

In [218... *# 1 - BIBLIOTECAS e PACOTES*

```
import numpy as np
import seaborn as sns
import pandas as pd
from pandas import DataFrame
from pandas.util.testing import assert_frame_equal
from pandas_datareader import data
import matplotlib.pyplot as plt
import datetime
import math
```

In [219... `from scipy import stats`In [220... 

```
import plotly.express as px
import plotly.figure_factory as ff
from copy import copy
```

In [221... 

```
from sklearn.linear_model import LinearRegression
from sklearn.model_selection import train_test_split
```

In [222... 

```
import sklearn.metrics
from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score
```

In [223... `from tensorflow import keras`In [224... *# 2 - PROCESSAMENTO/TRATAMENTO DOS DADOS*In [225... 

```
milho_df = pd.read_excel('Milho-CEPEA-ESALQ.xlsx')
milho_df
```

Out[225...

	INDICADOR DO MILHO ESALQ/BM&FBOVESPA	Unnamed: 1	Unnamed: 2	Unnamed: 3	Unnamed: 4	Unnamed: 5	Unnamed: 6	Unn:
0	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
1	Fonte: Cepea	NaN	NaN	NaN	NaN	NaN	NaN	
2	Data	À vista R\$	À vista US\$	NaN	NaN	NaN	NaN	
3	02/08/2004	18.24	5.98	NaN	NaN	NaN	NaN	
4	03/08/2004	18.04	5.91	NaN	NaN	NaN	NaN	
...	...	...	...	...	...	...	...	
4125	25/02/2021	85.59	15.55	NaN	NaN	NaN	NaN	
4126	26/02/2021	85.41	15.3	NaN	NaN	NaN	NaN	
4127	01/03/2021	85.59	15.29	NaN	NaN	NaN	NaN	
4128	02/03/2021	86.11	15.2	NaN	NaN	NaN	NaN	
4129	03/03/2021	87.06	15.14	NaN	NaN	NaN	NaN	

4130 rows × 10 columns

```
In [226... milho_df = milho_df.drop(milho_df.index[0:3])
milho_df = milho_df.drop(columns=milho_df.columns[3:])
milho_df
```

```
Out[226... INDICADOR DO MILHO ESALQ/BM&FBOVESPA Unnamed: 1 Unnamed: 2
```

	INDICADOR DO MILHO ESALQ/BM&FBOVESPA	Unnamed: 1	Unnamed: 2
3	02/08/2004	18.24	5.98
4	03/08/2004	18.04	5.91
5	04/08/2004	18.02	5.9
6	05/08/2004	18.06	5.89
7	06/08/2004	18.13	5.98
...	...	...	...
4125	25/02/2021	85.59	15.55
4126	26/02/2021	85.41	15.3
4127	01/03/2021	85.59	15.29
4128	02/03/2021	86.11	15.2
4129	03/03/2021	87.06	15.14

4127 rows × 3 columns

```
In [227... milho_df.rename(columns= {'INDICADOR DO MILHO ESALQ/BM&FBOVESPA': 'Data'}, inplace=True)
milho_df.rename(columns= {'Unnamed: 1': 'milho_reais'}, inplace=True)
milho_df.rename(columns= {'Unnamed: 2': 'milho_dolares'}, inplace=True)
milho_df
```

```
Out[227... Data milho_reais milho_dolares
```

	Data	milho_reais	milho_dolares
3	02/08/2004	18.24	5.98
4	03/08/2004	18.04	5.91
5	04/08/2004	18.02	5.9
6	05/08/2004	18.06	5.89
7	06/08/2004	18.13	5.98
...	...	...	...
4125	25/02/2021	85.59	15.55
4126	26/02/2021	85.41	15.3
4127	01/03/2021	85.59	15.29
4128	02/03/2021	86.11	15.2
4129	03/03/2021	87.06	15.14

4127 rows × 3 columns

In [228... milho\_df.info()

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 4127 entries, 3 to 4129
Data columns (total 3 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Data             4127 non-null   object
1   milho_reais      4127 non-null   object
2   milho_dolares    4127 non-null   object
dtypes: object(3)
memory usage: 129.0+ KB
```

```
In [229... milho_df['milho_reais'] = milho_df['milho_reais'].astype(float)
milho_df['milho_dolares'] = milho_df['milho_dolares'].astype(float)
milho_df
```

Out[229... Data milho\_reais milho\_dolares

3	02/08/2004	18.24	5.98
4	03/08/2004	18.04	5.91
5	04/08/2004	18.02	5.90
6	05/08/2004	18.06	5.89
7	06/08/2004	18.13	5.98
...	...	...	...
4125	25/02/2021	85.59	15.55
4126	26/02/2021	85.41	15.30
4127	01/03/2021	85.59	15.29
4128	02/03/2021	86.11	15.20
4129	03/03/2021	87.06	15.14

4127 rows × 3 columns

In [230... milho\_df.info()

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 4127 entries, 3 to 4129
Data columns (total 3 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Data             4127 non-null   object
1   milho_reais      4127 non-null   float64
2   milho_dolares    4127 non-null   float64
dtypes: float64(2), object(1)
memory usage: 129.0+ KB
```

In [231... milho\_df['Data']

```
Out[231... 3      02/08/2004
4      03/08/2004
5      04/08/2004
6      05/08/2004
7      06/08/2004
...
```

```

4125    25/02/2021
4126    26/02/2021
4127    01/03/2021
4128    02/03/2021
4129    03/03/2021
Name: Data, Length: 4127, dtype: object

```

```

In [232...] milho_df['Data'] = pd.to_datetime(milho_df['Data'], dayfirst=True)
milho_df = milho_df.sort_values(by = ['Data'])
milho_df['Data']

```

```

Out[232...] 3    2004-08-02
4    2004-08-03
5    2004-08-04
6    2004-08-05
7    2004-08-06
...
4125    2021-02-25
4126    2021-02-26
4127    2021-03-01
4128    2021-03-02
4129    2021-03-03
Name: Data, Length: 4127, dtype: datetime64[ns]

```

```

In [233...] milho_df.index

```

```

Out[233...] Int64Index([ 3, 4, 5, 6, 7, 8, 9, 10, 11, 12,
...
4120, 4121, 4122, 4123, 4124, 4125, 4126, 4127, 4128, 4129],
dtype='int64', length=4127)

```

```

In [234...] milho_df.index = pd.to_datetime(milho_df.Data)
milho_df.index.to_period('D')
milho_df.index

```

```

Out[234...] DatetimeIndex(['2004-08-02', '2004-08-03', '2004-08-04', '2004-08-05',
...
'2004-08-12', '2004-08-13',
...
'2021-02-18', '2021-02-19', '2021-02-22', '2021-02-23',
'2021-02-24', '2021-02-25', '2021-02-26', '2021-03-01',
'2021-03-02', '2021-03-03'],
dtype='datetime64[ns]', name='Data', length=4127, freq=None)

```

```

In [235...] milho_df.isnull().sum()

```

```

Out[235...] Data      0
milho_reais    0
milho_dolares  0
dtype: int64

```

```

In [236...] ccmfut_df = pd.read_excel('CCMFUT-ProfitChart.xlsx')
ccmfut_df

```

```

Out[236...]

```

	Data	Abertura	Máxima	Mínima	Fechamento	Volume Financeiro
0	2021-03-05	94.10	96.46	94.10	95.85	380377381.5
1	2021-03-04	91.00	94.72	90.40	94.20	325967202.0
2	2021-03-03	89.00	91.10	88.86	91.05	211768164.0
3	2021-03-02	88.80	89.06	88.54	88.94	95795617.5

	Data	Abertura	Máxima	Mínima	Fechamento	Volume Financeiro
<b>4</b>	2021-03-01	88.92	89.18	88.00	88.70	133054398.0
...	...	...	...	...	...	...
<b>2914</b>	2008-10-02	22.64	22.64	22.64	22.64	11295.0
<b>2915</b>	2008-09-30	22.55	22.55	22.55	22.55	112500.0
<b>2916</b>	2008-09-26	22.68	22.68	22.68	22.68	565875.0
<b>2917</b>	2008-09-22	22.64	22.64	22.64	22.64	112950.0
<b>2918</b>	2008-09-19	22.64	22.64	22.64	22.64	1129500.0

2919 rows × 6 columns

```
In [237... ccmfut_df.rename(columns= {'Abertura': 'ccmfut_abertura'}, inplace=True)
ccmfut_df.rename(columns= {'Máxima': 'ccmfut_máxima'}, inplace=True)
ccmfut_df.rename(columns= {'Mínima': 'ccmfut_mínima'}, inplace=True)
ccmfut_df.rename(columns= {'Fechamento': 'ccmfut_fechamento'}, inplace=True)
ccmfut_df.rename(columns= {'Volume Financeiro': 'ccmfut_volume_fin'}, inplace=True)
ccmfut_df
```

```
Out[237... Data ccmfut_abertura ccmfut_máxima ccmfut_mínima ccmfut_fechamento ccmfut_volume_fin
```

<b>0</b>	2021-03-05	94.10	96.46	94.10	95.85	380377381.5
<b>1</b>	2021-03-04	91.00	94.72	90.40	94.20	325967202.0
<b>2</b>	2021-03-03	89.00	91.10	88.86	91.05	211768164.0
<b>3</b>	2021-03-02	88.80	89.06	88.54	88.94	95795617.5
<b>4</b>	2021-03-01	88.92	89.18	88.00	88.70	133054398.0
...	...	...	...	...	...	...
<b>2914</b>	2008-10-02	22.64	22.64	22.64	22.64	11295.0
<b>2915</b>	2008-09-30	22.55	22.55	22.55	22.55	112500.0
<b>2916</b>	2008-09-26	22.68	22.68	22.68	22.68	565875.0
<b>2917</b>	2008-09-22	22.64	22.64	22.64	22.64	112950.0
<b>2918</b>	2008-09-19	22.64	22.64	22.64	22.64	1129500.0

2919 rows × 6 columns

```
In [238... ccmfut_df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2919 entries, 0 to 2918
Data columns (total 6 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Data                   2919 non-null   datetime64[ns]
1   ccmfut_abertura        2919 non-null   float64
2   ccmfut_máxima          2919 non-null   float64
3   ccmfut_mínima          2919 non-null   float64
4   ccmfut_fechamento     2919 non-null   float64
5   ccmfut_volume_fin      2919 non-null   float64
dtypes: datetime64[ns](1), float64(5)
memory usage: 137.0 KB
```

```
In [239... ccmfut_df = ccmfut_df.drop(ccmfut_df.index[0:2])
ccmfut_df
```

```
Out[239...      Data  ccmfut_abertura  ccmfut_máxima  ccmfut_mínima  ccmfut_fechamento  ccmfut_volume_fin
2  2021-03-03          89.00          91.10          88.86          91.05          211768164.0
3  2021-03-02          88.80          89.06          88.54          88.94          95795617.5
4  2021-03-01          88.92          89.18          88.00          88.70          133054398.0
5  2021-02-26          89.61          89.64          88.86          88.86          69200707.5
6  2021-02-25          89.45          89.72          88.81          89.63          106361550.0
...   ...
2914 2008-10-02          22.64          22.64          22.64          22.64          11295.0
2915 2008-09-30          22.55          22.55          22.55          22.55          112500.0
2916 2008-09-26          22.68          22.68          22.68          22.68          565875.0
2917 2008-09-22          22.64          22.64          22.64          22.64          112950.0
2918 2008-09-19          22.64          22.64          22.64          22.64          1129500.0
```

2917 rows × 6 columns

```
In [240... ccmfut_df = ccmfut_df.sort_values(by = ['Data'])
ccmfut_df
```

```
Out[240...      Data  ccmfut_abertura  ccmfut_máxima  ccmfut_mínima  ccmfut_fechamento  ccmfut_volume_fin
2918 2008-09-19          22.64          22.64          22.64          22.64          1129500.0
2917 2008-09-22          22.64          22.64          22.64          22.64          112950.0
```

	Data	ccmfut_abertura	ccmfut_máxima	ccmfut_mínima	ccmfut_fechamento	ccmfut_volume_fin
<b>2916</b>	2008-09-26	22.68	22.68	22.68	22.68	565875.0
<b>2915</b>	2008-09-30	22.55	22.55	22.55	22.55	112500.0
<b>2914</b>	2008-10-02	22.64	22.64	22.64	22.64	11295.0
...	...	...	...	...	...	...
<b>6</b>	2021-02-25	89.45	89.72	88.81	89.63	106361550.0
<b>5</b>	2021-02-26	89.61	89.64	88.86	88.86	69200707.5
<b>4</b>	2021-03-01	88.92	89.18	88.00	88.70	133054398.0
<b>3</b>	2021-03-02	88.80	89.06	88.54	88.94	95795617.5
<b>2</b>	2021-03-03	89.00	91.10	88.86	91.05	211768164.0

2917 rows × 6 columns

In [241...] `ccmfut_df.index`

Out[241...] `Int64Index([2918, 2917, 2916, 2915, 2914, 2913, 2912, 2911, 2910, 2909, ...  
11, 10, 9, 8, 7, 6, 5, 4, 3, 2],  
dtype='int64', length=2917)`

In [242...] `ccmfut_df.index = pd.to_datetime(ccmfut_df.Data)  
ccmfut_df.index.to_period('D')  
ccmfut_df.index`

Out[242...] `DatetimeIndex(['2008-09-19', '2008-09-22', '2008-09-26', '2008-09-30',  
'2008-10-02', '2008-10-03', '2008-10-06', '2008-10-07',  
'2008-10-08', '2008-10-09',  
...  
'2021-02-18', '2021-02-19', '2021-02-22', '2021-02-23',  
'2021-02-24', '2021-02-25', '2021-02-26', '2021-03-01',  
'2021-03-02', '2021-03-03'],  
dtype='datetime64[ns]', name='Data', length=2917, freq=None)`

In [243...] `ccmfut_df.isnull().sum()`

Out[243...] `Data 0  
ccmfut_abertura 0  
ccmfut_máxima 0  
ccmfut_mínima 0  
ccmfut_fechamento 0  
ccmfut_volume_fin 0  
dtype: int64`

In [244...] `milho_df.rename(columns= {'Data': 'data'}, inplace=True)  
ccmfut_df.rename(columns= {'Data': 'data'}, inplace=True)`

In [245...

milho\_df

Out[245...

	data	milho_reais	milho_dolares
Data			
<b>2004-08-02</b>	2004-08-02	18.24	5.98
<b>2004-08-03</b>	2004-08-03	18.04	5.91
<b>2004-08-04</b>	2004-08-04	18.02	5.90
<b>2004-08-05</b>	2004-08-05	18.06	5.89
<b>2004-08-06</b>	2004-08-06	18.13	5.98
...	...	...	...
<b>2021-02-25</b>	2021-02-25	85.59	15.55
<b>2021-02-26</b>	2021-02-26	85.41	15.30
<b>2021-03-01</b>	2021-03-01	85.59	15.29
<b>2021-03-02</b>	2021-03-02	86.11	15.20
<b>2021-03-03</b>	2021-03-03	87.06	15.14

4127 rows × 3 columns

In [246...

ccmfut\_df

Out[246...

	data	ccmfut_abertura	ccmfut_máxima	ccmfut_mínima	ccmfut_fechamento	ccmfut_volume_fin
Data						
<b>2008-09-19</b>	2008-09-19	22.64	22.64	22.64	22.64	1129500.0
<b>2008-09-22</b>	2008-09-22	22.64	22.64	22.64	22.64	112950.0
<b>2008-09-26</b>	2008-09-26	22.68	22.68	22.68	22.68	565875.0
<b>2008-09-30</b>	2008-09-30	22.55	22.55	22.55	22.55	112500.0
<b>2008-10-02</b>	2008-10-02	22.64	22.64	22.64	22.64	11295.0
...	...	...	...	...	...	...
<b>2021-02-25</b>	2021-02-25	89.45	89.72	88.81	89.63	106361550.0
<b>2021-02-26</b>	2021-02-26	89.61	89.64	88.86	88.86	69200707.5
<b>2021-03-01</b>	2021-03-01	88.92	89.18	88.00	88.70	133054398.0
<b>2021-03-02</b>	2021-03-02	88.80	89.06	88.54	88.94	95795617.5



	data	ccmfut_abertura	ccmfut_máxima	ccmfut_mínima	ccmfut_fechamento	ccmfut_volume_fin
<b>Data</b>						
<b>2021-03-03</b>	2021-03-03	89.00	91.10	88.86	91.05	211768164.0

2917 rows × 6 columns



```
In [247... mc_df = ccmfut_df.merge(
            milho_df.set_index('data'), how='left', on='data'
        )
```

```
In [248... mc_df
```

	data	ccmfut_abertura	ccmfut_máxima	ccmfut_mínima	ccmfut_fechamento	ccmfut_volume_fin
<b>0</b>	2008-09-19	22.64	22.64	22.64	22.64	1129500.0
<b>1</b>	2008-09-22	22.64	22.64	22.64	22.64	112950.0
<b>2</b>	2008-09-26	22.68	22.68	22.68	22.68	565875.0
<b>3</b>	2008-09-30	22.55	22.55	22.55	22.55	112500.0
<b>4</b>	2008-10-02	22.64	22.64	22.64	22.64	11295.0
...	...	...	...	...	...	...
<b>2912</b>	2021-02-25	89.45	89.72	88.81	89.63	106361550.0
<b>2913</b>	2021-02-26	89.61	89.64	88.86	88.86	69200707.5
<b>2914</b>	2021-03-01	88.92	89.18	88.00	88.70	133054398.0
<b>2915</b>	2021-03-02	88.80	89.06	88.54	88.94	95795617.5
<b>2916</b>	2021-03-03	89.00	91.10	88.86	91.05	211768164.0

2917 rows × 8 columns



```
In [249... mc_df.isnull().sum()
```

```
Out[249... data          0
ccmfut_abertura    0
ccmfut_máxima      0
ccmfut_mínima      0
ccmfut_fechamento 0
```

```
ccmfut_volume_fin    0
milho_reais          2
milho_dolares        2
dtype: int64
```

```
In [250... mc_df_mvmilho_reais = mc_df["milho_reais"].rolling(5).mean().shift(-5).round(0)
mc_df_mvmilho_dolares = mc_df["milho_dolares"].rolling(5).mean().shift(-5).round(0)
mc_df["milho_reais"].fillna(mc_df_mvmilho_reais, inplace=True)
mc_df["milho_dolares"].fillna(mc_df_mvmilho_dolares, inplace=True)
```

```
In [251... mc_df.isnull().sum()
```

```
Out[251... data                0
ccmfut_abertura          0
ccmfut_máxima            0
ccmfut_mínima            0
ccmfut_fechamento       0
ccmfut_volume_fin        0
milho_reais              1
milho_dolares            1
dtype: int64
```

```
In [252... mc_df_mvmilho_reais = mc_df["milho_reais"].rolling(5).mean().shift(-5).round(0)
mc_df_mvmilho_dolares = mc_df["milho_dolares"].rolling(5).mean().shift(-5).round(0)
mc_df["milho_reais"].fillna(mc_df_mvmilho_reais, inplace=True)
mc_df["milho_dolares"].fillna(mc_df_mvmilho_dolares, inplace=True)
```

```
In [253... mc_df.isnull().sum()
```

```
Out[253... data                0
ccmfut_abertura          0
ccmfut_máxima            0
ccmfut_mínima            0
ccmfut_fechamento       0
ccmfut_volume_fin        0
milho_reais              0
milho_dolares            0
dtype: int64
```

```
In [254... mc_df = mc_df.sort_values(by = ['data'])
mc_df
```

```
Out[254...
```

	data	ccmfut_abertura	ccmfut_máxima	ccmfut_mínima	ccmfut_fechamento	ccmfut_volume_fin
0	2008-09-19	22.64	22.64	22.64	22.64	1129500.0
1	2008-09-22	22.64	22.64	22.64	22.64	112950.0
2	2008-09-26	22.68	22.68	22.68	22.68	565875.0
3	2008-09-30	22.55	22.55	22.55	22.55	112500.0
4	2008-10-02	22.64	22.64	22.64	22.64	11295.0
...	...	...	...	...	...	...

	data	ccmfut_abertura	ccmfut_máxima	ccmfut_mínima	ccmfut_fechamento	ccmfut_volume_fin
<b>2912</b>	2021-02-25	89.45	89.72	88.81	89.63	106361550.0
<b>2913</b>	2021-02-26	89.61	89.64	88.86	88.86	69200707.5
<b>2914</b>	2021-03-01	88.92	89.18	88.00	88.70	133054398.0
<b>2915</b>	2021-03-02	88.80	89.06	88.54	88.94	95795617.5
<b>2916</b>	2021-03-03	89.00	91.10	88.86	91.05	211768164.0

2917 rows × 8 columns



```
In [255... mc_df.index = pd.to_datetime(mc_df.data)
mc_df.index.to_period('D')
mc_df.index
```

```
Out[255... DatetimeIndex(['2008-09-19', '2008-09-22', '2008-09-26', '2008-09-30',
                    '2008-10-02', '2008-10-03', '2008-10-06', '2008-10-07',
                    '2008-10-08', '2008-10-09',
                    ...,
                    '2021-02-18', '2021-02-19', '2021-02-22', '2021-02-23',
                    '2021-02-24', '2021-02-25', '2021-02-26', '2021-03-01',
                    '2021-03-02', '2021-03-03'],
                    dtype='datetime64[ns]', name='data', length=2917, freq=None)
```

```
In [256... mc_df
```

	data	ccmfut_abertura	ccmfut_máxima	ccmfut_mínima	ccmfut_fechamento	ccmfut_volume_fin
<b>data</b>						
<b>2008-09-19</b>	2008-09-19	22.64	22.64	22.64	22.64	1129500.0
<b>2008-09-22</b>	2008-09-22	22.64	22.64	22.64	22.64	112950.0
<b>2008-09-26</b>	2008-09-26	22.68	22.68	22.68	22.68	565875.0
<b>2008-09-30</b>	2008-09-30	22.55	22.55	22.55	22.55	112500.0
<b>2008-10-02</b>	2008-10-02	22.64	22.64	22.64	22.64	11295.0
...	...	...	...	...	...	...
<b>2021-02-25</b>	2021-02-25	89.45	89.72	88.81	89.63	106361550.0
<b>2021-02-26</b>	2021-02-26	89.61	89.64	88.86	88.86	69200707.5

	data	ccmfut_abertura	ccmfut_máxima	ccmfut_mínima	ccmfut_fechamento	ccmfut_volume_fin
<b>data</b>						
<b>2021-03-01</b>	2021-03-01	88.92	89.18	88.00	88.70	133054398.0
<b>2021-03-02</b>	2021-03-02	88.80	89.06	88.54	88.94	95795617.5
<b>2021-03-03</b>	2021-03-03	89.00	91.10	88.86	91.05	211768164.0

2917 rows × 8 columns



In [257...] *# 3 - Analise e Exploracao dos Dados*

In [258...] `milho_df.describe()`

Out[258...] 

	milho_reais	milho_dolares
<b>count</b>	4127.000000	4127.000000
<b>mean</b>	30.409537	11.535544
<b>std</b>	12.350048	3.276651
<b>min</b>	13.320000	5.890000
<b>25%</b>	21.325000	9.250000
<b>50%</b>	27.770000	10.720000
<b>75%</b>	35.375000	13.750000
<b>max</b>	87.060000	19.960000

In [259...] `milho_df[milho_df['milho_reais']==milho_df['milho_reais'].max()]`

Out[259...] 

	data	milho_reais	milho_dolares
--	------	-------------	---------------

<b>Data</b>			
<b>2021-03-03</b>	2021-03-03	87.06	15.14

In [260...] `milho_df[milho_df['milho_dolares']==milho_df['milho_dolares'].max()]`

Out[260...] 

	data	milho_reais	milho_dolares
--	------	-------------	---------------

<b>Data</b>			
<b>2011-07-01</b>	2011-07-01	31.08	19.96

In [261...] `milho_df[milho_df['milho_reais']==milho_df['milho_reais'].min()]`

Out[261...] 

	data	milho_reais	milho_dolares
--	------	-------------	---------------

Data	data	milho_reais	milho_dolares
Data			
2006-03-30	2006-03-30	13.32	6.08

In [262... milho\_df[milho\_df['milho\_dolares']==milho\_df['milho\_dolares'].min()]]

Out[262... data milho\_reais milho\_dolares

Data			
2004-08-05	2004-08-05	18.06	5.89

In [263... plt.figure(figsize=(10,7))  
 sns.set\_context('notebook', font\_scale=1.5, rc={'font.size':20, 'axes.titlesize':20, 'a  
 sns.distplot(milho\_df['milho\_reais'], rug=True, color= 'green')  
 sns.set\_style('darkgrid')  
 plt.title('Distribuição do Preço do Milho em Reais')

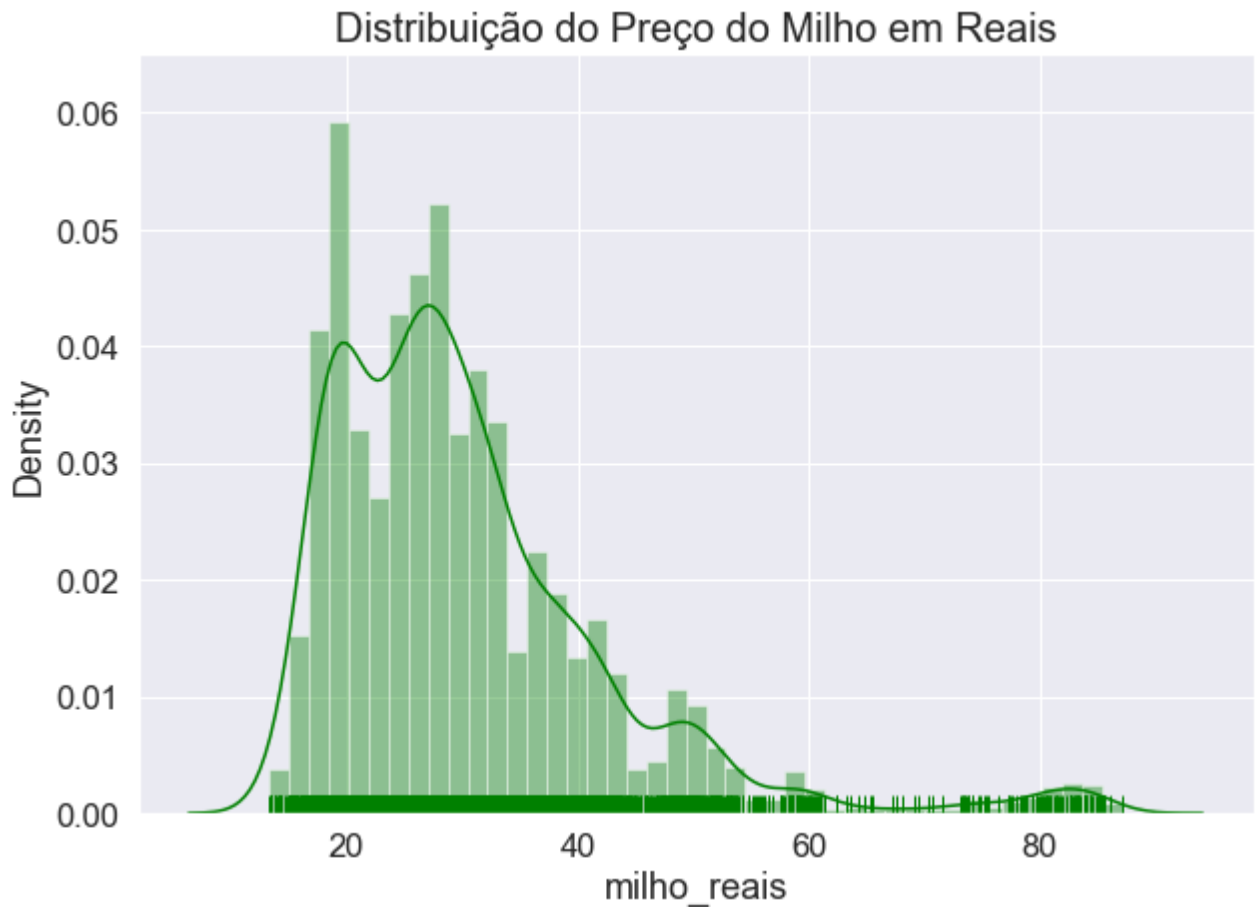
C:\Users\Jonathan Lincher\anaconda3\lib\site-packages\seaborn\distributions.py:2551: FutureWarning:

`distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

C:\Users\Jonathan Lincher\anaconda3\lib\site-packages\seaborn\distributions.py:2055: FutureWarning:

The `axis` variable is no longer used and will be removed. Instead, assign variables directly to `x` or `y`.

Out[263... Text(0.5, 1.0, 'Distribuição do Preço do Milho em Reais')



```
In [264... plt.figure(figsize=(10,7))
sns.set_context('notebook', font_scale=1.5, rc={'font.size':20, 'axes.titlesize':20, 'a
sns.distplot(milho_df['milho_dolares'], rug=True, color= 'green')
sns.set_style('darkgrid')
plt.title('Distribuição do Preço do Milho em Dolares')
```

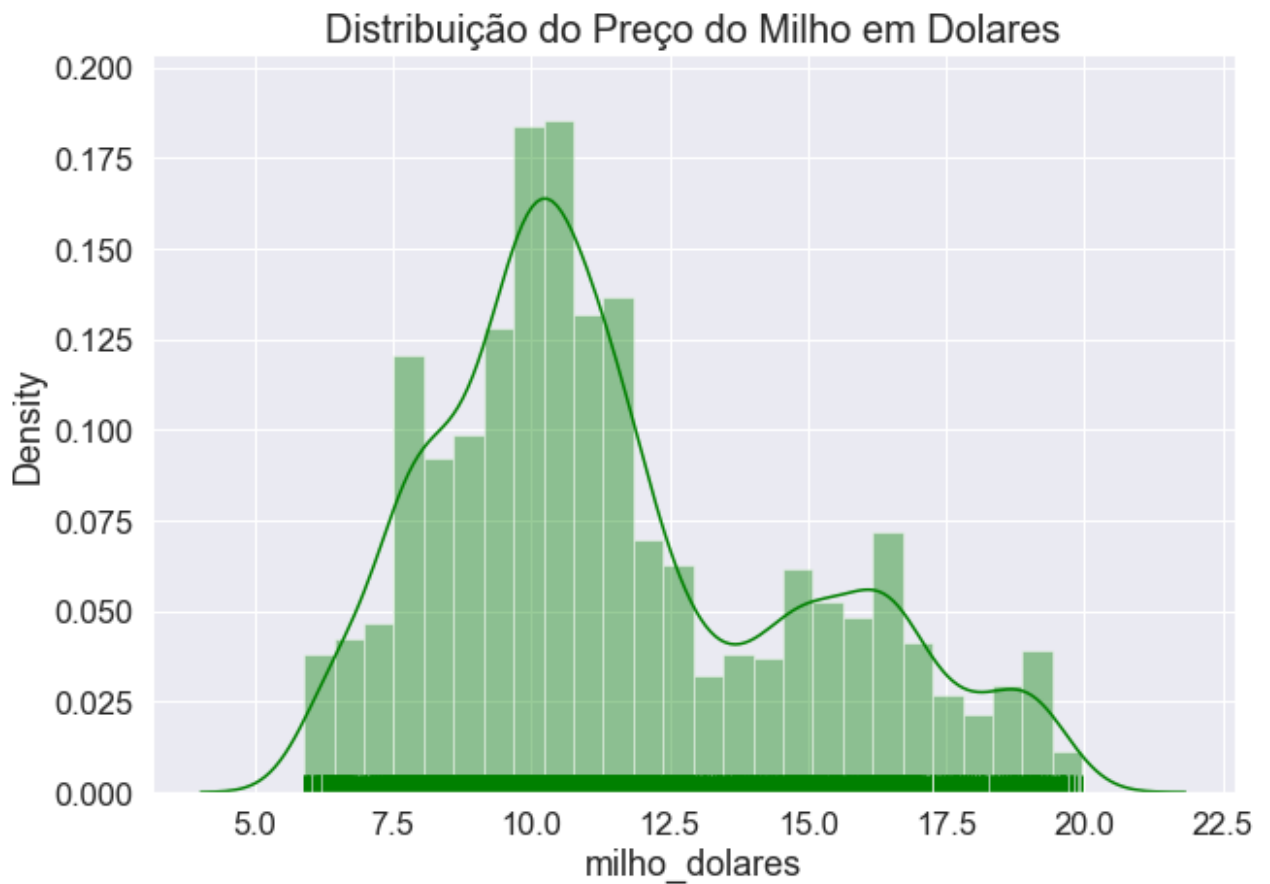
C:\Users\Jonathan Lincher\anaconda3\lib\site-packages\seaborn\distributions.py:2551: FutureWarning:

`distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

C:\Users\Jonathan Lincher\anaconda3\lib\site-packages\seaborn\distributions.py:2055: FutureWarning:

The `axis` variable is no longer used and will be removed. Instead, assign variables directly to `x` or `y`.

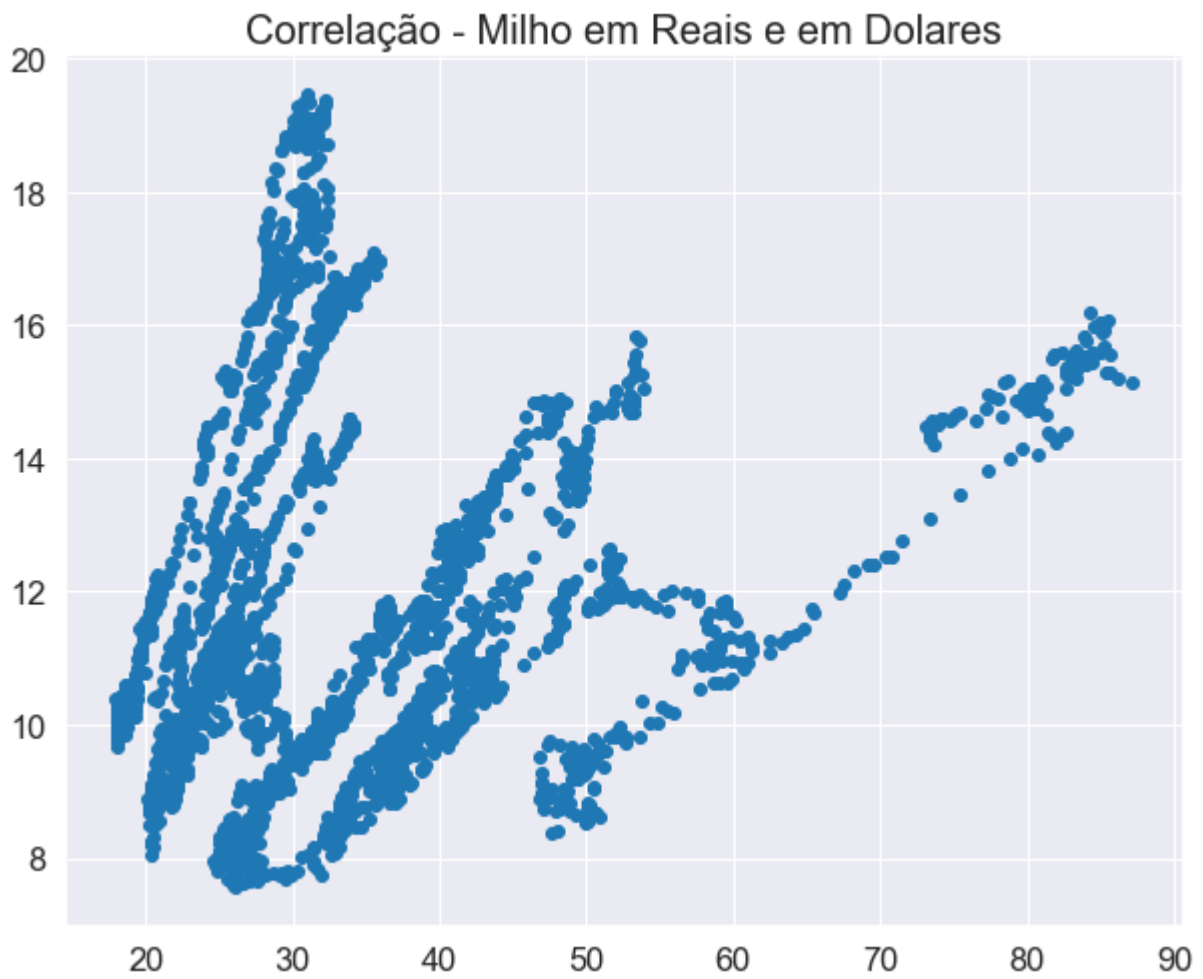
```
Out[264... Text(0.5, 1.0, 'Distribuição do Preço do Milho em Dolares')
```



```
In [265...] mc_df['milho_reais'].corr(mc_df['milho_dolares'])
```

```
Out[265...] 0.1816166279649749
```

```
In [266...] data1=mc_df['milho_reais']  
data2=mc_df['milho_dolares']  
plt.scatter(data1, data2)  
plt.title('Correlação - Milho em Reais e em Dolares')  
plt.gcf().set_size_inches(10, 8)  
plt.show()
```



```
In [267... def interactive_plot(df, title):  
    fig = px.line(title = title)  
    for i in df.columns[1:]:  
        fig.add_scatter(x = df['data'], y = df[i], name = i)  
    fig.show()
```

```
In [268... interactive_plot(milho_df, "Milho em Reais e Milho em Dolares")
```



```
In [269...] dolar_df = pd.read_csv('USD_BRL Dados Históricos.csv', decimal=",")
dolar_df
```

```
Out[269...]
      Data  Último  Abertura  Máxima  Mínima  Var%
0  03.03.2021  5.6193    5.6872    5.7729    5.5806  -1,01%
1  02.03.2021  5.6764    5.6386    5.7327    5.6386   0,61%
2  01.03.2021  5.6418    5.5870    5.6427    5.5553   0,77%
3  26.02.2021  5.5986    5.5340    5.6093    5.4905   1,23%
4  25.02.2021  5.5308    5.4450    5.5390    5.4173   2,30%
...      ...      ...      ...      ...      ...      ...
4319 06.08.2004  3.0330    3.0722    3.0780    3.0300  -1,25%
4320 05.08.2004  3.0713    3.0540    3.0713    3.0500   0,58%
4321 04.08.2004  3.0537    3.0500    3.0660    3.0460   0,12%
4322 03.08.2004  3.0500    3.0450    3.0620    3.0440   0,11%
4323 02.08.2004  3.0465    3.0365    3.0585    3.0365   0,30%
```

4324 rows × 6 columns

```
In [270...] dolar_df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4324 entries, 0 to 4323
Data columns (total 6 columns):
#   Column      Non-Null Count  Dtype
---  -
0   Data        4324 non-null   object
1   Último      4324 non-null   float64
2   Abertura    4324 non-null   float64
3   Máxima      4324 non-null   float64
4   Mínima      4324 non-null   float64
5   Var%        4324 non-null   object
dtypes: float64(4), object(2)
memory usage: 202.8+ KB
```

```
In [271...] dolar_df = dolar_df.drop(columns=dolar_df.columns[2:])
dolar_df
```

```
Out[271...]
      Data  Último
```

	Data	Último
0	03.03.2021	5.6193
1	02.03.2021	5.6764
2	01.03.2021	5.6418
3	26.02.2021	5.5986
4	25.02.2021	5.5308
...	...	...
4319	06.08.2004	3.0330
4320	05.08.2004	3.0713
4321	04.08.2004	3.0537
4322	03.08.2004	3.0500
4323	02.08.2004	3.0465

4324 rows × 2 columns

```
In [272...] dolar_df['Data'] = pd.to_datetime(dolar_df['Data'], dayfirst=True)
dolar_df = dolar_df.sort_values(by = ['Data'])
dolar_df['Data']
```

```
Out[272...] 4323    2004-08-02
4322    2004-08-03
4321    2004-08-04
4320    2004-08-05
4319    2004-08-06
...
4        2021-02-25
3        2021-02-26
2        2021-03-01
1        2021-03-02
0        2021-03-03
Name: Data, Length: 4324, dtype: datetime64[ns]
```

```
In [273...] dolar_df.index = pd.to_datetime(dolar_df.Data)
dolar_df.index.to_period('D')
dolar_df.index
```

```
Out[273...] DatetimeIndex(['2004-08-02', '2004-08-03', '2004-08-04', '2004-08-05',
                          '2004-08-06', '2004-08-09', '2004-08-10', '2004-08-11',
                          '2004-08-12', '2004-08-13',
                          ...,
                          '2021-02-18', '2021-02-19', '2021-02-22', '2021-02-23',
                          '2021-02-24', '2021-02-25', '2021-02-26', '2021-03-01',
                          '2021-03-02', '2021-03-03'],
                          dtype='datetime64[ns]', name='Data', length=4324, freq=None)
```

```
In [274...] dolar_df.rename(columns= {'Data': 'data'}, inplace=True)
dolar_df.rename(columns= {'Último': 'Dolar_Último'}, inplace=True)
```

```
In [275...] dolar_df
```

```
Out[275...]      data  Dolar_Último
```

Data	data	Dolar_Último
Data		
2004-08-02	2004-08-02	3.0465
2004-08-03	2004-08-03	3.0500
2004-08-04	2004-08-04	3.0537
2004-08-05	2004-08-05	3.0713
2004-08-06	2004-08-06	3.0330
...	...	...
2021-02-25	2021-02-25	5.5308
2021-02-26	2021-02-26	5.5986
2021-03-01	2021-03-01	5.6418
2021-03-02	2021-03-02	5.6764
2021-03-03	2021-03-03	5.6193

4324 rows × 2 columns

In [276...] milho\_df

Out[276...] data milho\_reais milho\_dolares

Data			
Data			
2004-08-02	2004-08-02	18.24	5.98
2004-08-03	2004-08-03	18.04	5.91
2004-08-04	2004-08-04	18.02	5.90
2004-08-05	2004-08-05	18.06	5.89
2004-08-06	2004-08-06	18.13	5.98
...	...	...	...
2021-02-25	2021-02-25	85.59	15.55
2021-02-26	2021-02-26	85.41	15.30
2021-03-01	2021-03-01	85.59	15.29
2021-03-02	2021-03-02	86.11	15.20
2021-03-03	2021-03-03	87.06	15.14

4127 rows × 3 columns

In [277...] dolar\_milho\_df = pd.merge(milho\_df,dolar\_df, how='inner', on=['Data'],suffixes=('\_M', '\_D'),dolar\_milho\_df

Out[277...] data\_M milho\_reais milho\_dolares data\_D Dolar\_Último

Data	data_M	milho_reais	milho_dolares	data_D	Dolar_Último
Data					
2004-08-02	2004-08-02	18.24	5.98	2004-08-02	3.0465
2004-08-03	2004-08-03	18.04	5.91	2004-08-03	3.0500
2004-08-04	2004-08-04	18.02	5.90	2004-08-04	3.0537
2004-08-05	2004-08-05	18.06	5.89	2004-08-05	3.0713
2004-08-06	2004-08-06	18.13	5.98	2004-08-06	3.0330
...	...	...	...	...	...
2021-02-25	2021-02-25	85.59	15.55	2021-02-25	5.5308
2021-02-26	2021-02-26	85.41	15.30	2021-02-26	5.5986
2021-03-01	2021-03-01	85.59	15.29	2021-03-01	5.6418
2021-03-02	2021-03-02	86.11	15.20	2021-03-02	5.6764
2021-03-03	2021-03-03	87.06	15.14	2021-03-03	5.6193

4127 rows × 5 columns

```
In [278... dolar_milho_df = dolar_milho_df.drop(columns='data_D')
dolar_milho_df
```

```
Out[278...
```

	data_M	milho_reais	milho_dolares	Dolar_Último
Data				
2004-08-02	2004-08-02	18.24	5.98	3.0465
2004-08-03	2004-08-03	18.04	5.91	3.0500
2004-08-04	2004-08-04	18.02	5.90	3.0537
2004-08-05	2004-08-05	18.06	5.89	3.0713
2004-08-06	2004-08-06	18.13	5.98	3.0330
...	...	...	...	...
2021-02-25	2021-02-25	85.59	15.55	5.5308
2021-02-26	2021-02-26	85.41	15.30	5.5986
2021-03-01	2021-03-01	85.59	15.29	5.6418
2021-03-02	2021-03-02	86.11	15.20	5.6764
2021-03-03	2021-03-03	87.06	15.14	5.6193

4127 rows × 4 columns

```
In [279... dolar_milho_df['milho_reais'].corr(dolar_milho_df['Dolar_Último'])
```

```
Out[279... 0.7824021362013128
```

In [280... `ccmfut_df.describe()`

Out[280...

	<code>ccmfut_abertura</code>	<code>ccmfut_máxima</code>	<code>ccmfut_mínima</code>	<code>ccmfut_fechamento</code>	<code>ccmfut_volume_fin</code>
<b>count</b>	2917.000000	2917.000000	2917.000000	2917.000000	2.917000e+03
<b>mean</b>	26.723809	26.988920	26.474443	26.742749	3.300606e+07
<b>std</b>	13.506372	13.711821	13.312827	13.540517	4.274579e+07
<b>min</b>	12.680000	12.800000	12.310000	12.710000	0.000000e+00
<b>25%</b>	19.620000	19.780000	19.500000	19.640000	1.002070e+07
<b>50%</b>	22.560000	22.760000	22.350000	22.560000	2.244742e+07
<b>75%</b>	30.140000	30.510000	29.820000	30.140000	3.881796e+07
<b>max</b>	89.730000	91.120000	89.280000	91.050000	6.528013e+08

In [281... `plt.figure(figsize=(10,7))`  
`sns.set_context('notebook', font_scale=1.5, rc={'font.size':20, 'axes.titlesize':20, 'a`  
`sns.distplot(ccmfut_df['ccmfut_fechamento'], rug=True, color='green')`  
`sns.set_style('darkgrid')`  
`plt.title('Distribuição do Preço de Fechamento do CCMFUT')`

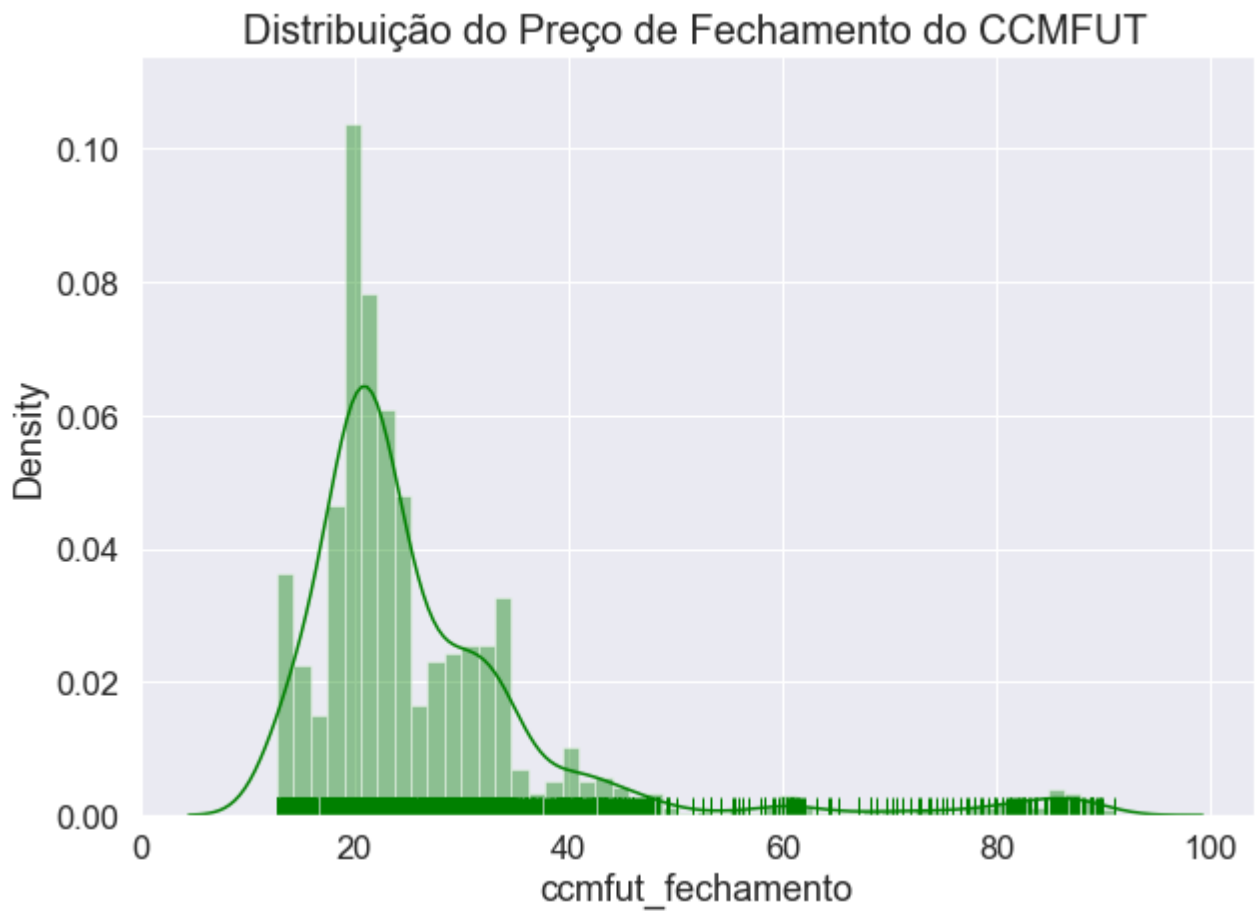
C:\Users\Jonathan Lincher\anaconda3\lib\site-packages\seaborn\distributions.py:2551: FutureWarning:

`distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

C:\Users\Jonathan Lincher\anaconda3\lib\site-packages\seaborn\distributions.py:2055: FutureWarning:

The `axis` variable is no longer used and will be removed. Instead, assign variables directly to `x` or `y`.

Out[281... Text(0.5, 1.0, 'Distribuição do Preço de Fechamento do CCMFUT')



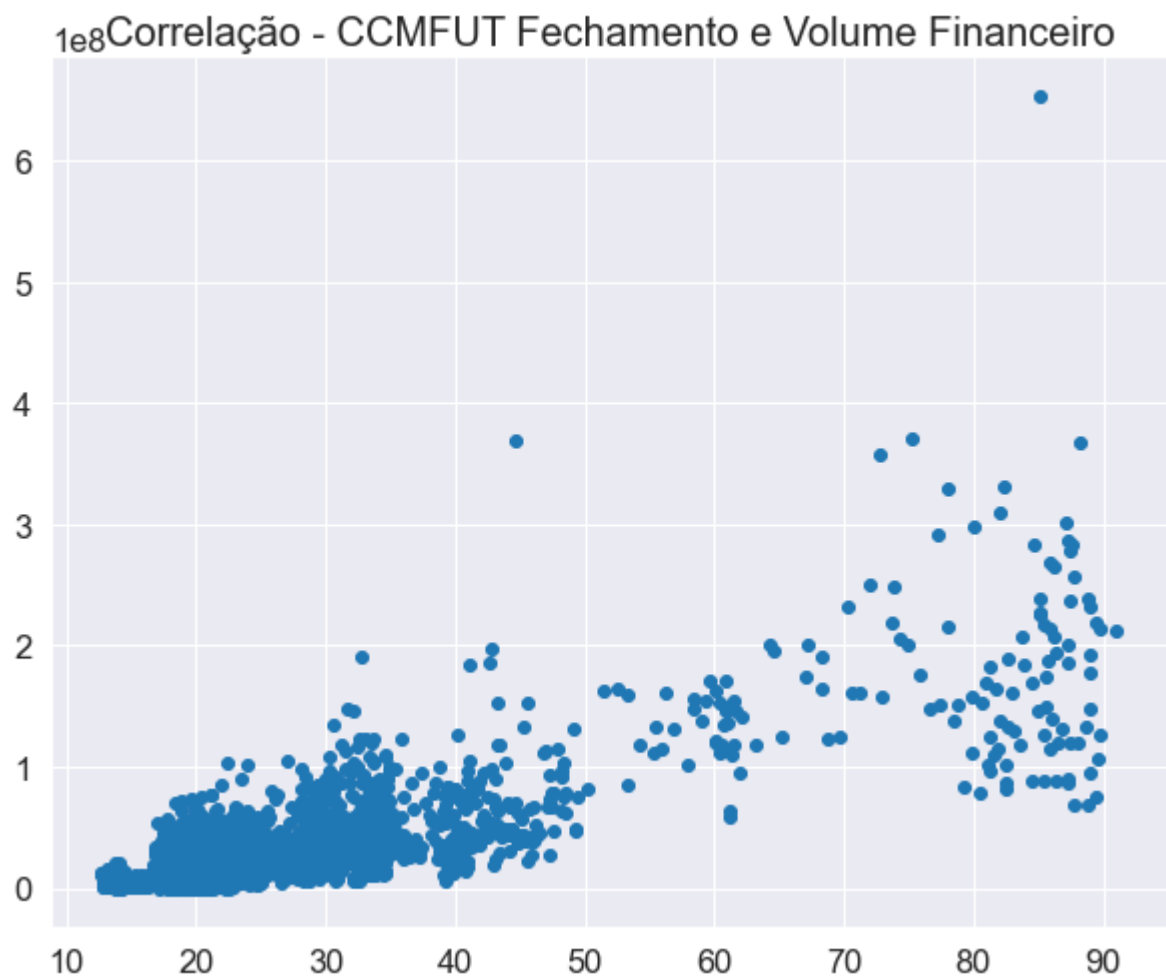
```
In [282... def interactive_plot(df, title):  
    fig = px.line(title = title)  
    for i in df.columns[1:]:  
        fig.add_scatter(x = df['data'], y = df[i], name = i)  
    fig.show()
```

```
In [283... interactive_plot(ccmfut_df, "Historico de Precos - CCMFUT")
```

```
In [284... ccmfut_df['ccmfut_fechamento'].corr(ccmfut_df['ccmfut_volume_fin'])
```

```
Out[284... 0.7999203525936038
```

```
In [285... data1=ccmfut_df['ccmfut_fechamento']  
data2=ccmfut_df['ccmfut_volume_fin']  
plt.scatter(data1, data2)  
plt.title('Correlação - CCMFUT Fechamento e Volume Financeiro')  
plt.gcf().set_size_inches(10, 8)  
plt.show()
```



```
In [286... b = ccmfut_df["data"]  
data1 = ccmfut_df["ccmfut_fechamento"]  
data2 = ccmfut_df["ccmfut_volume_fin"]  
  
fig, ax1 = plt.subplots()
```

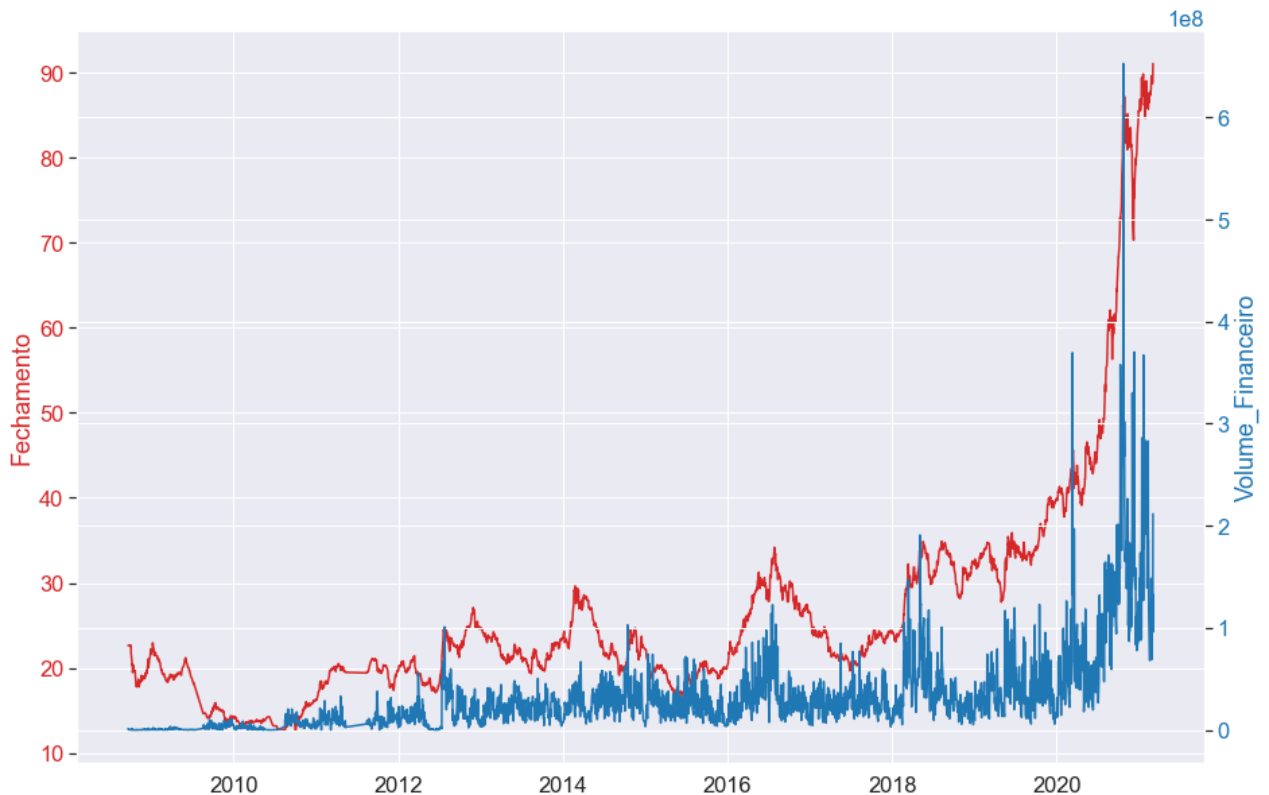
```

color = 'tab:red'
ax1.set_xlabel('')
ax1.set_ylabel('Fechamento', color=color)
ax1.plot(b, data1, color=color)
ax1.tick_params(axis='y', labelcolor=color)

ax2 = ax1.twinx()

color = 'tab:blue'
ax2.set_ylabel('Volume_Financeiro', color=color)
ax2.plot(b, data2, color=color)
ax2.tick_params(axis='y', labelcolor=color)
plt.gcf().set_size_inches(15, 10)
plt.show()

```



```
In [287...] mc_df['milho_reais'].corr(mc_df['ccmfut_fechamento'])
```

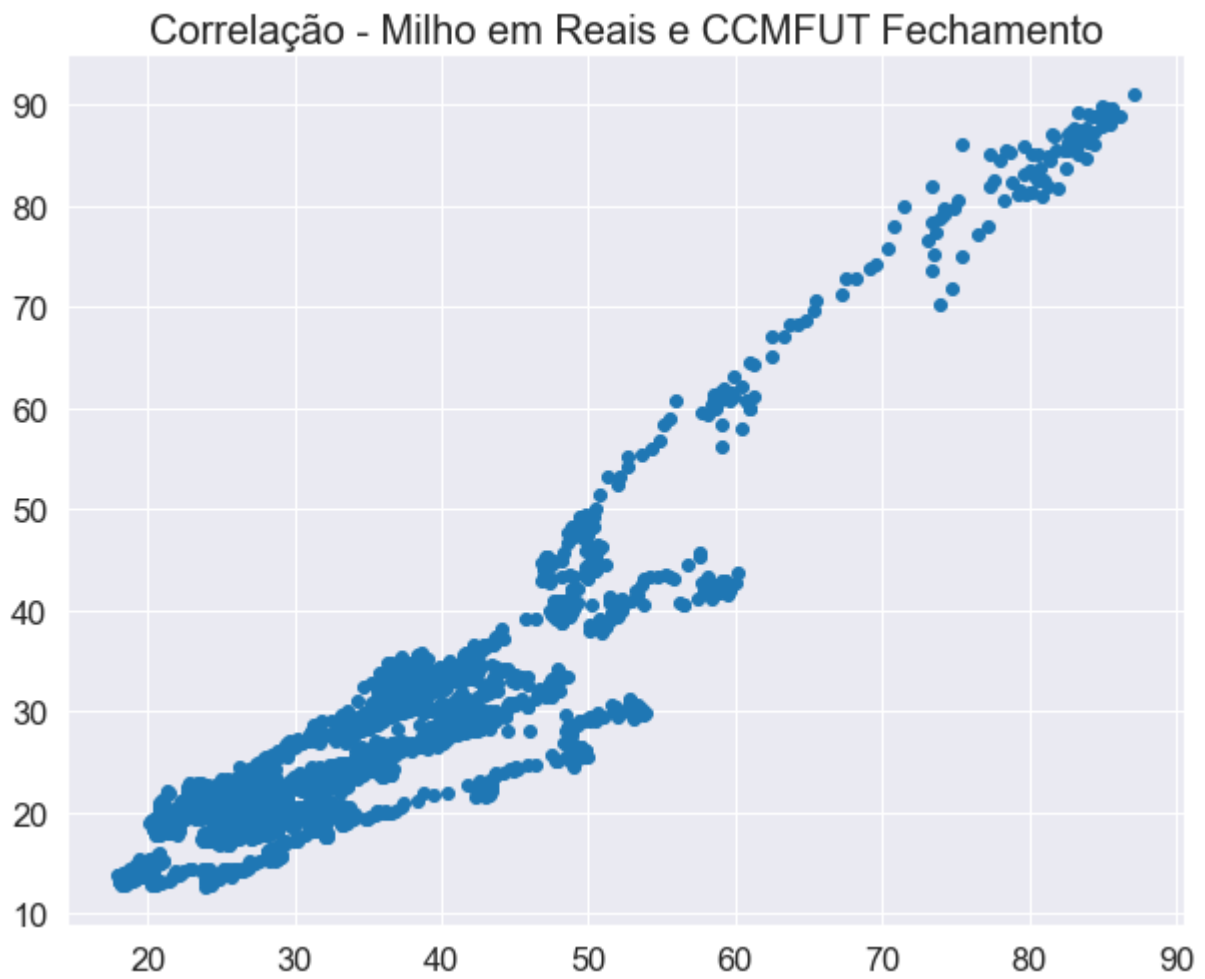
```
Out[287...] 0.928074900682255
```

```

In [288...] data1=mc_df['milho_reais']
data2=mc_df['ccmfut_fechamento']
plt.scatter(data1, data2)
plt.title('Correlação - Milho em Reais e CCMFUT Fechamento')
plt.gcf().set_size_inches(10, 8)
plt.show()

```



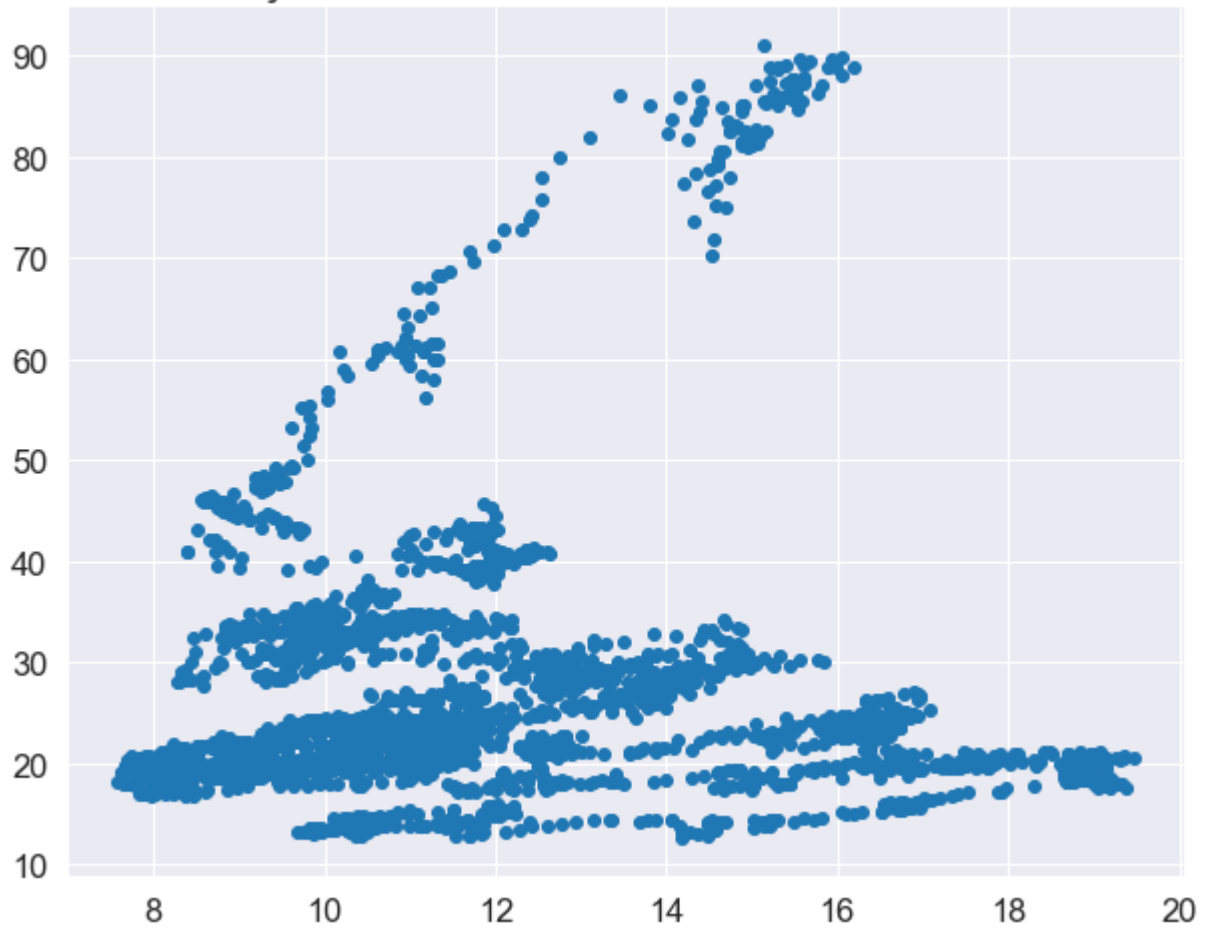


```
In [289...] mc_df['milho_dolares'].corr(mc_df['ccmfut_fechamento'])
```

```
Out[289...] 0.065512943089332
```

```
In [290...] data1=mc_df['milho_dolares']  
data2=mc_df['ccmfut_fechamento']  
plt.scatter(data1, data2)  
plt.title('Correlação - Milho em Dolares e CCMFUT Fechamento')  
plt.gcf().set_size_inches(10, 8)  
plt.show()
```

## Correlação - Milho em Dolares e CCMFUT Fechamento



```
In [291...] mc_df = mc_df[['data', 'ccmfut_abertura', 'ccmfut_máxima', 'ccmfut_mínima', 'ccmfut_vo
```

```
In [292...] def interactive_plot(df, title):
    fig = px.line(title = title)
    for i in df.columns[5:]:
        fig.add_scatter(x = df['data'], y = df[i], name = i)
    fig.show()
```

```
In [293...] interactive_plot(mc_df, "CCMFUT Fechamento, Milho em Reais e Milho em Dolares")
```

In [294...] *# 4 - Ridge Regression*

```
In [295...] mc_df['ccmfut_fechamento_alvo'] = mc_df[['ccmfut_fechamento']].shift(-1)
mc_df = mc_df[:-1]
mc_df
```

Out[295...] **data ccmfut\_abertura ccmfut\_máxima ccmfut\_mínima ccmfut\_volume\_fin ccmfut\_fechamento**

data		ccmfut_abertura	ccmfut_máxima	ccmfut_mínima	ccmfut_volume_fin	ccmfut_fechamento
<b>2008-09-19</b>	2008-09-19	22.64	22.64	22.64	1129500.0	22.64
<b>2008-09-22</b>	2008-09-22	22.64	22.64	22.64	112950.0	22.64
<b>2008-09-26</b>	2008-09-26	22.68	22.68	22.68	565875.0	22.68
<b>2008-09-30</b>	2008-09-30	22.55	22.55	22.55	112500.0	22.55
<b>2008-10-02</b>	2008-10-02	22.64	22.64	22.64	11295.0	22.64
...	...	...	...	...	...	...
<b>2021-02-24</b>	2021-02-24	89.08	89.47	88.42	75830647.5	89.47
<b>2021-02-25</b>	2021-02-25	89.45	89.72	88.81	106361550.0	89.63
<b>2021-02-26</b>	2021-02-26	89.61	89.64	88.86	69200707.5	88.86
<b>2021-03-01</b>	2021-03-01	88.92	89.18	88.00	133054398.0	88.70
<b>2021-03-02</b>	2021-03-02	88.80	89.06	88.54	95795617.5	88.94

2916 rows × 9 columns

In [296...] **from** sklearn.preprocessing **import** MinMaxScaler

```
sc = MinMaxScaler(feature_range = (0, 1))
mc_df_scaled = sc.fit_transform(mc_df.drop(columns = ['data']))
```

In [297... mc\_df\_scaled

```
Out[297... array([[0.12926671, 0.12563841, 0.13420813, ..., 0.08071617, 0.44107744,
        0.12675517],
       [0.12926671, 0.12563841, 0.13420813, ..., 0.07836807, 0.45622896,
        0.12726576],
       [0.12978585, 0.12614913, 0.13472782, ..., 0.07734077, 0.41750842,
        0.12560633],
       ...,
       [0.99844257, 0.98110317, 0.99454333, ..., 0.98972703, 0.64983165,
        0.97000255],
       [0.98948735, 0.97522983, 0.98337014, ..., 0.99236865, 0.6489899 ,
        0.97306612],
       [0.98792992, 0.97369765, 0.99038586, ..., 1.          , 0.64141414,
        1.          ]])
```

In [298... mc\_df\_scaled.shape

Out[298... (2916, 8)

```
In [299... X = mc_df_scaled[:, :7]
y = mc_df_scaled[:, 7:]
```

In [300... X

```
Out[300... array([[0.12926671, 0.12563841, 0.13420813, ..., 0.12877707, 0.08071617,
        0.44107744],
       [0.12926671, 0.12563841, 0.13420813, ..., 0.12877707, 0.07836807,
        0.45622896],
       [0.12978585, 0.12614913, 0.13472782, ..., 0.12929581, 0.07734077,
        0.41750842],
       ...,
       [0.99844257, 0.98110317, 0.99454333, ..., 0.98755025, 0.98972703,
        0.64983165],
       [0.98948735, 0.97522983, 0.98337014, ..., 0.9854753 , 0.99236865,
        0.6489899 ],
       [0.98792992, 0.97369765, 0.99038586, ..., 0.98858773, 1.          ,
        0.64141414]])
```

In [301... X.shape

Out[301... (2916, 7)

In [302... X[0,6]

Out[302... 0.44107744107744096

In [303... y

```
Out[303... array([[0.12675517],
        [0.12726576],
        [0.12560633],
        ...,
        [0.97000255],
        [0.97306612],
        [1.          ]])
```

```
In [304... X = np.asarray(X)
y = np.asarray(y)
X.shape, y.shape
```

```
Out[304... ((2916, 7), (2916, 1))
```

```
In [305... X
```

```
Out[305... array([[0.12926671, 0.12563841, 0.13420813, ..., 0.12877707, 0.08071617,
        0.44107744],
        [0.12926671, 0.12563841, 0.13420813, ..., 0.12877707, 0.07836807,
        0.45622896],
        [0.12978585, 0.12614913, 0.13472782, ..., 0.12929581, 0.07734077,
        0.41750842],
        ...,
        [0.99844257, 0.98110317, 0.99454333, ..., 0.98755025, 0.98972703,
        0.64983165],
        [0.98948735, 0.97522983, 0.98337014, ..., 0.9854753 , 0.99236865,
        0.6489899 ],
        [0.98792992, 0.97369765, 0.99038586, ..., 0.98858773, 1.          ,
        0.64141414]])
```

```
In [306... y
```

```
Out[306... array([[0.12675517],
        [0.12726576],
        [0.12560633],
        ...,
        [0.97000255],
        [0.97306612],
        [1.          ]])
```

```
In [307... split = int(0.70 * len(X))
X_treino = X[:split]
y_treino = y[:split]
X_teste = X[split:]
y_teste = y[split:]
```

```
In [308... X_treino.shape, y_treino.shape
```

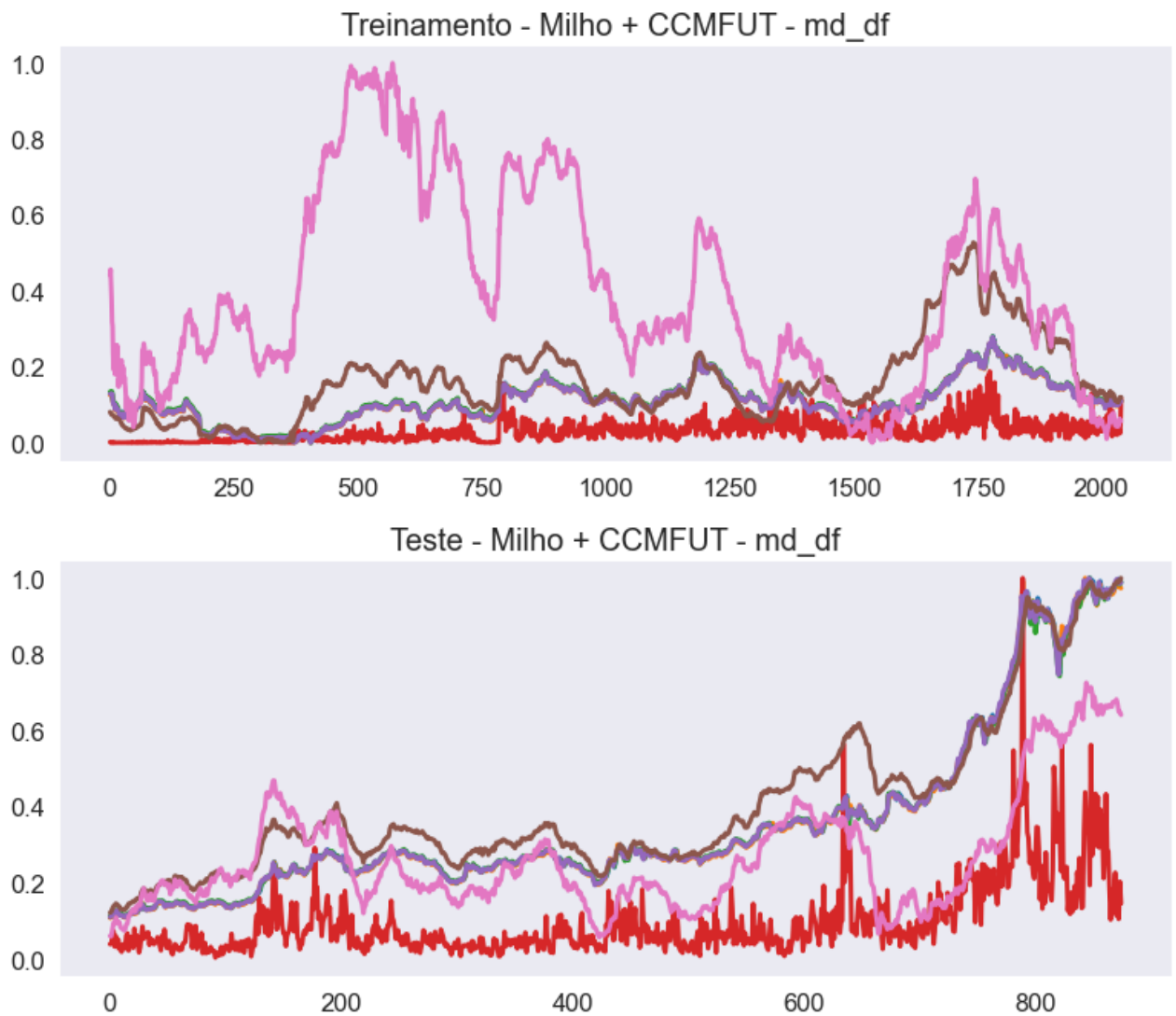
```
Out[308... ((2041, 7), (2041, 1))
```

```
In [309... X_teste.shape, y_teste.shape
```

```
Out[309... ((875, 7), (875, 1))
```

```
In [310... def show_plot_mc(data, title):
    plt.figure(figsize = (13, 5))
    plt.plot(data, linewidth = 3)
    plt.title(title)
    plt.grid()

show_plot_mc(X_treino, 'Treinamento - Milho + CCMFUT - md_df')
show_plot_mc(X_teste, 'Teste - Milho + CCMFUT - md_df')
```



```
In [311... from sklearn.linear_model import Ridge
regression_model = Ridge (alpha=1)
regression_model.fit(X_treino, y_treino)
```

```
Out[311... Ridge(alpha=1)
```

```
In [312... lr_accuracy = regression_model.score(X_teste, y_teste)
print("Linear Regression Score: ", lr_accuracy)
```

```
Linear Regression Score: 0.9874692042284041
```

```
In [313... predicted_prices_mc = regression_model.predict(X)
predicted_prices_mc
```

```
Out[313... array([[0.12060122],
        [0.12045201],
        [0.12093868],
        ...,
        [0.91668274],
        [0.91338118],
        [0.91366199]])
```

```
In [314... predicted_mc = []
for i in predicted_prices_mc:
    predicted_mc.append(i[0])
```

```
In [315... mc_fechamento = []
for i in mc_df_scaled:
    mc_fechamento.append(i[4])
```

```
In [316... mc_predicao = pd.DataFrame(columns = ['data' , 'mc_fechamento', 'mc_fechamento_predito']
mc_predicao['data'] = mc_df['data']
mc_predicao['mc_fechamento'] = mc_fechamento
mc_predicao['mc_fechamento_predito'] = predicted_mc
mc_predicao
```

Out[316... data mc\_fechamento mc\_fechamento\_predito

data			
<b>2008-09-19</b>	2008-09-19	0.128777	0.120601
<b>2008-09-22</b>	2008-09-22	0.128777	0.120452
<b>2008-09-26</b>	2008-09-26	0.129296	0.120939
<b>2008-09-30</b>	2008-09-30	0.127610	0.119349
<b>2008-10-02</b>	2008-10-02	0.128777	0.120427
...	...	...	...
<b>2021-02-24</b>	2021-02-24	0.995461	0.915434
<b>2021-02-25</b>	2021-02-25	0.997536	0.920280
<b>2021-02-26</b>	2021-02-26	0.987550	0.916683
<b>2021-03-01</b>	2021-03-01	0.985475	0.913381
<b>2021-03-02</b>	2021-03-02	0.988588	0.913662

2916 rows × 3 columns

```
In [317... def interactive_plot(data, title):
    fig = px.line(title = title)
    for i in data.columns[1:]:
        fig.add_scatter(x = data['data'], y = data[i], name = i)
    fig.show()
```

```
In [318... interactive_plot(mc_predicao, "CCMFUT Fechamento e CCMFUT Fechamento Predito")
```

```
In [319... #Cálculo do erro
mse = mean_squared_error(mc_predicao['mc_fechamento'], mc_predicao['mc_fechamento_predi
print('MSE: '+str(mse))
mae = mean_absolute_error(mc_predicao['mc_fechamento'], mc_predicao['mc_fechamento_pred
print('MAE: '+str(mae))
rmse = math.sqrt(mean_squared_error(mc_predicao['mc_fechamento'], mc_predicao['mc_fecha
print('RMSE: '+str(rmse))
```

```
MSE: 0.0002551720660963173
MAE: 0.008955258899856846
RMSE: 0.015974106112591004
```

```
In [320... mc_df
```

```
Out[320... data ccmfut_abertura ccmfut_máxima ccmfut_mínima ccmfut_volume_fin ccmfut_fechamento
```

	data	ccmfut_abertura	ccmfut_máxima	ccmfut_mínima	ccmfut_volume_fin	ccmfut_fechamento
	data					
2008-09-19	2008-09-19	22.64	22.64	22.64	1129500.0	22.64
2008-09-22	2008-09-22	22.64	22.64	22.64	112950.0	22.64
2008-09-26	2008-09-26	22.68	22.68	22.68	565875.0	22.68
2008-09-30	2008-09-30	22.55	22.55	22.55	112500.0	22.55
2008-10-02	2008-10-02	22.64	22.64	22.64	11295.0	22.64
...	...	...	...	...	...	...
2021-02-24	2021-02-24	89.08	89.47	88.42	75830647.5	89.47
2021-02-25	2021-02-25	89.45	89.72	88.81	106361550.0	89.63
2021-02-26	2021-02-26	89.61	89.64	88.86	69200707.5	88.86



	data	ccmfut_abertura	ccmfut_máxima	ccmfut_mínima	ccmfut_volume_fin	ccmfut_fechamento
<b>data</b>						
<b>2021-03-01</b>	2021-03-01	88.92	89.18	88.00	133054398.0	88.70
<b>2021-03-02</b>	2021-03-02	88.80	89.06	88.54	95795617.5	88.94

2916 rows × 9 columns



```
In [321... mc_df = mc_df.drop(columns='ccmfut_abertura')
mc_df = mc_df.drop(columns='ccmfut_máxima')
mc_df = mc_df.drop(columns='ccmfut_mínima')
mc_df = mc_df.drop(columns='ccmfut_volume_fin')
mc_df = mc_df.drop(columns='milho_reais')
mc_df = mc_df.drop(columns='milho_dolares')
mc_df
```

```
Out[321... data ccmfut_fechamento ccmfut_fechamento_alvo
```

<b>data</b>			
<b>2008-09-19</b>	2008-09-19	22.64	22.64
<b>2008-09-22</b>	2008-09-22	22.64	22.68
<b>2008-09-26</b>	2008-09-26	22.68	22.55
<b>2008-09-30</b>	2008-09-30	22.55	22.64
<b>2008-10-02</b>	2008-10-02	22.64	21.65
...	...	...	...
<b>2021-02-24</b>	2021-02-24	89.47	89.63
<b>2021-02-25</b>	2021-02-25	89.63	88.86
<b>2021-02-26</b>	2021-02-26	88.86	88.70
<b>2021-03-01</b>	2021-03-01	88.70	88.94
<b>2021-03-02</b>	2021-03-02	88.94	91.05

2916 rows × 3 columns

```
In [322... mc_df_scaled_fech = sc.fit_transform(mc_df.drop(columns = ['data']))
```

```
In [323... X = mc_df_scaled_fech[:, :1]
y = mc_df_scaled_fech[:, 1:]
```

```
In [324... X = np.asarray(X)
y = np.asarray(y)
```

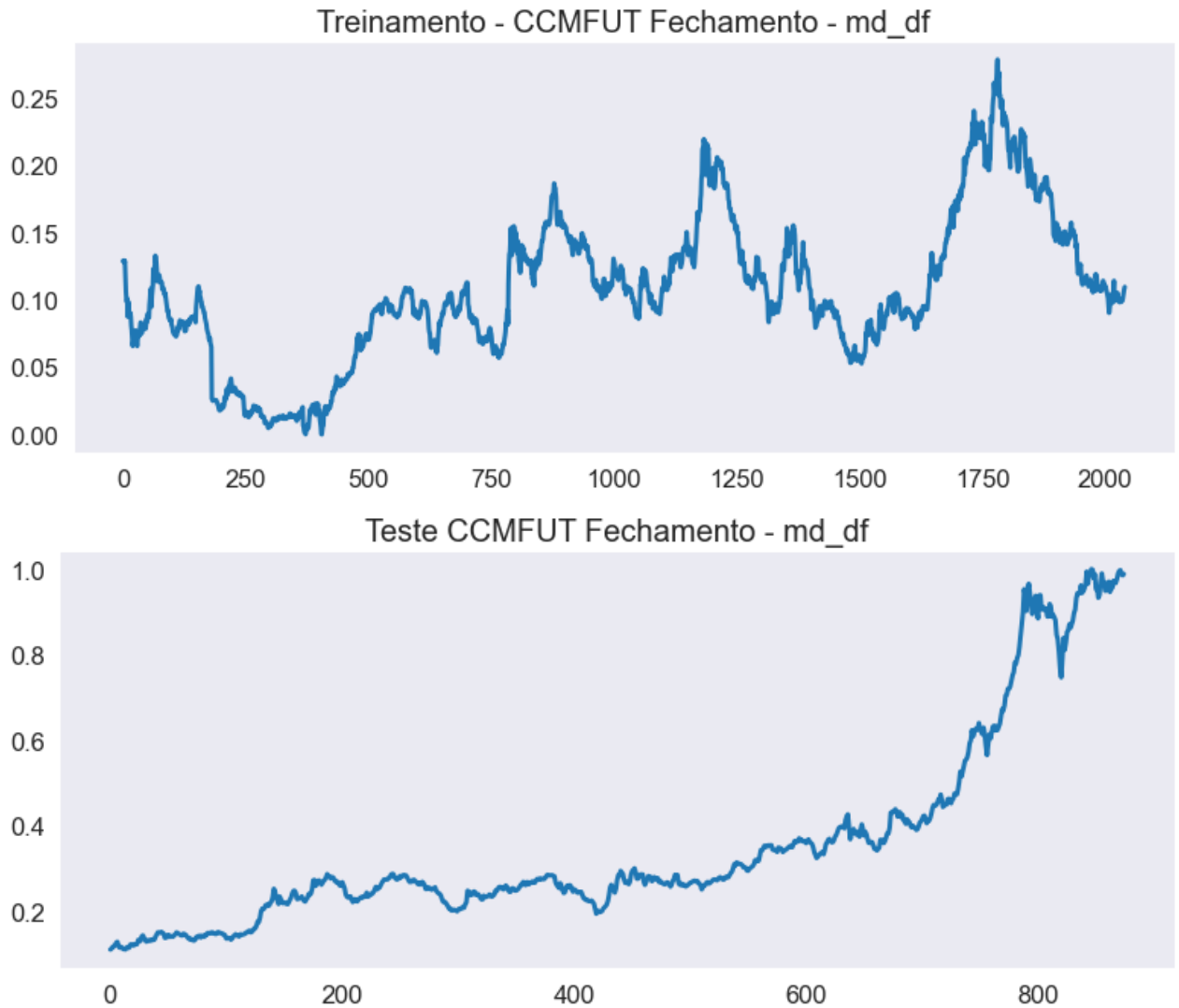
```
In [325... split = int(0.70 * len(X))
X_treino = X[:split]
y_treino = y[:split]
```

```
X_teste = X[split:]
y_teste = y[split:]
```

In [326...

```
def show_plot_mc(data, title):
    plt.figure(figsize = (13, 5))
    plt.plot(data, linewidth = 3)
    plt.title(title)
    plt.grid()

show_plot_mc(X_treino, 'Treinamento - CCMFUT Fechamento - md_df')
show_plot_mc(X_teste, 'Teste CCMFUT Fechamento - md_df')
```

In [327... `regression_model.fit(X_treino, y_treino)`Out[327... `Ridge(alpha=1)`In [328... `lr_accuracy = regression_model.score(X_teste, y_teste)`  
`print("Linear Regression Score: ", lr_accuracy)`

Linear Regression Score: 0.9484580850390733

In [329... `predicted_prices_mc = regression_model.predict(X)`  
`predicted_prices_mc`Out[329... `array([[0.12355197],`

```
[0.12355197],
[0.12398756],
...,
[0.84466647],
[0.84292412],
[0.84553764]])
```

```
In [330... predicted_mc = []
for i in predicted_prices_mc:
    predicted_mc.append(i[0])
```

```
In [331... mc_predicao
```

```
Out[331... data mc_fechamento mc_fechamento_predito
```

data			
<b>2008-09-19</b>	2008-09-19	0.128777	0.120601
<b>2008-09-22</b>	2008-09-22	0.128777	0.120452
<b>2008-09-26</b>	2008-09-26	0.129296	0.120939
<b>2008-09-30</b>	2008-09-30	0.127610	0.119349
<b>2008-10-02</b>	2008-10-02	0.128777	0.120427
...	...	...	...
<b>2021-02-24</b>	2021-02-24	0.995461	0.915434
<b>2021-02-25</b>	2021-02-25	0.997536	0.920280
<b>2021-02-26</b>	2021-02-26	0.987550	0.916683
<b>2021-03-01</b>	2021-03-01	0.985475	0.913381
<b>2021-03-02</b>	2021-03-02	0.988588	0.913662

2916 rows × 3 columns

```
In [332... mc_predicao = mc_predicao.drop(columns='mc_fechamento_predito')
mc_predicao
```

```
Out[332... data mc_fechamento
```

data		
<b>2008-09-19</b>	2008-09-19	0.128777
<b>2008-09-22</b>	2008-09-22	0.128777
<b>2008-09-26</b>	2008-09-26	0.129296
<b>2008-09-30</b>	2008-09-30	0.127610
<b>2008-10-02</b>	2008-10-02	0.128777
...	...	...
<b>2021-02-24</b>	2021-02-24	0.995461
<b>2021-02-25</b>	2021-02-25	0.997536

**data mc\_fechamento****data**

<b>2021-02-26</b>	2021-02-26	0.987550
<b>2021-03-01</b>	2021-03-01	0.985475
<b>2021-03-02</b>	2021-03-02	0.988588

2916 rows × 2 columns

In [333... `mc_predicao['mc_fechamento_predito'] = predicted_mc`In [334... `mc_predicao`Out[334... **data mc\_fechamento mc\_fechamento\_predito****data**

<b>2008-09-19</b>	2008-09-19	0.128777	0.123552
<b>2008-09-22</b>	2008-09-22	0.128777	0.123552
<b>2008-09-26</b>	2008-09-26	0.129296	0.123988
<b>2008-09-30</b>	2008-09-30	0.127610	0.122572
<b>2008-10-02</b>	2008-10-02	0.128777	0.123552
...	...	...	...
<b>2021-02-24</b>	2021-02-24	0.995461	0.851309
<b>2021-02-25</b>	2021-02-25	0.997536	0.853052
<b>2021-02-26</b>	2021-02-26	0.987550	0.844666
<b>2021-03-01</b>	2021-03-01	0.985475	0.842924
<b>2021-03-02</b>	2021-03-02	0.988588	0.845538

2916 rows × 3 columns

In [335... `interactive_plot(mc_predicao, "CCMFUT Fechamento e CCMFUT Fechamento Predito")`

```
In [336... #Cálculo do erro
mse = mean_squared_error(mc_predicao['mc_fechamento'], mc_predicao['mc_fechamento_predi
print('MSE: '+str(mse))
mae = mean_absolute_error(mc_predicao['mc_fechamento'], mc_predicao['mc_fechamento_pred
print('MAE: '+str(mae))
rmse = math.sqrt(mean_squared_error(mc_predicao['mc_fechamento'], mc_predicao['mc_fecha
print('RMSE: '+str(rmse))
```

```
MSE: 0.0009740986387099513
MAE: 0.017127976172408
RMSE: 0.03121055332271364
```

```
In [337... # 5 - RNN - LSTM
```

```
In [338... mc_df
```

```
Out[338... data ccmfut_fechamento ccmfut_fechamento_alvo
```

data			
<b>2008-09-19</b>	2008-09-19	22.64	22.64
<b>2008-09-22</b>	2008-09-22	22.64	22.68
<b>2008-09-26</b>	2008-09-26	22.68	22.55
<b>2008-09-30</b>	2008-09-30	22.55	22.64
<b>2008-10-02</b>	2008-10-02	22.64	21.65
...	...	...	...
<b>2021-02-24</b>	2021-02-24	89.47	89.63
<b>2021-02-25</b>	2021-02-25	89.63	88.86
<b>2021-02-26</b>	2021-02-26	88.86	88.70
<b>2021-03-01</b>	2021-03-01	88.70	88.94
<b>2021-03-02</b>	2021-03-02	88.94	91.05

2916 rows × 3 columns

```
In [339... mc_df = mc_df.drop(columns='ccmfut_fechamento_alvo')
mc_df
```

```
Out[339... data ccmfut_fechamento
```

	data	
<b>2008-09-19</b>	2008-09-19	22.64
<b>2008-09-22</b>	2008-09-22	22.64
<b>2008-09-26</b>	2008-09-26	22.68
<b>2008-09-30</b>	2008-09-30	22.55
<b>2008-10-02</b>	2008-10-02	22.64
...	...	...
<b>2021-02-24</b>	2021-02-24	89.47
<b>2021-02-25</b>	2021-02-25	89.63
<b>2021-02-26</b>	2021-02-26	88.86
<b>2021-03-01</b>	2021-03-01	88.70
<b>2021-03-02</b>	2021-03-02	88.94

2916 rows × 2 columns

```
In [340... training_data = mc_df.iloc[:, 1:].values
training_data
```

```
Out[340... array([[22.64],
        [22.64],
        [22.68],
        ...,
        [88.86],
        [88.7 ],
        [88.94]])
```

```
In [341... from sklearn.preprocessing import MinMaxScaler
sc = MinMaxScaler(feature_range = (0, 1))
training_set_scaled = sc.fit_transform(training_data)
```

```
In [342... training_set_scaled
```

```
Out[342... array([[0.12877707],
        [0.12877707],
        [0.12929581],
        ...,
        [0.98755025],
        [0.9854753 ],
        [0.98858773]])
```

```
In [343... X = []
y = []
for i in range(1, len(mc_df)):
    X.append(training_set_scaled [i-1:i, 0])
    y.append(training_set_scaled [i, 0])
```

In [344...] X

```
Out[344...] [array([0.12877707]),
array([0.12877707]),
array([0.12929581]),
array([0.12760991]),
array([0.12877707]),
array([0.11593827]),
array([0.10647127]),
array([0.10076514]),
array([0.09830113]),
array([0.10063546]),
array([0.0874076]),
array([0.09843081]),
array([0.09246531]),
array([0.09013098]),
array([0.08896382]),
array([0.09013098]),
array([0.08727791]),
array([0.08079367]),
array([0.07729218]),
array([0.0719751]),
array([0.06562054]),
array([0.06912203]),
array([0.06912203]),
array([0.06795487]),
array([0.0719751]),
array([0.07262352]),
array([0.07729218]),
array([0.07145636]),
array([0.06678771]),
array([0.06562054]),
array([0.0702892]),
array([0.07612502]),
array([0.07262352]),
array([0.07495785]),
array([0.07392037]),
array([0.07962651]),
array([0.083128]),
array([0.08260926]),
array([0.07495785]),
array([0.07962651]),
array([0.08196084]),
array([0.08079367]),
array([0.07975619]),
array([0.08092336]),
array([0.08196084]),
array([0.08079367]),
array([0.07845934]),
array([0.0814421]),
array([0.08611075]),
array([0.08377642]),
array([0.0874076]),
array([0.08974193]),
array([0.08662949]),
array([0.08675918]),
array([0.0959668]),
array([0.10050577]),
array([0.10673064]),
array([0.10828686]),
array([0.09479964]),
array([0.10296978]),
array([0.10776812]),
array([0.1113993]),
array([0.12294125]),
```

```
array([0.11593827]),
array([0.12060693]),
array([0.12644274]),
array([0.1327973]),
array([0.12877707]),
array([0.12760991]),
array([0.12177409]),
array([0.11360394]),
array([0.11477111]),
array([0.11710543]),
array([0.11528985]),
array([0.1182726]),
array([0.11593827]),
array([0.11373363]),
array([0.11295552]),
array([0.11243678]),
array([0.11295552]),
array([0.10776812]),
array([0.10776812]),
array([0.10893529]),
array([0.10776812]),
array([0.10530411]),
array([0.10361821]),
array([0.10478537]),
array([0.10180262]),
array([0.09830113]),
array([0.0959668]),
array([0.09363247]),
array([0.09129815]),
array([0.09194657]),
array([0.08779665]),
array([0.08611075]),
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...]
```

```
In [346... X = np.asarray(X)
           y = np.asarray(y)
```

```
In [347... X.shape , y.shape
```

```
Out[347... ((2915, 1), (2915,))
```

```
In [348... split = int(0.7 * len(X))
X_train = X[:split]
y_train = y[:split]
X_test = X[split:]
y_test = y[split:]
```

```
In [349... X_train.shape , y_train.shape, X_test.shape, y_test.shape
```

```
Out[349... ((2040, 1), (2040,), (875, 1), (875,))
```

```
In [350... X_train = np.reshape(X_train, (X_train.shape[0], X_train.shape[1], 1))
X_test = np.reshape(X_test, (X_test.shape[0], X_test.shape[1], 1))
X_train.shape, X_test.shape
```

```
Out[350... ((2040, 1, 1), (875, 1, 1))
```

```
In [351... X_train
```

```
Out[351... array([[0.12877707]],
        [[0.12877707]],
        [[0.12929581]],
        ...,
        [[0.10361821]],
        [[0.1054338 ]],
        [[0.10815718]]])
```

```
In [352... X_train.shape
```

```
Out[352... (2040, 1, 1)
```

```
In [353... inputs = keras.layers.Input(shape=(X_train.shape[1], X_train.shape[2]))
x = keras.layers.LSTM(150, return_sequences=True)(inputs)
x = keras.layers.Dropout(0.3)(x)
x = keras.layers.LSTM(150, return_sequences=True)(x)
x = keras.layers.Dropout(0.3)(x)
x = keras.layers.LSTM(150)(x)
outputs = keras.layers.Dense(1, activation='linear')(x)

model = keras.Model(inputs=inputs, outputs=outputs)
model.compile(optimizer='adam', loss="mse")
model.summary()
```

Model: "model\_2"

Layer (type)	Output Shape	Param #
input_3 (InputLayer)	[(None, 1, 1)]	0
lstm_6 (LSTM)	(None, 1, 150)	91200
dropout_4 (Dropout)	(None, 1, 150)	0
lstm_7 (LSTM)	(None, 1, 150)	180600

dropout_5 (Dropout)	(None, 1, 150)	0
lstm_8 (LSTM)	(None, 150)	180600
dense_2 (Dense)	(None, 1)	151
=====		
Total params: 452,551		
Trainable params: 452,551		
Non-trainable params: 0		

In [354... `X_train.shape, y_train.shape`

Out[354... `((2040, 1, 1), (2040,))`

In [355... `history = model.fit(  
 X_train, y_train,  
 epochs = 20,  
 batch_size = 32,  
 validation_split = 0.2  
)`

```
Epoch 1/20
51/51 [=====] - 3s 51ms/step - loss: 0.0030 - val_loss: 0.0057
Epoch 2/20
51/51 [=====] - 1s 11ms/step - loss: 0.0010 - val_loss: 4.1449e-05
Epoch 3/20
51/51 [=====] - 1s 11ms/step - loss: 6.6296e-05 - val_loss: 2.5949e-05
Epoch 4/20
51/51 [=====] - 1s 11ms/step - loss: 4.0457e-05 - val_loss: 2.4025e-05
Epoch 5/20
51/51 [=====] - 0s 10ms/step - loss: 3.9753e-05 - val_loss: 3.7790e-05
Epoch 6/20
51/51 [=====] - 1s 10ms/step - loss: 4.1150e-05 - val_loss: 2.5174e-05
Epoch 7/20
51/51 [=====] - 0s 8ms/step - loss: 3.2069e-05 - val_loss: 4.0390e-05
Epoch 8/20
51/51 [=====] - 0s 9ms/step - loss: 3.2752e-05 - val_loss: 2.4666e-05
Epoch 9/20
51/51 [=====] - 0s 8ms/step - loss: 3.4260e-05 - val_loss: 2.8544e-05
Epoch 10/20
51/51 [=====] - 0s 7ms/step - loss: 3.5984e-05 - val_loss: 2.3459e-05
Epoch 11/20
51/51 [=====] - 0s 7ms/step - loss: 3.1575e-05 - val_loss: 2.9729e-05
Epoch 12/20
51/51 [=====] - 0s 8ms/step - loss: 3.0410e-05 - val_loss: 3.6740e-05
Epoch 13/20
51/51 [=====] - 0s 9ms/step - loss: 3.1194e-05 - val_loss: 2.8755e-05
Epoch 14/20
51/51 [=====] - 0s 8ms/step - loss: 2.9855e-05 - val_loss: 2.9955e-05
```

```

Epoch 15/20
51/51 [=====] - 0s 7ms/step - loss: 3.0825e-05 - val_loss: 2.44
23e-05
Epoch 16/20
51/51 [=====] - 0s 9ms/step - loss: 2.8862e-05 - val_loss: 2.40
80e-05
Epoch 17/20
51/51 [=====] - 0s 8ms/step - loss: 3.6636e-05 - val_loss: 2.47
57e-05
Epoch 18/20
51/51 [=====] - 0s 8ms/step - loss: 3.3046e-05 - val_loss: 2.37
31e-05
Epoch 19/20
51/51 [=====] - 0s 8ms/step - loss: 2.9455e-05 - val_loss: 2.38
31e-05
Epoch 20/20
51/51 [=====] - 0s 8ms/step - loss: 3.6684e-05 - val_loss: 4.77
91e-05

```

```
In [356... predicted_LSTM = model.predict(X)
```

```
In [357... predicted_LSTM
```

```
Out[357... array([[0.12406403],
        [0.12406403],
        [0.12458012],
        ...,
        [0.9513022 ],
        [0.9426287 ],
        [0.9408232 ]], dtype=float32)
```

```
In [358... predicted.shape
```

```
Out[358... (2915, 1)
```

```
In [359... test_predicted_LSTM = []

for i in predicted_LSTM:
    test_predicted_LSTM.append(i[0])
```

```
In [360... len(test_predicted_LSTM)
```

```
Out[360... 2915
```

```
In [361... df_predicao_LSTM = pd.DataFrame(columns = ['data' , 'Fechamento', 'Fechamento_Predito'])
```

```
In [362... df_predicao_LSTM
```

```
Out[362...    data  Fechamento  Fechamento_Predito
```

```
In [363... df_predicao_LSTM['data'] = mc_df[1:]['data']
```

```
In [364... df_predicao_LSTM
```

```
Out[364...    data  Fechamento  Fechamento_Predito

    data
```



	<b>data</b>	<b>Fechamento</b>	<b>Fechamento_Predito</b>
	<b>data</b>		
<b>2008-09-22</b>	2008-09-22	NaN	NaN
<b>2008-09-26</b>	2008-09-26	NaN	NaN
<b>2008-09-30</b>	2008-09-30	NaN	NaN
<b>2008-10-02</b>	2008-10-02	NaN	NaN
<b>2008-10-03</b>	2008-10-03	NaN	NaN
...	...	...	...
<b>2021-02-24</b>	2021-02-24	NaN	NaN
<b>2021-02-25</b>	2021-02-25	NaN	NaN
<b>2021-02-26</b>	2021-02-26	NaN	NaN
<b>2021-03-01</b>	2021-03-01	NaN	NaN
<b>2021-03-02</b>	2021-03-02	NaN	NaN

2915 rows × 3 columns

```
In [365... Fechamento_Scaled = []
for i in training_set_scaled:
    Fechamento_Scaled.append(i[0])
```

```
In [366... len(Fechamento_Scaled)
```

```
Out[366... 2916
```

```
In [367... df_predicao_LSTM
```

```
Out[367...
```

	<b>data</b>	<b>Fechamento</b>	<b>Fechamento_Predito</b>
	<b>data</b>		
<b>2008-09-22</b>	2008-09-22	NaN	NaN
<b>2008-09-26</b>	2008-09-26	NaN	NaN
<b>2008-09-30</b>	2008-09-30	NaN	NaN
<b>2008-10-02</b>	2008-10-02	NaN	NaN
<b>2008-10-03</b>	2008-10-03	NaN	NaN
...	...	...	...
<b>2021-02-24</b>	2021-02-24	NaN	NaN
<b>2021-02-25</b>	2021-02-25	NaN	NaN
<b>2021-02-26</b>	2021-02-26	NaN	NaN
<b>2021-03-01</b>	2021-03-01	NaN	NaN
<b>2021-03-02</b>	2021-03-02	NaN	NaN

2915 rows × 3 columns

In [368... df\_predicao\_LSTM['Fechamento'] = Fechamento\_Scaled[1:]

In [369... df\_predicao\_LSTM

Out[369... data Fechamento Fechamento\_Predito

data			
<b>2008-09-22</b>	2008-09-22	0.128777	NaN
<b>2008-09-26</b>	2008-09-26	0.129296	NaN
<b>2008-09-30</b>	2008-09-30	0.127610	NaN
<b>2008-10-02</b>	2008-10-02	0.128777	NaN
<b>2008-10-03</b>	2008-10-03	0.115938	NaN
...	...	...	...
<b>2021-02-24</b>	2021-02-24	0.995461	NaN
<b>2021-02-25</b>	2021-02-25	0.997536	NaN
<b>2021-02-26</b>	2021-02-26	0.987550	NaN
<b>2021-03-01</b>	2021-03-01	0.985475	NaN
<b>2021-03-02</b>	2021-03-02	0.988588	NaN

2915 rows × 3 columns

In [370... df\_predicao\_LSTM['Fechamento\_Predito'] = test\_predicted\_LSTM

In [371... df\_predicao\_LSTM

Out[371... data Fechamento Fechamento\_Predito

data			
<b>2008-09-22</b>	2008-09-22	0.128777	0.124064
<b>2008-09-26</b>	2008-09-26	0.129296	0.124064
<b>2008-09-30</b>	2008-09-30	0.127610	0.124580
<b>2008-10-02</b>	2008-10-02	0.128777	0.122903
<b>2008-10-03</b>	2008-10-03	0.115938	0.124064
...	...	...	...
<b>2021-02-24</b>	2021-02-24	0.995461	0.944320
<b>2021-02-25</b>	2021-02-25	0.997536	0.949502
<b>2021-02-26</b>	2021-02-26	0.987550	0.951302
<b>2021-03-01</b>	2021-03-01	0.985475	0.942629
<b>2021-03-02</b>	2021-03-02	0.988588	0.940823

2915 rows × 3 columns

```
In [372... interactive_plot(df_predicao_LSTM, "LSTM - CCMFUT Fechamento e CCMFUT Fechamento Predit
```

```
In [373... #Cálculo do erro
mse = mean_squared_error(df_predicao_LSTM['Fechamento'], df_predicao_LSTM['Fechamento_P
print('MSE: '+str(mse))
mae = mean_absolute_error(df_predicao_LSTM['Fechamento'], df_predicao_LSTM['Fechamento_
print('MAE: '+str(mae))
rmse = math.sqrt(mean_squared_error(df_predicao_LSTM['Fechamento'], df_predicao_LSTM['F
print('RMSE: '+str(rmse))
```

```
MSE: 0.00011615309602060521
MAE: 0.0069011389543802695
RMSE: 0.010777434575102055
```

```
In [ ]:
```