```
In [1]: import datacleaner
import config
import os
import sys
import pandas as pd
import numpy as np
from fbprophet import Prophet
import matplotlib.pyplot as plt
import fbprophet

# "high resolution"
%config InlineBackend.figure_form
```

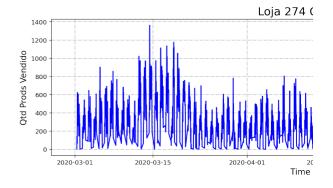
Análise dados para loja

Leitura do ficheiro para cri as variáveis número de ver (sales, clients)

```
In [2]: df_274_time_sale = datacleaner.ge
    df_274_time_cli = datacleaner.get
    df_274_time_sale.reset_index(inpl
    df_274_time_cli.reset_index(inpla)
```

Visualizando os dados reais para vendas

```
In [3]: plt.figure(figsize=(16,4))
    plt.grid(linestyle='-.')
    plt.plot(df_274_time_sale.ds, df_
    plt.title('Loja 274 Qtd Produtos/
    plt.ylabel('Qtd Prods Vendido', f
    plt.xlabel('Time (aaaa-mm-dd)', f
    plt.show()
```

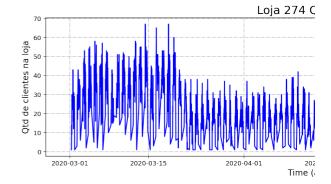


```
In [4]: # Obtendo o valor máximo de produ
df_274_time_sale['y'].max()
```

Out[4]: 1362

Visualizando os dados reais para clientes

In [5]: plt.figure(figsize=(16,4))
 plt.grid(linestyle='-.')
 plt.plot(df_274_time_cli.ds, df_2
 plt.title('Loja 274 Qtd Clientes/
 plt.ylabel('Qtd de clientes na lo
 plt.xlabel('Time (aaaa-mm-dd)', f
 plt.show()



In [6]: # Obtendo o número máximo de clie
df_274_time_cli['y'].max()

Out[6]: 67

In [7]:	# definindo o cap (carrying capac
	$df_274_time_sale['cap'] = 1362$
	<pre>df_274_time_sale.reset_index(drop</pre>
	df_274_time_sale

Out[7]:

	ds	У	cap
0	2020-03-01 07:30:00	2	1362
1	2020-03-01 08:00:00	144	1362
2	2020-03-01 08:30:00	117	1362
3	2020-03-01 09:00:00	65	1362
4	2020-03-01 09:30:00	271	1362
2550	2020-05-31 18:30:00	268	1362
2551	2020-05-31 19:00:00	173	1362
2552	2020-05-31 19:30:00	136	1362
2553	2020-05-31 20:00:00	23	1362
2554	2020-05-31 20:30:00	80	1362

2555 rows × 3 columns

In [8]: # df_274_time_sale['floor'] = ?? df_274_time_cli['cap'] = 67 df_274_time_cli.reset_index(drop= # df_274_time_cli['floor'] = ??

Out[8]:

ds	У	сар
2020-03-01 07:30:00	1	67
2020-03-01 08:00:00	11	67
2020-03-01 08:30:00	14	67
2020-03-01 09:00:00	9	67
2020-03-01 09:30:00	30	67
2020-05-31 18:30:00	18	67
2020-05-31 19:00:00	9	67
2020-05-31 19:30:00	9	67
2020-05-31 20:00:00	3	67
2020-05-31 20:30:00	7	67
	2020-03-01 07:30:00 2020-03-01 08:00:00 2020-03-01 08:30:00 2020-03-01 09:00:00 2020-03-01 09:30:00 2020-05-31 18:30:00 2020-05-31 19:00:00 2020-05-31 20:00:00	2020-03-01 07:30:00 1 2020-03-01 08:00:00 11 2020-03-01 08:30:00 14 2020-03-01 09:00:00 9 2020-03-01 09:30:00 30 2020-05-31 18:30:00 18 2020-05-31 19:00:00 9 2020-05-31 19:30:00 9 2020-05-31 20:00:00 3

2555 rows × 3 columns

Inicializando e ajustan modelos

Paramêtros:

- growth: linear/logistic
- seasonality: additive/multiplicative
- holidays:
- · changepoints:
- interval_width: assume que o futuro verá taxas de mudança do passado.

Por padrão o Prophet retornará apena observação.

Para obter a incerteza na sazonalidad bayesiana completa. Isso é feito usan padrão é 0)

* mcmc = Markov chain Monte Carlo ı

```
In [9]:
        path = getattr(config, 'path', 'def
        old_stdout = sys.stdout
        sys.stdout = open(path+'/logs/sis
        m 274 time sale = Prophet(growth=
                                   interva
                                   changep
                                   yearly_
                                   weekly_
                                   holiday
        m_274_time_sale.add_country_holid
        m_274_time_sale.fit(df_274_time_s
        m 274 time cli = Prophet(growth='
                                  interval
                                  changepo
                                  yearly_s
                                  weekly s
                                  holidays
        m_274_time_cli.add_country_holida
        m_274_time_cli.fit(df_274_time_cl
```

Out[9]: <fbprophet.forecaster.Prophet at</pre>

Parâmetro growth (saturação)

- Quando a previsão cresce, alguns ponto um teto, isso é chamado de carrying cap mercados, cujos limites podem ser atino possibilidades como novas tecnologias, econômicas, mudanças de hábitos e eto da previsão para que o modelo comport
- É possível definir o carrying capacity (ca do dataframe.

Por padrão o Prophet usa um modelo de s o modelo logistico.

Parâmetro interval_width

 O interval_width de confiança = 95%; is: produzir um intervalo de confiança em to

Parâmetros Trend Changepoints

- Para os dados que estamos a analisar, o momento ou índice de tempo que define mudar sua direção, quer seja crescente de inflexão.
- Prophet usa um entre dois métodos para
 - Especificar a flexibilidade das tendê
 - Especificar o local dos changepoint próprio dataframe como uma série

Parâmetro changepoint_prior_scale

Representa o quão flexível o modelo irá

Parâmetros de sazonalidade

 Definem o período de sazonalidade a se desejamos que o algoritmo considere se um evento) anual e mensal.

Parâmetro holiday

 Os feriados e eventos influenciam no co das pessoas. Para esse estudo foram a conhecidos, entretanto, cabe destacar c municipais que impactam o comércio lo

```
In [10]: m_274_time_sale.train_holiday_nam
Out[10]: 0
                                    parana
         1
                           New Year's Day
         2
                               Tiradentes
         3
                             Worker's Day
         4
                         Independence Day
         5
               Our Lady of the Apparition
         6
                           All Souls' Day
         7
                Republic Proclamation Day
         8
                                Christmas
         dtype: object
```

Criação do dataframe para

- Identificando os valores criados fora do
- Frequência horária
- Cada 30 minutos de cada hora
- Carry Capacity (teto) = 1362 (valor carec
 - Carry Capacity (base) = Não utilizac

In [11]: future_274_time_sale future 274 time sale

future_274_time_sale = m_274_time
future_274_time_sale['cap'] = 136
teste_future_274_time_sale = futu
teste_future_274_time_sale

Out[11]:

	ds	сар
2558	2020-06-01 00:30:00	1362
2559	2020-06-01 01:30:00	1362
2560	2020-06-01 02:30:00	1362
2561	2020-06-01 03:30:00	1362
2562	2020-06-01 04:30:00	1362
4350	2020-08-14 16:30:00	1362
4351	2020-08-14 17:30:00	1362
4352	2020-08-14 18:30:00	1362
4353	2020-08-14 19:30:00	1362
4354	2020-08-14 20:30:00	1362

1797 rows × 2 columns

O prophet está projetando dados para hor isto é, após 20:30 até 07:00, como não há podem impactar nas análises. Dessa form necessário remover do dataframe, projeta

In	[12]:	<pre>future_274_time_sale_adjusted = f</pre>
		<pre>future_274_time_sale_adjusted['ds</pre>
		<pre>future_274_time_sale_adjusted = f</pre>
		<pre>future_274_time_sale_adjusted = f</pre>
		#future_274_time_sale_adjusted.re
		future_274_time_sale_adjusted

Out[12]:

ds		
2020-03-01 07:30:00	2020-03-01 07:30:00	1362
2020-03-01 08:00:00	2020-03-01 08:00:00	1362
2020-03-01 08:30:00	2020-03-01 08:30:00	1362
2020-03-01 09:00:00	2020-03-01 09:00:00	1362
2020-03-01 09:30:00	2020-03-01 09:30:00	1362
2020-08-14 16:30:00	2020-08-14 16:30:00	1362
2020-08-14 17:30:00	2020-08-14 17:30:00	1362
2020-08-14 18:30:00	2020-08-14 18:30:00	1362
2020-08-14 19:30:00	2020-08-14 19:30:00	1362
2020-08-14 20:30:00	2020-08-14 20:30:00	1362

ds

ca

3473 rows × 2 columns

```
In [13]: # Confirmando que os valores fora
teste2_future_274_time_sale = fut
teste2_future_274_time_sale
```

Out[13]:

ds		
2020-06-01 07:30:00	2020-06-01 07:30:00	136
2020-06-01 08:30:00	2020-06-01 08:30:00	136
2020-06-01 09:30:00	2020-06-01 09:30:00	136
2020-06-01 10:30:00	2020-06-01 10:30:00	136
2020-06-01 11:30:00	2020-06-01 11:30:00	136
		-
2020-08-14 16:30:00	2020-08-14 16:30:00	136
2020-08-14 17:30:00	2020-08-14 17:30:00	136
2020-08-14 18:30:00	2020-08-14 18:30:00	136
2020-08-14 19:30:00	2020-08-14 19:30:00	136

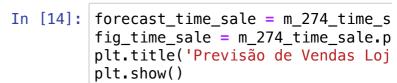
ds

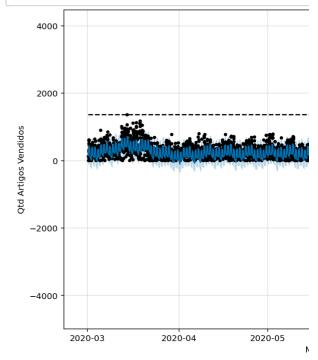
car

1050 rows × 2 columns

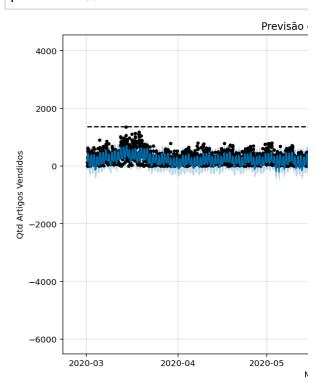
Visualizando o gráfico da ε vendas

2020-08-14 20:30:00 2020-08-14 20:30:00 136%

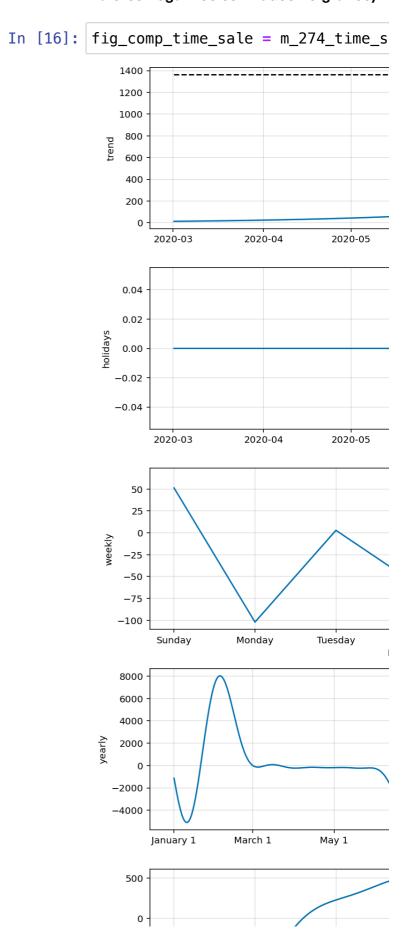


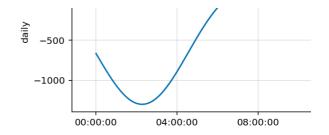


In [15]: # Previsão sem a remoção dos valo
forecast_teste = m_274_time_sale.
fig_time_gap_sale = m_274_time_sa
plt.title('Previsão de Vendas Loj
plt.show()



O dataframe projetado para um futuro cor minutos) passa a projetar valores mais dis em função das incertezas que precisam s uncertainty=True assume uma média da fi verificada no passado, que seja a mesma valores negativos estimados no gráfico).





Curiosamente o componente diário (daily) preciso verificar isso

Verificando a quantidade d previstos

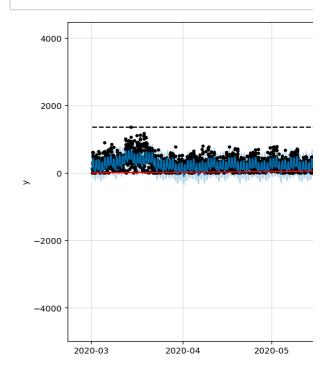
In [17]: #forecast_time_sale = m_274_time_ future_no_art_sales_pred = foreca future_no_art_sales_pred = future future_no_art_sales_pred.sample(5

Out [17]:

	ds	yhat	ył
3340	2020-08-05 14:30:00	-2047.538233	-230
2505	2020-06-06 19:30:00	-246.577136	-49
3379	2020-08-08 11:30:00	-2236.730794	-250
3090	2020-07-18 16:30:00	3782.366027	353
2571	2020-06-11 15:30:00	-1087.363411	-133

Trend Changepoints

In [18]: from fbprophet.plot import add_ch fig_274_sales_changepoint = m_274 a = add_changepoints_to_plot(fig_



ANÁLISE DOS CLIENT

In [20]:

future_274_time_cli = m_274_time_
future_274_time_cli['cap'] = 67
teste_future_274_time_cli = futur
teste_future_274_time_cli

Out [20]:

	ds	сар
2558	2020-06-01 00:30:00	67
2559	2020-06-01 01:30:00	67
2560	2020-06-01 02:30:00	67
2561	2020-06-01 03:30:00	67
2562	2020-06-01 04:30:00	67
4350	2020-08-14 16:30:00	67
4351	2020-08-14 17:30:00	67
4352	2020-08-14 18:30:00	67
4353	2020-08-14 19:30:00	67
4354	2020-08-14 20:30:00	67

1797 rows × 2 columns

O prophet está projetando dados para hoi isto é, após 20:30 até 07:00, como não há podem impactar nas análises. Dessa form necessário remover do dataframe, projeta

In [21]: future_274_time_cli_adjusted = fu future_274_time_cli_adjusted['ds' future_274_time_cli_adjusted = fu future_274_time_cli_adjusted = fu #future_274_time_cli_adjusted.res future_274_time_cli_adjusted

Out [21]:

ds cap

67

ds		
2020-03-01 07:30:00	2020-03-01 07:30:00	67
2020-03-01 08:00:00	2020-03-01 08:00:00	67
2020-03-01 08:30:00	2020-03-01 08:30:00	67
2020-03-01 09:00:00	2020-03-01 09:00:00	67
2020-03-01 09:30:00	2020-03-01 09:30:00	67
2020-08-14 16:30:00	2020-08-14 16:30:00	67
2020-08-14 17:30:00	2020-08-14 17:30:00	67
2020-08-14 18:30:00	2020-08-14 18:30:00	67
2020-08-14 19:30:00	2020-08-14 19:30:00	67

2020-08-14 20:30:00 2020-08-14 20:30:00

3473 rows × 2 columns

In [22]: # Confirmando que os valores fora
teste2_future_274_time_cli = futu
teste2_future_274_time_cli

Out[22]:

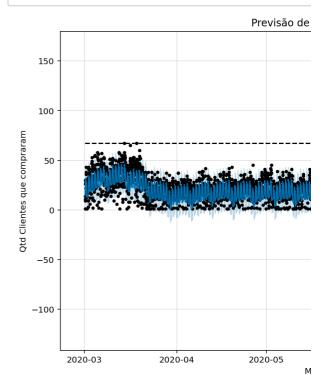
ds cap

ds		
2020-06-01 07:30:00	2020-06-01 07:30:00	67
2020-06-01 08:30:00	2020-06-01 08:30:00	67
2020-06-01 09:30:00	2020-06-01 09:30:00	67
2020-06-01 10:30:00	2020-06-01 10:30:00	67
2020-06-01 11:30:00	2020-06-01 11:30:00	67
2020-08-14 16:30:00	2020-08-14 16:30:00	67
2020-08-14 17:30:00	2020-08-14 17:30:00	67
2020-08-14 18:30:00	2020-08-14 18:30:00	67
2020-08-14 19:30:00	2020-08-14 19:30:00	67
2020-08-14 20:30:00	2020-08-14 20:30:00	67

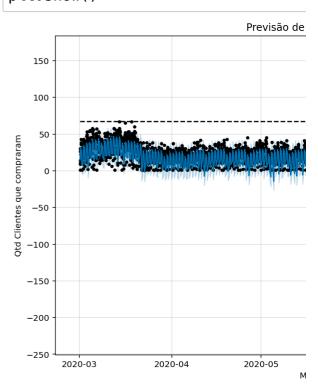
1050 rows × 2 columns

Visualizando o gráfico da ε vendas

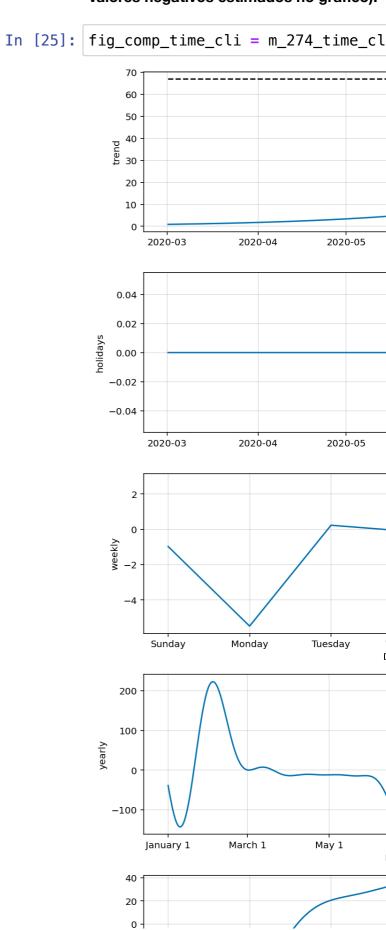
In [23]: forecast_time_cli = m_274_time_cl
fig_time_cli = m_274_time_cli.plo
plt.title('Previsão de Clientes n
plt.show()

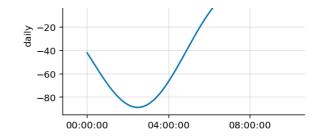


In [24]: # Previsão sem a remoção dos valo
forecast_teste_cli = m_274_time_c
fig_time_gap_cli = m_274_time_cli
plt.title('Previsão de Clientes n
plt.show()



O dataframe projetado para um futuro cor minutos) passa a projetar valores mais dis em função das incertezas que precisam s uncertainty=True assume uma média da fi verificada no passado, que seja a mesma valores negativos estimados no gráfico).





Curiosamente o componente diário (daily) preciso verificar isso

Verificando a quantidade d previstos de comprarem na

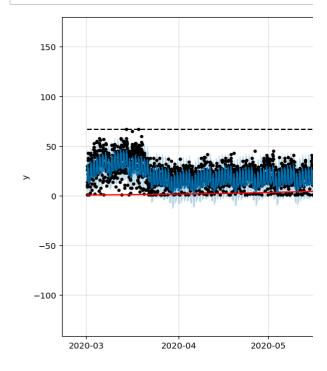
In [26]: #forecast_time_cli = m_274_time_c
future_no_cli_sales_pred = foreca
future_no_cli_sales_pred = future
future_no_cli_sales_pred.sample(5)

Out [26]:

	ds	yhat	yhat_l
2888	2020-07-04 10:30:00	-28.607160	-41.03
2473	2020-06-04 15:30:00	16.336670	3.34
2969	2020-07-10 07:30:00	48.615955	36.09
2834	2020-06-30 12:30:00	-71.806913	-84.09
3004	2020-07-12 14:30:00	88.972830	74.53

Trend Changepoints

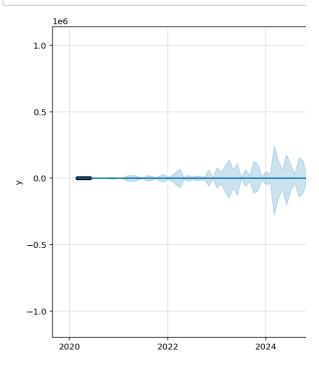
In [27]: from fbprophet.plot import add_ch
fig_274_cli_changepoint = m_274_t
a = add_changepoints_to_plot(fig_



Sales monthly

In [29]: future_month_time_sale = m_month_

In [30]: fcast_m = m_month_time_sale.predi
fig_fcast_m = m_month_time_sale.p



Visualizando os componentes do gráfico

In [31]: | fig_fcast_m_comp = m_month_time_s

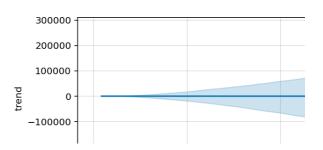
/Users/clonyjr/Library/Mobile Doc/UA/CLONY/Bolsas/Forecast/prophet ophet/plot.py:413: UserWarning: F together with FixedLocator

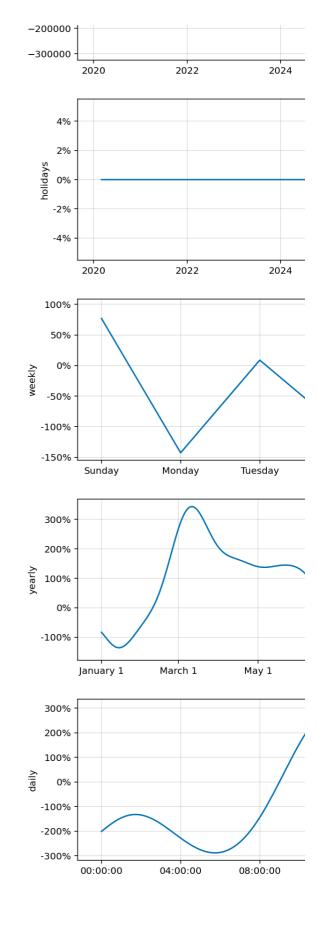
ax.set_yticklabels(yticklabels)
/Users/clonyjr/Library/Mobile Doc
/UA/CLONY/Bolsas/Forecast/prophet
ophet/plot.py:413: UserWarning: F
together with FixedLocator

ax.set_yticklabels(yticklabels)
/Users/clonyjr/Library/Mobile Doc
/UA/CLONY/Bolsas/Forecast/prophet
ophet/plot.py:413: UserWarning: F
together with FixedLocator

ax.set_yticklabels(yticklabels)
/Users/clonyjr/Library/Mobile Doc
/UA/CLONY/Bolsas/Forecast/prophet
ophet/plot.py:413: UserWarning: F
together with FixedLocator

ax.set_yticklabels(yticklabels)





Estimando a quatidade de

In [32]: | future_274_time_cli = m_274_time_

```
In [33]: mean = df_274_time_sale['y'].mean
    stdev = df_274_time_sale['y'].std
    quantile1 = df_274_time_sale['y']
    quantile2 = df_274_time_sale['y']
    iqr = quantile2 - quantile1
    high = mean + stdev
    low = mean - stdev

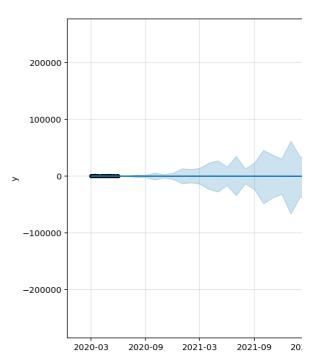
In [34]: # Essa é a definição dos changepo
    # desvio padrão
    df_time_sale_filtered = df_274_ti
    df_time_sale_filtered_changepoint

# Essa é a definição dos changepo
    filtered_iqr = df_274_time_sale[(
```

TESTE PARA OS PARÂ SAZONALIDADE

In [35]: m_274_sale_test = Prophet(interva m_274_sale_test.fit(df_274_time_s future_274_sales_test = m_274_sal forecast_274_sales_test = m_274_s fig_274_sales_test = m_274_sale_t

> INFO:fbprophet:Disabling yearly s ly_seasonality=True to override t



In [17]: m_274_sale_test = Prophet(seasona m_274_sale_test.fit(df_274_time_s future_274_sales_test = m_274_sal forecast_274_sales_test = m_274_s fig_274_sales_test = m_274_sale_t

> INFO:fbprophet:Disabling yearly s ly_seasonality=True to override t

