

Bachelor's Thesis

The Role of SAP Analytics Cloud Smart Predict Time Series Feature in Evaluating Energy Consumption Trends and Environmental Impact of the SAP Metronom Business Center Office Location

Academic Supervisor: prof. Ing. Martin Hanel, Ph.D.

Supervisor at SAP: Ing. Pavel Strnad

Author: Vera Vavan, SAP

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Agenda

Thesis Objectives

- SAP Perspective
- Data Analysis Perspective

Research Methodology

- Data Wrangling
- Visualizations
- Forecasts

Results and Discussion

- Energy Consumption and CO2 Production
- Time Series Chart Forecast and Predictive Scenarios

Conclusions

Thesis Objectives

SAP Objectives

- Environmental impact of office buildings monitoring
- Data Democratization (DDEM) empowerment with SAP BTP solutions
- The chances of becoming carbon neutral in 2023

Data Analysis Objectives

- Investigating the energy consumption changes and their environmental impact in the context of COVID-19 shutdown in one of the Prague SAP offices
- Assessing the accuracy and reliability of smart predict forecast features in SAP Analytics Cloud

Research Methodology

Data Wrangling

- Data acquirement through the SAP Metronom Business Center landlord (White Star Real Estate) for the period between January 1, 2019 and March 31, 2022
- Importing the energy consumption data for the heating, cooling, water and electricity consumption into the SAC Modeler tool and editing the dataset semantic
- Data quality checks and calculations for the corresponding CO2 productions using the given emission factors:
 - 0.382/1000 for electricity
 - 0.055*1000 for cooling and heating
 - 0.00038*1000 for water

Research Methodology

Sensor	Sensor Type	Unit	Floor	Building	Location	TimeStampNew	Consumption
2RSC.D_PW_23	electricity meter	Wh	2.NP	C	Bucharova 2817/11	1/1/2020 0:00	768
8RSB.D_PW_83	electricity meter	Wh	8.NP	B	Bucharova 2817/11	8/11/2021 11:00	140
8WM.B.67_160422517A	water meter	m3	8.NP	B	Bucharova 2817/11	10/23/2021 23:00	0
2WM.A.33_16210433	water meter	m3	2.NP	A	Bucharova 2817/11	8/16/2020 1:00	2.66
5CM.B.1_51722749	calorimeter cooling	GJ	5.NP	B	Bucharova 2817/11	1/7/2022 14:00	0
9CM.A.1_62016499	calorimeter cooling	GJ	9.NP	A	Bucharova 2817/11	3/25/2020 23:00	0.09
9HM.B.1_53872591	calorimeter heating	GJ	9.NP	B	Bucharova 2817/11	1/25/2022 16:00	0.11
6HM.C.1_61107075	calorimeter heating	GJ	6.NP	C	Bucharova 2817/11	1/27/2022 19:00	0.1

Table 1: Data Sample of 8 Rows for the Provided CSV data

Research Methodology

Visualizations

- Creation of two pages using the SAC Story tool, one focused on energy consumption and one on CO2 production
- Finding the individual consumptions/productions per variable for each year and finding their variance compared to the previous year
- Producing time series charts, separating the consumption/production per variable for different buildings (A, B and C)

Research Methodology

Forecasts

- Applying the time series chart forecast functionality to CO2 production visualizations in SAC Story on each variable (heating, cooling, electricity, water and total production)
- Training a predictive model using the Predictive Scenario tool in SAC, with the target variable being consumption separated by the sensor types and suppressing negative values
 - Forecast periods: 92
 - Predicted date range: April 1, 2022 until July 1, 2022

Results and Discussion



Figure 1: Energy Consumption SAC Story Page (Cooling and Heating)

Results and Discussion



Figure 2: Energy Consumption SAC Story Page (Water and Electricity)

Results and Discussion – Energy Consumption

- Negative trend for cooling and water consumption kept over the years
- Slight increase in cooling, attributed to the catch-up effect and low temperatures
- High increase in electricity, connected to the missing sensor data before 2021
- High seasonal changes for cooling and heating, steady trend for electricity and slight yearly trend for water consumption with an increase between June and October
- Highest usage in building B, followed by building A
- Overall, the consumption is lower than the pre-pandemic level

Results and Discussion

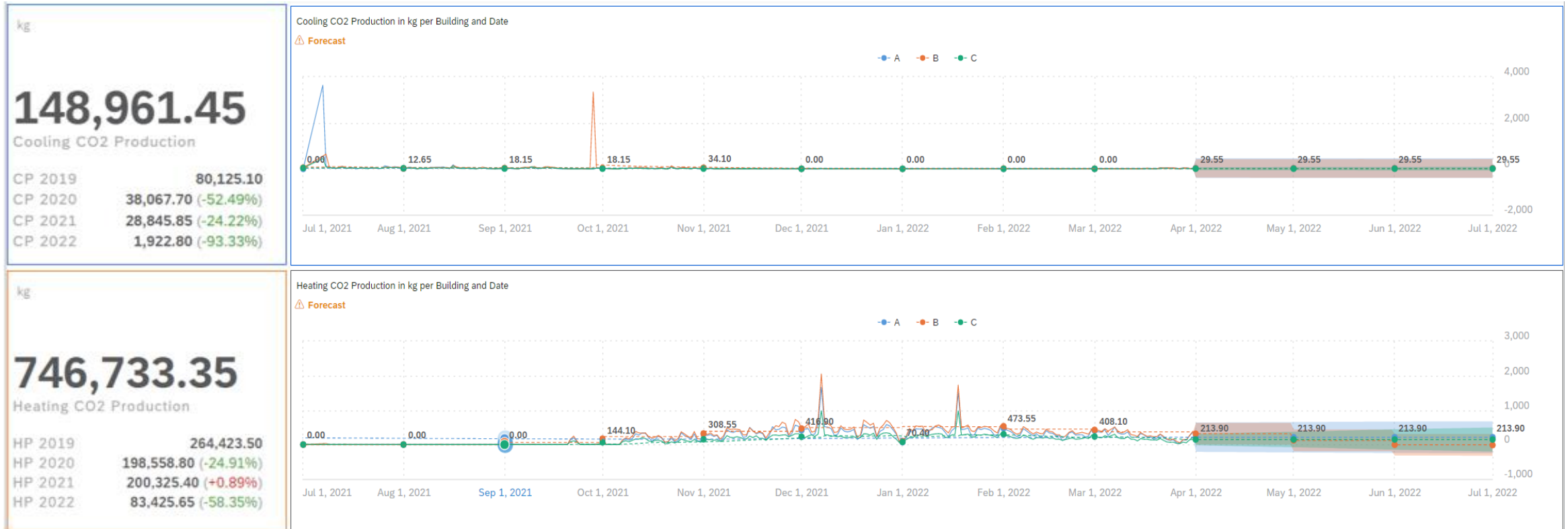


Figure 3: CO2 Production SAC Story Page with Time Series Chart Forecasting (Cooling and Heating)

Results and Discussion

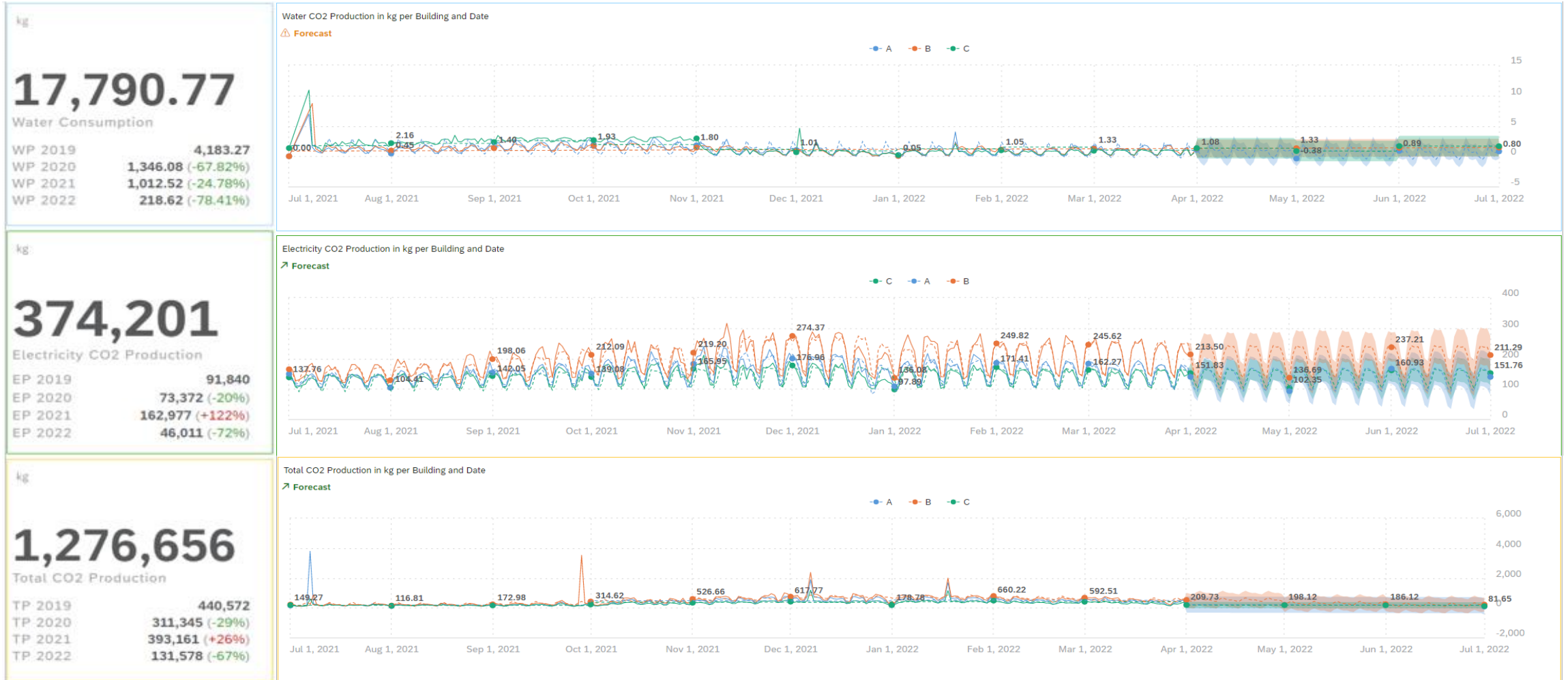


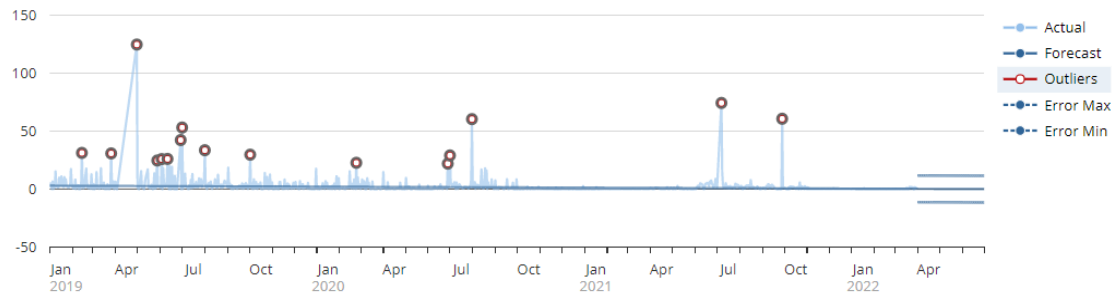
Figure 4: CO2 Production SAC Story Page with Time Series Chart Forecasting (Water, Electricity and Total)

Results and Discussion – CO2 Production Forecast

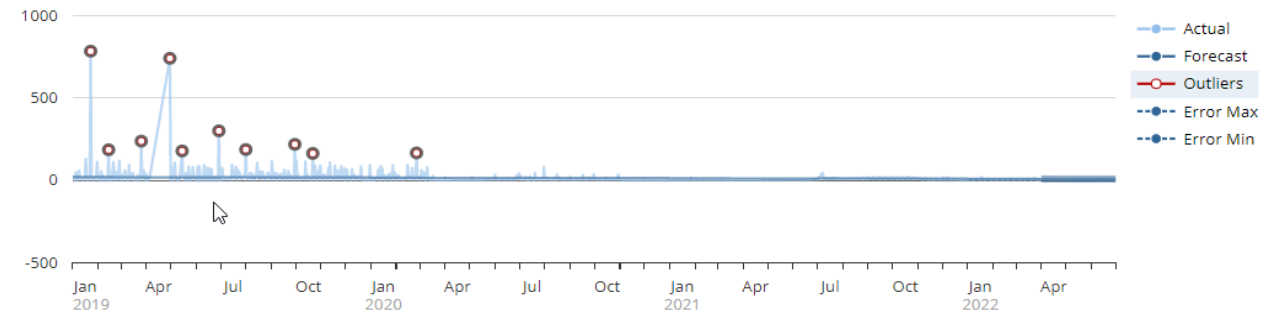
- Direct relation of CO2 production to energy consumption, but the units are the same (kg)
- Low confidence in heating, cooling and water consumption
- Highest confidence in electricity consumption where the weekly patterns are recognized
- Total CO2 production heavily influenced by heating consumption
- Decreasing trends interpreted as future negative values, making the predictions unrealistic
- The tool is easy to use and interpret

Results and Discussion

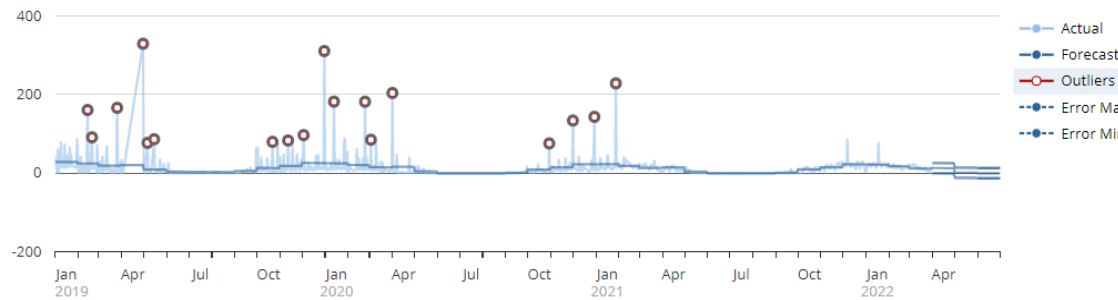
Forecast vs. Actual for Calorimeter Cooling



Forecast vs. Actual for Water Meter



Forecast vs. Actual for Calorimeter Heating



Forecast vs. Actual for Electricity Meter

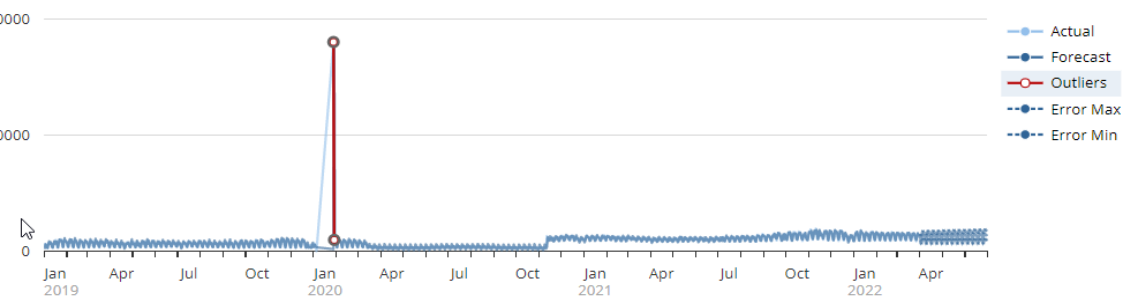


Figure 5: Predictive Scenario Charts for Cooling, Heating, Water and Electricity

Results and Discussion

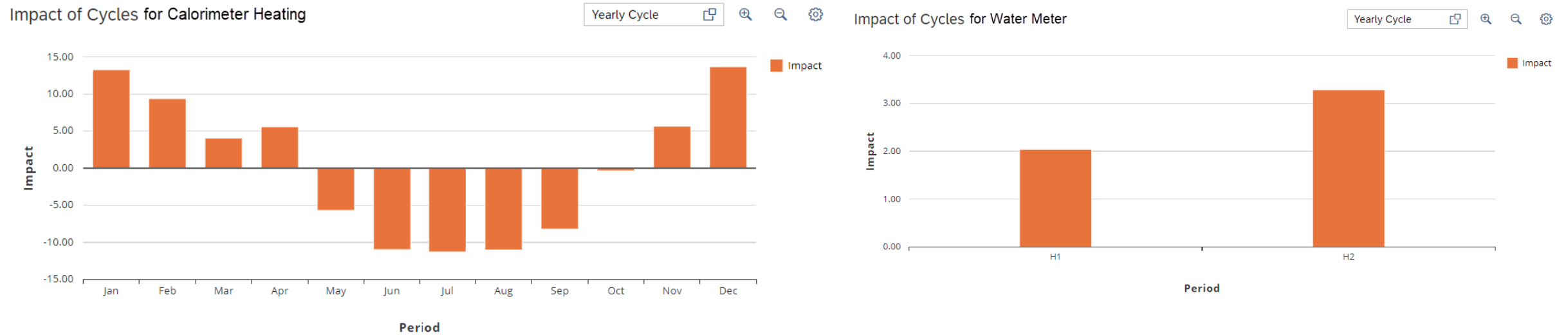


Figure 6: Predictive Scenario Cycle Impact for Heating and Water Consumption

Results and Discussion – Energy Consumption Responsibility Scenario

- MAPE median expected value of 193.15% and the MAPE average of 163.36%
- MAPE 12.13% for electricity, 131.15% for water, 254.79% for cooling and 255.01% for heating
- No negative values, only confidence intervals which are quite large
- Poor weekly trend recognition, resulting in bad data fittings for the prediction model and high MAPE
- Yearly trend recognized only for heating and water consumption, and the impact is low
- Limited results insights and personalization
- Missing data and COVID-19 shutdown not impacting results

Results and Discussion

- General negative trend is recognized in comparison to 2019, with a catch-up effect only in heating and electricity consumption, due to employee preference of the remote working model
- Limited information due to missing and restricted data which also affects CO2 production
- Poor confidence and prediction reliability due to scarce personalization options and weekly pattern recognition inconsistencies
- Easy to use, however, addition of data insights and data semantic options would increase reliability and data-driven decision making
- Metronom Business Center implements energy usage restrictions based on occupancy

Conclusions

- Overall, historical data shows a negative trend in consumption with a slight catch-up effect which is not higher than the consumption before COVID-19
- Highest influence on CO2 production is from heating consumption, hence implementing new policies and investing in greener energy sources can help SAP become carbon neutral
- Future predictions have low confidence in most variables due to poor recognition of daily patterns, but the trend seems to be decreasing on a higher monthly scale
- While both tools empower DDEM, Predictive Scenarios are more accurate but time series chart forecasts are easier to use, which can result in ill-informed user misinterpreting data
- Including data insights and suggestion can highly improve user experience and DDEM

Thank you.

