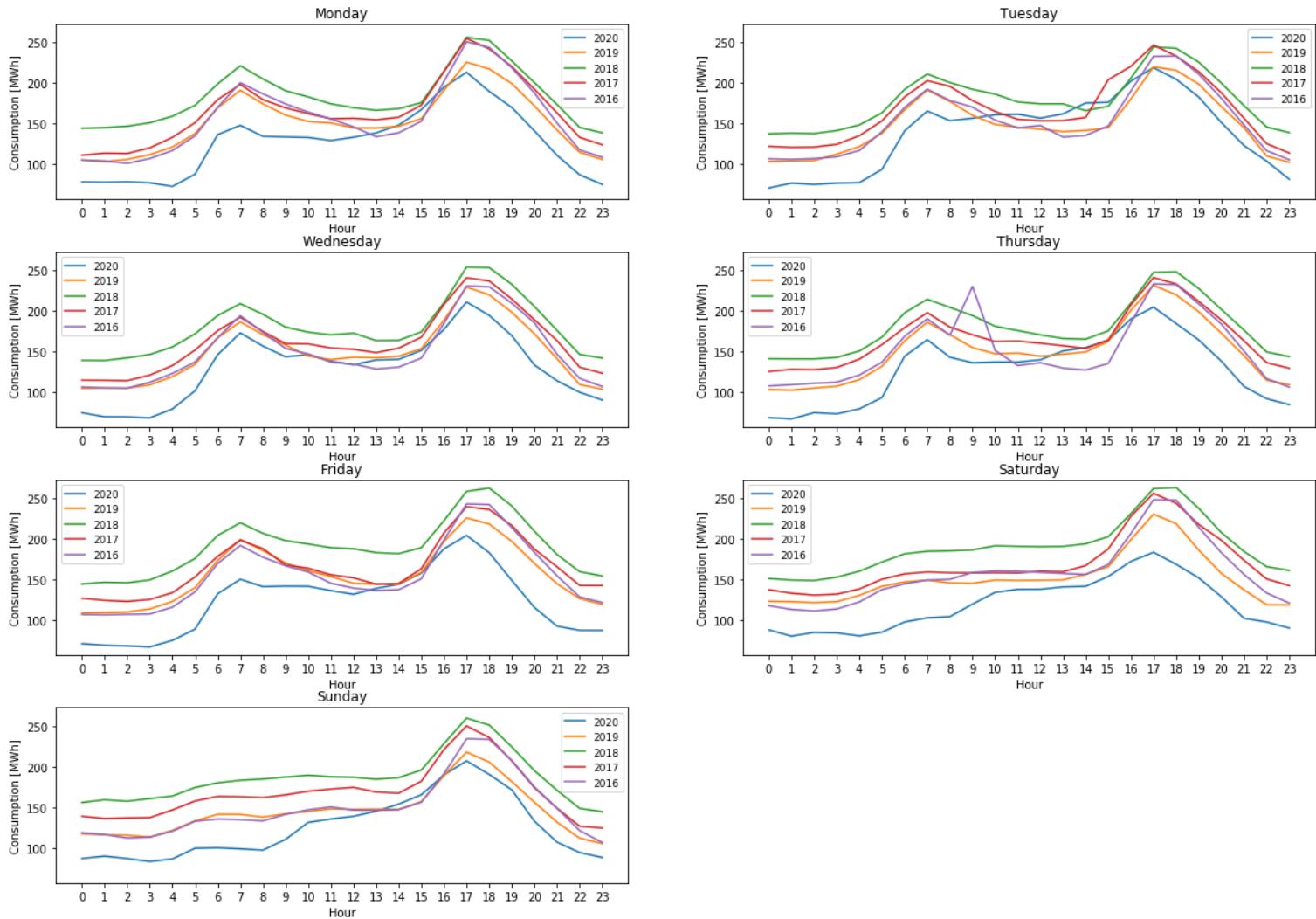


Figure 16:

Simple Moving Average for 14 days window

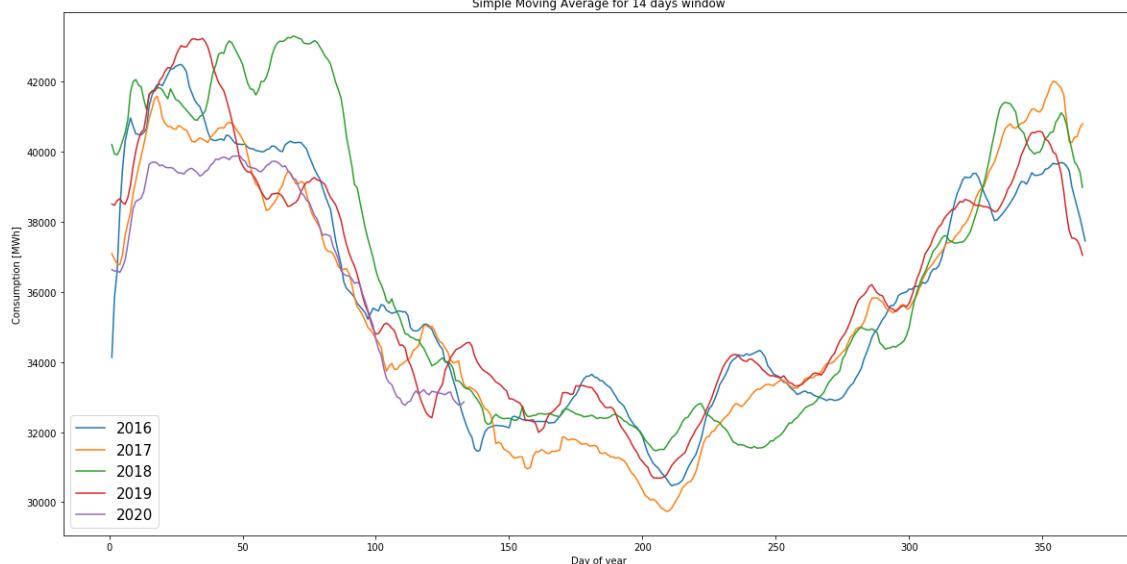


Figure 17:

3.1.1 Load data prognosis

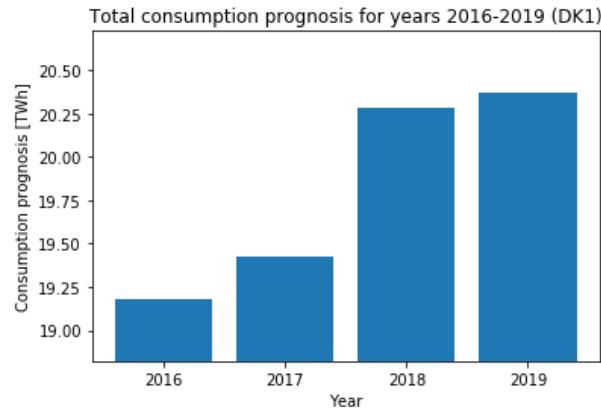


Figure 18:

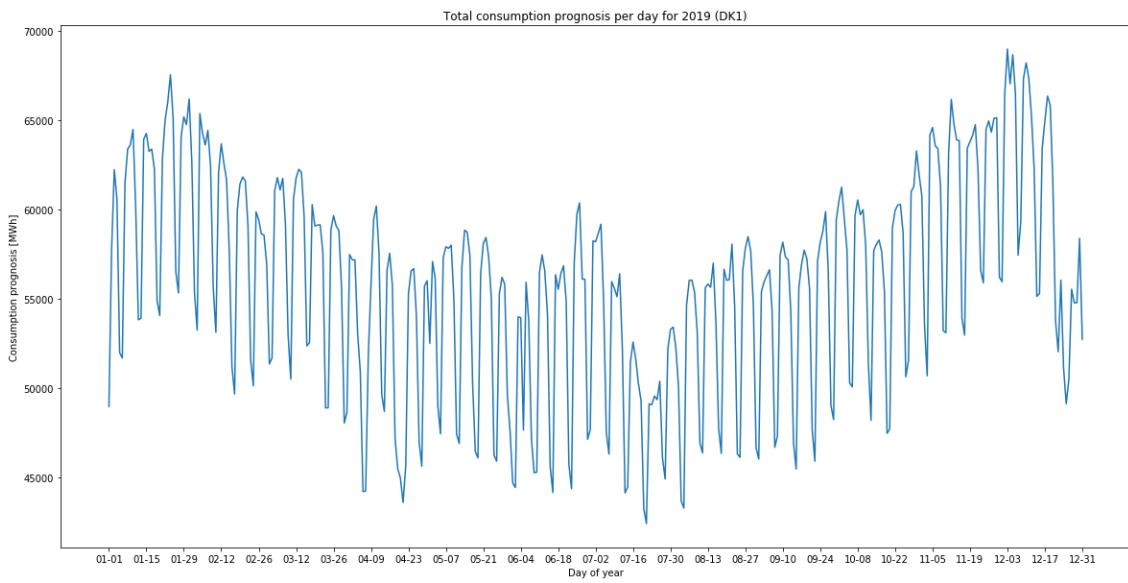


Figure 19:

Average consumption prognosis for years 2016-2020 (DK1)

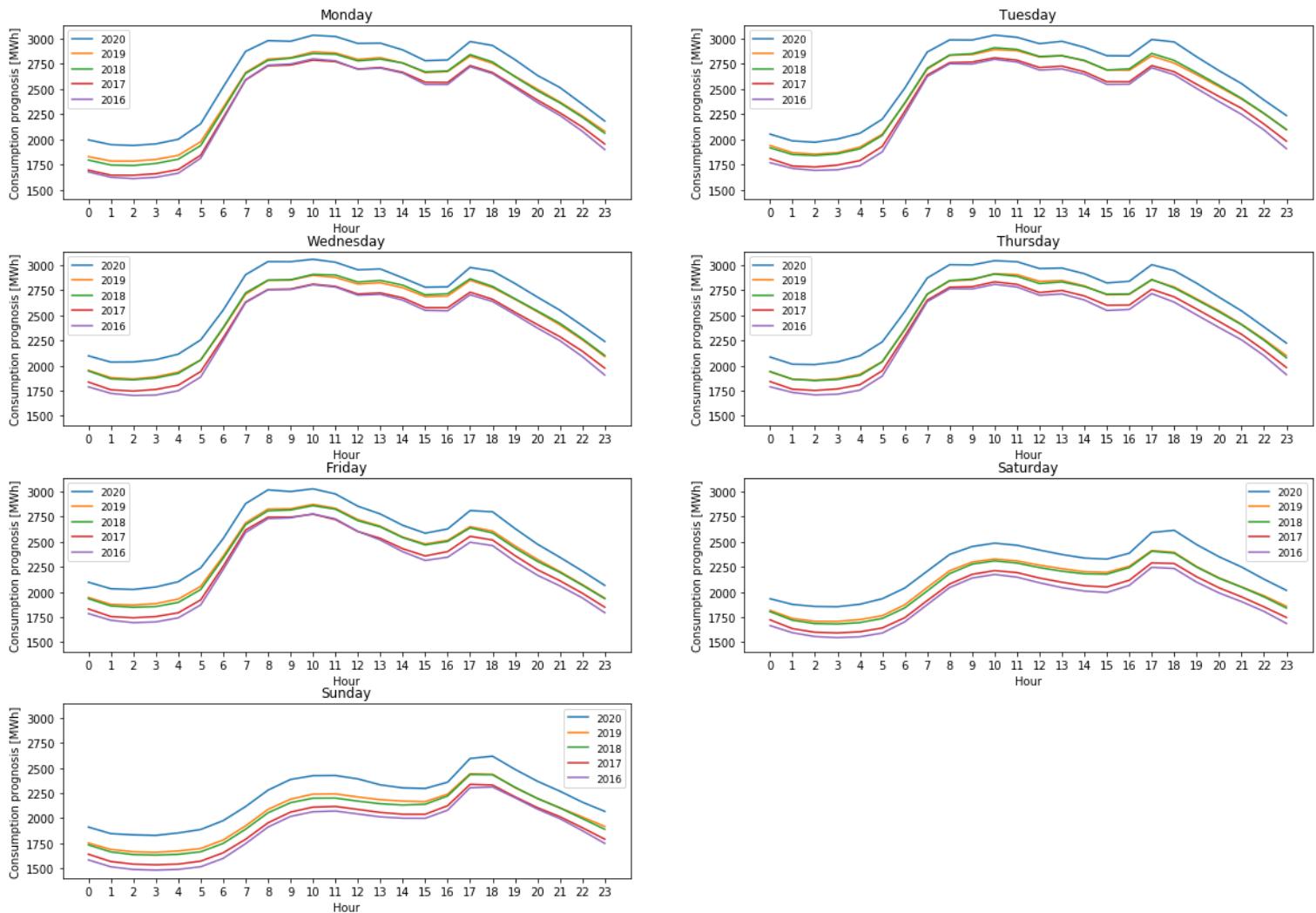


Figure 20:

Average consumption prognosis for each hour for years 2016-2020 (DK1)

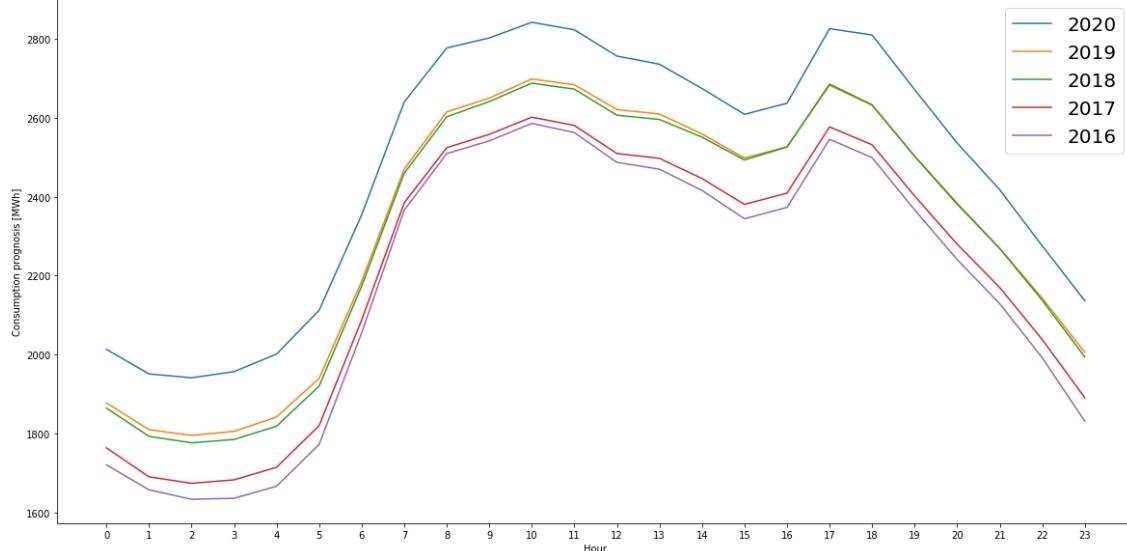


Figure 21:

Average consumption prognosis for years 2016-2020 (DK1)

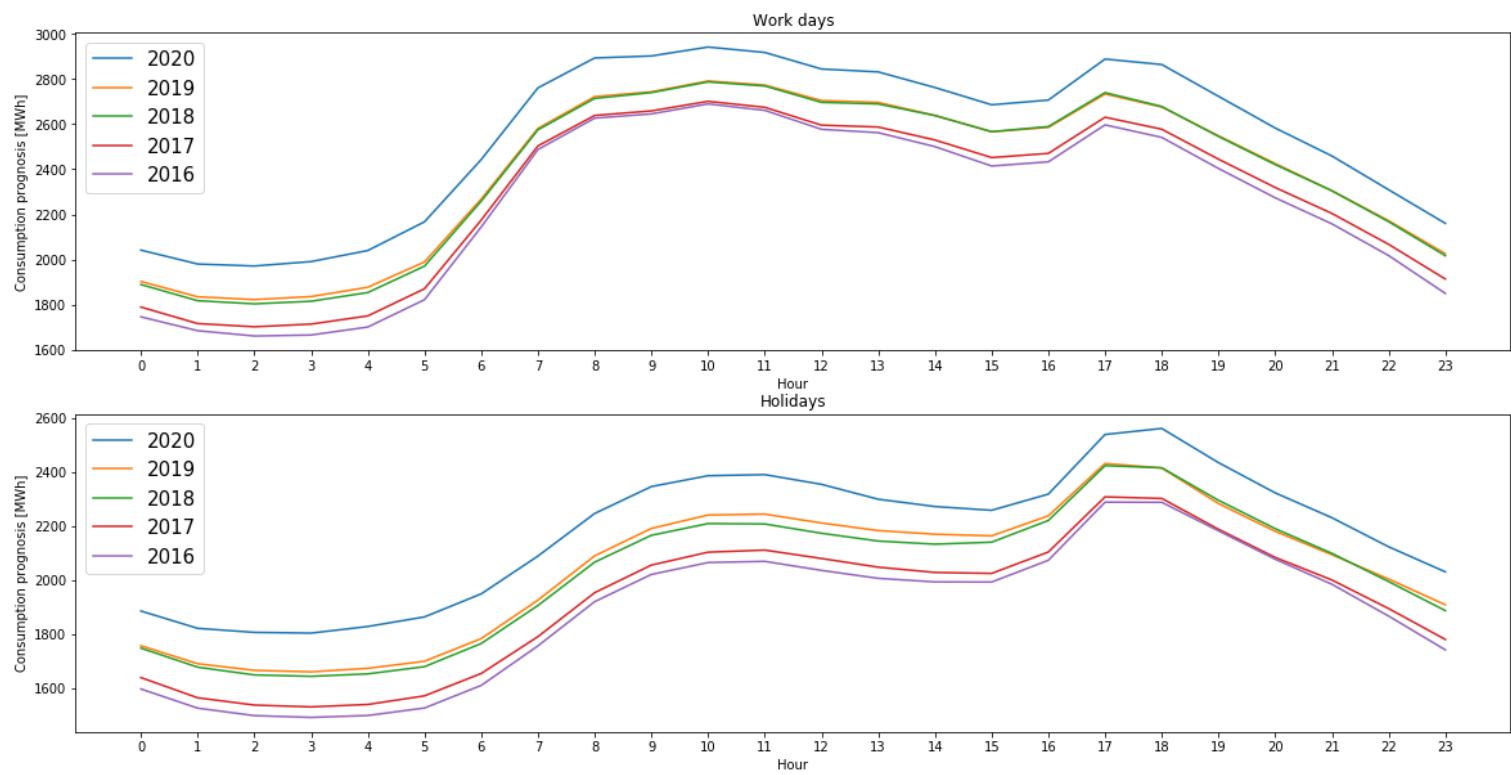


Figure 22:

Median consumption prognosis for years 2016-2020 (DK1)

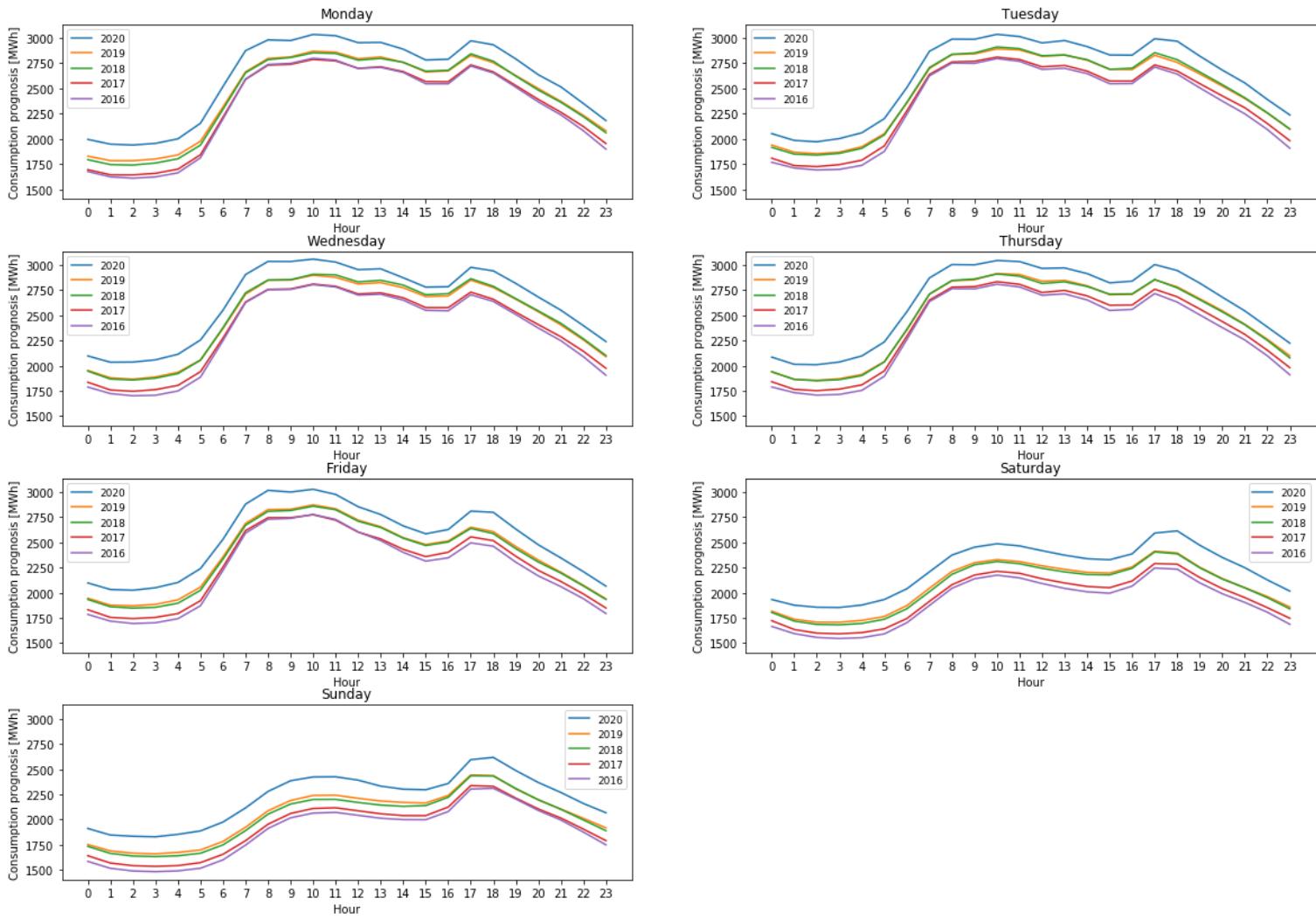


Figure 23:

Standard deviation of consumption prognosis for years 2016-2020 (DK1)

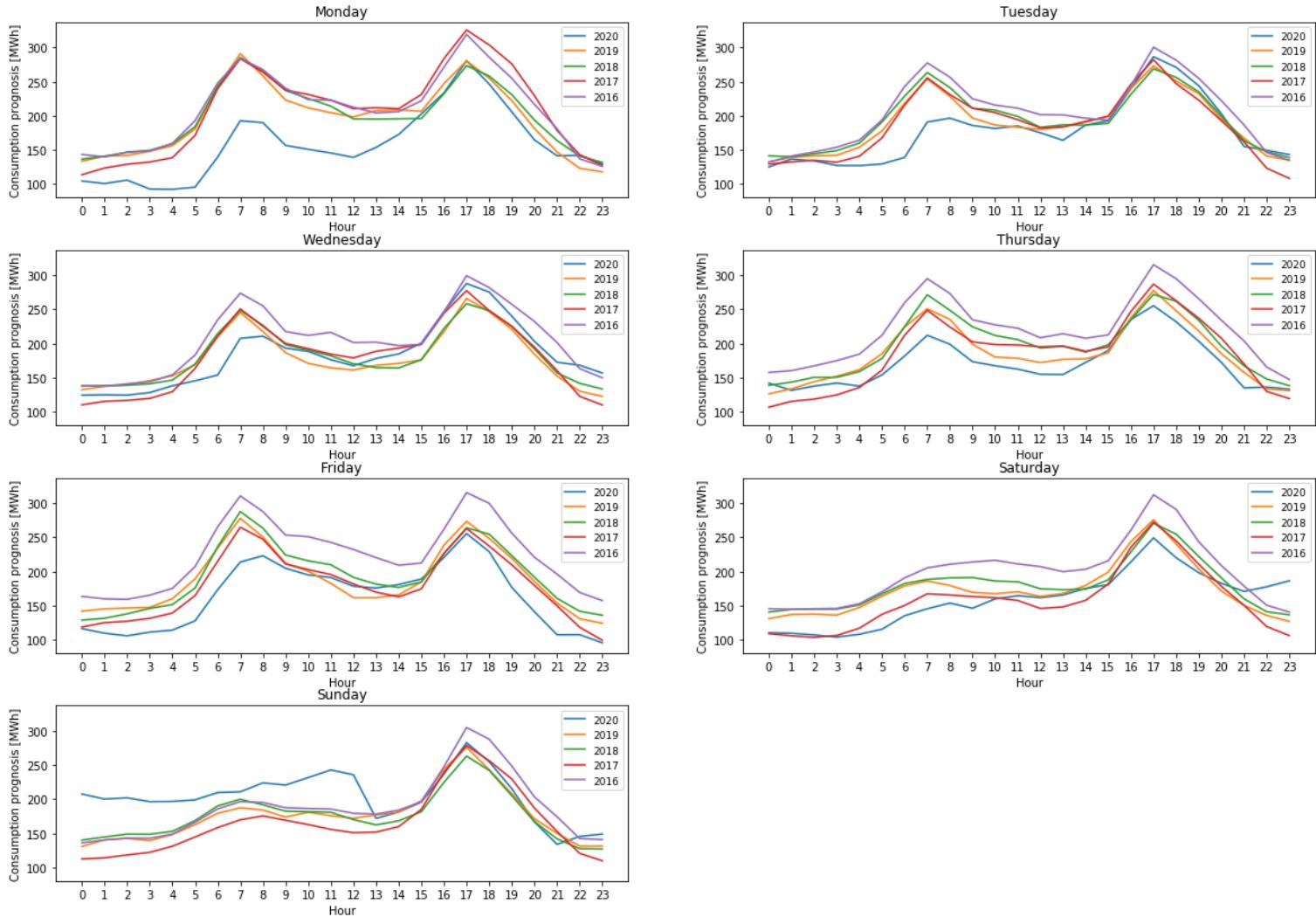


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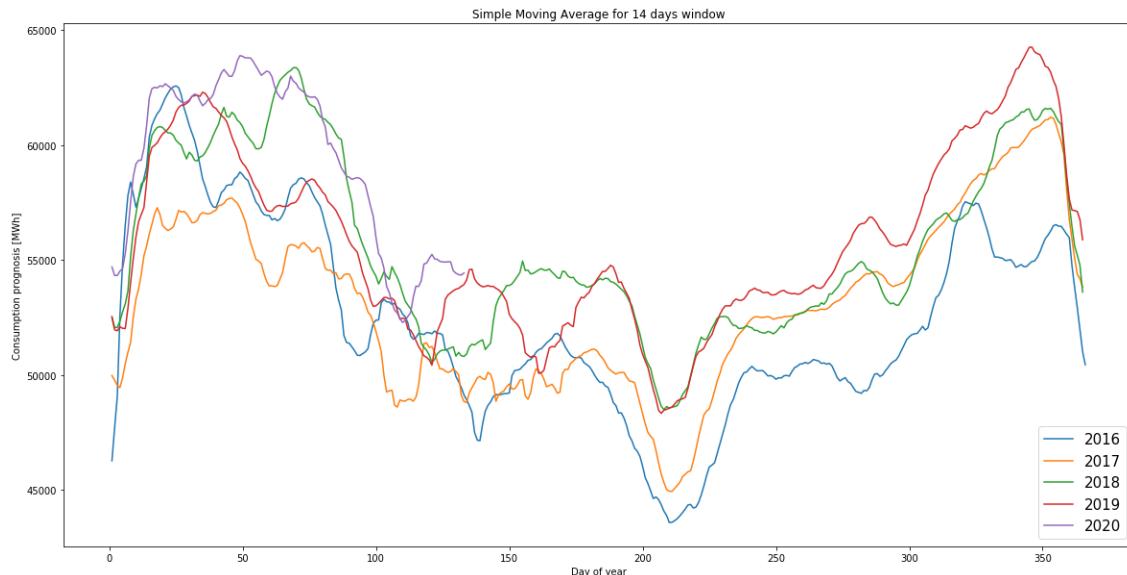


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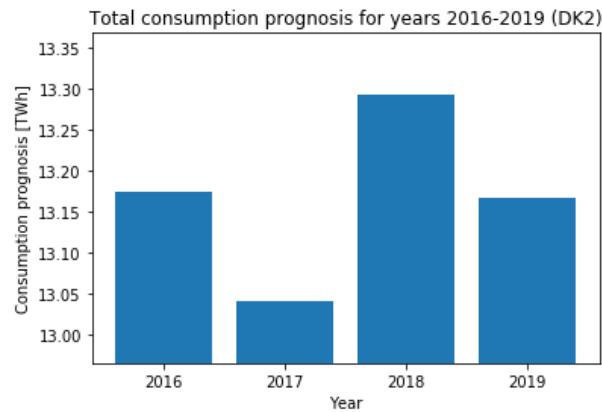


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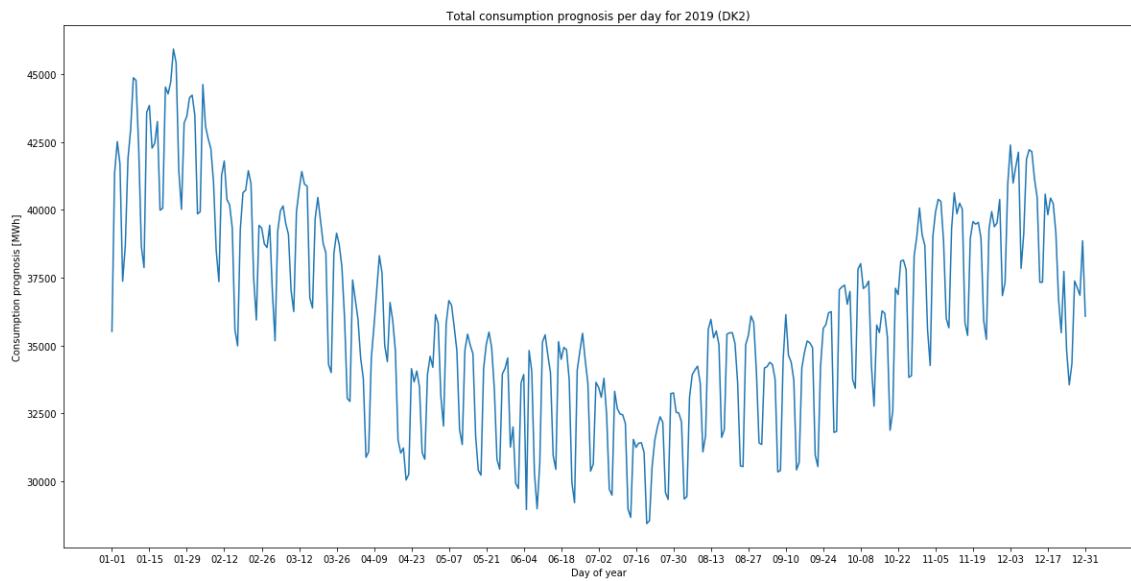


Figure 27:

Average consumption prognosis for years 2016-2020 (DK2)

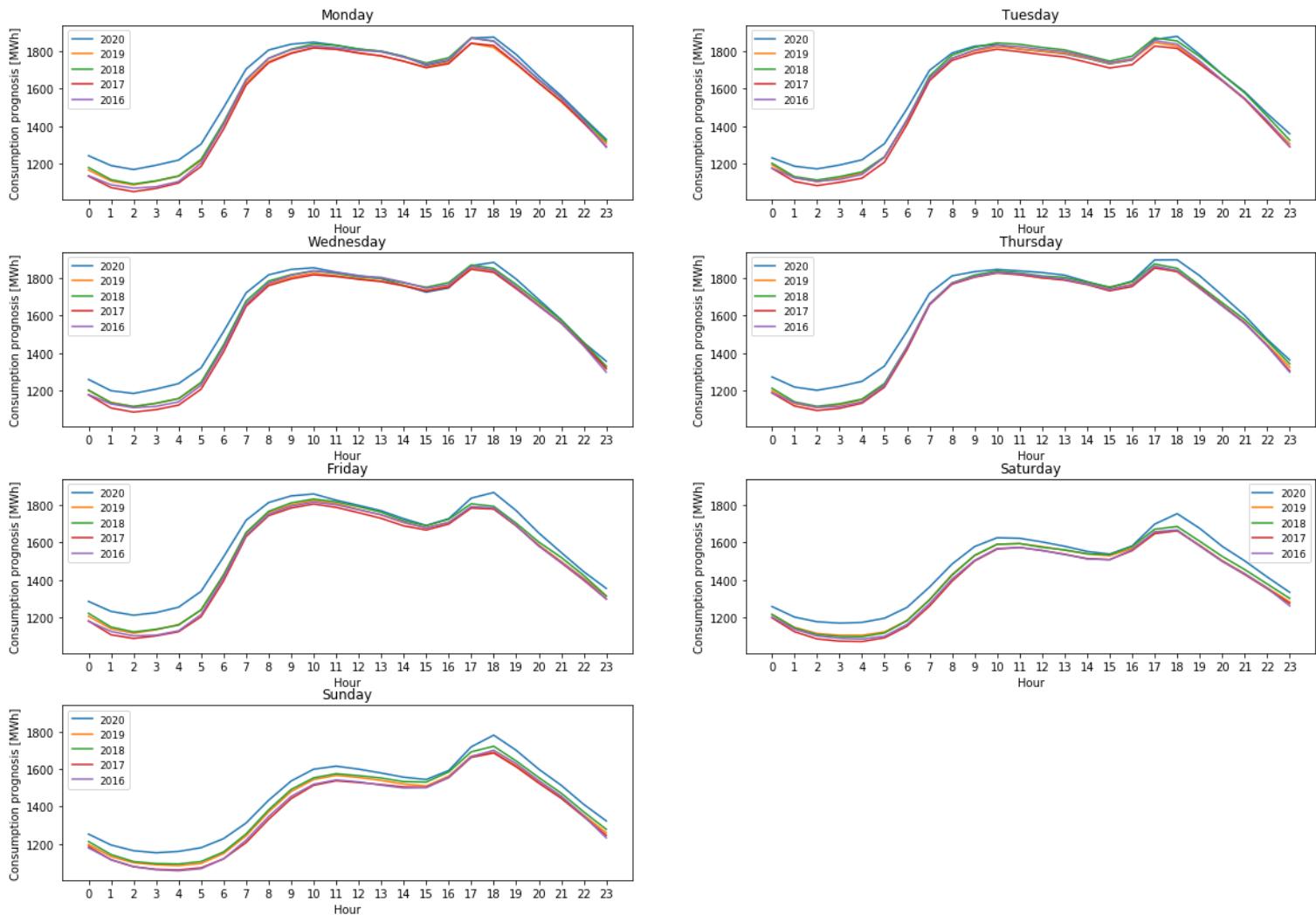


Figure 28:

Average consumption prognosis for each hour for years 2016-2020 (DK2)

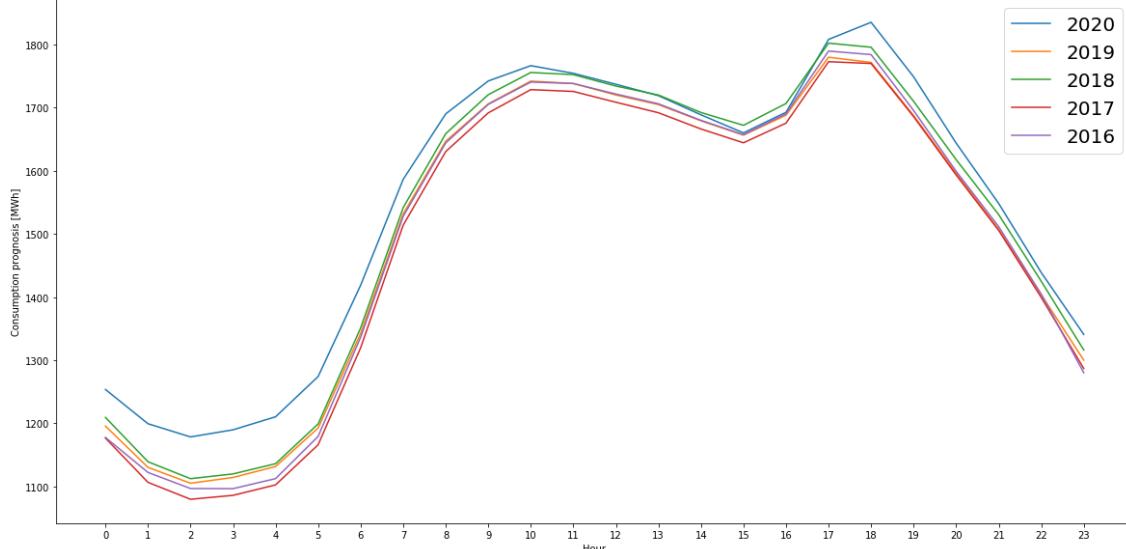


Figure 29:

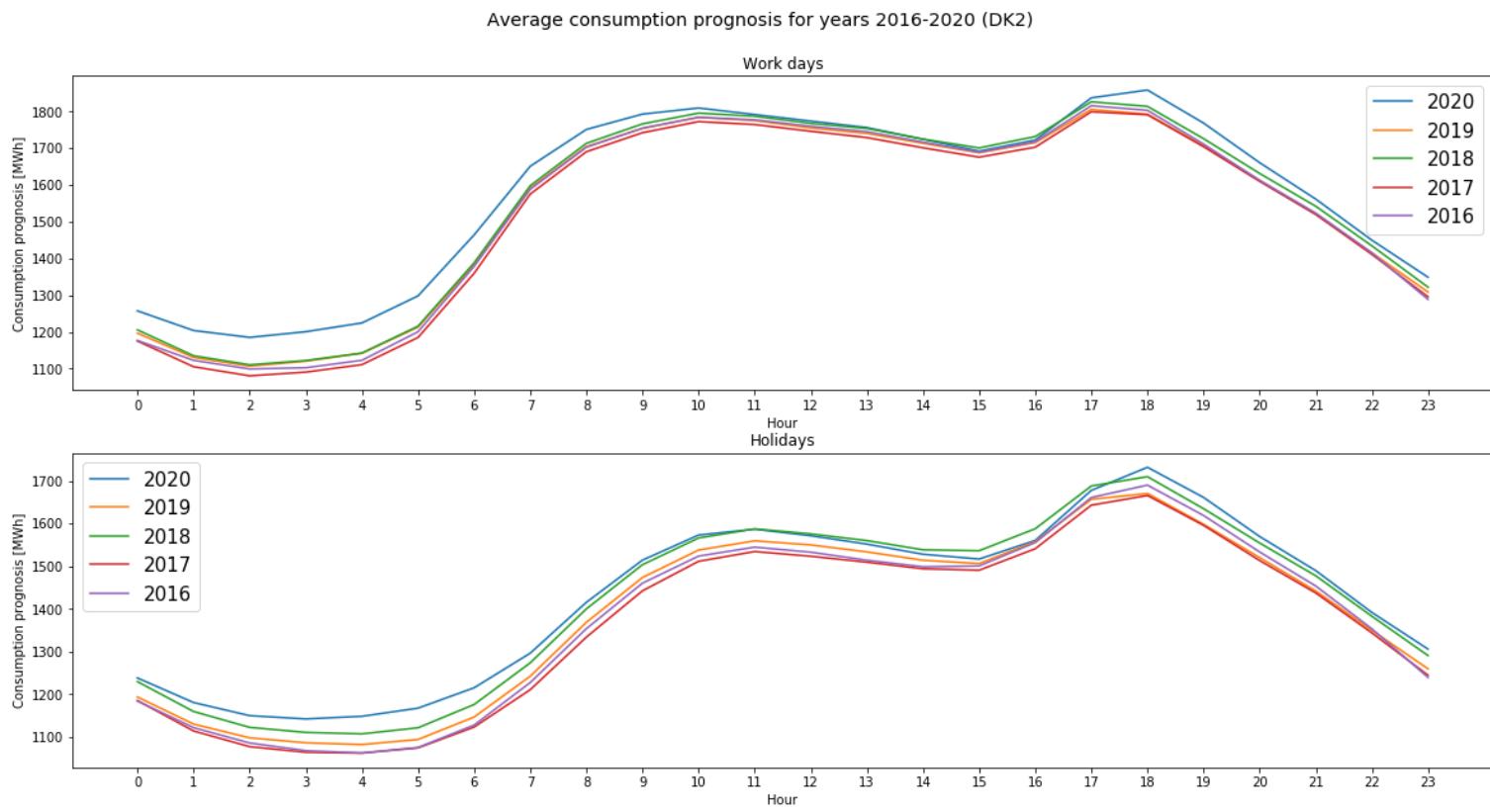


Figure 30:

Median consumption prognosis for years 2016-2020 (DK2)

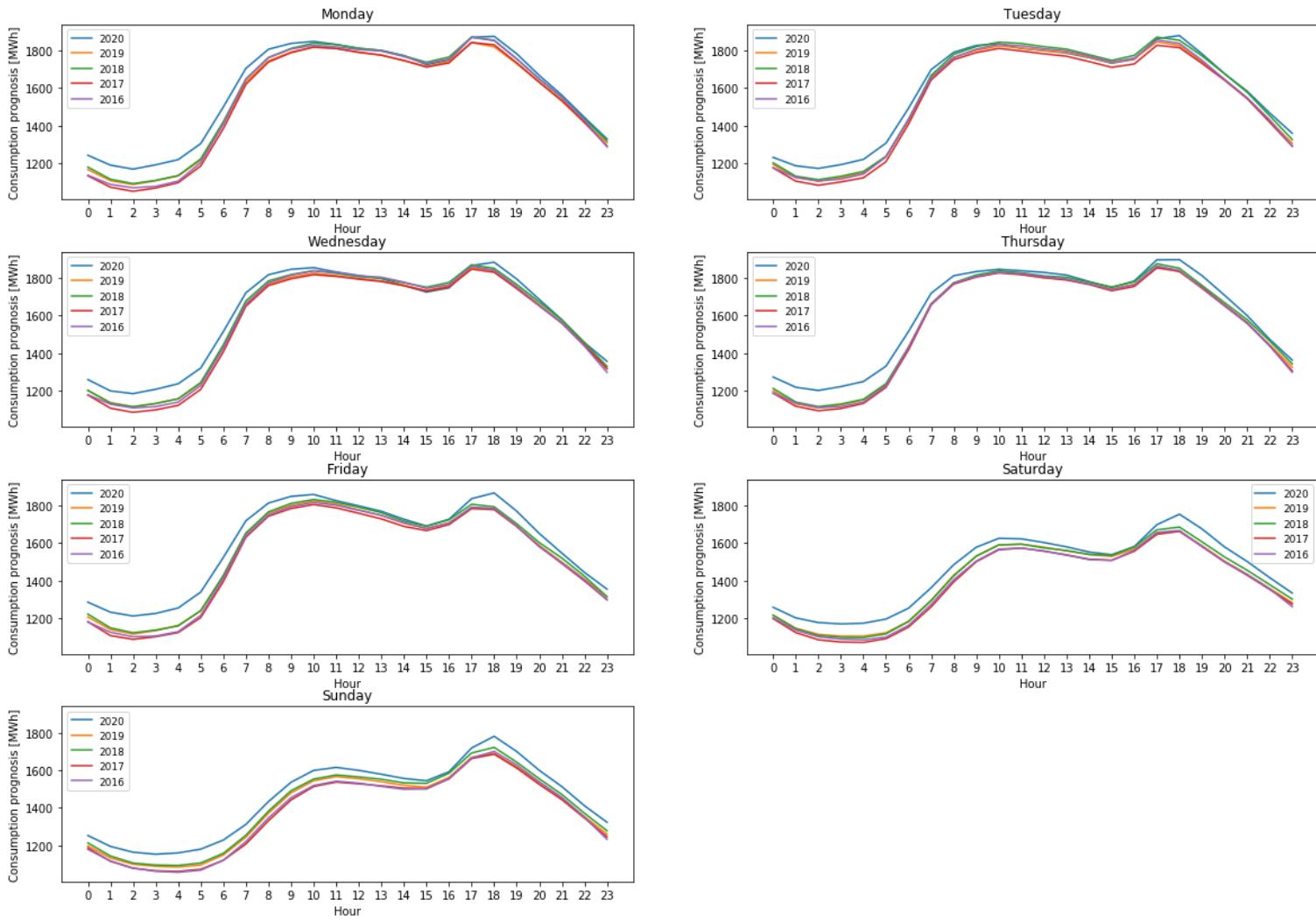


Figure 31:

Standard deviation of consumption prognosis for years 2016-2020 (DK2)

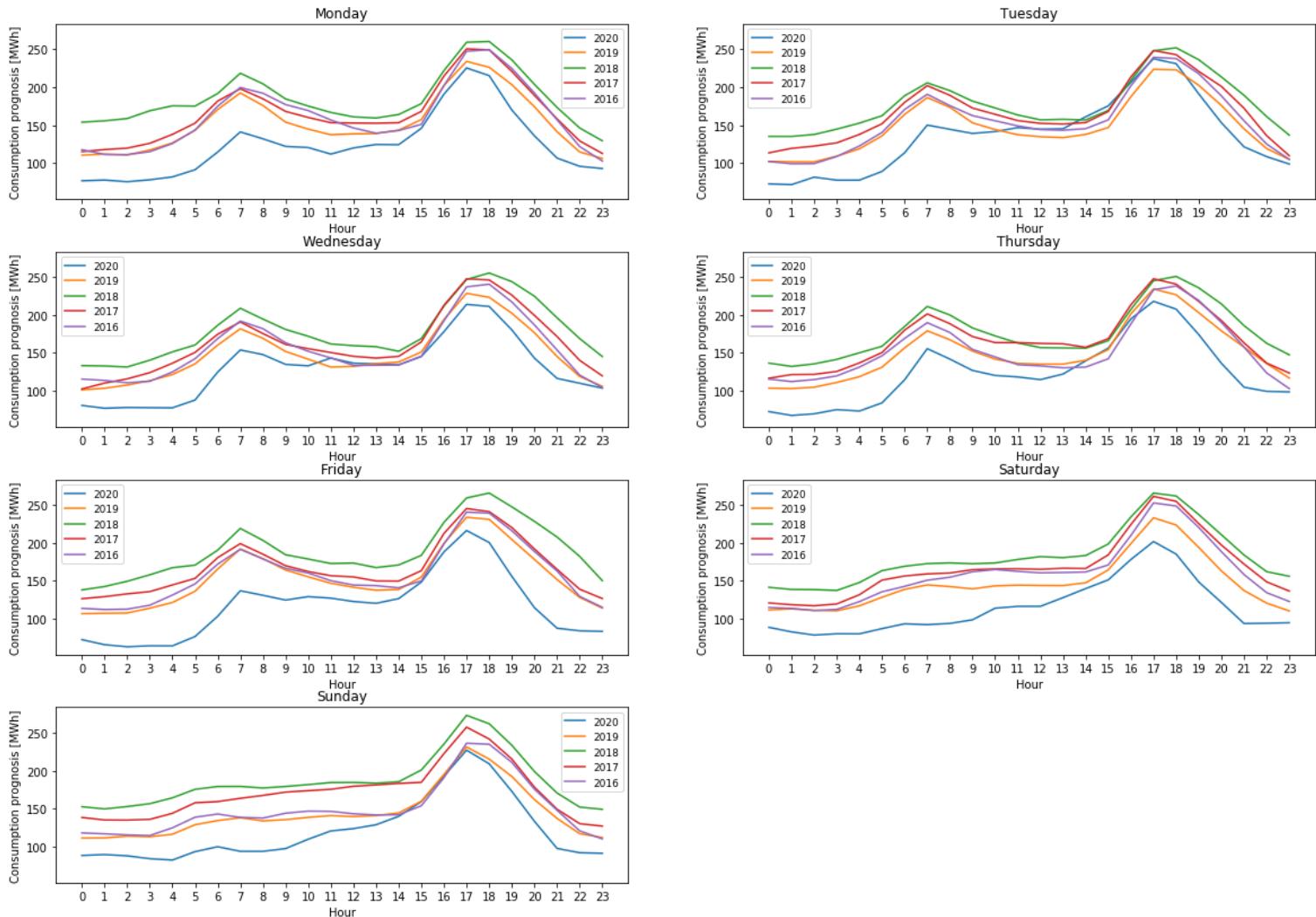


Figure 32:

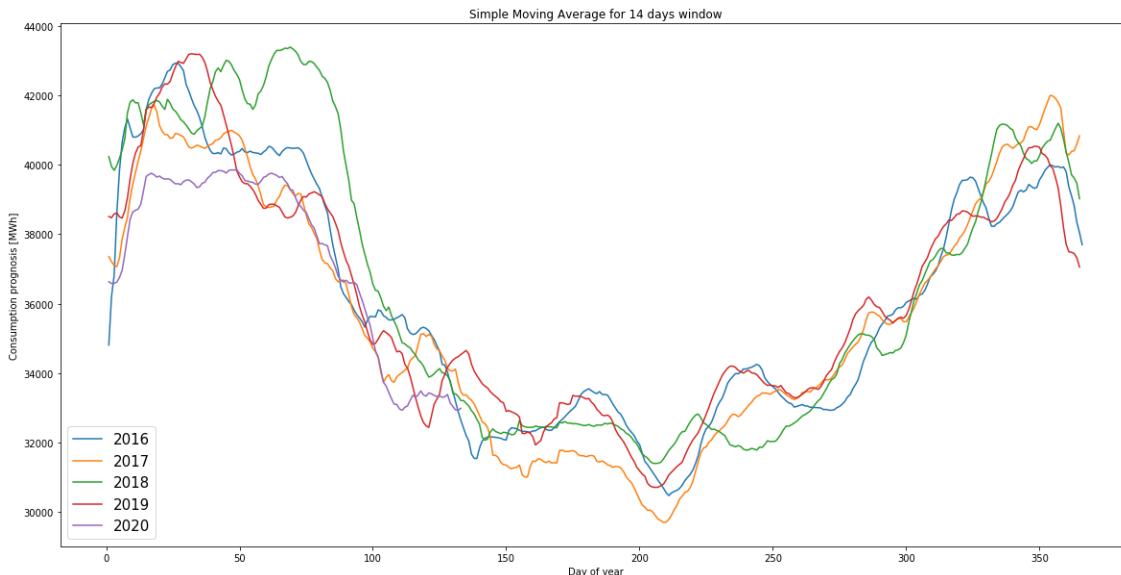


Figure 33:

3.1.2 Load data prognosis metrics

$$RMSE_t = \sqrt{\frac{1}{24} \sum_h (y_{t,h} - \hat{y}_{t,h})^2} \quad (1)$$

$$RMSE_h = \sqrt{\frac{1}{T} \sum_t (y_{t,h} - \hat{y}_{t,h})^2} \quad (2)$$

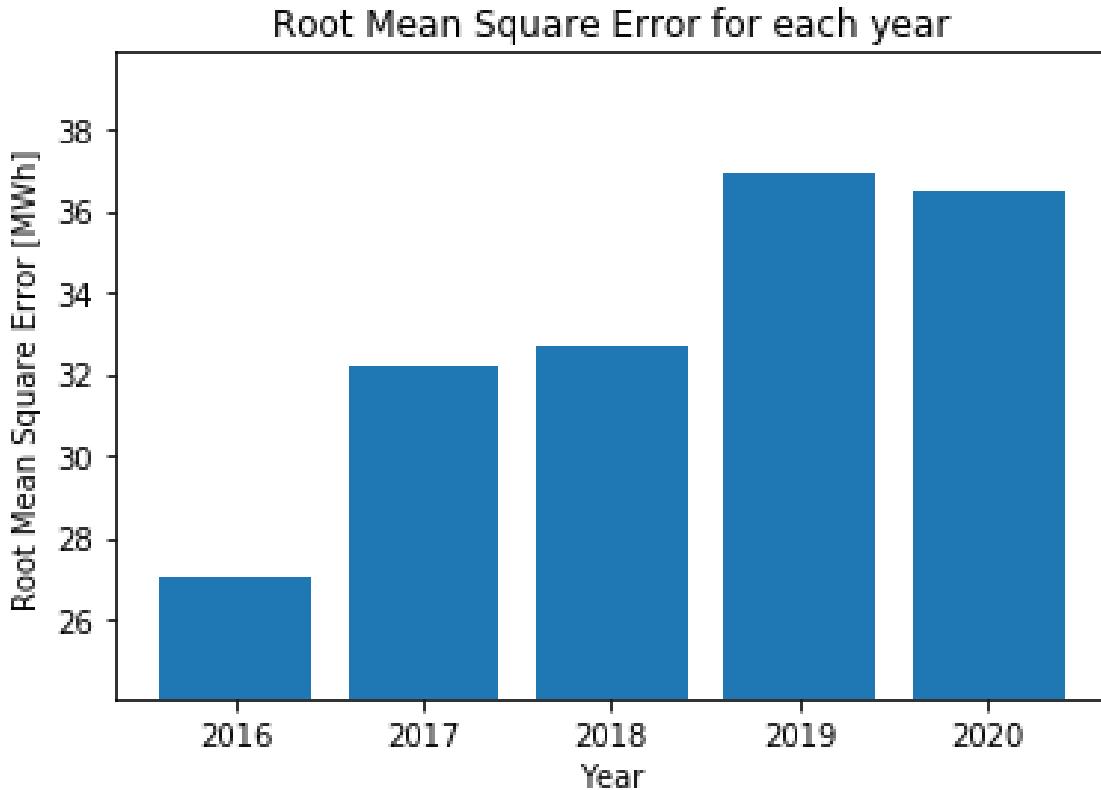


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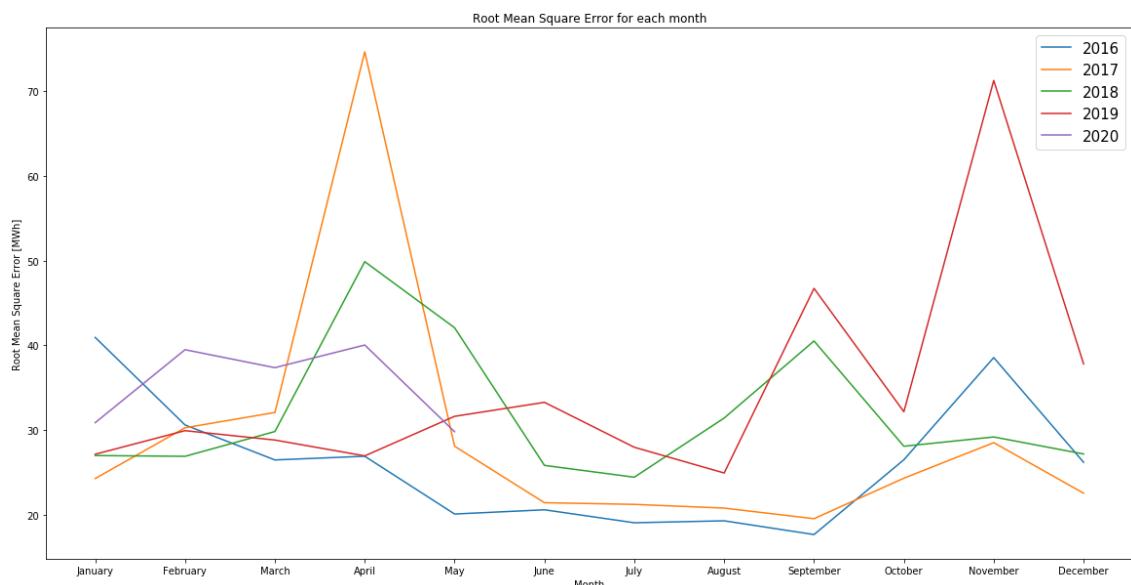


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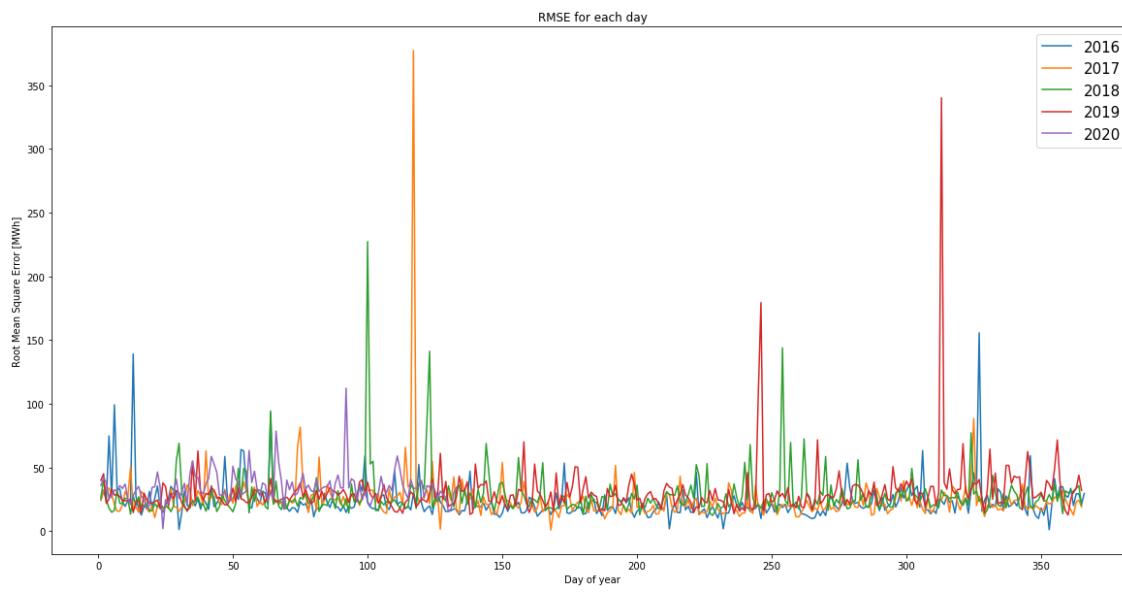


Figure 36:

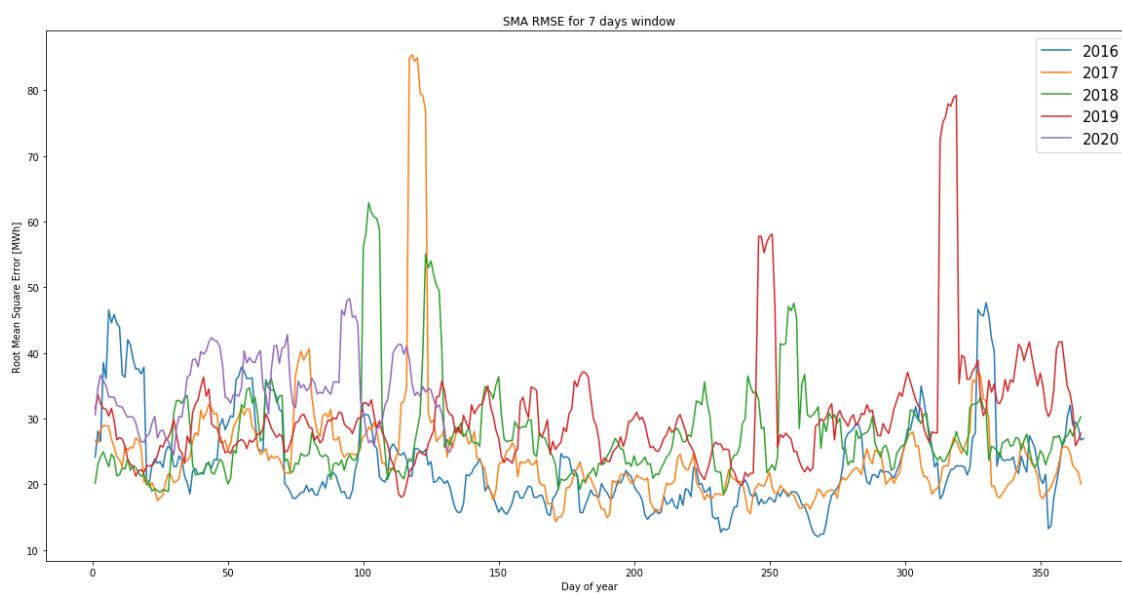


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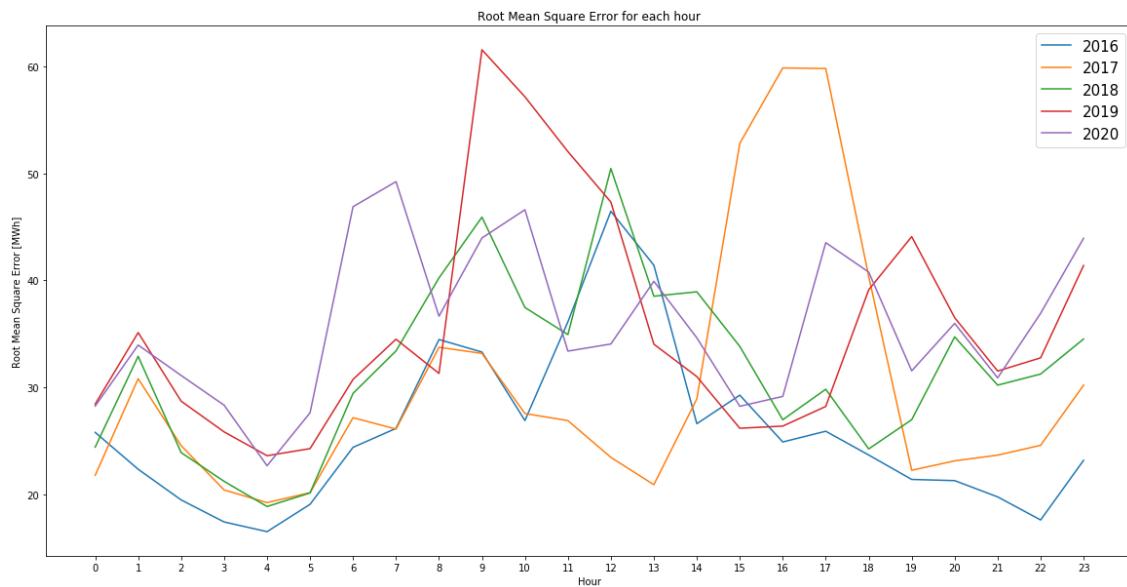


Figure 38:

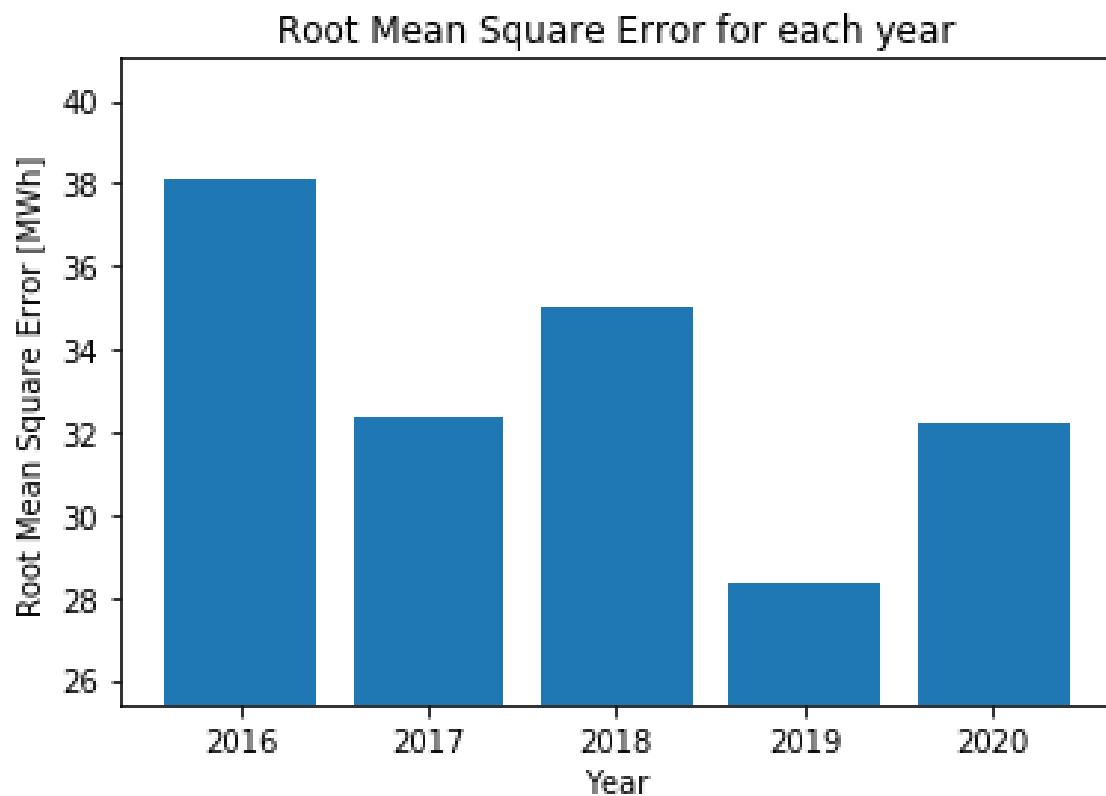


Figure 39:

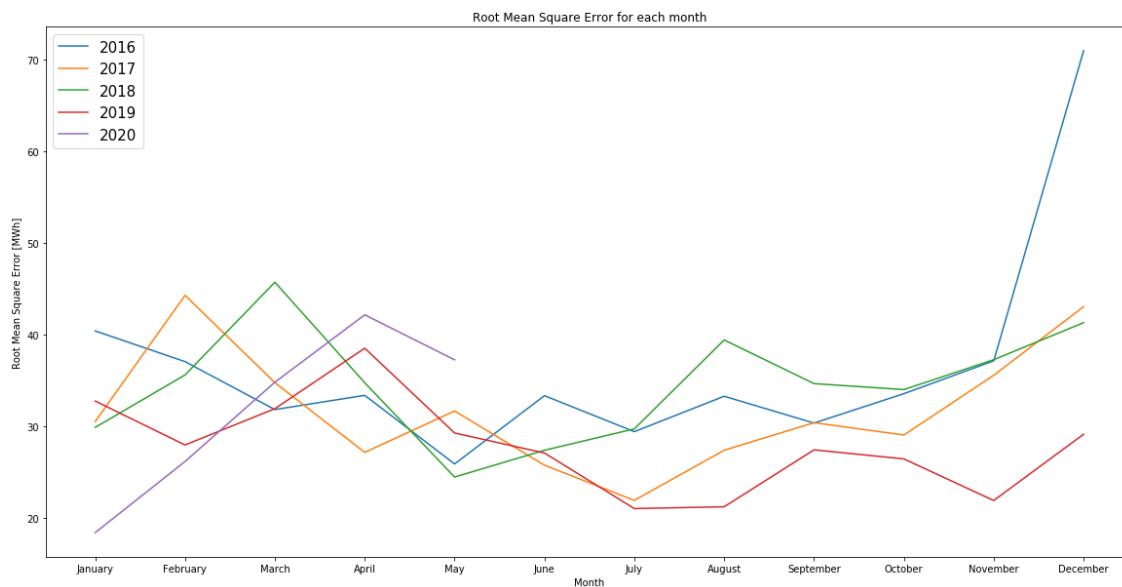


Figure 40:

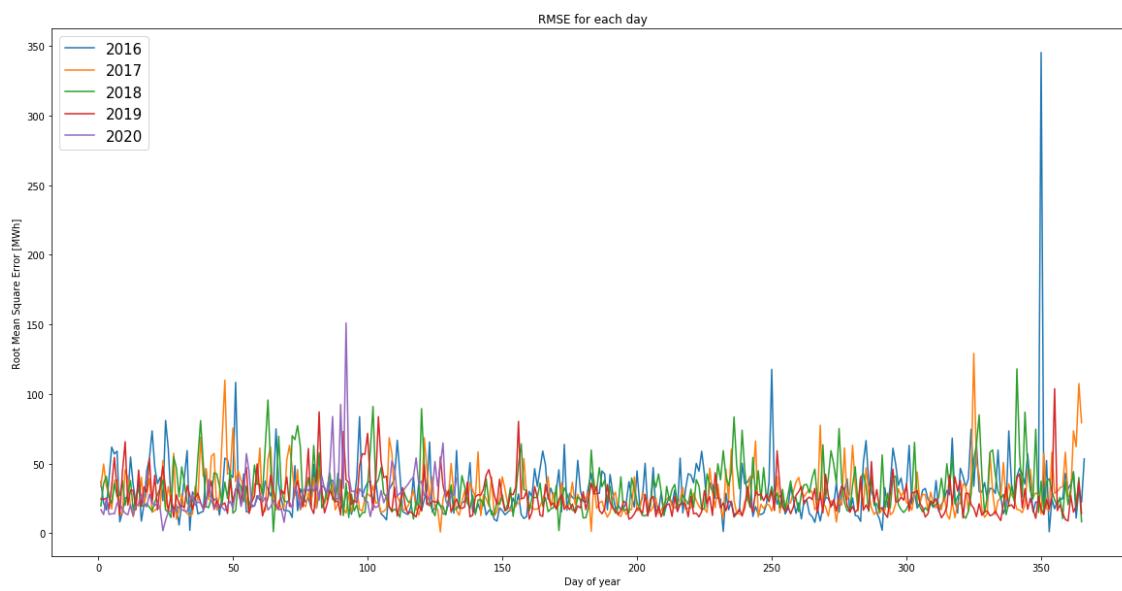


Figure 41:

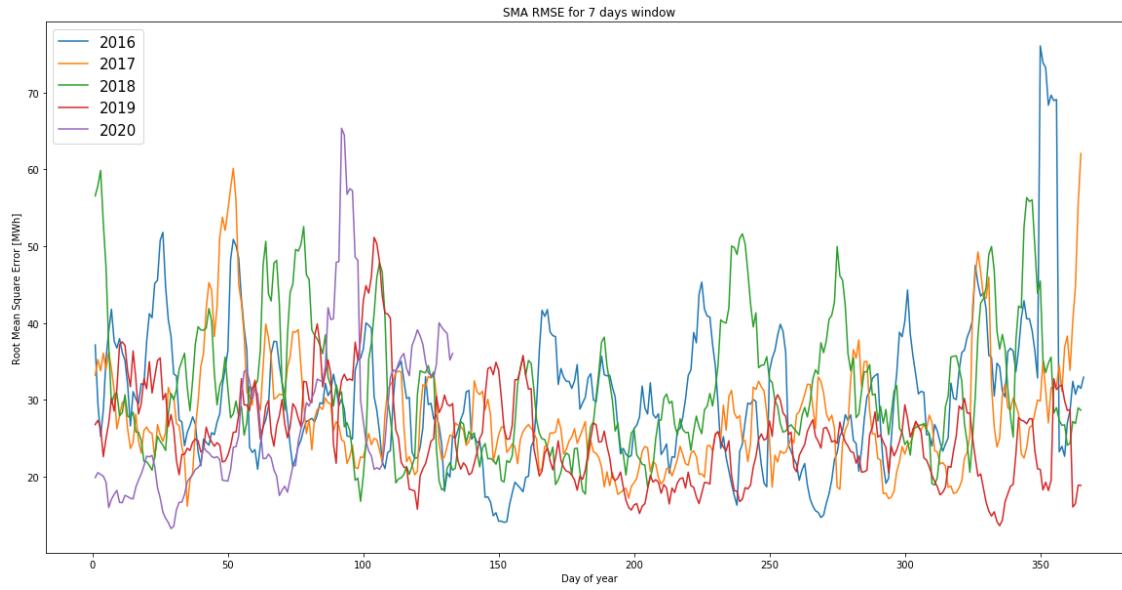


Figure 42:

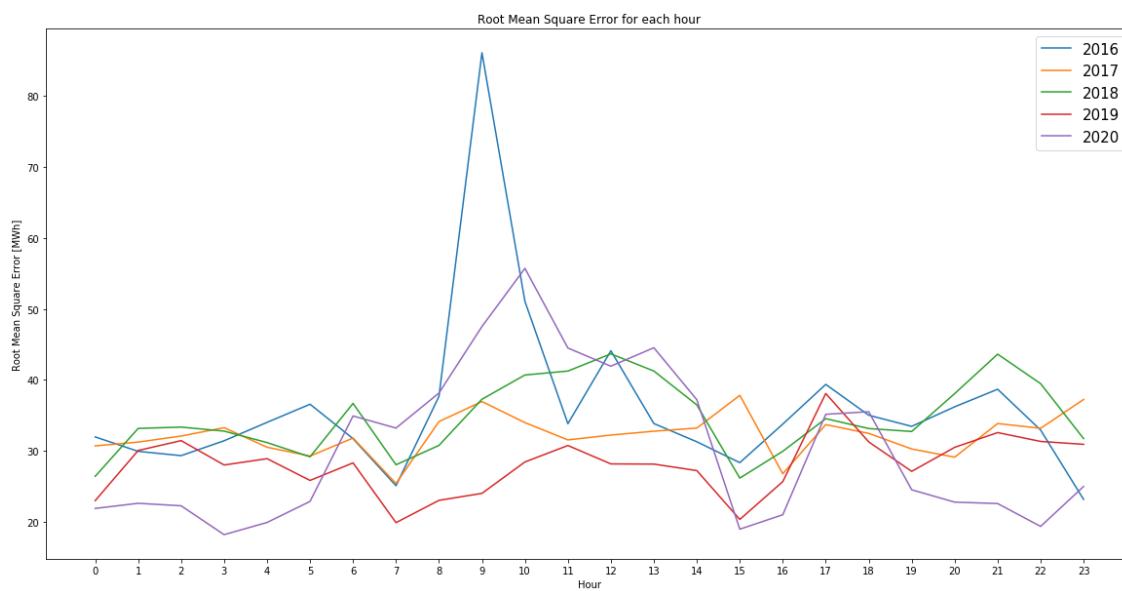


Figure 43:

3.2 Wind power data

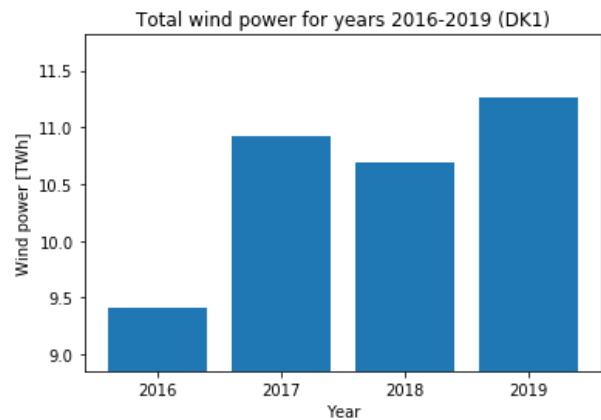


Figure 44:

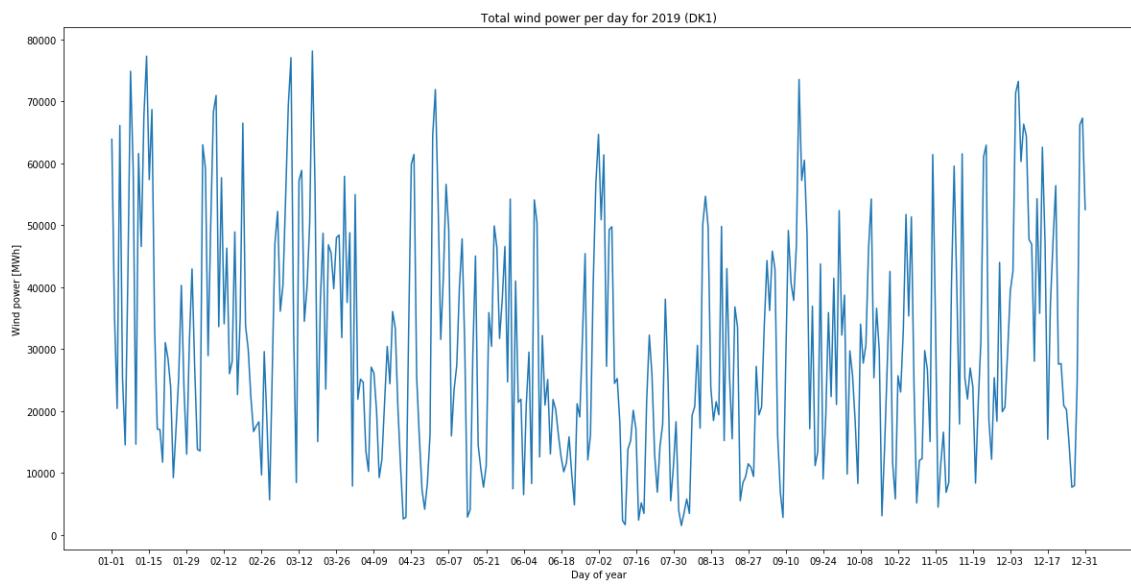


Figure 45:

Average wind power for years 2016-2020 (DK1)

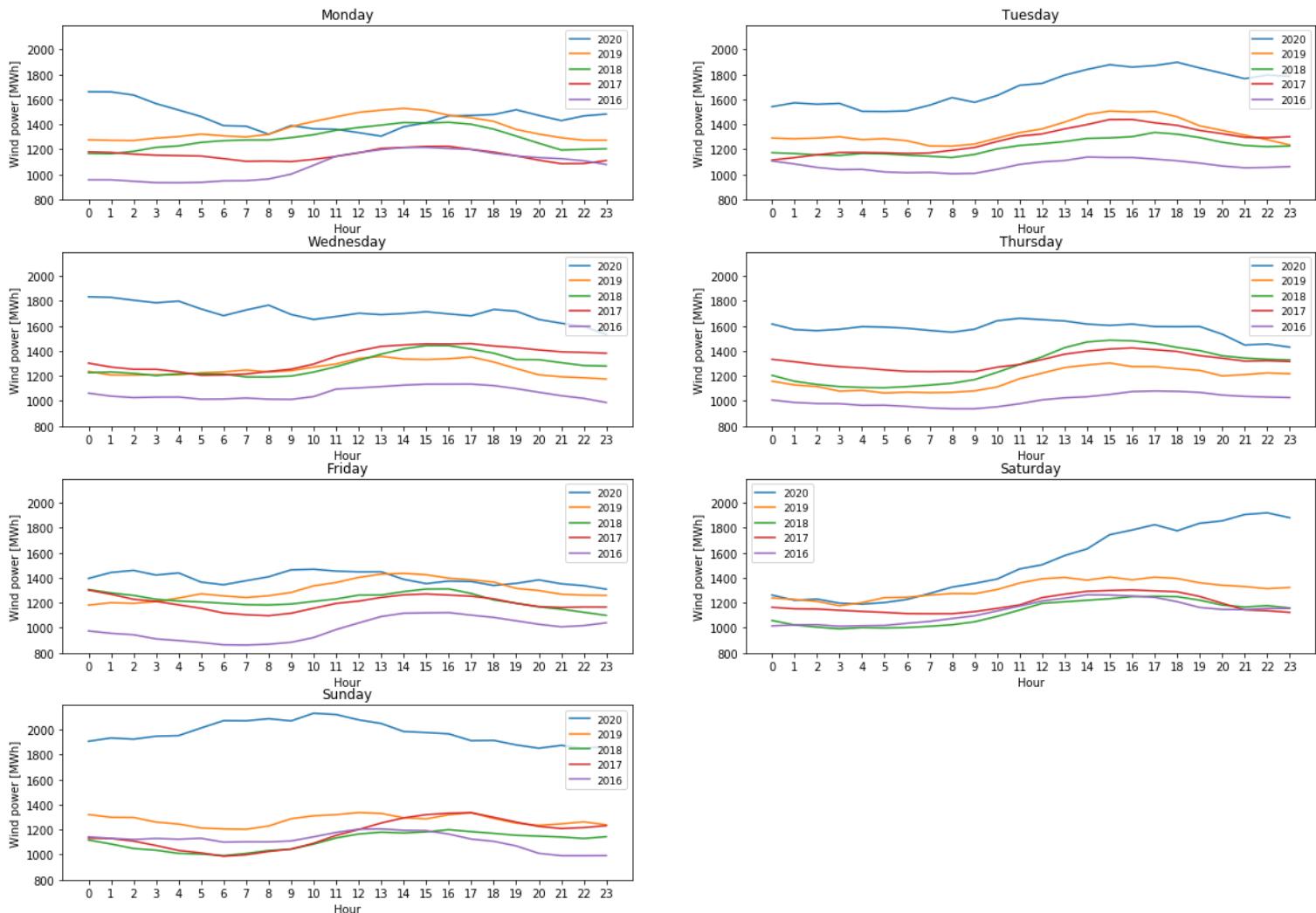


Figure 46:

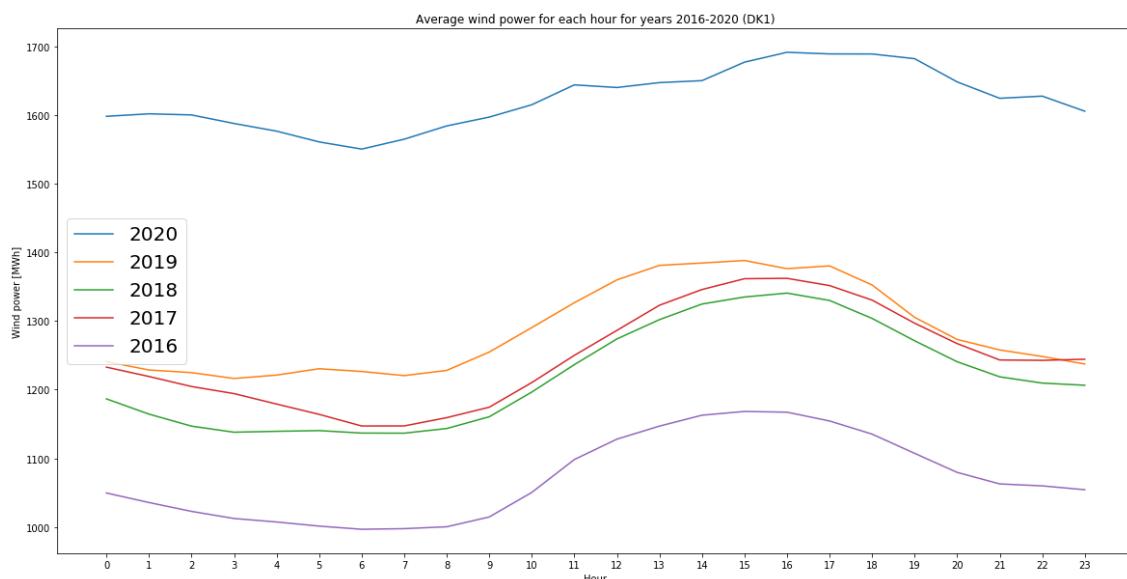


Figure 47:

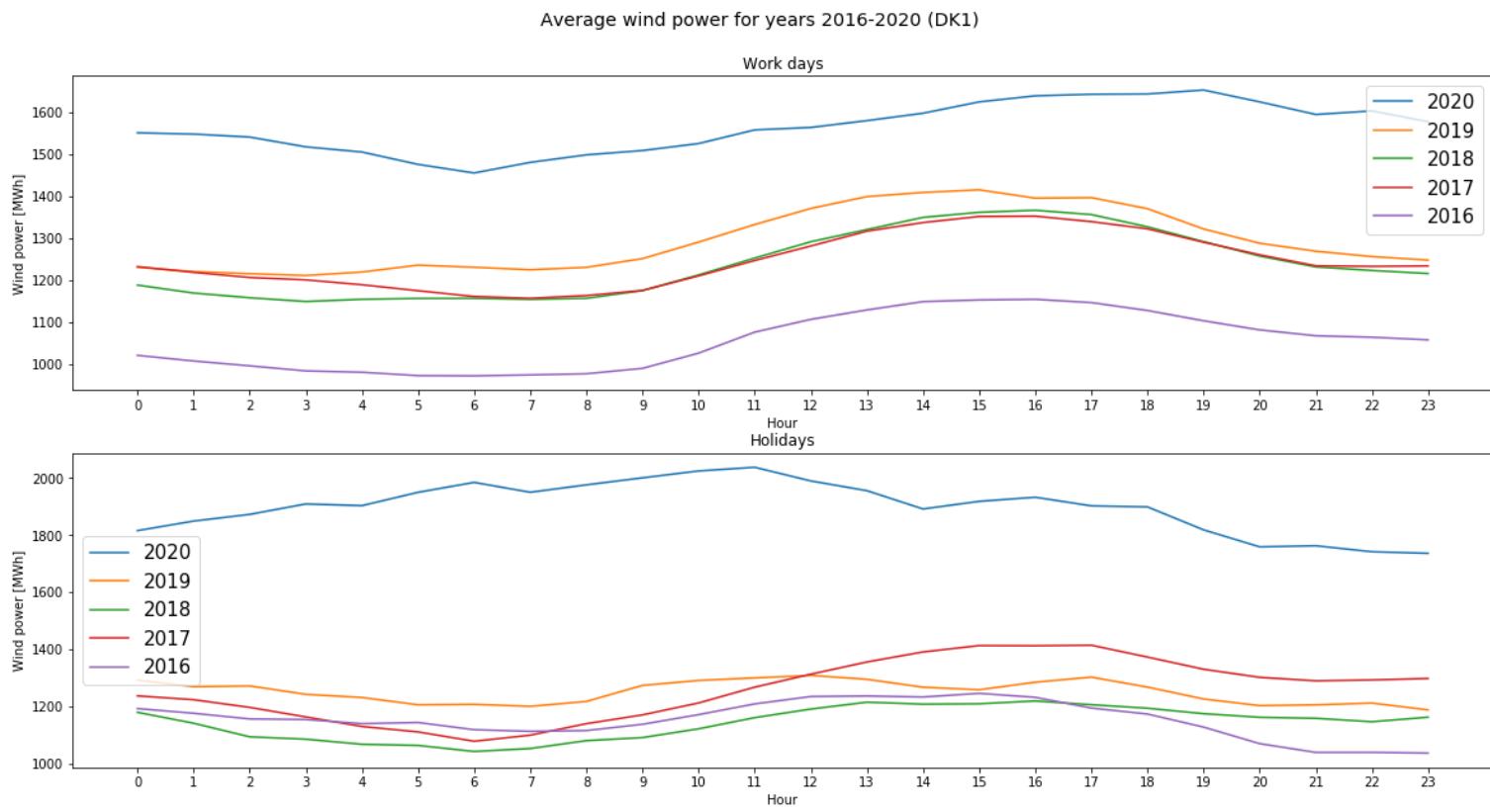


Figure 48:

Median wind power for years 2016-2020 (DK1)

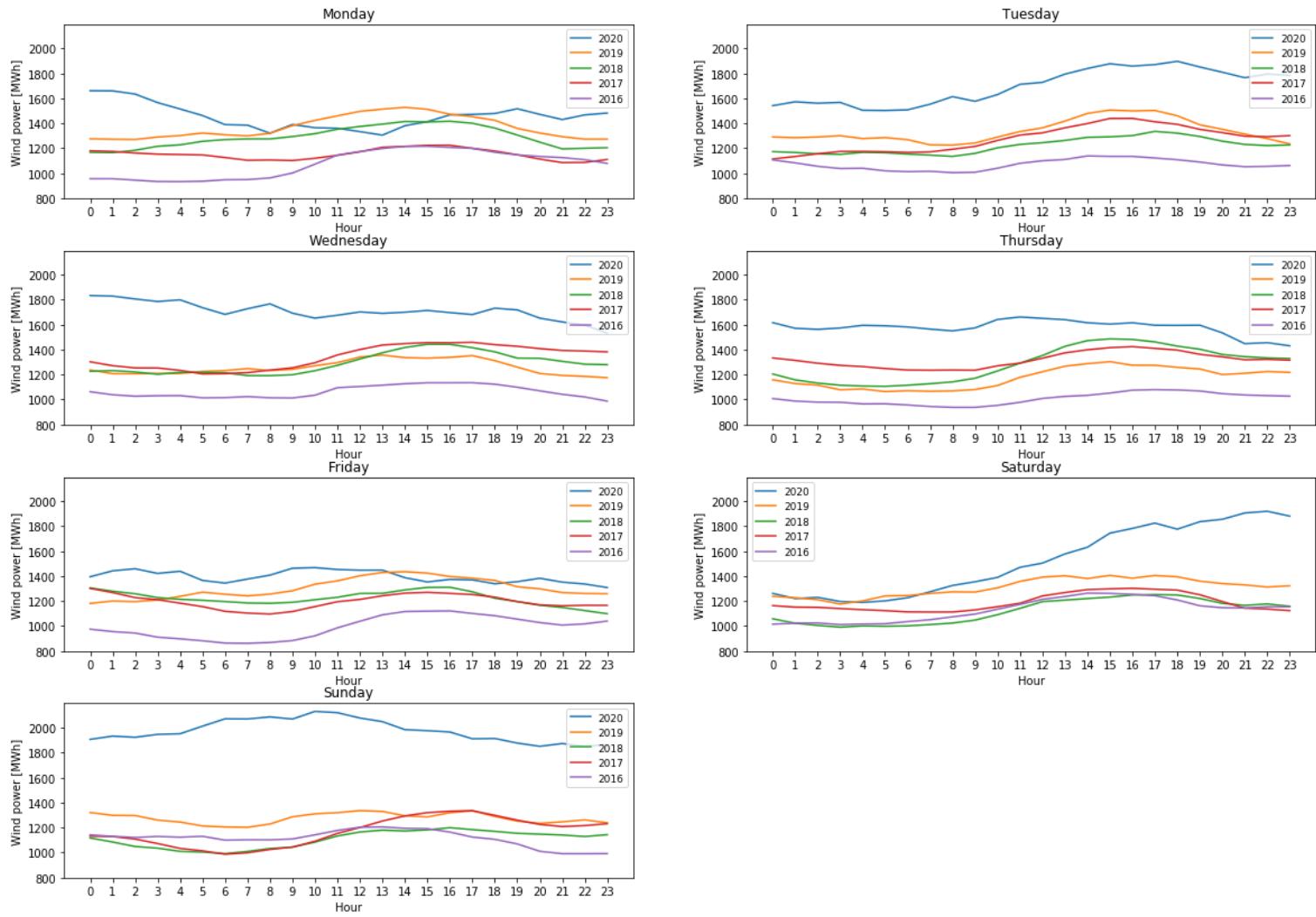


Figure 49:

Standard deviation of wind power for years 2016-2020 (DK1)

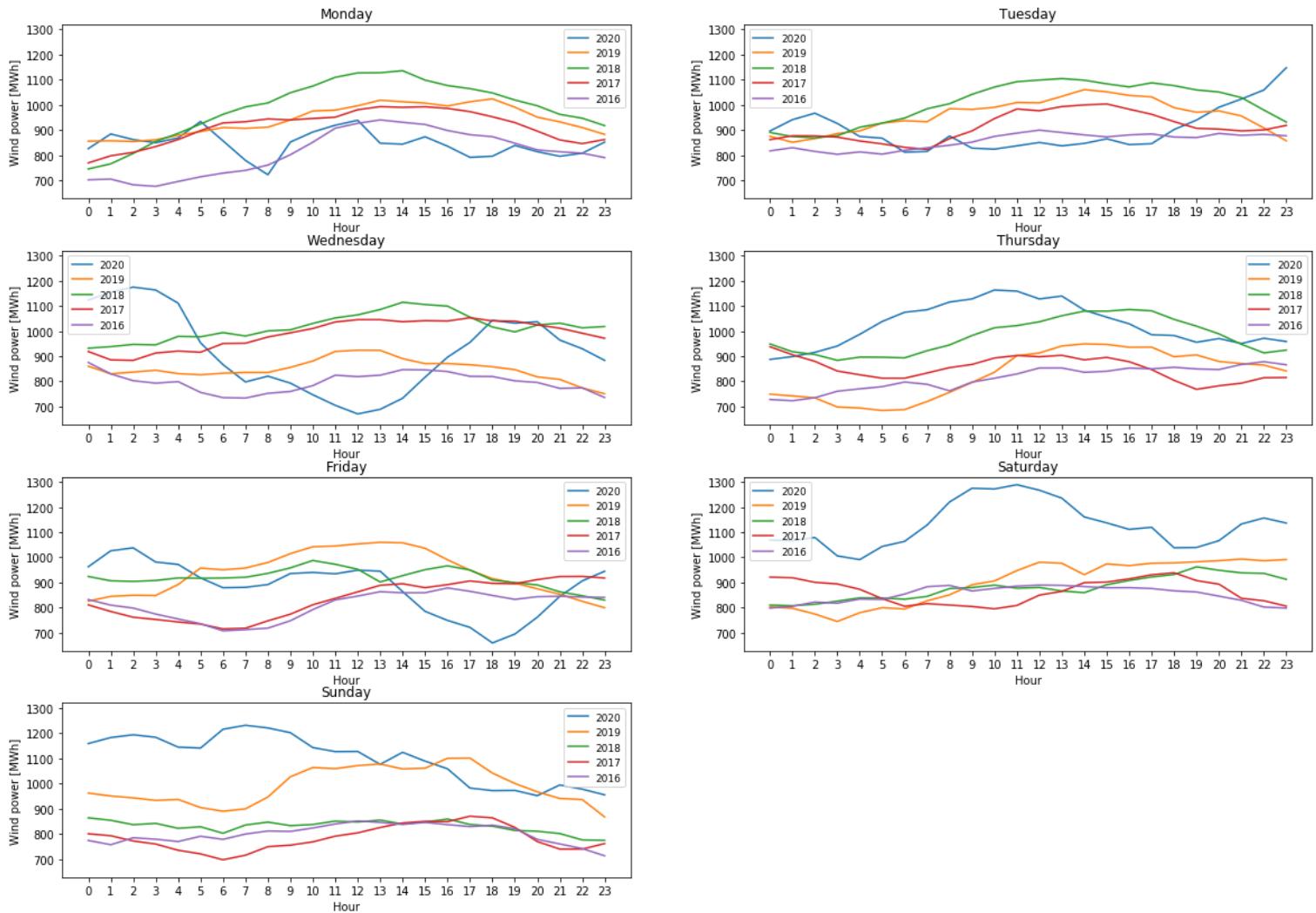


Figure 50:

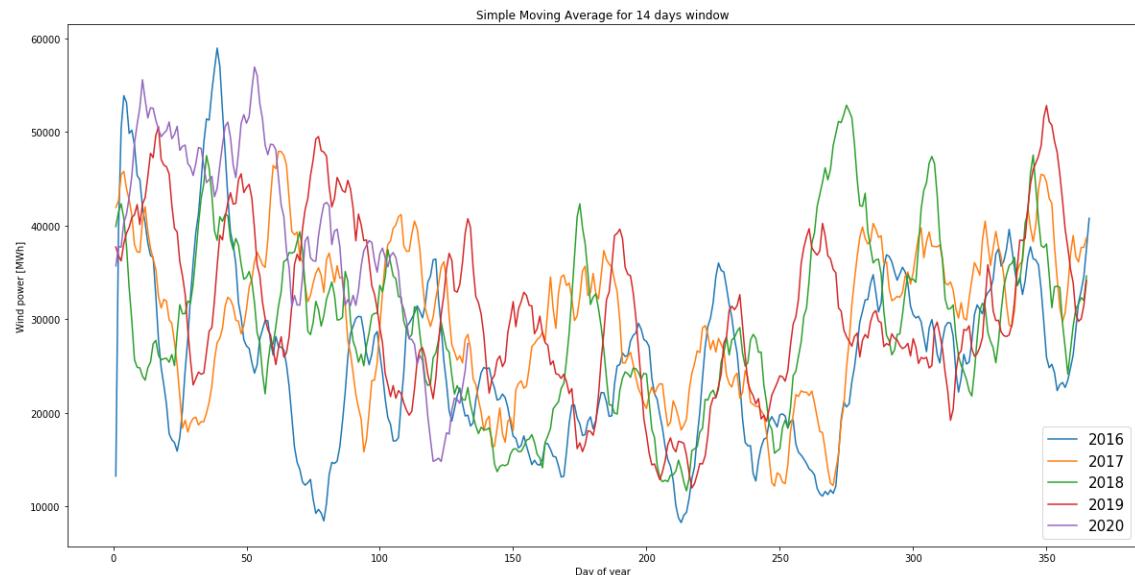


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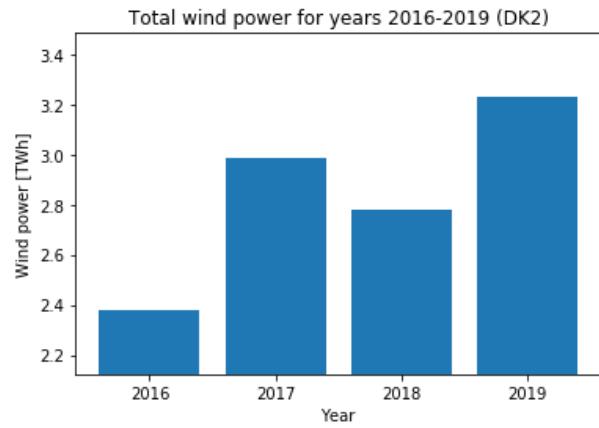


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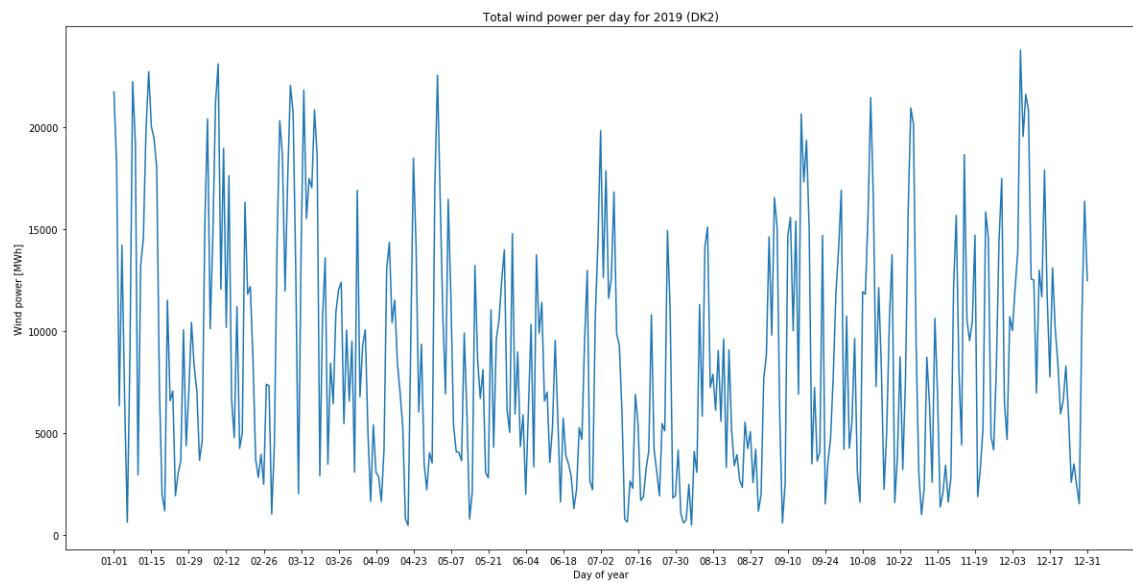


Figure 53:

Average wind power for years 2016-2020 (DK2)

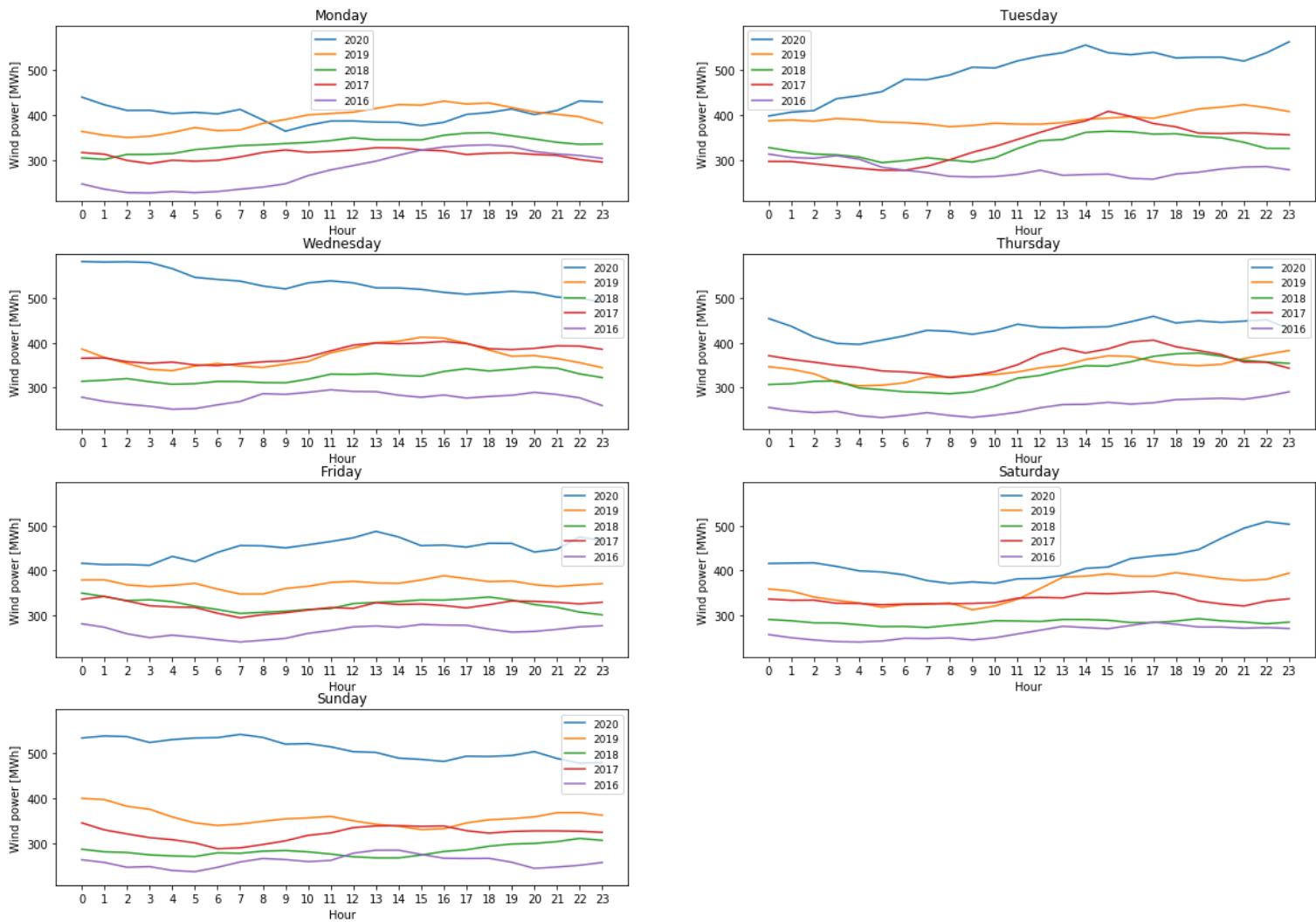


Figure 54:

Average wind power for each hour for years 2016-2020 (DK2)

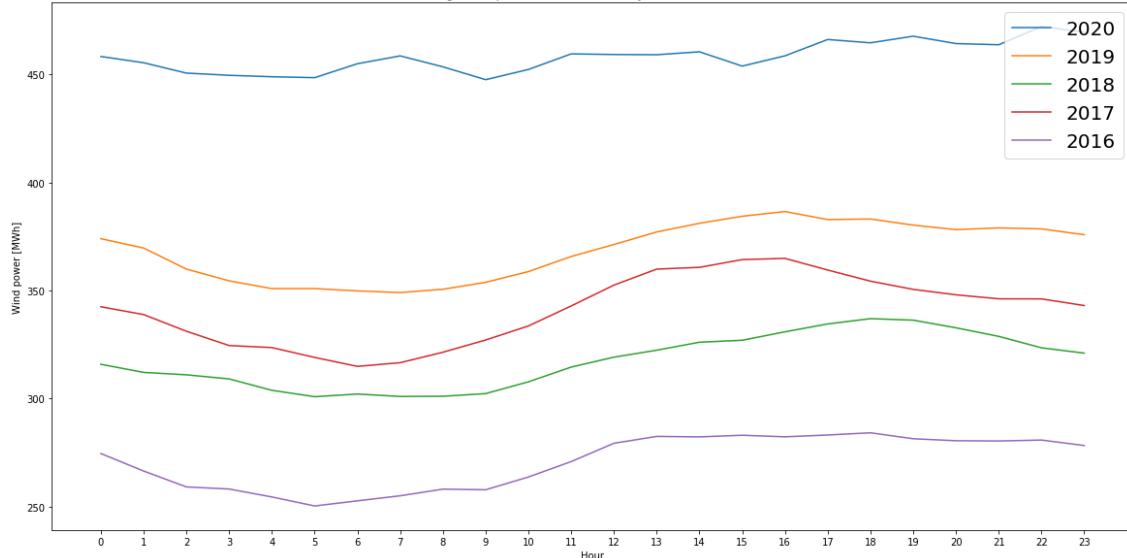


Figure 55:

Average wind power for years 2016-2020 (DK2)

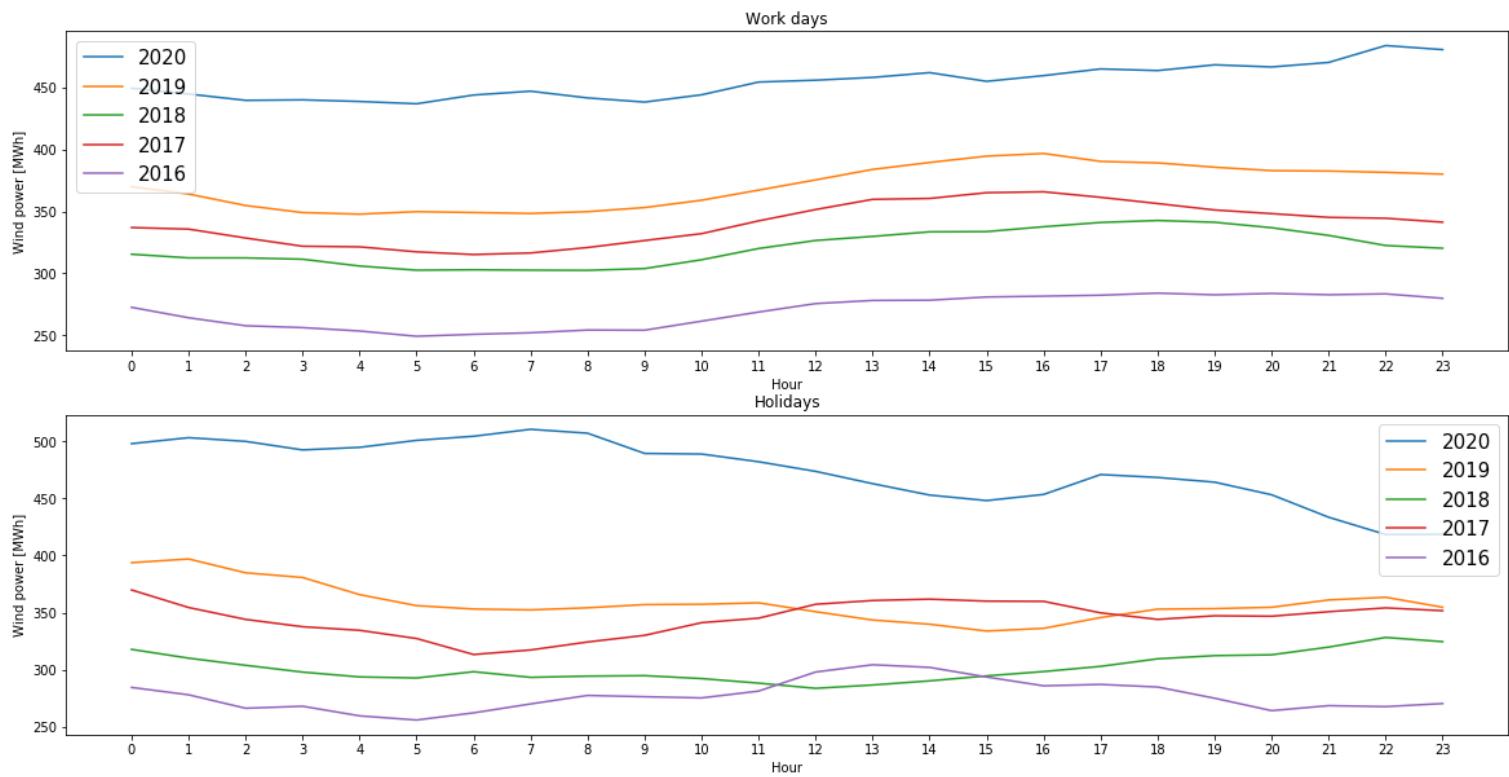


Figure 56:

Median wind power for years 2016-2020 (DK2)

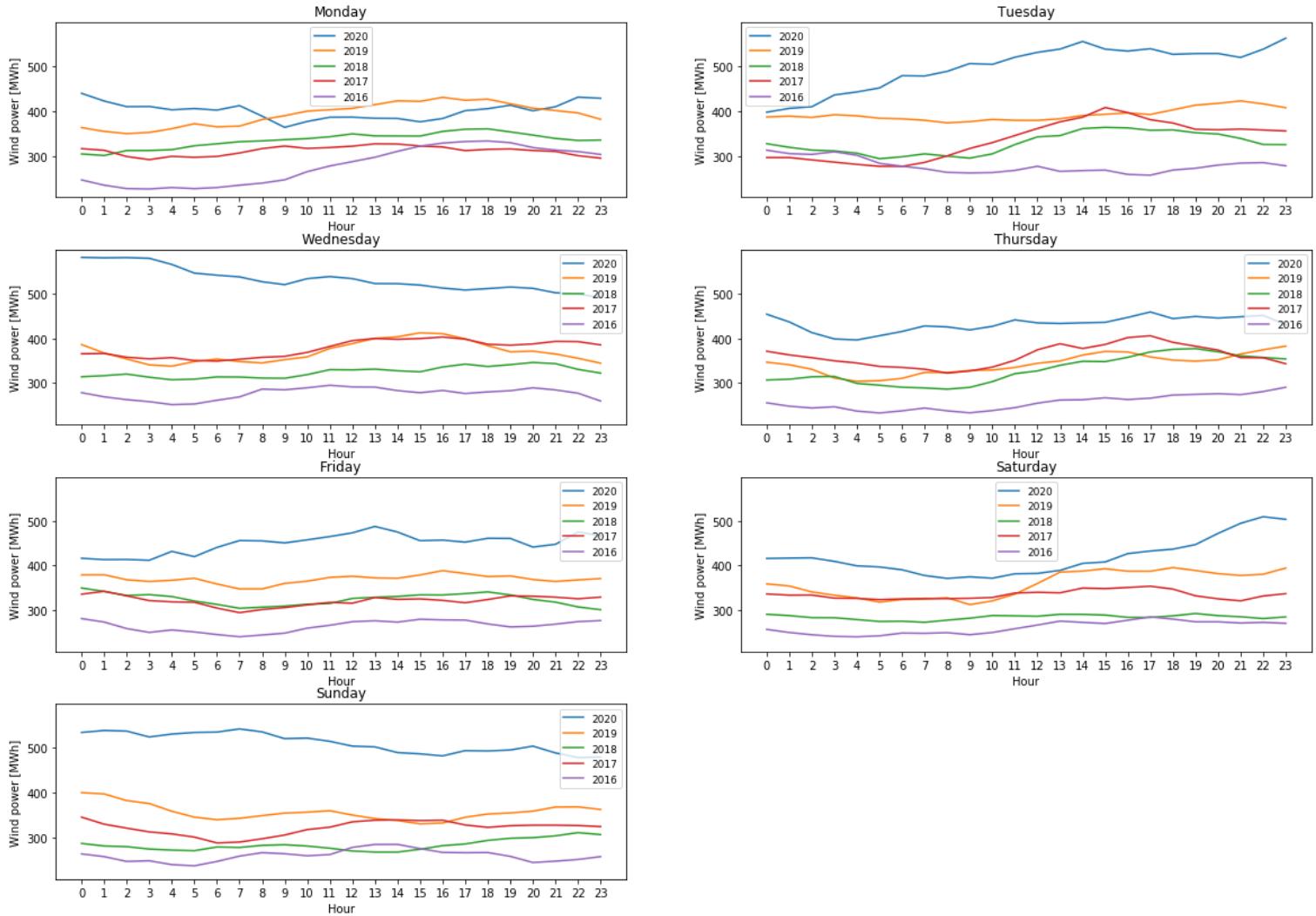


Figure 57:

Standard deviation of wind power for years 2016-2020 (DK2)

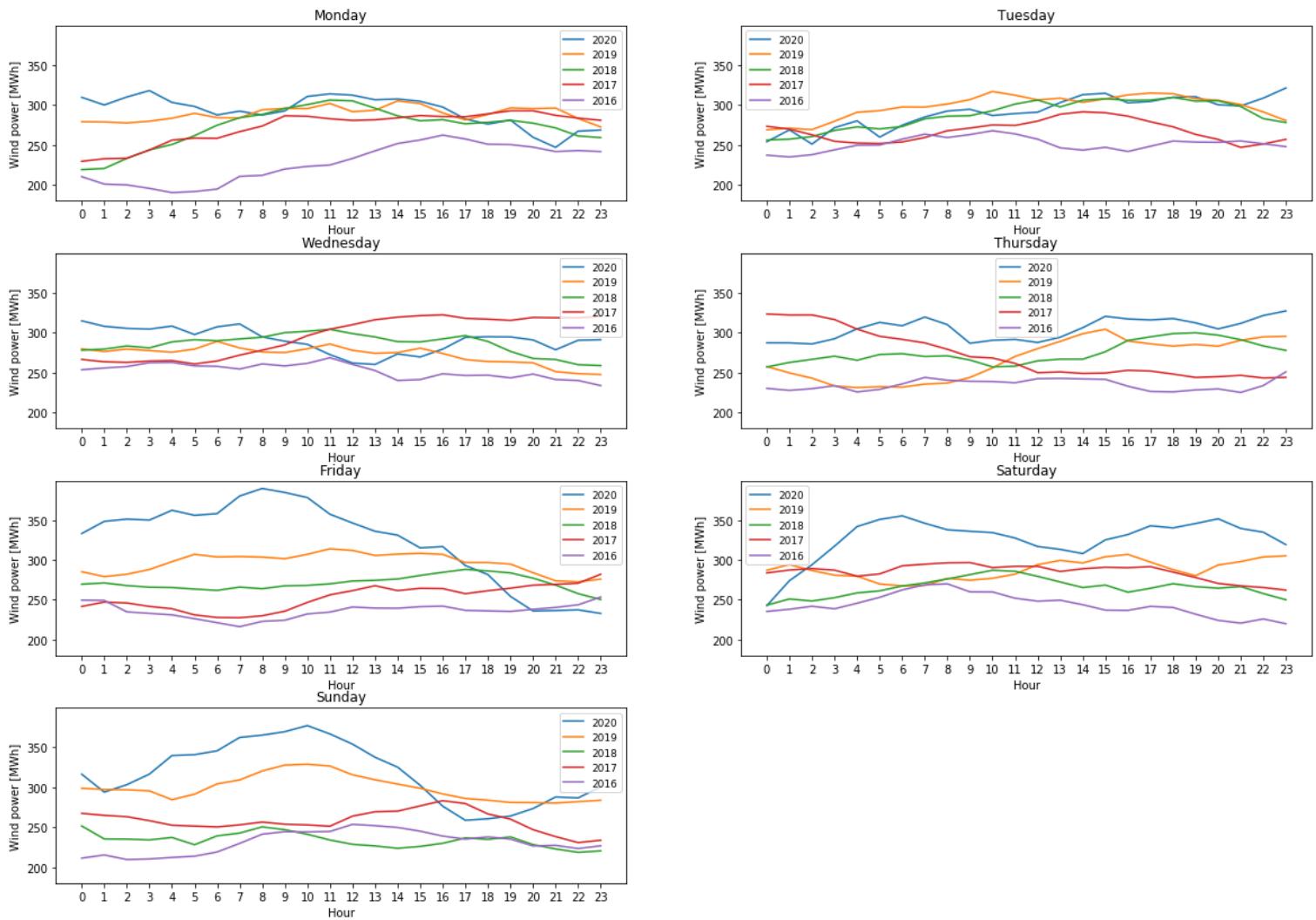


Figure 58:

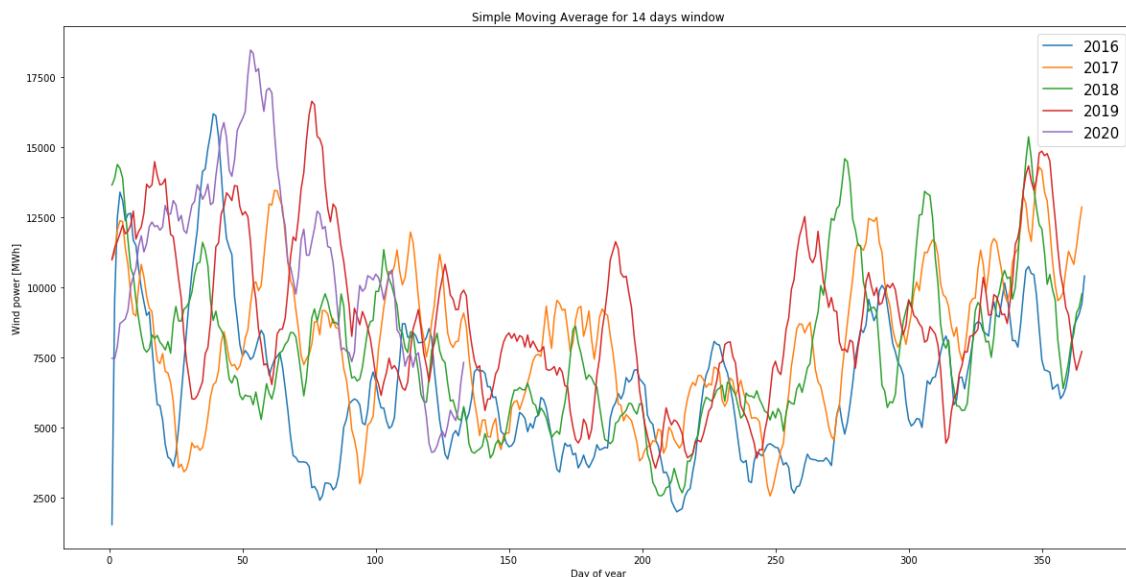


Figure 59:

3.2.1 Wind power data prognosis

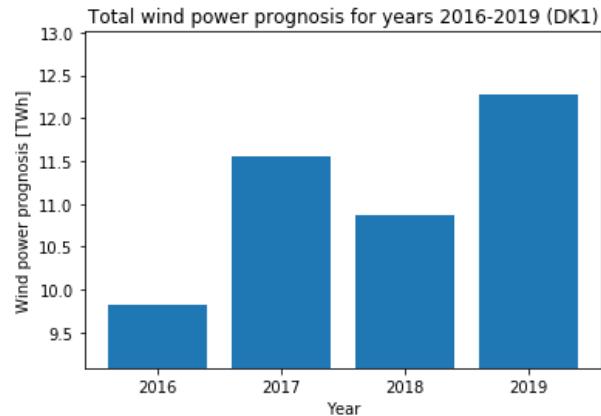


Figure 60:

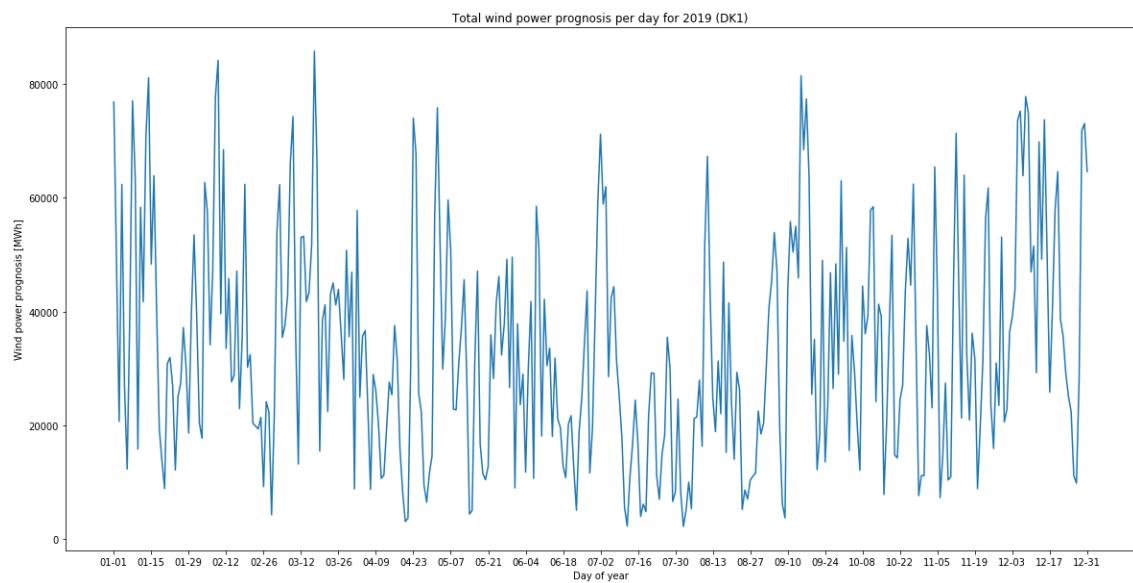


Figure 61:

Average wind power prognosis for years 2016-2020 (DK1)

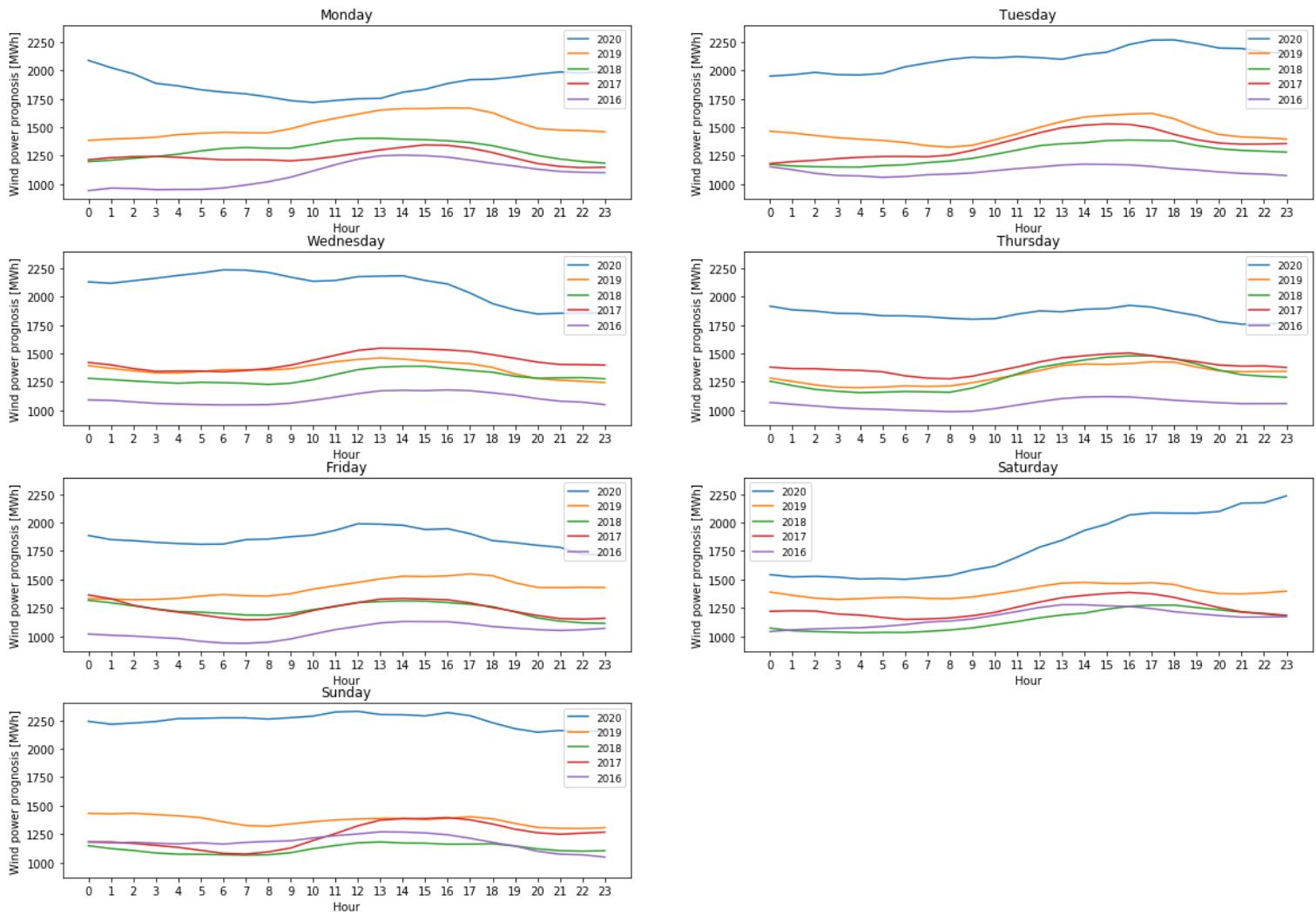


Figure 62:

Average wind power prognosis for each hour for years 2016-2020 (DK1)

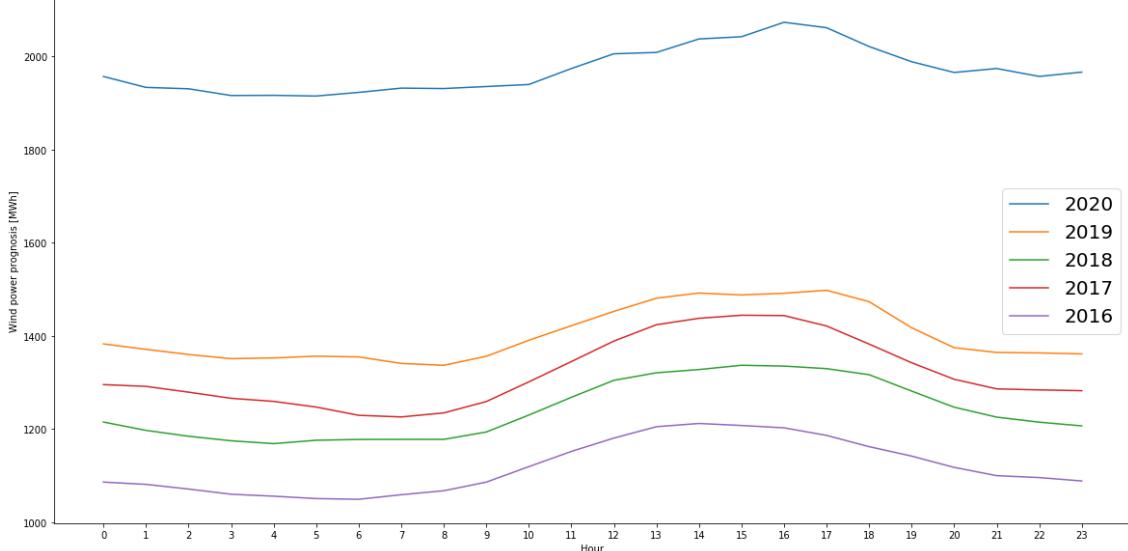


Figure 63:

Average wind power prognosis for years 2016-2020 (DK1)

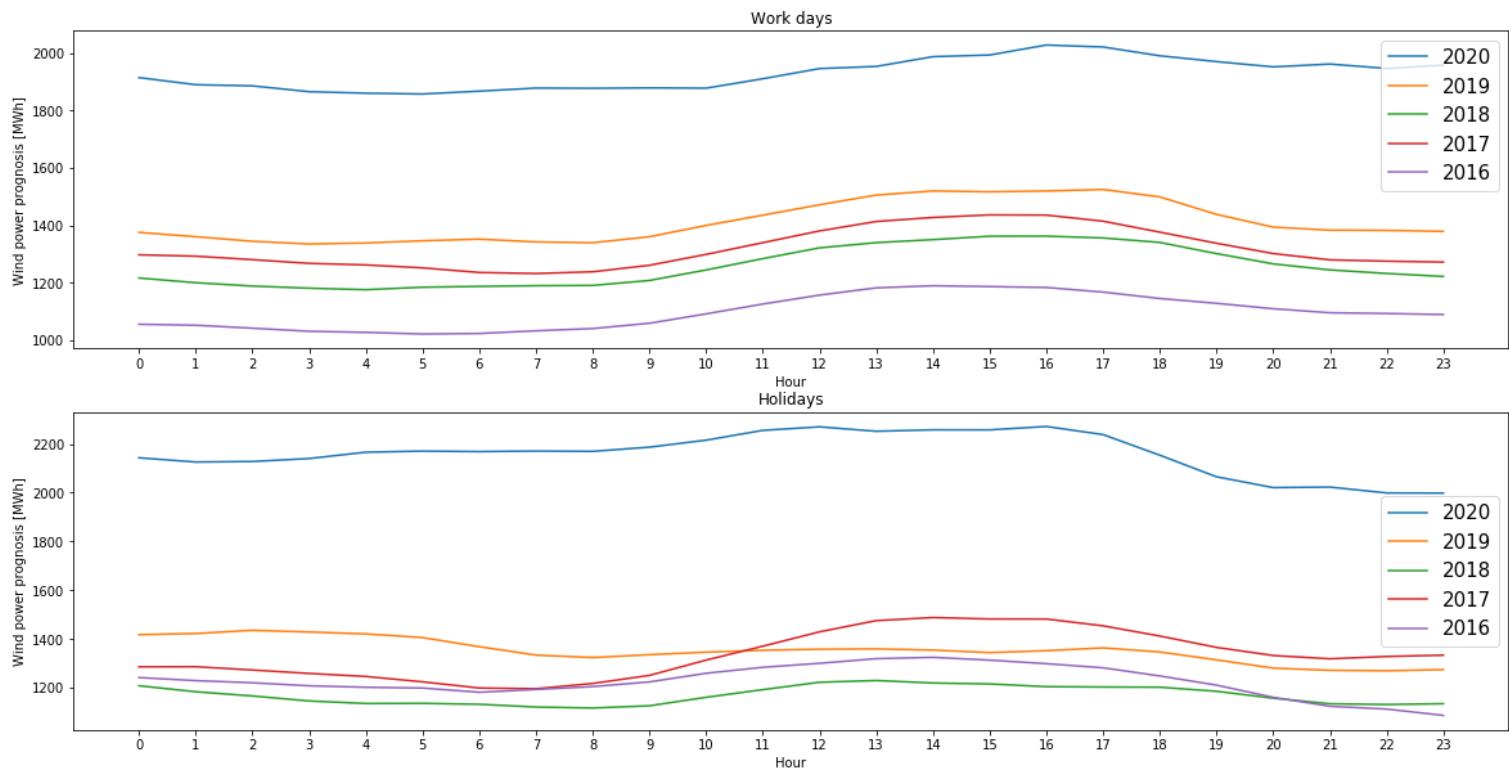


Figure 64:

Median wind power prognosis for years 2016-2020 (DK1)

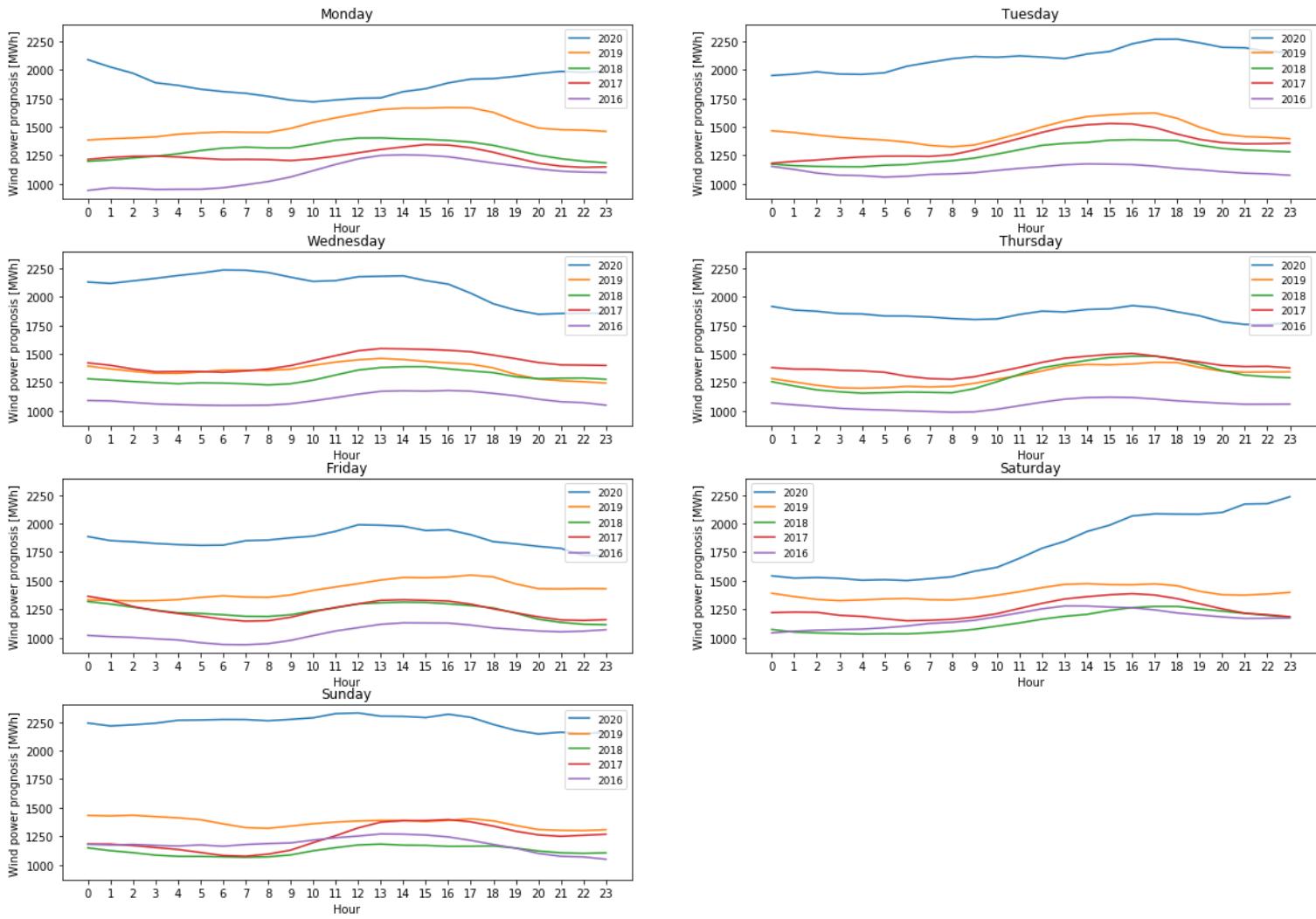


Figure 65:

Standard deviation of wind power prognosis for years 2016-2020 (DK1)

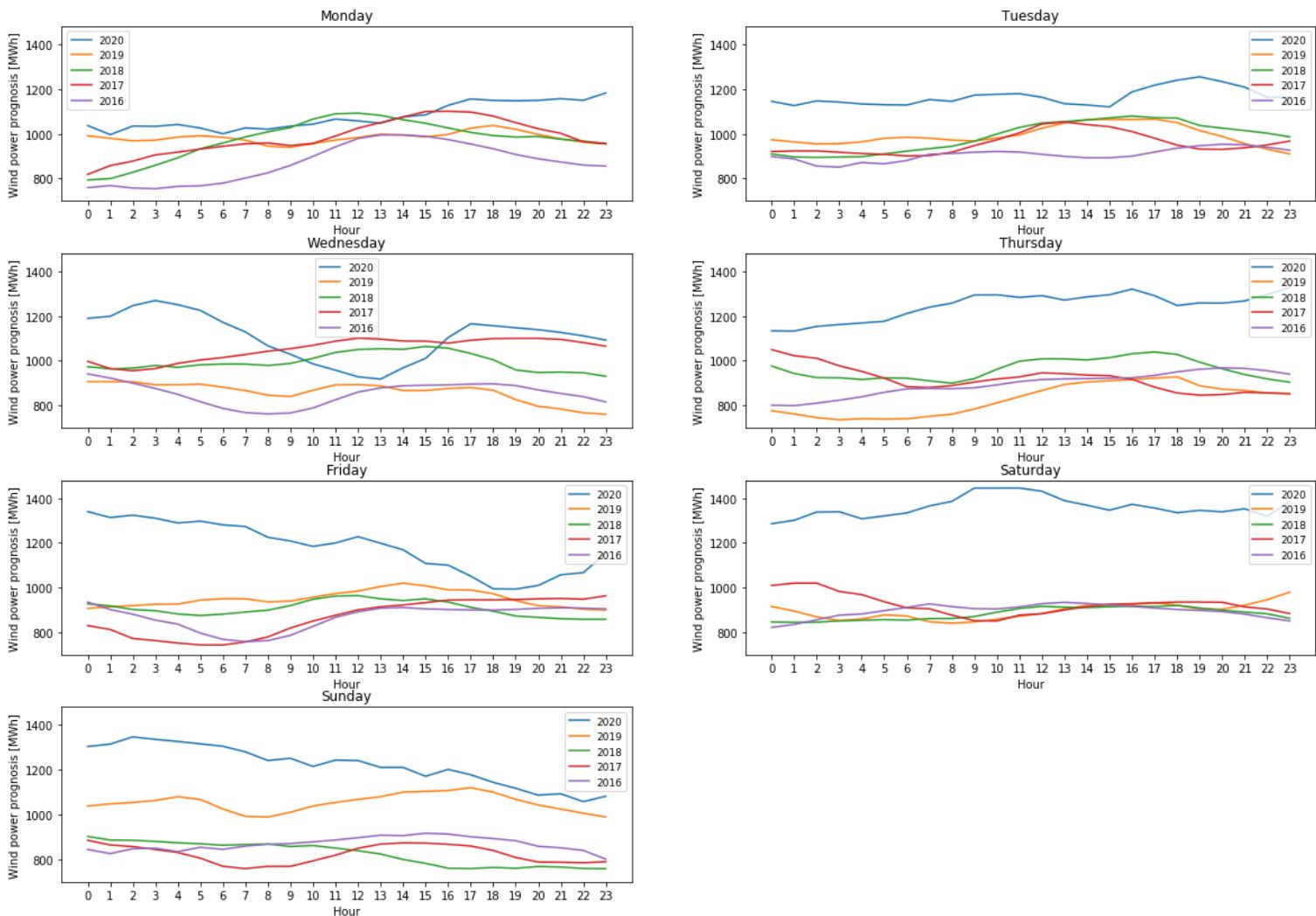


Figure 66:

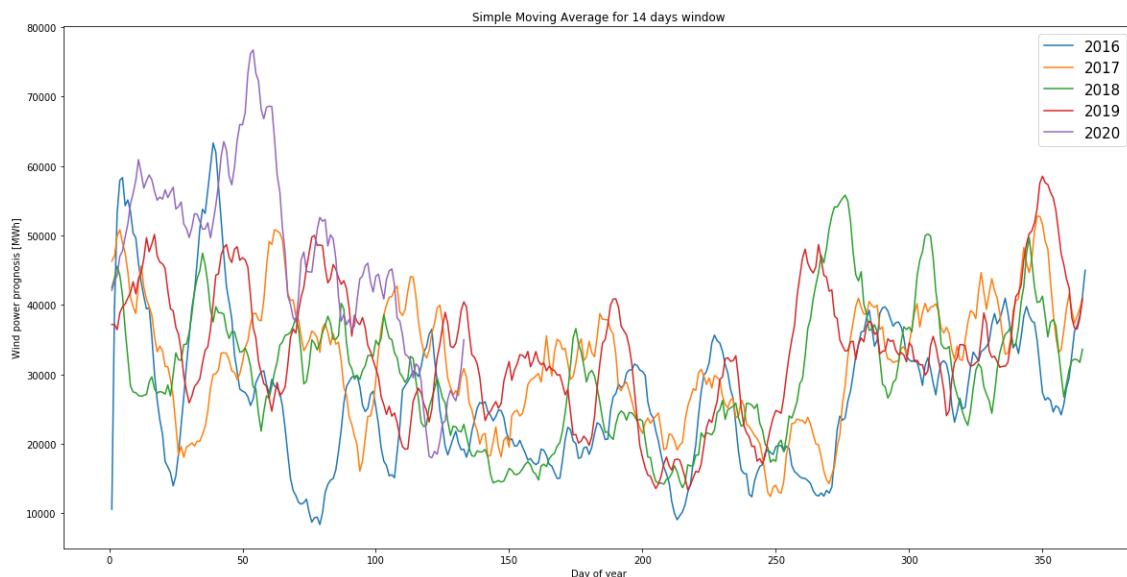


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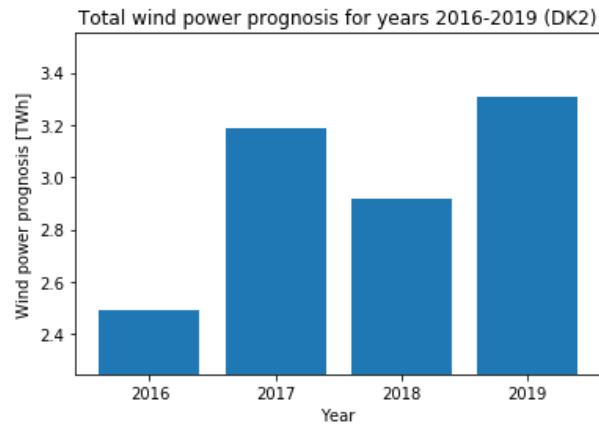


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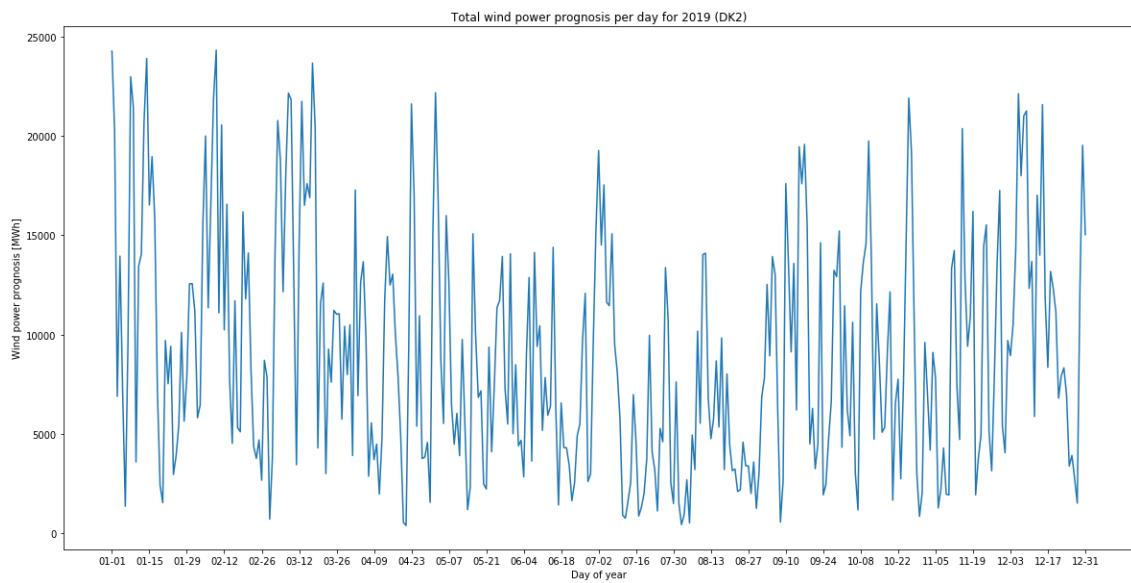


Figure 69:

Average wind power prognosis for years 2016-2020 (DK2)

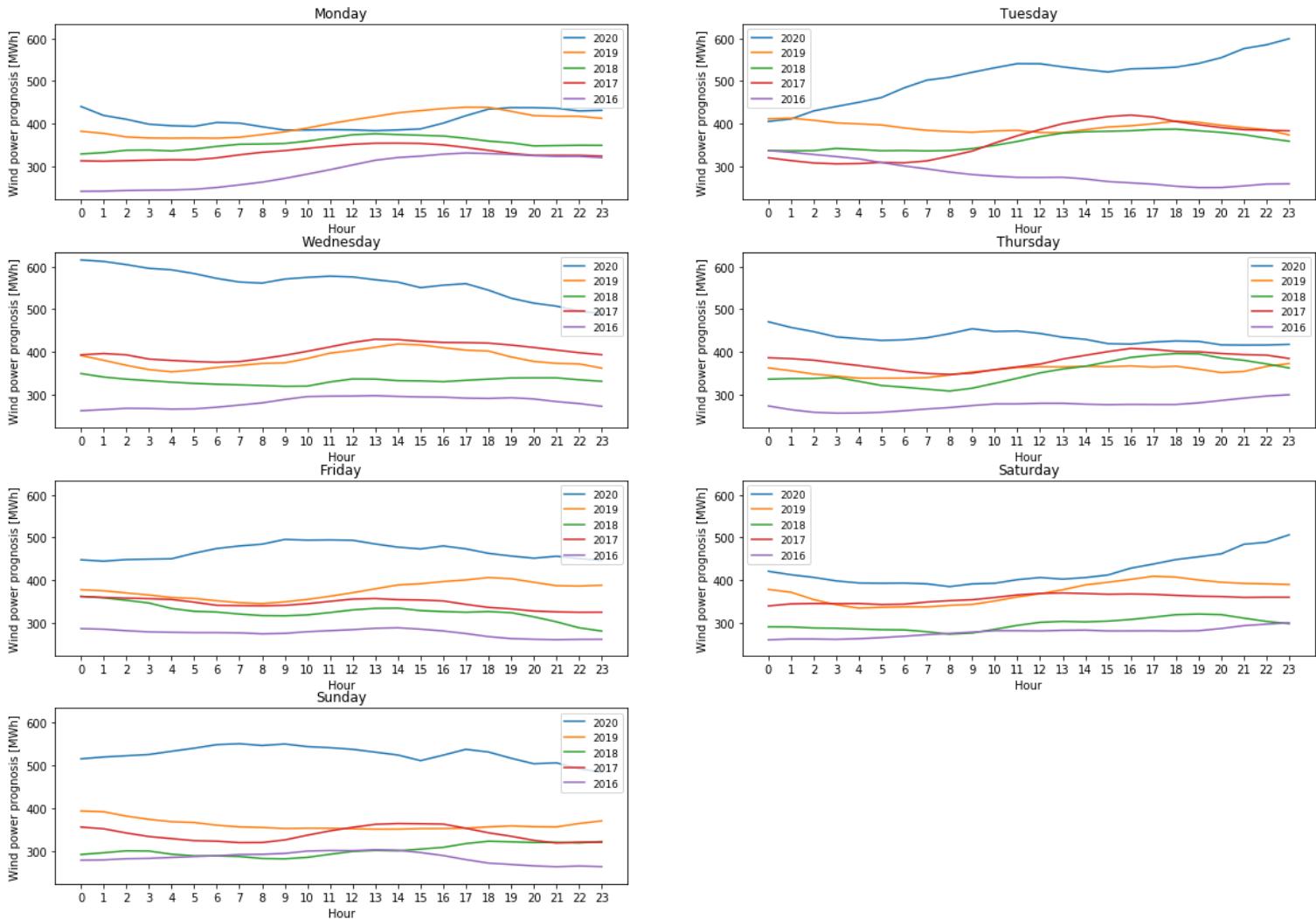


Figure 70:

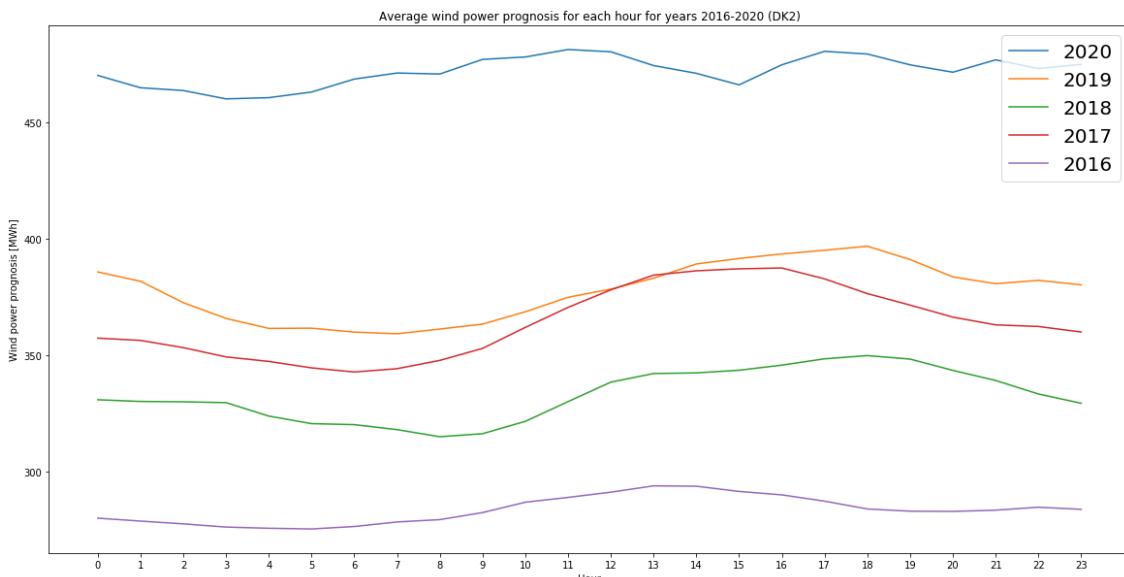


Figure 71:

Average wind power prognosis for years 2016-2020 (DK2)

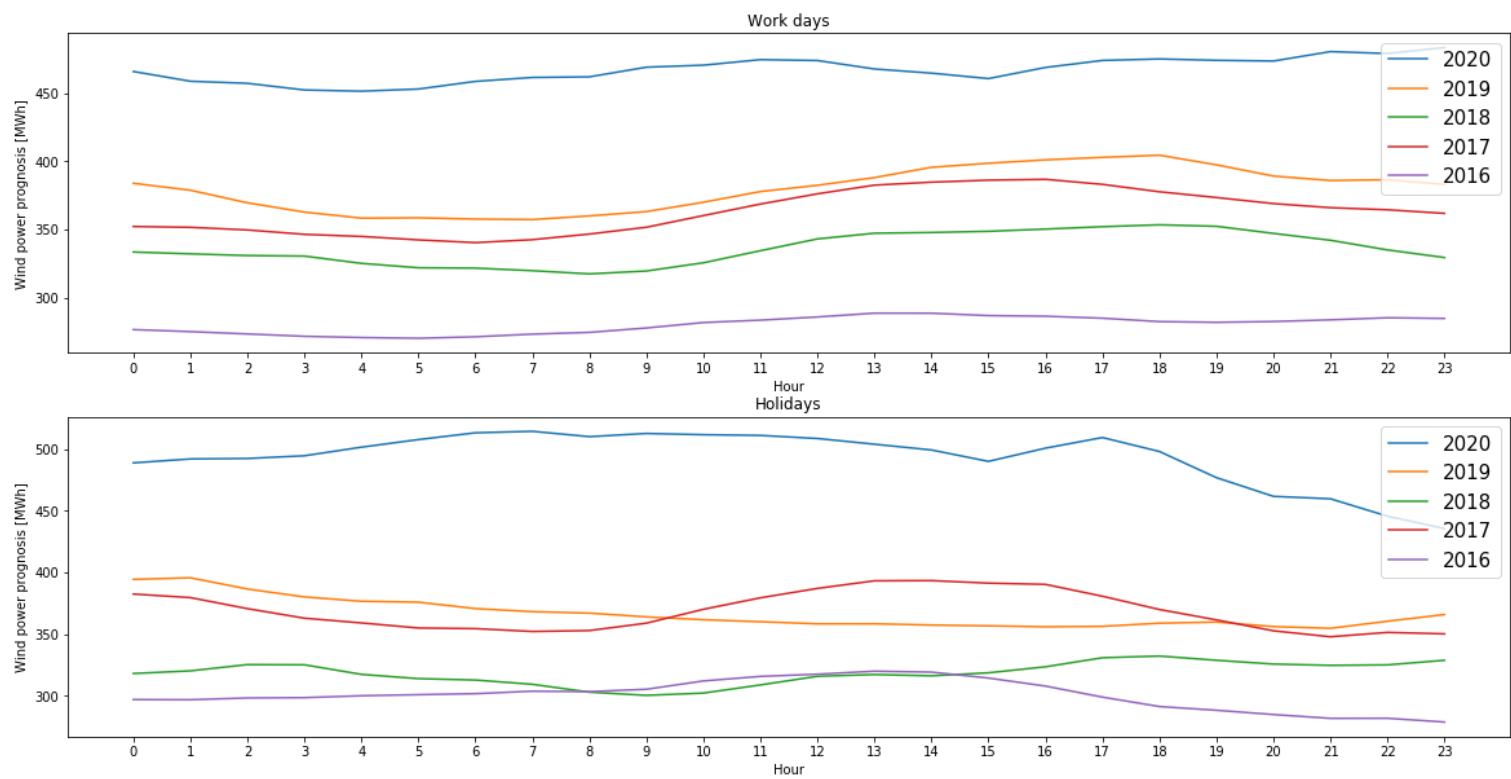


Figure 72:

Median wind power prognosis for years 2016-2020 (DK2)

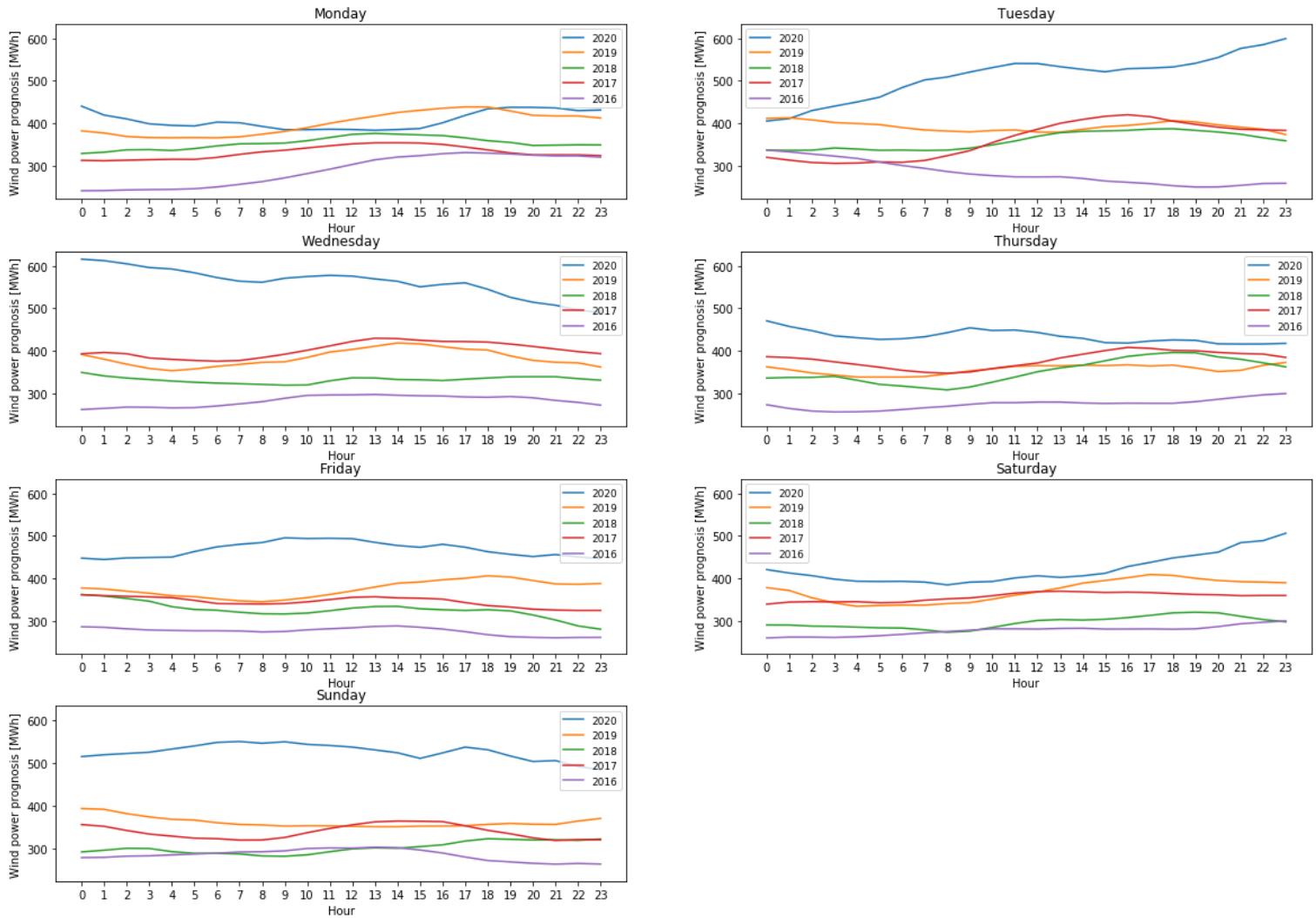


Figure 73:

Standard deviation of wind power prognosis for years 2016-2020 (DK2)

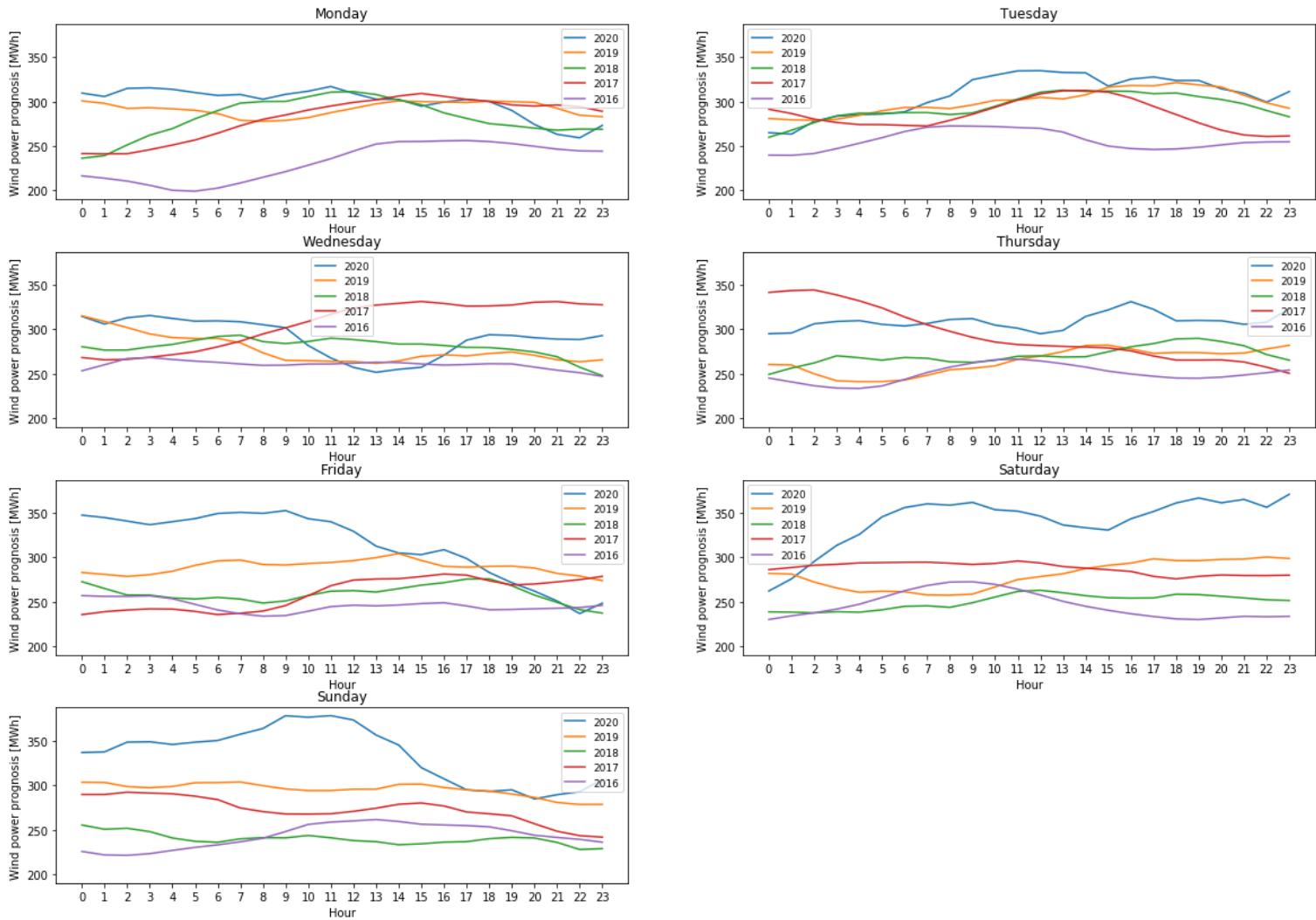


Figure 74:

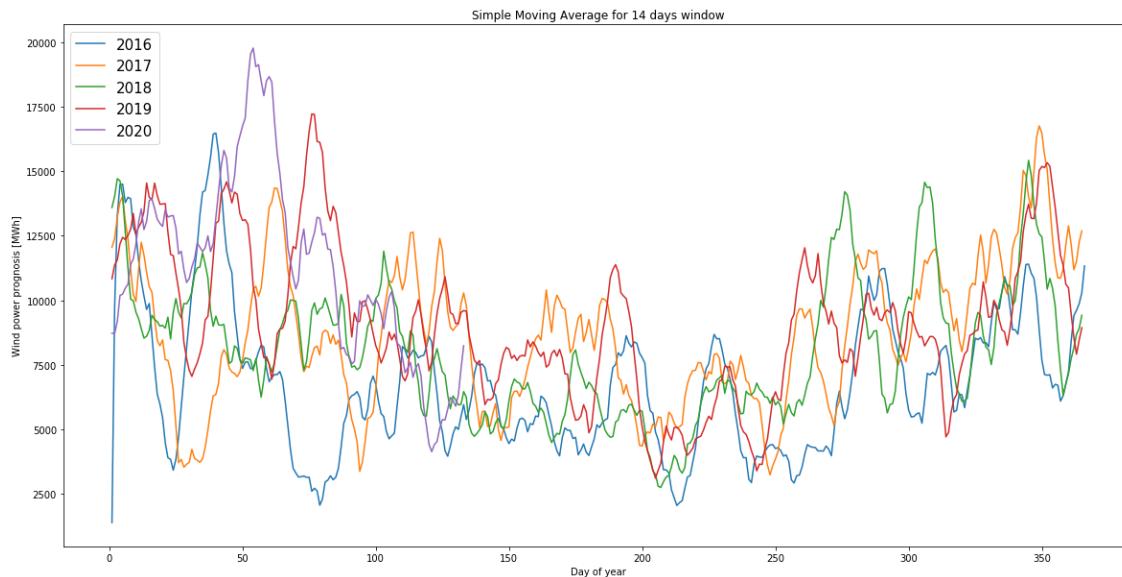


Figure 75:

3.2.2 Wind power data prognosis metrics

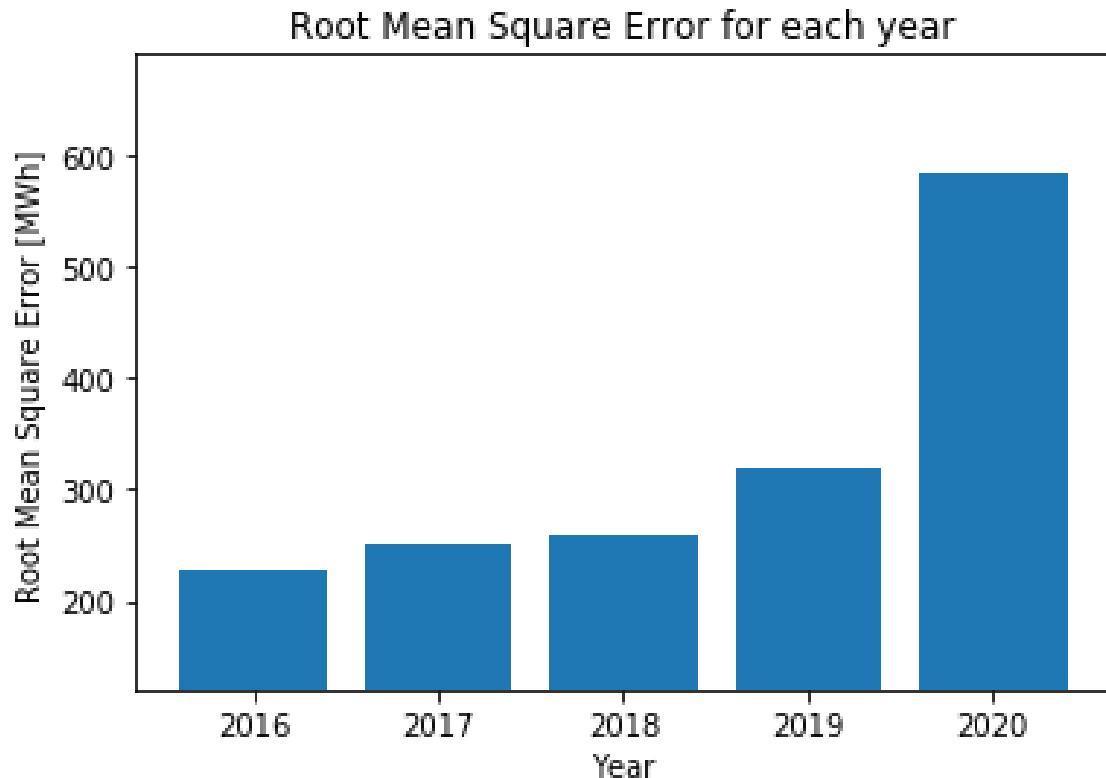


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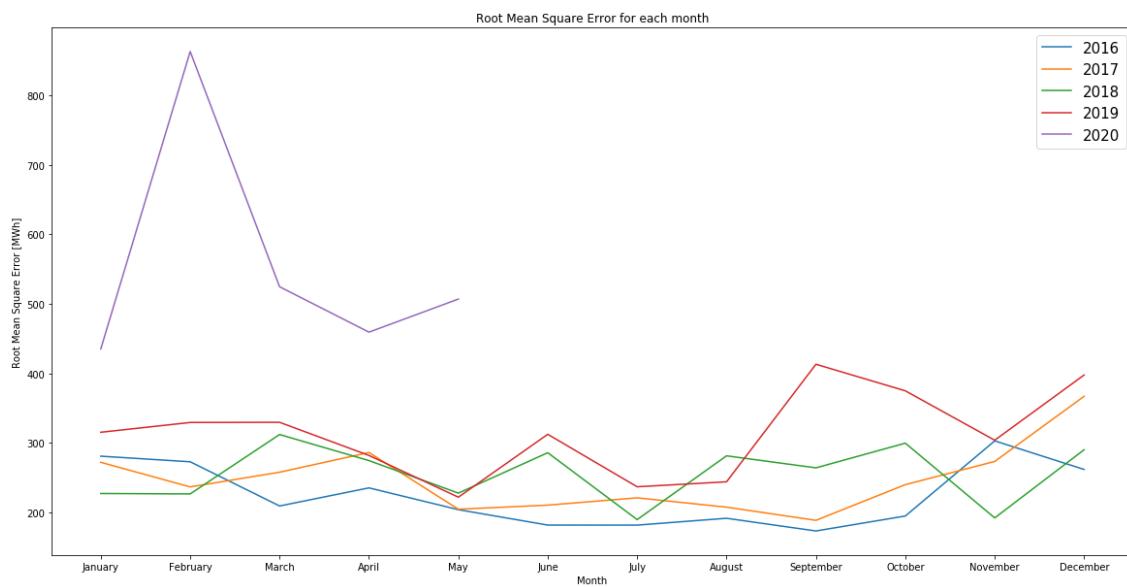


Figure 77:

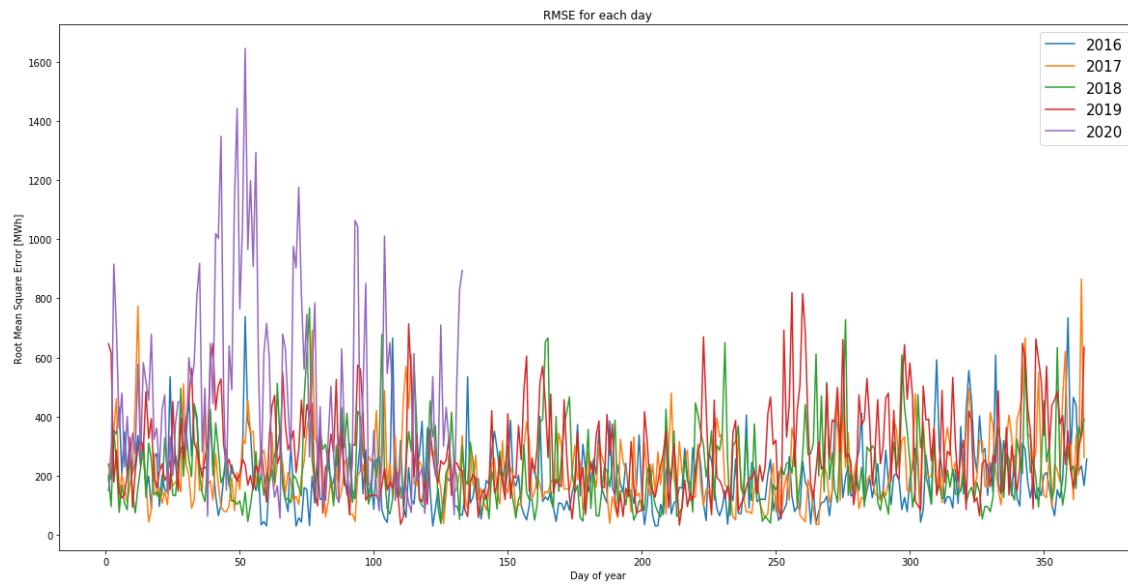


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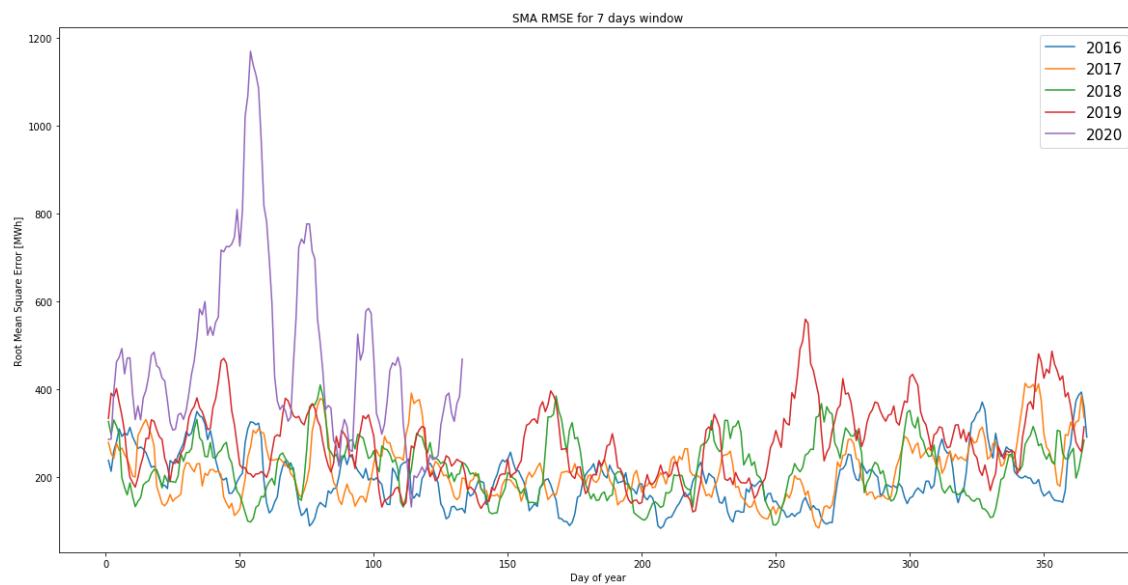


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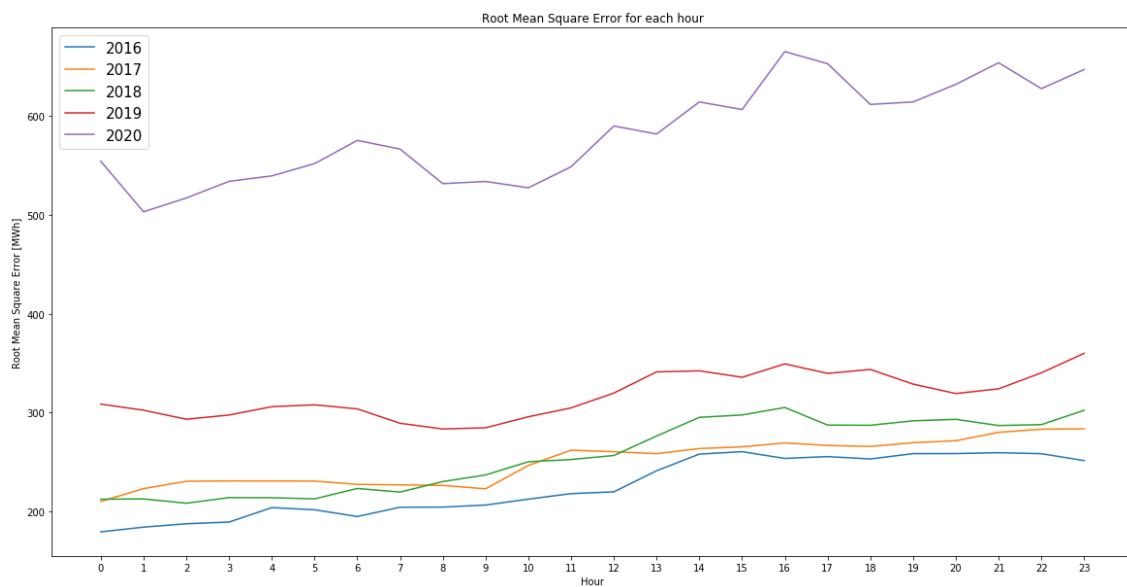


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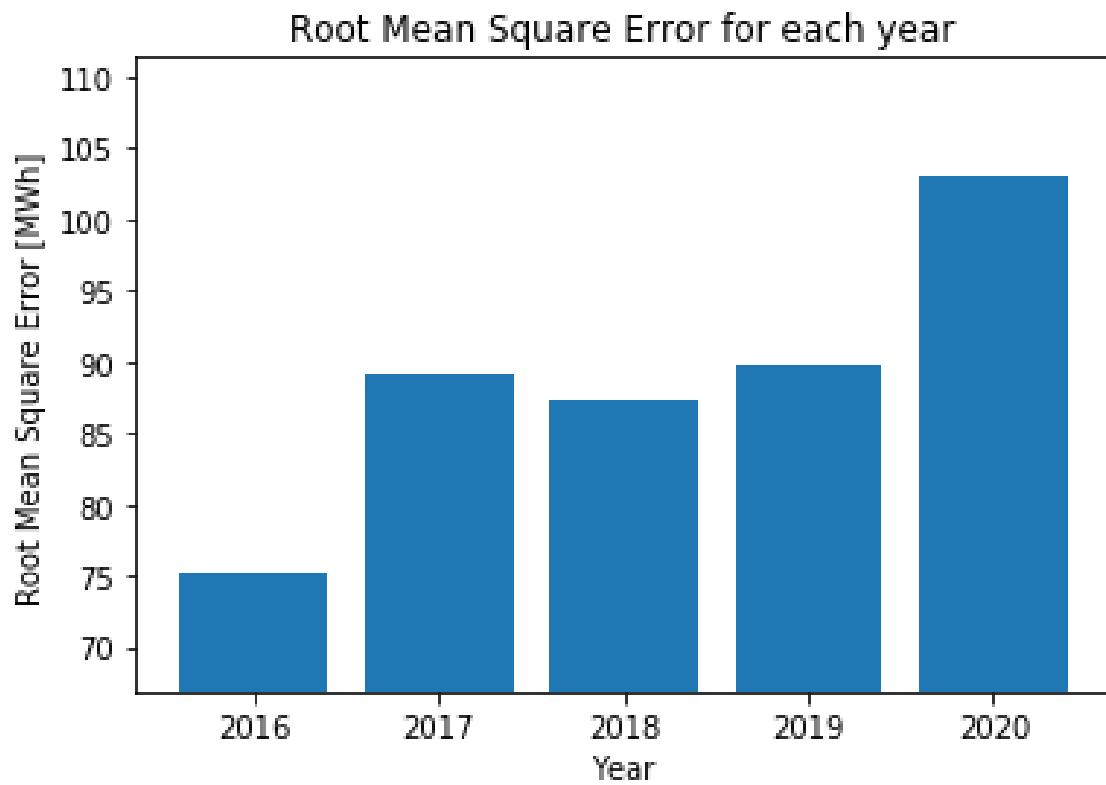


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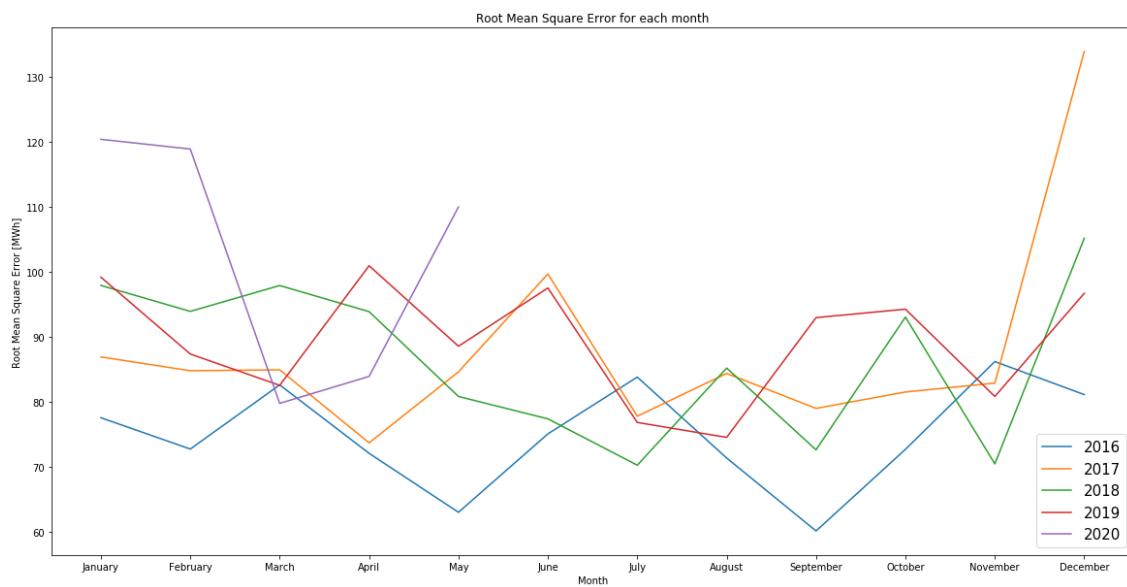


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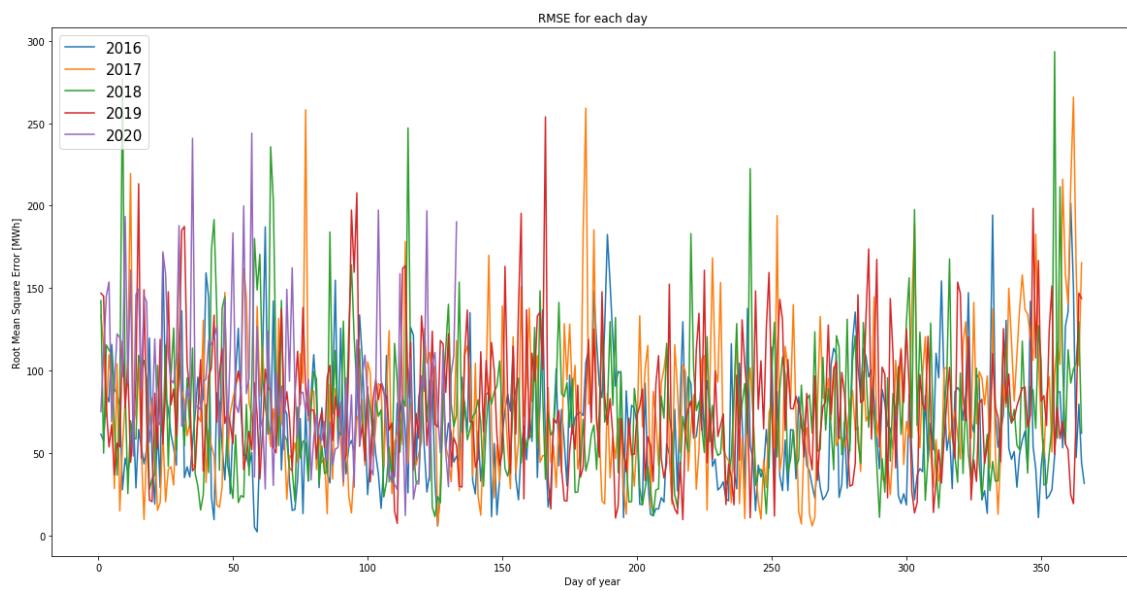


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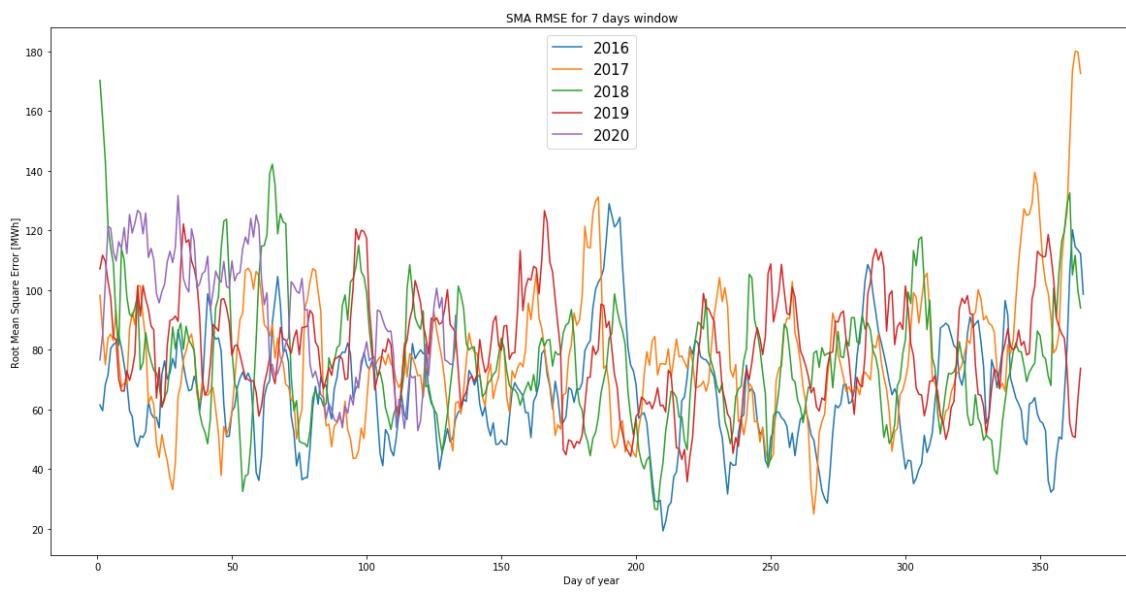


Figure 84:

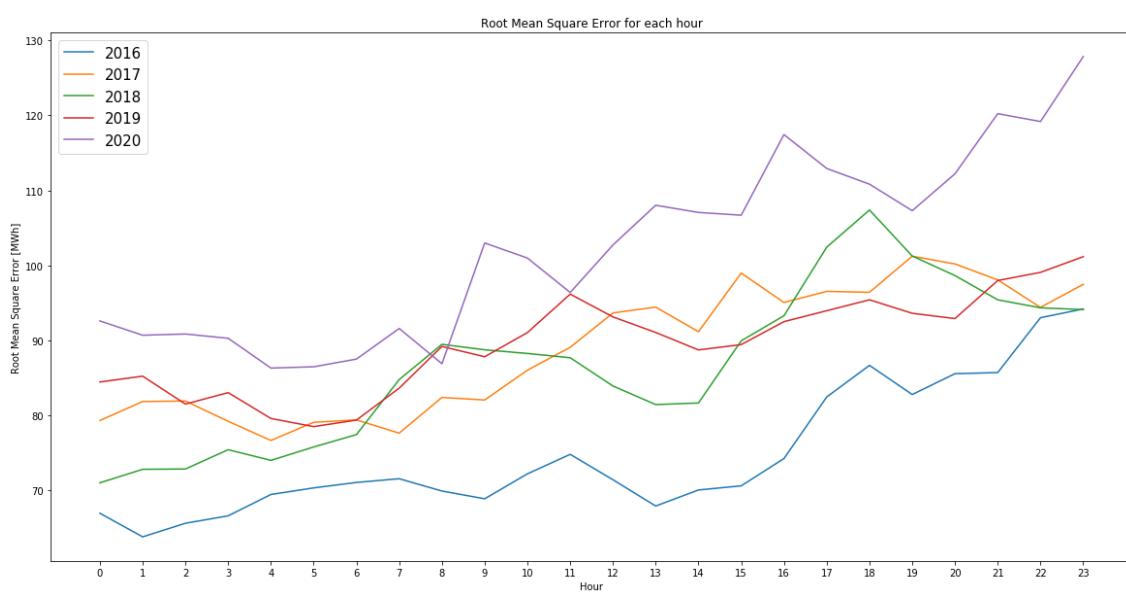


Figure 85:

3.3 Price data

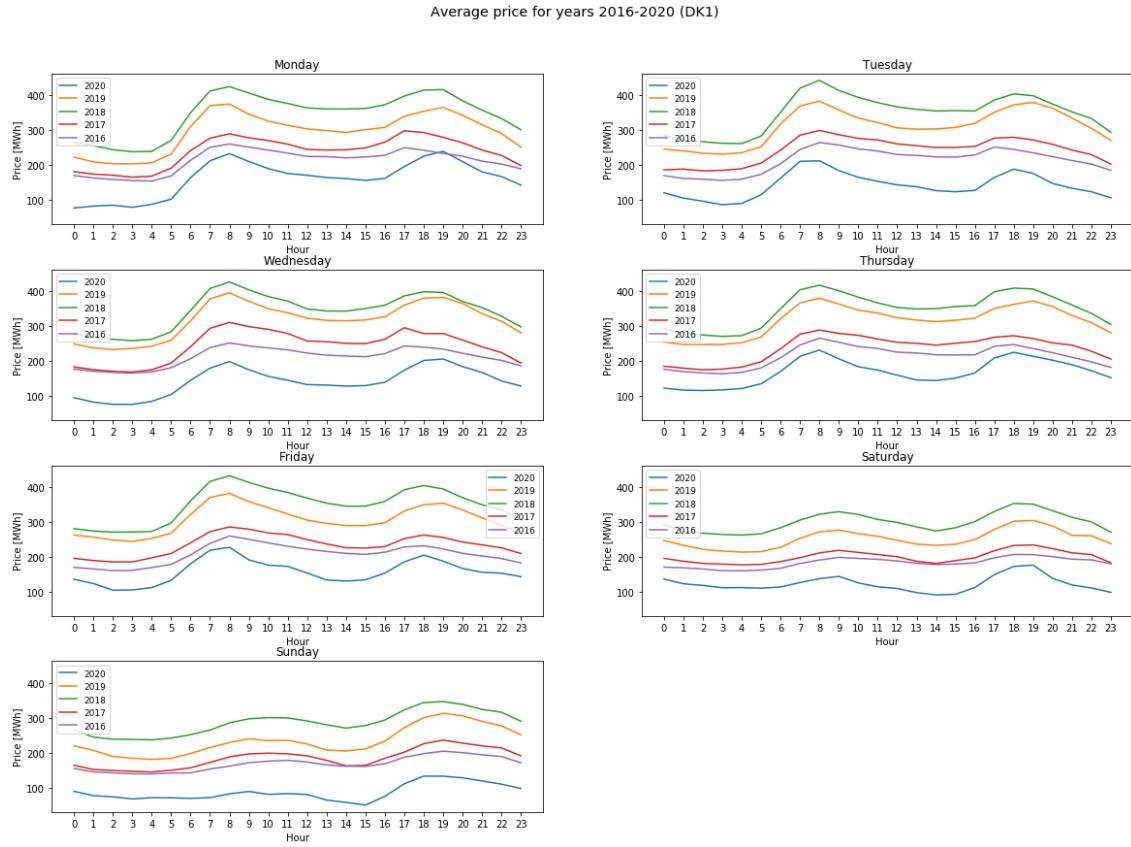


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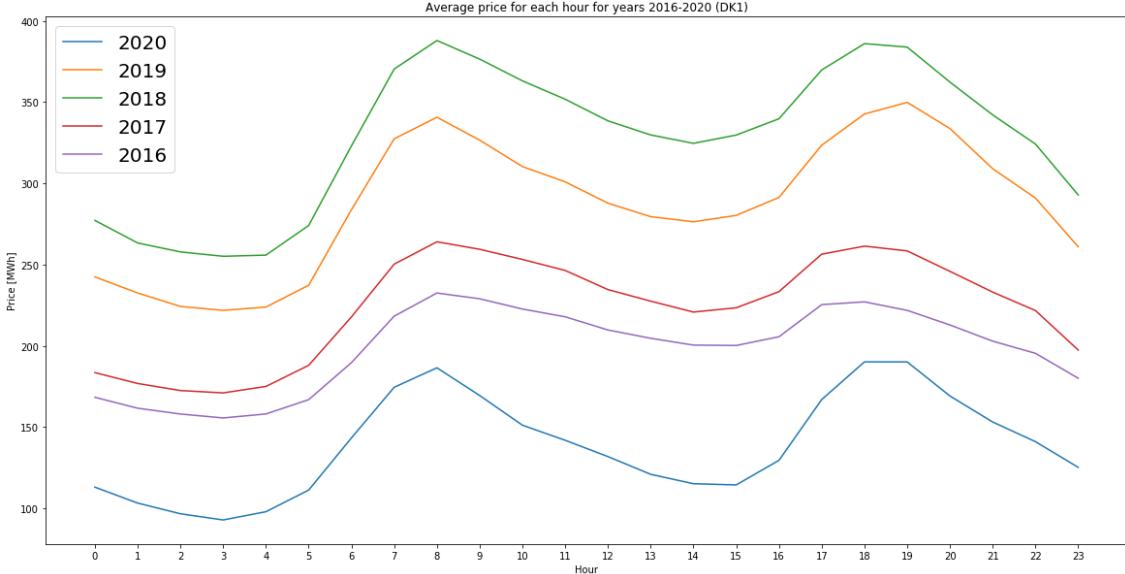


Figure 87:

$$RMSE_t = \sqrt{\frac{1}{24} \sum_h (y_{t,h} - \hat{y}_{t,h})^2} \quad (3)$$

$$RMSE_h = \sqrt{\frac{1}{T} \sum_t (y_{t,h} - \hat{y}_{t,h})^2} \quad (4)$$

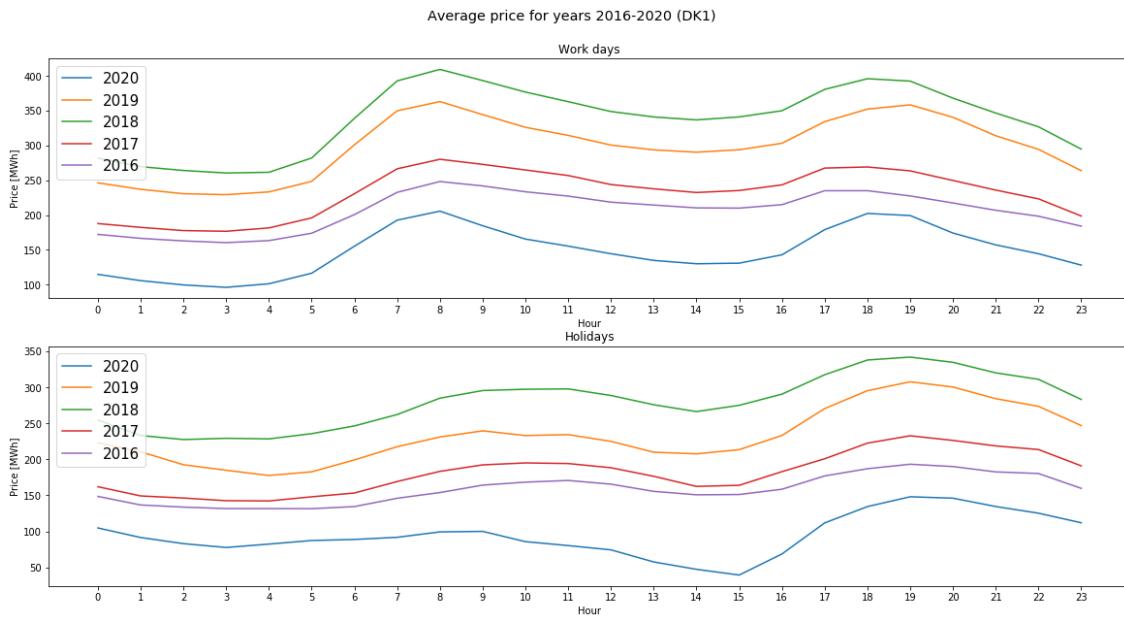


Figure 88:

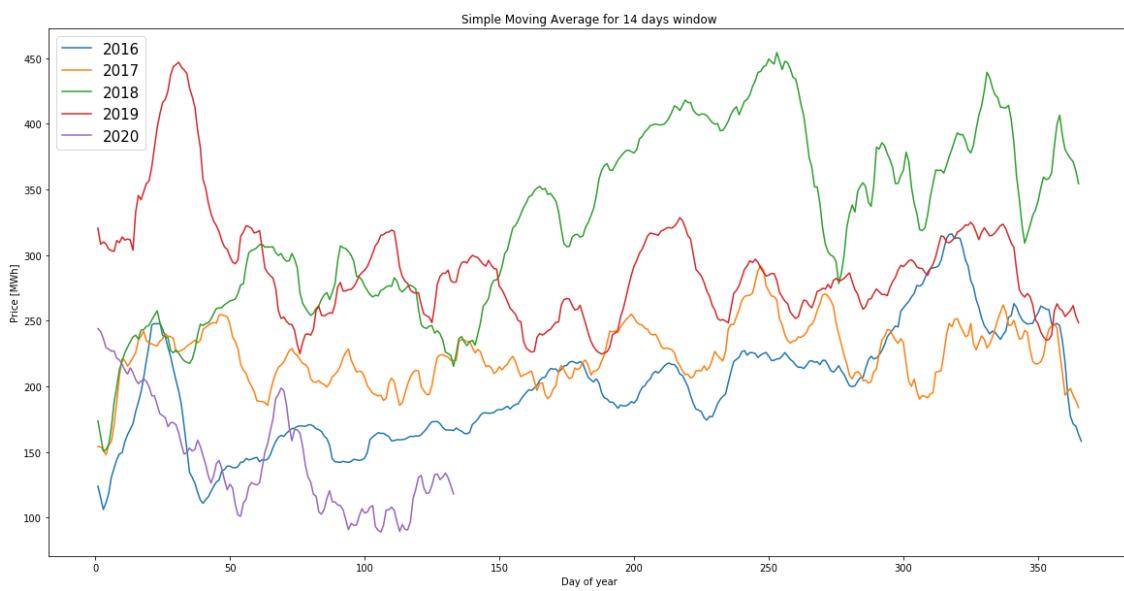


Figure 89:

Median price for years 2016-2020 (DK1)

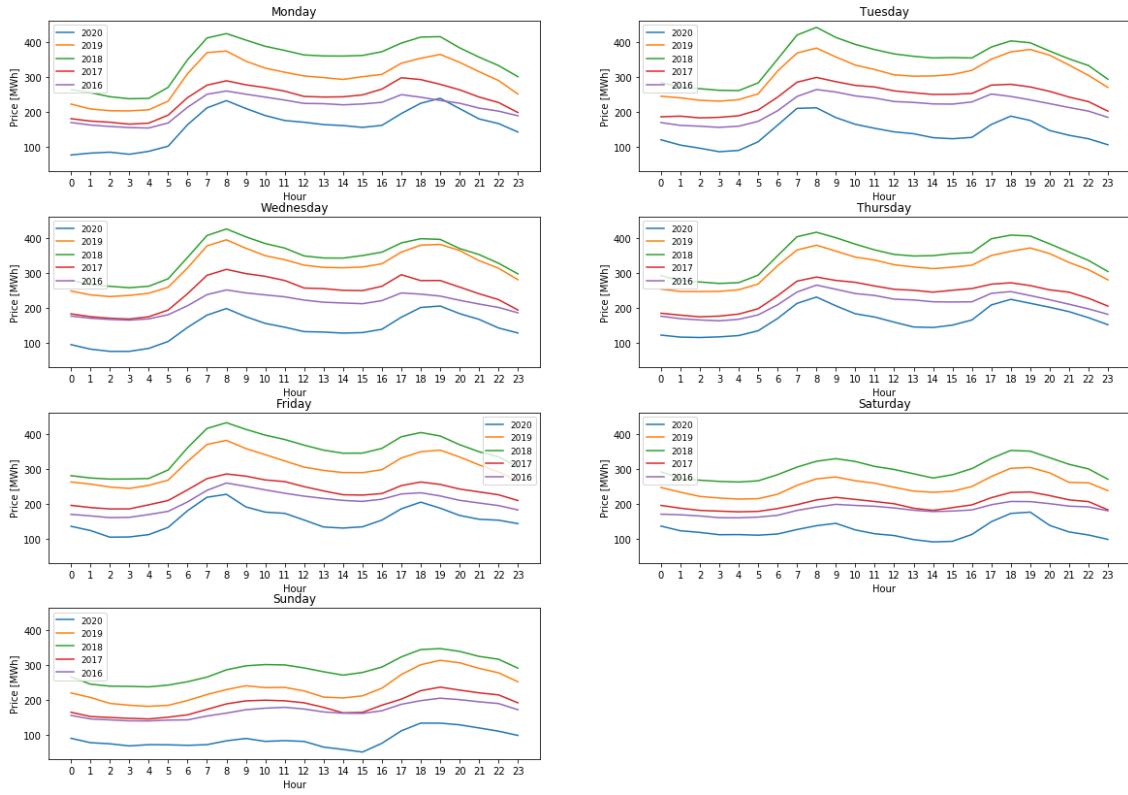


Figure 90:

Standard deviation of price for years 2016-2020 (DK1)

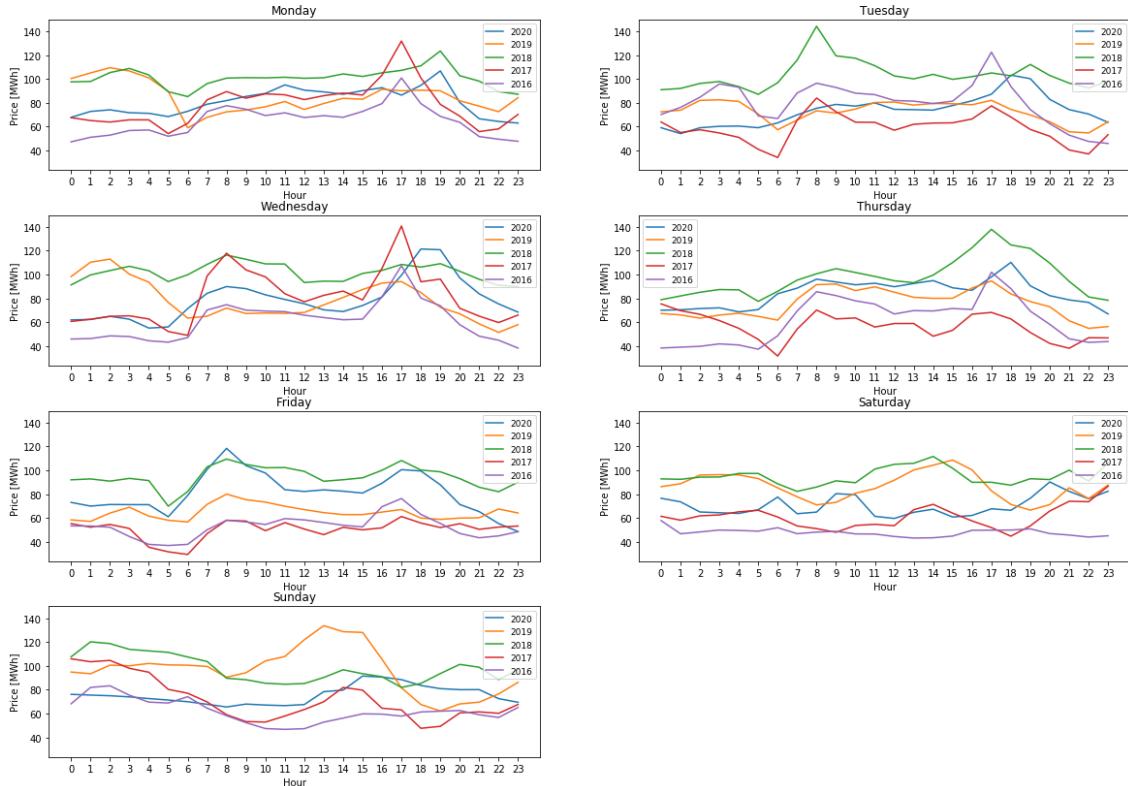


Figure 91:

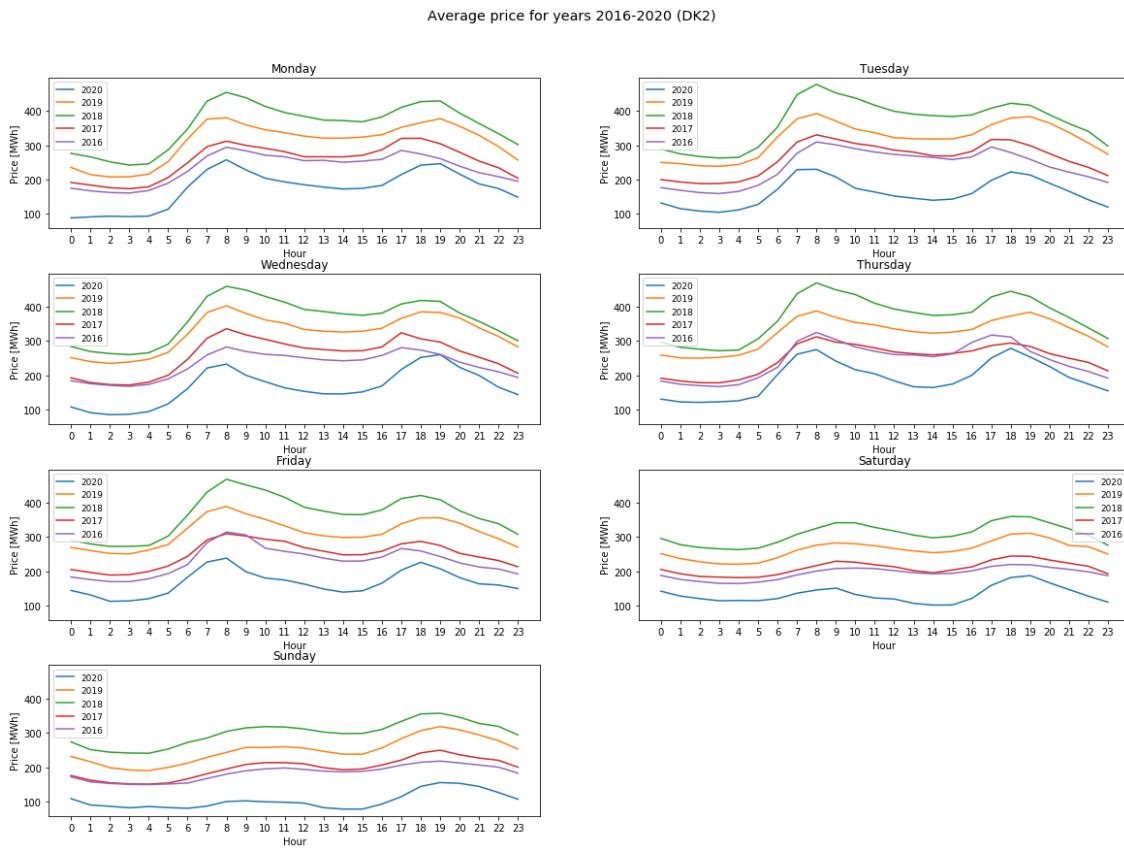


Figure 92:

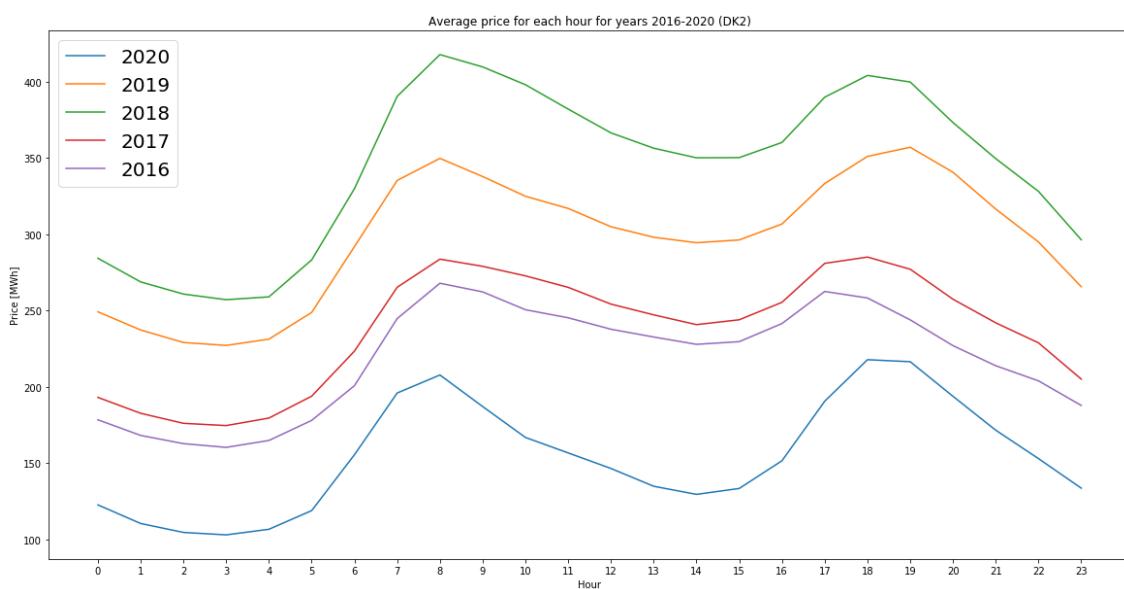


Figure 93:

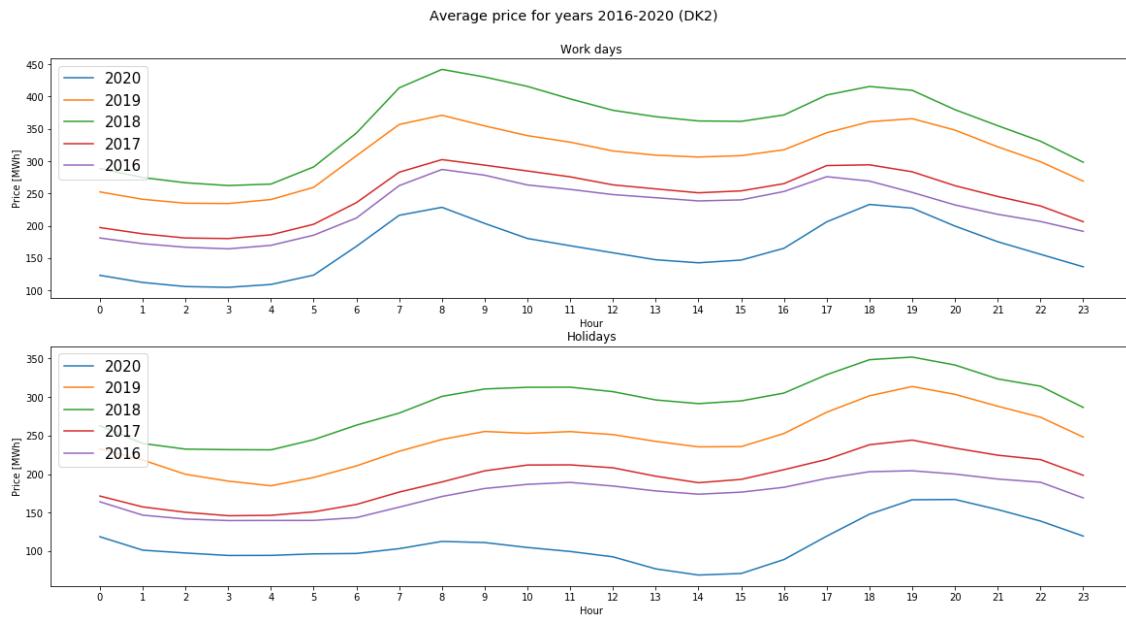


Figure 94:

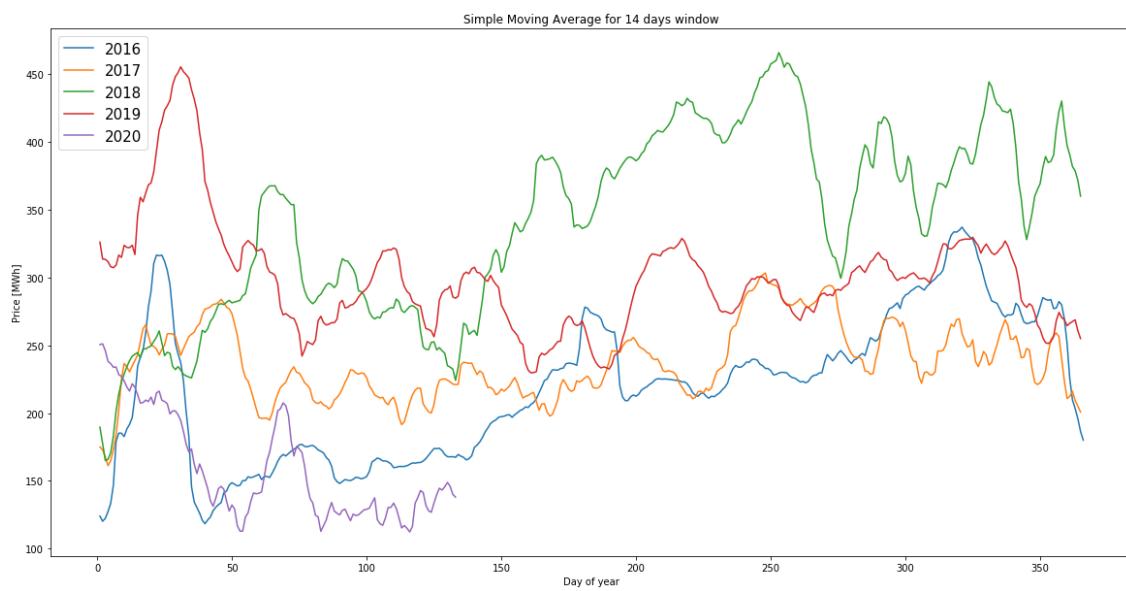


Figure 95:

Median price for years 2016-2020 (DK2)

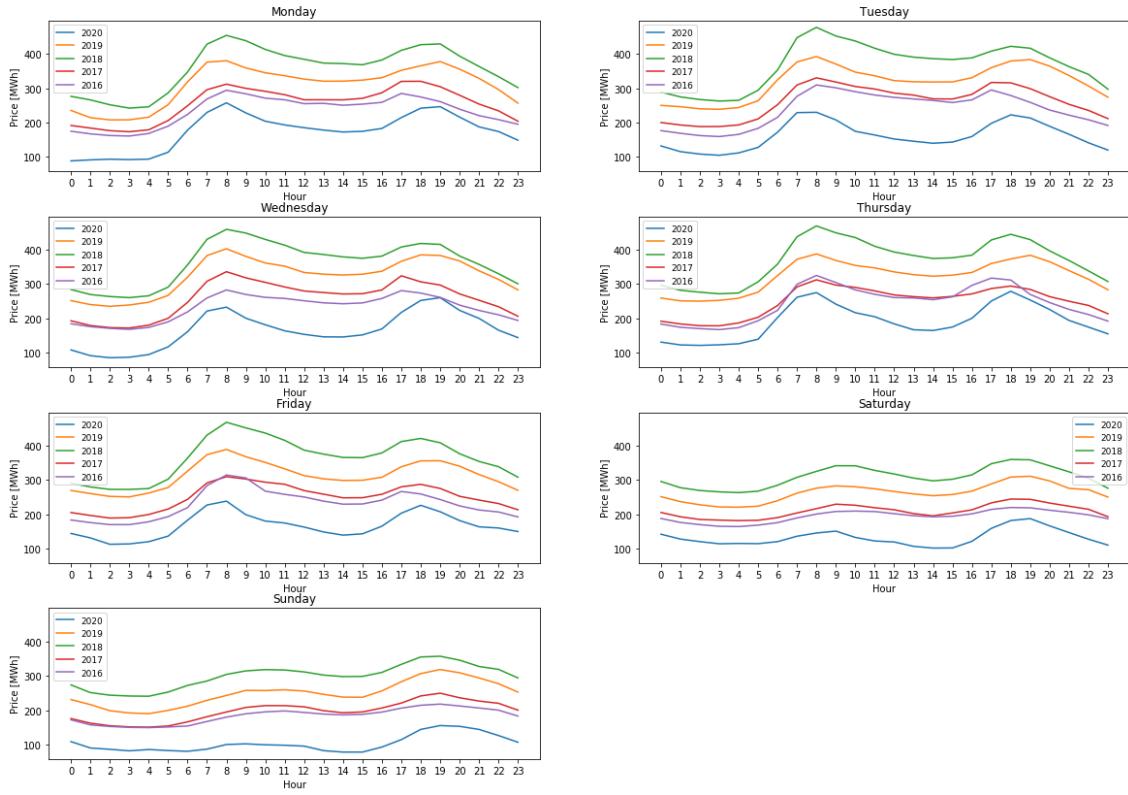


Figure 96:

Standard deviation of price for years 2016-2020 (DK2)

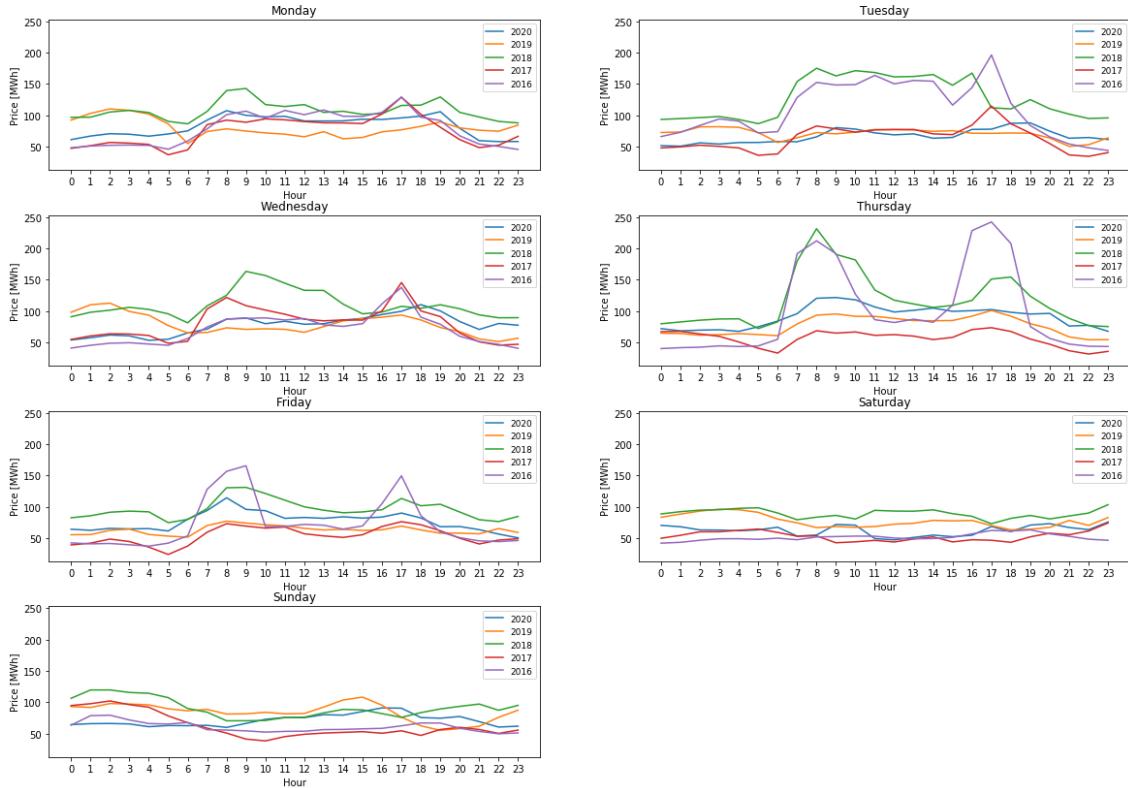


Figure 97:

3.4 Holidays

Usage of the electricity decreases during weekends and public holidays and this has significant effect on prediction especially during Christmas or Easter. Due to that fact, each day was aligned with variable holiday with following value based of occurrence of day of week or public holiday:

- 1 - National Holidays (e.g. Easter Monday)[2]
- 1 - Sundays
- 0 - Not a public holidays (e.g. New Year's Eve)
- 0 - Other days

4 Demand forecasting

$$L_{t,h} = \sum_{i=1}^7 L_i D_{it} + \phi_1 F L_{t,h} + \varepsilon_t \quad (5)$$

$$L_{t,h} = \sum_{i=1}^7 L_i D_{it} + \phi_1 F L_{t,h} + \beta_1 L_{t-1,h} + \beta_2 L_{t-2,h} + \beta_3 L_{t-7,h} + \varepsilon_t \quad (6)$$

$$L_{t,h} = \sum_{i=1}^7 L_i D_{it} + \phi_1 F L_{t,h} + \beta_1 L_{t-1,h} + \beta_2 L_{t-2,h} + \beta_3 L_{t-7,h} + \phi_1 F W_{t,h} + \varepsilon_t \quad (7)$$

$$L_{t,h} = F L_{t,h} \quad (8)$$

5 Analysis of demand forecast

6 Wind power forecasting

$$W_{t,h} = L_0 + \phi F W_{t,h} + \varepsilon_t \quad (9)$$

$$W_{t,h} = L_0 + \phi F W_{t,h} + \beta_1 W_{t-1,h} + \beta_2 W_{t-2,h} + \varepsilon_t \quad (10)$$

$$W_{t,h} = L_0 + \phi_1 F W_{t,h} + \beta_1 W_{t-1,h} + \beta_2 W_{t-2,h} + \phi_2 F L_{t,h} + \varepsilon_t \quad (11)$$

$$W_{t,h} = F W_{t,h} \quad (12)$$

7 Analysis of wind power forecast

8 Price forecasting

$$y_{t,h} = \sum_{i=1}^7 L_i D_{it} + \varepsilon_t \quad (13)$$

$$y_{t,h} = \sum_{i=1}^7 L_i D_{it} + \beta_1 y_{t-1,h} + \beta_2 y_{t-2,h} + \beta_3 y_{t-7,h} + \varepsilon_t \quad (14)$$

$$y_{t,h} = \sum_{i=1}^7 L_i D_{it} + \beta_1 y_{t-1,h} + \beta_2 y_{t-2,h} + \beta_3 y_{t-7,h} + \beta_4 y_{t-7,min} + \beta_5 y_{t-1,h} + \beta_6 y_{t-1,24} + \varepsilon_t \quad (15)$$

$$y_{t,h} = \sum_{i=1}^7 L_i D_{it} + \beta_1 y_{t-1,h} + \beta_2 y_{t-2,h} + \beta_3 y_{t-7,h} + \beta_4 y_{t-7,min} + \beta_5 y_{t-1,h} + \beta_6 y_{t-1,24} + \beta_7 F W_{t,h} + \beta_8 F L_{t,h} + \varepsilon_t \quad (16)$$

9 Analysis of price forecast

10 Conclusions

11 Data analysis

References

- [1] *Historical Market Data — Nordpool.* URL: <https://www.nordpoolgroup.com/historical-market-data/>. (accessed: 2020-05-10).
- [2] *National Holidays in Denmark.* URL: <https://www.officeholidays.com/countries/denmark>. (accessed: 2020-05-10).