

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
import seaborn as sns
from statsmodels.tsa.seasonal import seasonal_decompose
```

▼ Precio del Dolar

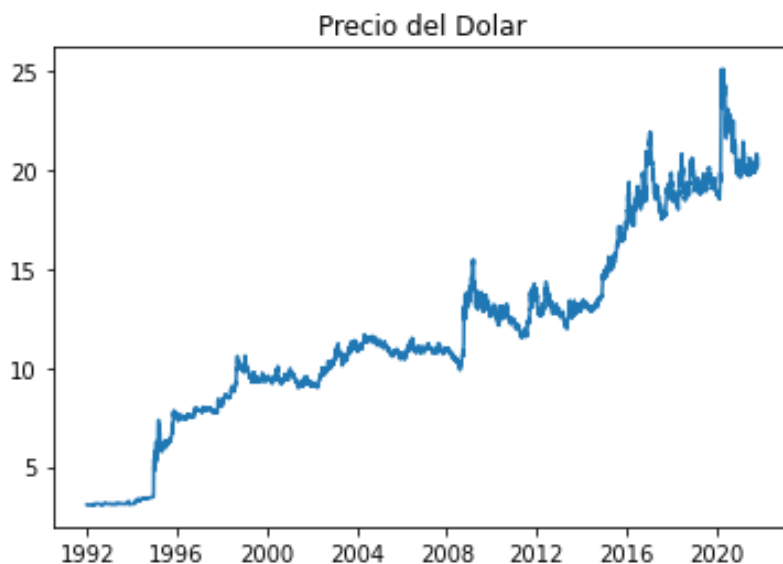
```
df_dolar = pd.read_excel(r'tipo-de-cambio.xlsx')
print(df_dolar.head())
```

```
df_dolar = df_dolar[["Fecha",
                    "Cambio a 48 horas Cierre Compra"]].rename(columns={df_dolar.co:
                    df_dolar.columns[1]: "dolar"})
```

```
plt.plot(df_dolar["Fecha"], df_dolar["dolar"])
plt.title("Precio del Dolar")
```

| | Fecha | Cambio a 48 horas Cierre Compra | Cambio a 48 horas Cierre Venta |
|---|------------|---------------------------------|--------------------------------|
| 0 | 1992-01-02 | 3.0745 | 3.0755 |
| 1 | 1992-01-03 | 3.0723 | 3.0733 |
| 2 | 1992-01-06 | 3.0673 | 3.0683 |
| 3 | 1992-01-07 | 3.0650 | 3.0660 |
| 4 | 1992-01-08 | 3.0650 | 3.0660 |

Text(0.5, 1.0, 'Precio del Dolar')



Estadísticas de estos datos:

```
stat_dolar = df_dolar['dolar'].describe()
```

```
stat_dolar
```

```
count    7500.000000
mean      11.823985
std        4.888754
min         3.060000
25%         9.245000
50%        11.037000
75%        13.687350
max        25.105000
Name: dolar, dtype: float64
```

```
# Tambien podemos sacar los datos del dolar pero promediados por mes para hacerlos m
```

```
df_dolar_mes = df_dolar.groupby(pd.PeriodIndex(df_dolar['Fecha'],
                                                freq="M"))['dolar'].mean().reset_index()
```

```
fecha_i = pd.Timestamp('2012-07-01')
```

```
fecha_f = pd.Timestamp('2020-11-01')
```

```
df_dolar_mes['Fecha'] = [pd.Timestamp(str(j)) for j in df_dolar_mes["Fecha"]]
```

```
df_dolar_mes = df_dolar_mes[(df_dolar_mes['Fecha']>=fecha_i) & (df_dolar_mes['Fecha'
```

```
plt.plot(df_dolar_mes['Fecha'],df_dolar_mes['dolar'])
```

```
[<matplotlib.lines.Line2D at 0x7fc802e4eb50>]
```



```
# Tambien podemos sacar los datos del dolar pero promediados por mes para hacerlos m
```

```
df_oil = pd.read_excel(r'Crude-Oil.xlsx')
```

```
print(df_oil.head())
```

```
df_oil_mes = df_oil.groupby(pd.PeriodIndex(df_oil['Fecha'],
```

```
freq="M"))['Crude Oil Price'].mean().reset_index()
```

```
fecha_i = pd.Timestamp('2012-07-01')
```

```
fecha_f = pd.Timestamp('2020-11-01')
```

```
df_oil_mes['Fecha'] = [pd.Timestamp(str(j)) for j in df_oil_mes["Fecha"]]
```

```
df_oil_mes = df_oil_mes[(df_oil_mes['Fecha']>=fecha_i) & (df_oil_mes['Fecha']<=fecha_f)]
```

```
plt.plot(df_oil_mes['Fecha'],df_oil_mes['Crude Oil Price'])
```

| | Fecha | Crude Oil Price |
|---|------------|-----------------|
| 0 | 1986-01-02 | 25.56 |
| 1 | 1986-01-03 | 26.00 |
| 2 | 1986-01-06 | 26.53 |
| 3 | 1986-01-07 | 25.85 |
| 4 | 1986-01-08 | 25.87 |

```
[<matplotlib.lines.Line2D at 0x7fc7fd04aa10>]
```



```
# Tambien podemos sacar los datos del dolar pero promediados por mes para hacerlos mas faciles de manejar
```

```
df_petroleo = pd.read_excel(r'Petroleo.xlsx')
```

```
df_petroleo.replace(to_replace = "N/E",value = np.nan, inplace=True)
```

```
print(df_petroleo.head())
```

```
df_petroleo_mes = df_petroleo.groupby(pd.PeriodIndex(df_petroleo['Fecha'],
freq="M"))['Precio del Petroleo Mexicano'].mean().reset_index()
```

```
fecha_i = pd.Timestamp('2012-07-01')
```

```
fecha_f = pd.Timestamp('2020-11-01')
```

```
df_petroleo_mes['Fecha'] = [pd.Timestamp(str(j)) for j in df_petroleo_mes["Fecha"]]
```

```
df_petroleo_mes = df_petroleo_mes[(df_petroleo_mes['Fecha']>=fecha_i) & (df_petroleo_mes['Fecha']<=fecha_f)]

plt.plot(df_petroleo_mes['Fecha'],df_petroleo_mes['Precio del Petroleo Mexicano'])
df_petroleo_mes.head()
```

| | Fecha | Precio del Petroleo Mexicano |
|---|------------|------------------------------|
| 0 | 1996-01-01 | NaN |
| 1 | 1996-01-02 | NaN |
| 2 | 1996-01-03 | 17.40 |
| 3 | 1996-01-04 | 17.41 |
| 4 | 1996-01-05 | 17.70 |

| | Fecha | Precio del Petroleo Mexicano |
|------------|------------|------------------------------|
| 198 | 2012-07-01 | 94.661905 |
| 199 | 2012-08-01 | 101.877500 |
| 200 | 2012-09-01 | 102.696842 |
| 201 | 2012-10-01 | 99.875000 |
| 202 | 2012-11-01 | 94.886667 |



```
# Tambien podemos sacar los datos del dolar pero promediados por mes para hacerlos mas faciles de manejar
df_interes = pd.read_excel(r'interes.xlsx')
# Reemplazamos los "N/E" que vienen en los datos por nan
# y nos quedamos solamente con la TIIE a 28 dias
df_interes.replace(to_replace = "N/E",value = np.nan, inplace=True)
df_interes = df_interes[['Fecha', 'TIIE a 28 dias']]

df_interes_mes = df_interes.groupby(pd.PeriodIndex(df_interes['Fecha'],
                                                    freq="M"))['TIIE a 28 dias'].mean().reset_index()

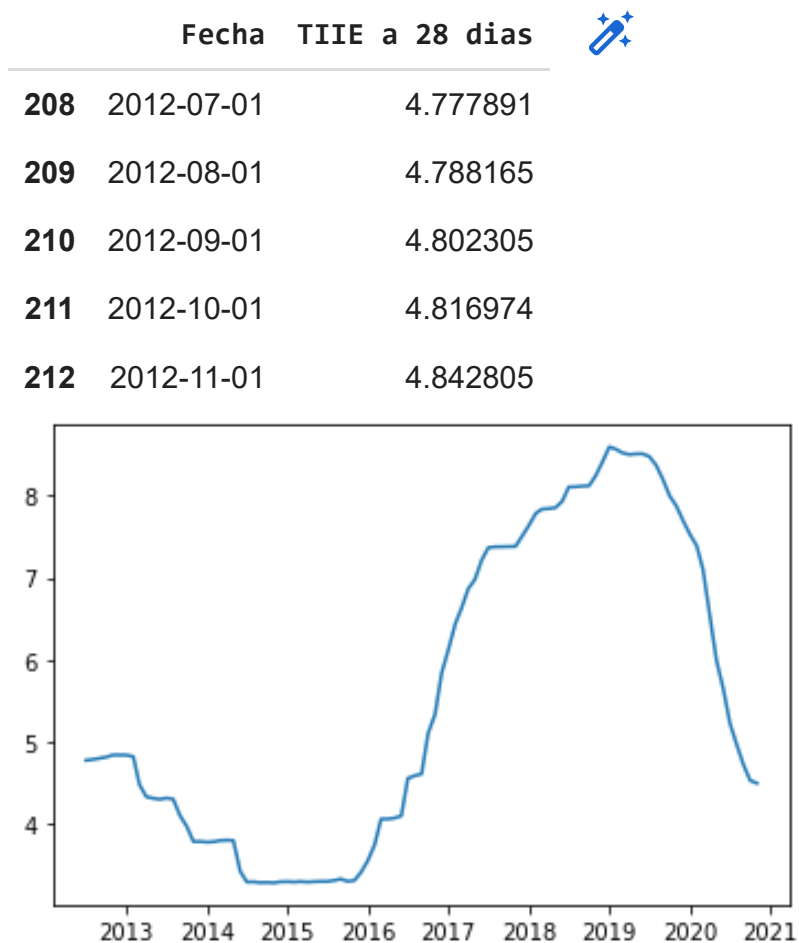
fecha_i = pd.Timestamp('2012-07-01')
```

```
fecha_f = pd.Timestamp('2020-11-01')
```

```
df_interes_mes['Fecha'] = [pd.Timestamp(str(j)) for j in df_interes_mes["Fecha"]]
```

```
df_interes_mes = df_interes_mes[(df_interes_mes['Fecha']>=fecha_i) & (df_interes_mes
```

```
plt.plot(df_interes_mes['Fecha'],df_interes_mes['TIIE a 28 dias'])
df_interes_mes.head()
```



```
df_cetes = pd.read_excel(r'cetes.xlsx')
```

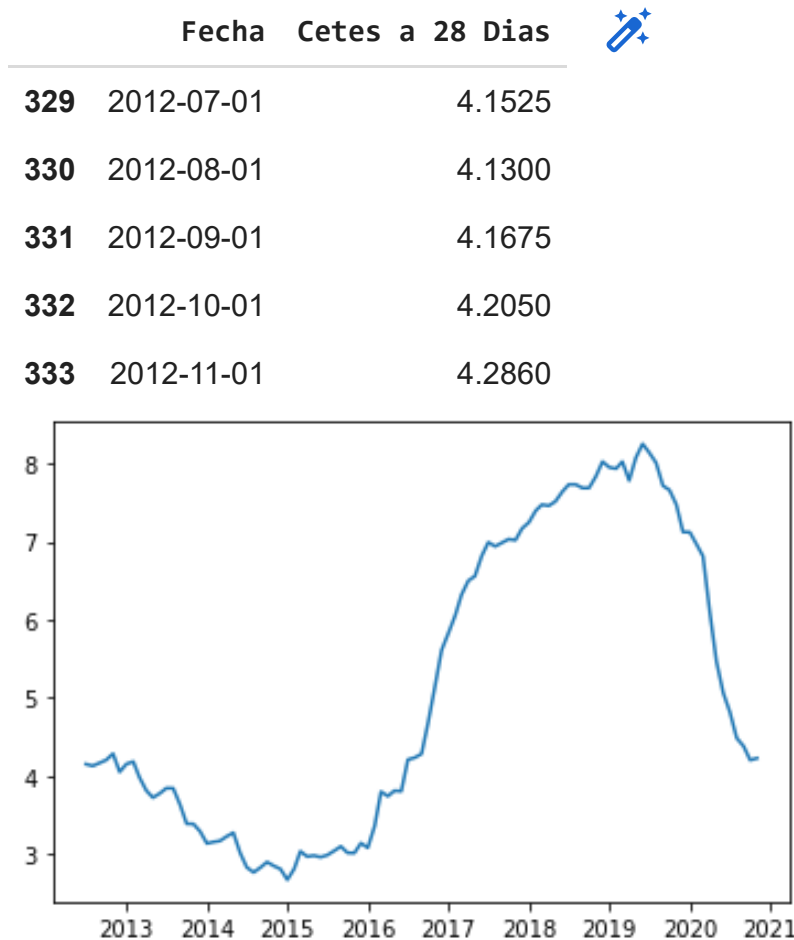
```
df_cetes_mes = df_cetes.groupby(pd.PeriodIndex(df_cetes['Fecha'],
                                                freq="M"))['Cetes a 28 Dias'].mean().reset_index()
```

```
fecha_i = pd.Timestamp('2012-07-01')
```

```
fecha_f = pd.Timestamp('2020-11-01')
```

```
df_cetes_mes['Fecha'] = [pd.Timestamp(str(j)) for j in df_cetes_mes["Fecha"]]
```

```
df_cetes_mes = df_cetes_mes[(df_cetes_mes['Fecha']>=fecha_i) & (df_cetes_mes['Fecha']
plt.plot(df_cetes_mes['Fecha'],df_cetes_mes['Cetes a 28 Dias'])
df_cetes_mes.head()
```



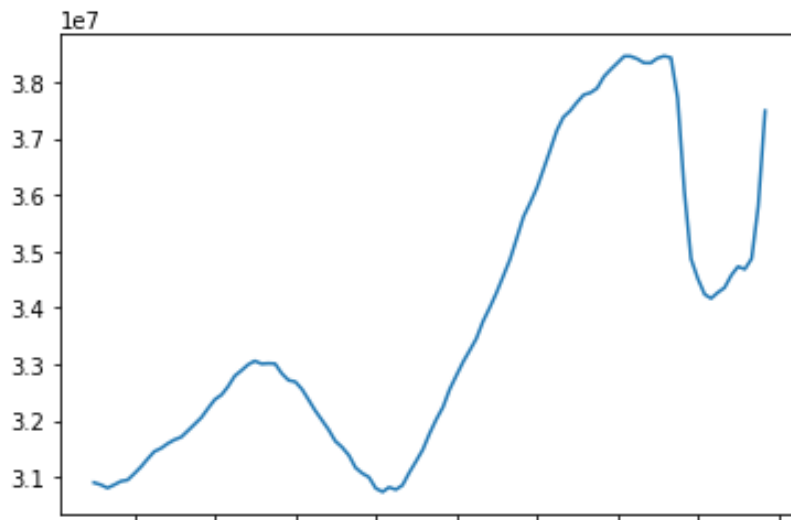
```
df_exportaciones = pd.read_excel(r'Exportaciones.xlsx')
result=seasonal_decompose(df_exportaciones['Exportaciones'], model='additive', period=12)
df_exportaciones['Trend Exportaciones'] = result.trend
df_exportaciones_trend_mes = df_exportaciones.groupby(pd.PeriodIndex(df_exportaciones.index,
freq="M"))['Trend Exportaciones'].mean().reset_index()
```

```
df_exportaciones_trend_mes['Fecha'] = [pd.Timestamp(str(j)) for j in df_exportaciones.index]
```

```
print(df_exportaciones_trend_mes.head())
df_exportaciones_trend_mes = df_exportaciones_trend_mes[(df_exportaciones_trend_mes['Fecha']>=fecha_i) & (df_exportaciones_trend_mes['Fecha']<=fecha_f)]
plt.plot(df_exportaciones_trend_mes['Fecha'],df_exportaciones_trend_mes['Trend Exportaciones'])
df_exportaciones_trend_mes.head()
```

| | Fecha | Trend Exportaciones |
|---|------------|---------------------|
| 0 | 1993-01-01 | NaN |
| 1 | 1993-02-01 | NaN |
| 2 | 1993-03-01 | NaN |
| 3 | 1993-04-01 | NaN |
| 4 | 1993-05-01 | NaN |

| | Fecha | Trend Exportaciones |
|------------|------------|---------------------|
| 234 | 2012-07-01 | 3.089820e+07 |
| 235 | 2012-08-01 | 3.086041e+07 |
| 236 | 2012-09-01 | 3.080217e+07 |
| 237 | 2012-10-01 | 3.085861e+07 |
| 238 | 2012-11-01 | 3.092272e+07 |



```
df_exportaciones = pd.read_excel(r'Exportaciones.xlsx')
result=seasonal_decompose(df_exportaciones['Exportaciones'], model='additive', period=12)
df_exportaciones['Trend Exportaciones'] = result.trend
df_exportaciones_trend_mes = df_exportaciones.groupby(pd.PeriodIndex(df_exportaciones['Fecha'], freq="M"))['Trend Exportaciones'].mean().reset_index()
```

```
df_exportaciones_trend_mes['Fecha'] = [pd.Timestamp(str(j)) for j in df_exportaciones_trend_mes.index]
```

```
print(df_exportaciones_trend_mes.head())
df_exportaciones_trend_mes = df_exportaciones_trend_mes[(df_exportaciones_trend_mes['Fecha'] >= '2012-07-01') && (df_exportaciones_trend_mes['Fecha'] <= '2012-11-01')]
plt.plot(df_exportaciones_trend_mes['Fecha'],df_exportaciones_trend_mes['Trend Exportaciones'])
df_exportaciones_trend_mes.head()
```

```
df_importaciones = pd.read_excel(r'Importaciones.xlsx')
```

```

result=seasonal_decompose(df_importaciones['Importaciones'], model='additive', period=12)
df_importaciones['Trend Importaciones'] = result.trend
df_importaciones_trend_mes = df_importaciones.groupby(pd.PeriodIndex(df_importaciones['Fecha'], freq="M"))['Trend Importaciones'].mean().reset_index()

```

```
df_importaciones_trend_mes['Fecha'] = [pd.Timestamp(str(j)) for j in df_importaciones['Fecha']]
```

```
print(df_importaciones_trend_mes.head())
```

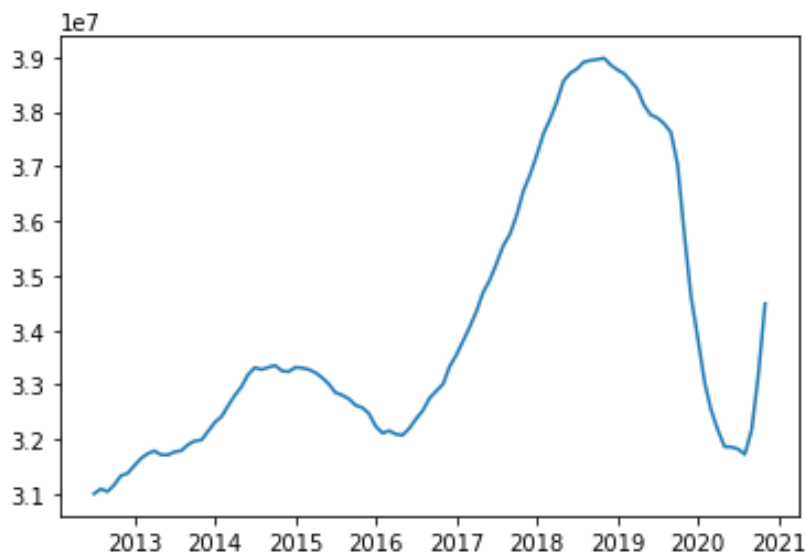
```

df_importaciones_trend_mes = df_importaciones_trend_mes[(df_importaciones_trend_mes['Fecha'] >= pd.Timestamp('2012-07-01'))]
plt.plot(df_importaciones_trend_mes['Fecha'],df_importaciones_trend_mes['Trend Importaciones'])
df_importaciones_trend_mes.head()

```

| | Fecha | Trend Importaciones |
|---|------------|---------------------|
| 0 | 1993-01-01 | NaN |
| 1 | 1993-02-01 | NaN |
| 2 | 1993-03-01 | NaN |
| 3 | 1993-04-01 | NaN |
| 4 | 1993-05-01 | NaN |

| | Fecha | Trend Importaciones |
|------------|------------|---------------------|
| 234 | 2012-07-01 | 3.100446e+07 |
| 235 | 2012-08-01 | 3.109416e+07 |
| 236 | 2012-09-01 | 3.104571e+07 |
| 237 | 2012-10-01 | 3.116630e+07 |
| 238 | 2012-11-01 | 3.133753e+07 |



Haz doble clic (o ingresa) para editar

▼ Variables

```
df = pd.read_excel(r'lab1 mensuales.xlsx')

df = df.rename(columns={df.columns[0]: "Fecha"})
df = df.set_index('Fecha')

df = pd.merge(df, df_oil_mes, on="Fecha", how="left")
df = pd.merge(df, df_petroleo_mes, on="Fecha", how="left")
df = pd.merge(df, df_cetes_mes, on="Fecha", how="left")
df = pd.merge(df, df_interes_mes, on="Fecha", how="left")
df = pd.merge(df, df_exportaciones_trend_mes, on="Fecha", how="left")
df = pd.merge(df, df_importaciones_trend_mes, on="Fecha", how="left")

df = pd.merge(df, df_dolar_mes, on="Fecha", how="left")

df = df.set_index('Fecha')

df.tail()
```

| | confianza | exportaciones | importaciones | incertidumbre política | INPC Mensual | Infla Subyac Men |
|--|-----------|---------------|---------------|---------------------------|-----------------|------------------------|
|--|-----------|---------------|---------------|---------------------------|-----------------|------------------------|

Fecha

```
#Normalizamos los datos
for a in df.columns:
    df[a] = (df[a]-np.min(df[a]))/ (np.max(df[a]) - np.min(df[a]))
df.head()
```

| | confianza | exportaciones | importaciones | incertidumbre política | INPC Mensual | Infla Subyac Mer |
|--|-----------|---------------|---------------|---------------------------|-----------------|------------------------|
|--|-----------|---------------|---------------|---------------------------|-----------------|------------------------|

Fecha

| | | | | | | |
|------------|----------|----------|----------|------|----------|------|
| 2012-07-01 | 0.584822 | 0.510949 | 0.401792 | 0.18 | 0.579336 | 0.41 |
| 2012-08-01 | 0.591384 | 0.568249 | 0.487114 | 0.20 | 0.483395 | 0.32 |
| 2012-09-01 | 0.525223 | 0.468375 | 0.329829 | 0.17 | 0.535055 | 0.27 |
| 2012-10-01 | 0.578894 | 0.663895 | 0.615805 | 0.13 | 0.560886 | 0.33 |
| 2012-11-01 | 0.566421 | 0.560227 | 0.490390 | 0.10 | 0.623616 | 0.13 |



```
# Estadistica descriptiva
df.describe()
```

| | confianza | exportaciones | importaciones | incertidumbre política | INPC Mensual | In Sub |
|--------------|------------|---------------|---------------|---------------------------|-----------------|-----------|
| count | 101.000000 | 101.000000 | 101.000000 | 101.000000 | 101.000000 | 101 |
| mean | 0.496308 | 0.657719 | 0.549298 | 0.227426 | 0.492894 | C |
| std | 0.179545 | 0.170270 | 0.161462 | 0.189165 | 0.136236 | C |
| min | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | C |
| 25% | 0.394694 | 0.560227 | 0.441206 | 0.080000 | 0.428044 | C |
| 50% | 0.463806 | 0.642770 | 0.540851 | 0.170000 | 0.512915 | C |
| 75% | 0.585466 | 0.787235 | 0.634260 | 0.320000 | 0.575646 | C |
| max | 1.000000 | 1.000000 | 1.000000 | 1.000000 | 1.000000 | 1 |

```

for col in df.columns:
    if col == 'dolar':
        continue
    plt.plot(df['dolar'])
    plt.plot(df[col])
    print('Variable: ' + col )
    print('\n Gráfica como función del tiempo')
    plt.show()

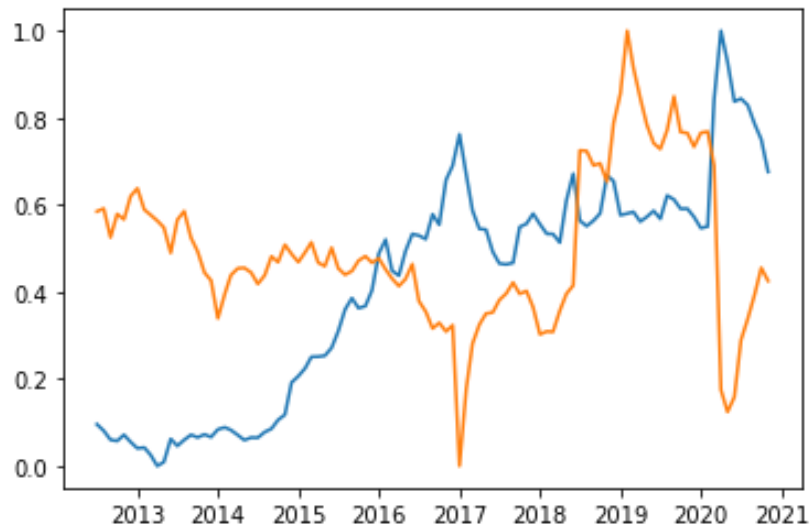
print('\n Grafica de variable vs dolar e histogramas')
sns.jointplot(x=df[col], y=df['dolar'])
plt.show()

print('\n\n\n')

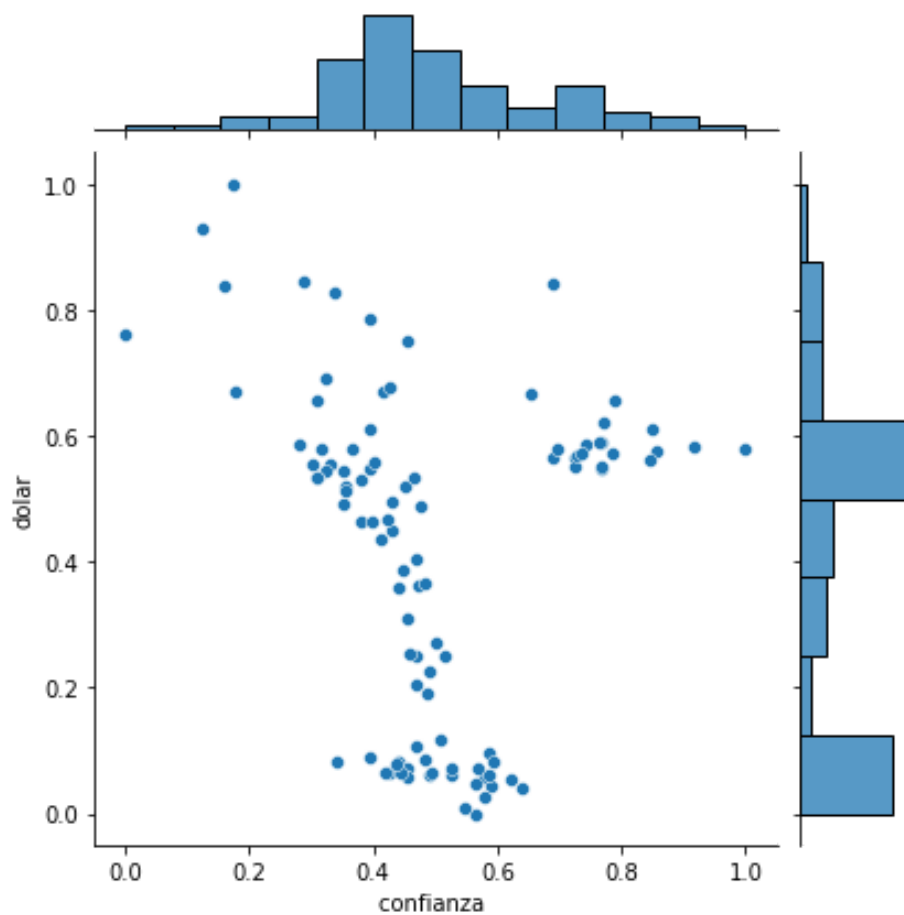
```

Variable: confianza

Gráfica como función del tiempo

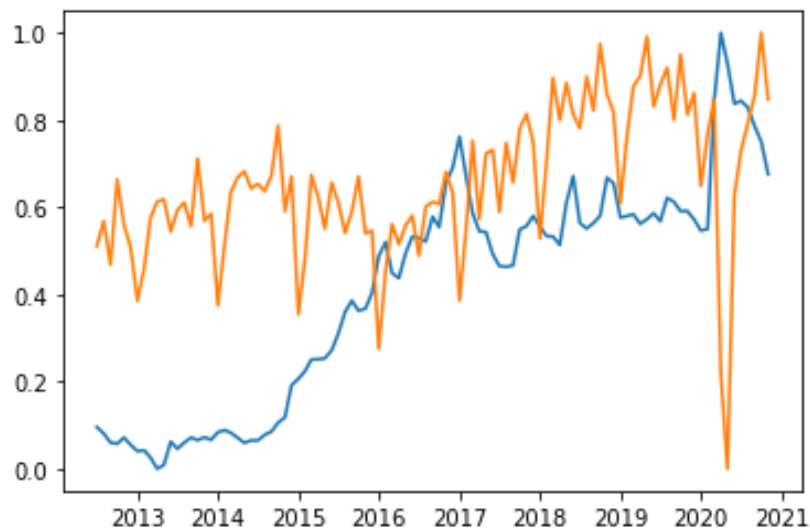


Grafica de variable vs dolar e histogramas

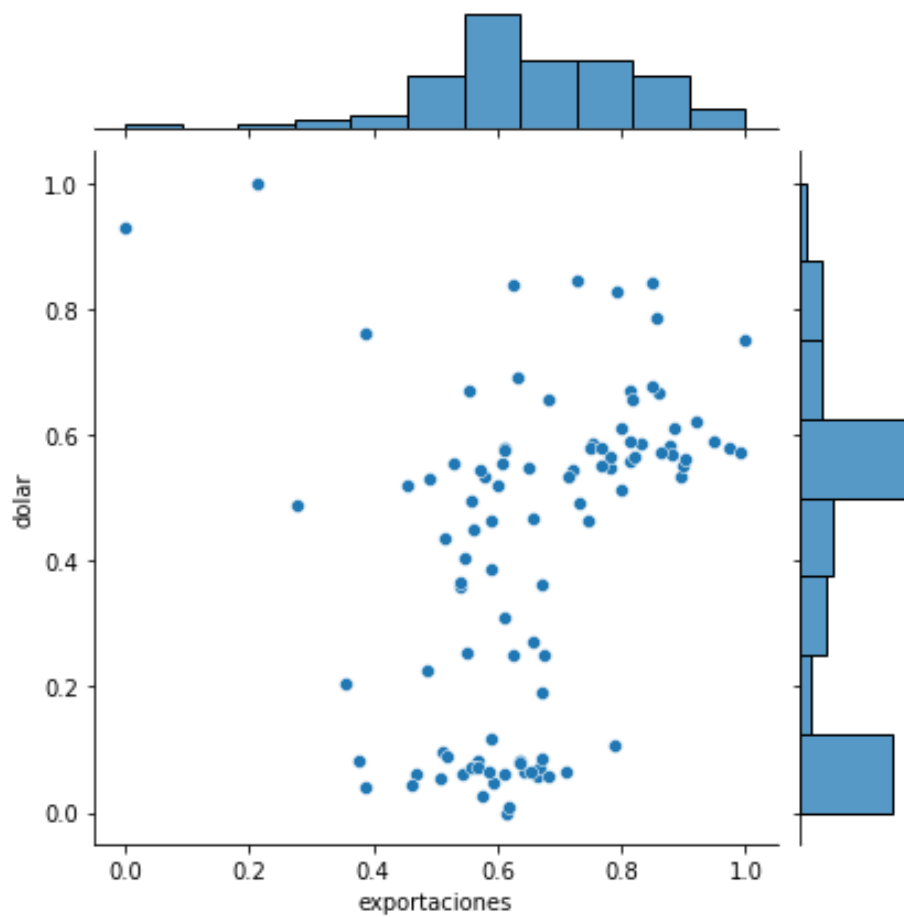


Variable: exportaciones

Gráfica como función del tiempo

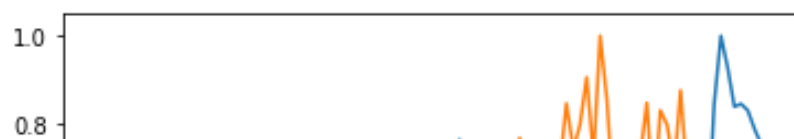


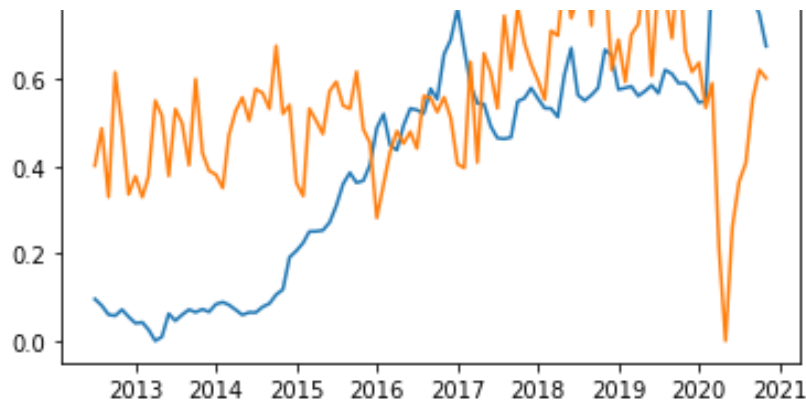
Grafica de variable vs dolar e histogramas



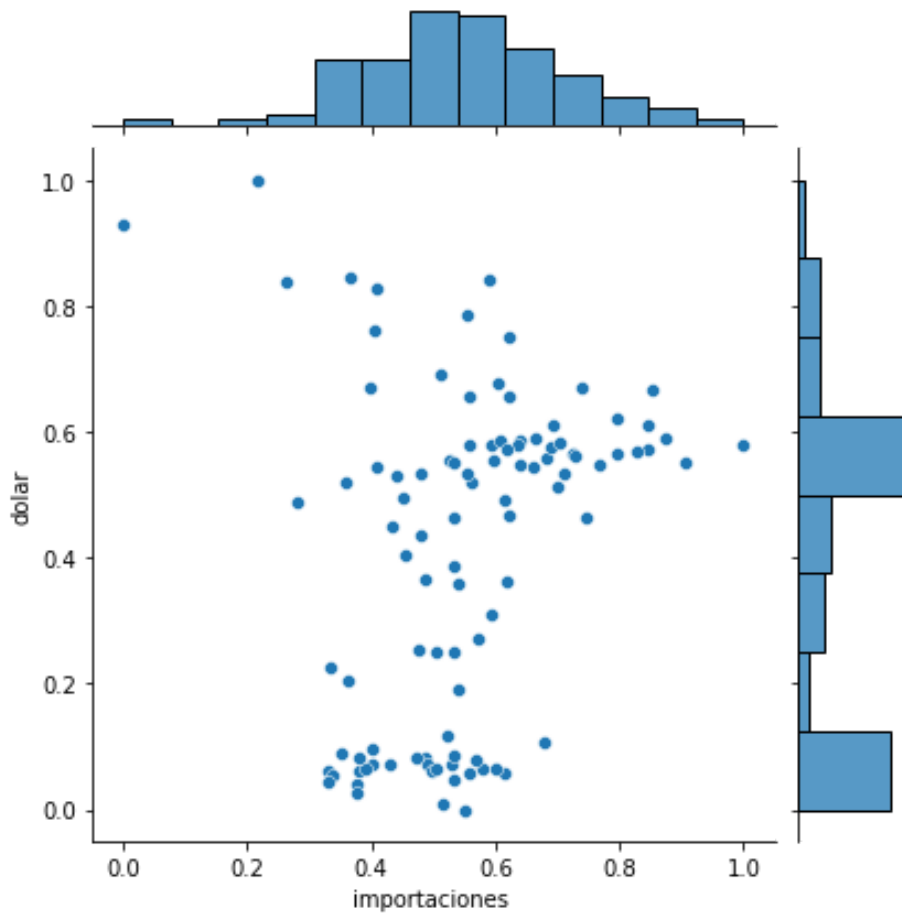
Variable: importaciones

Gráfica como función del tiempo



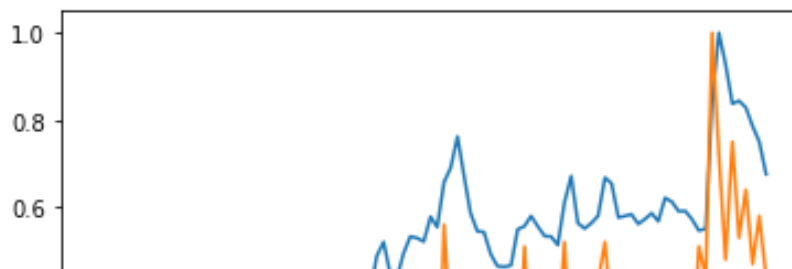


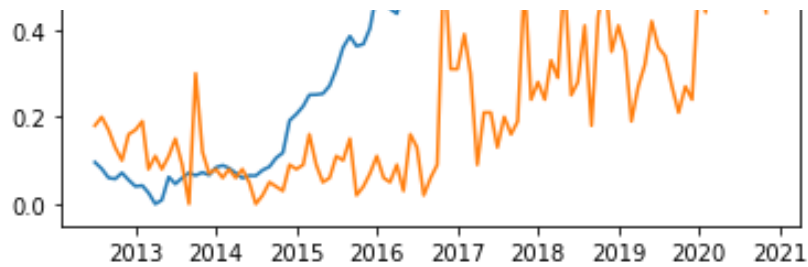
Grafica de variable vs dolar e histogramas



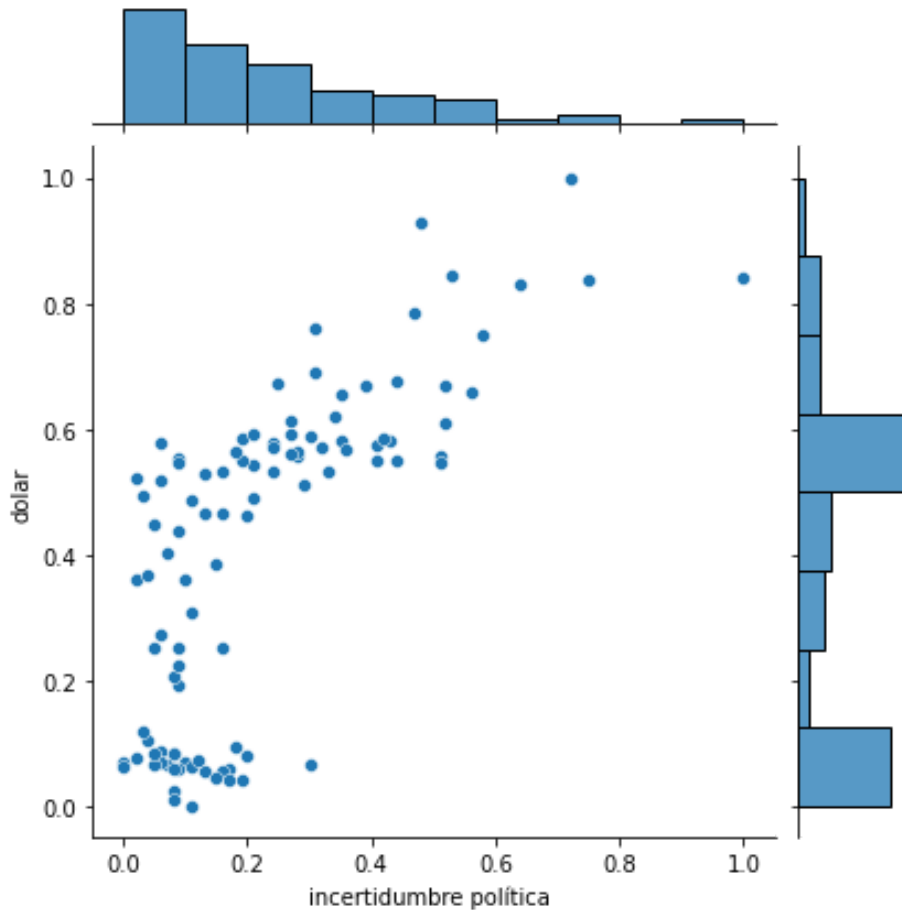
Variable: incertidumbre política

Gráfica como función del tiempo



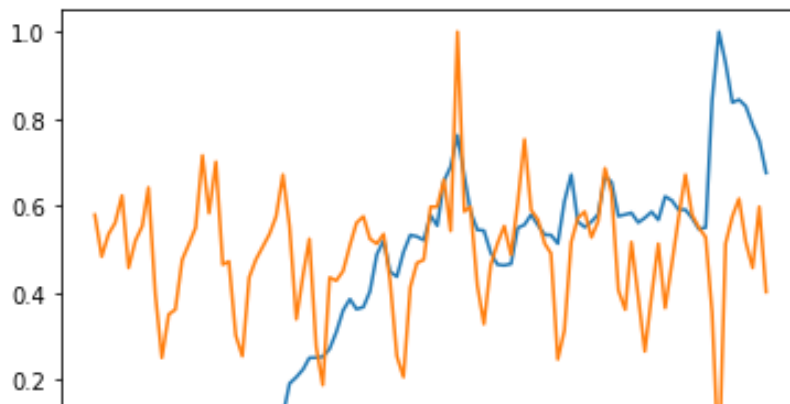


Grafica de variable vs dolar e histogramas



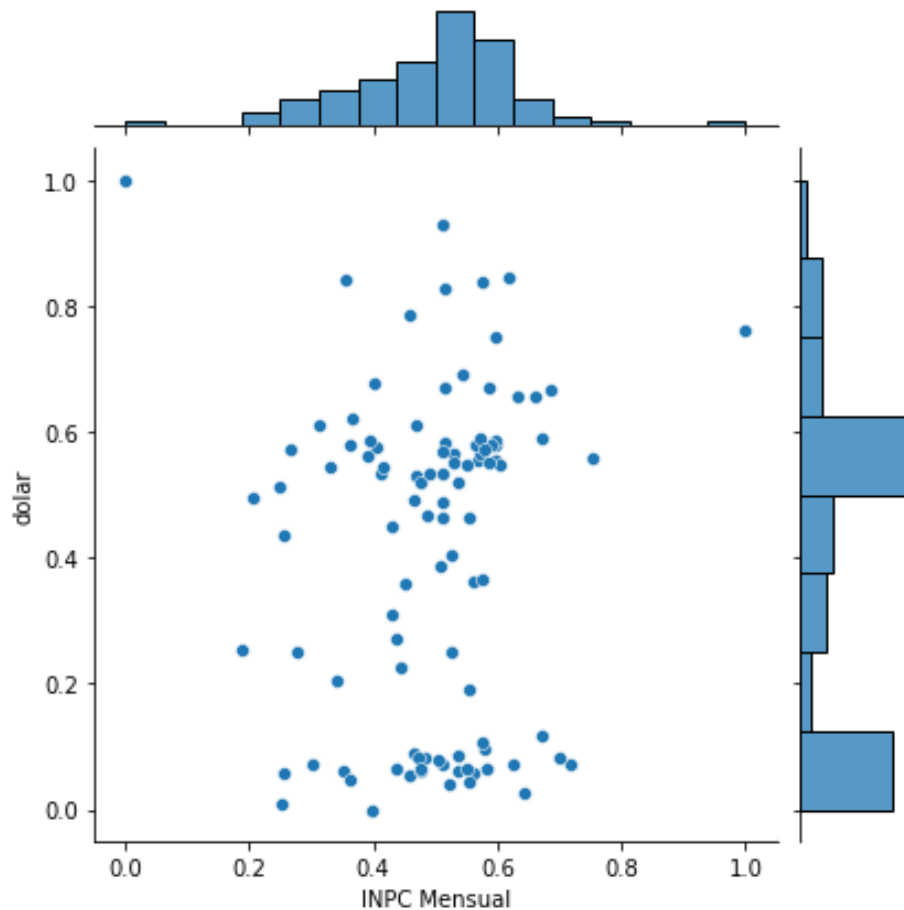
Variable: INPC Mensual

Gráfica como función del tiempo



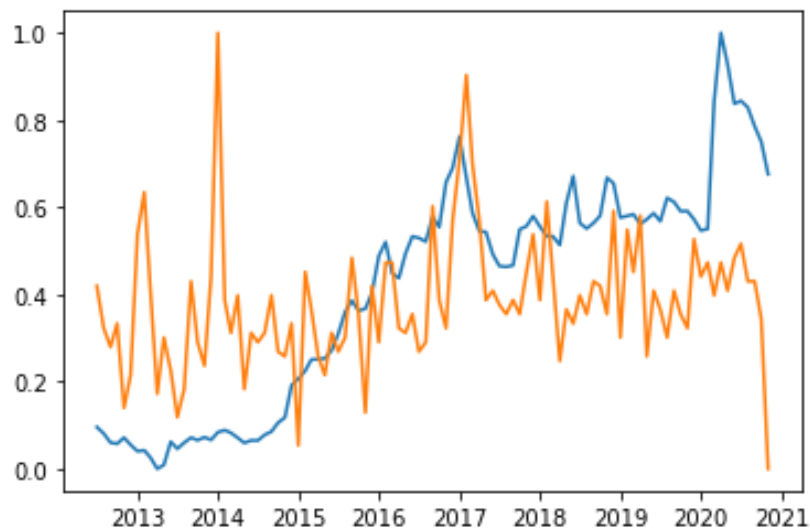


Grafica de variable vs dolar e histogramas

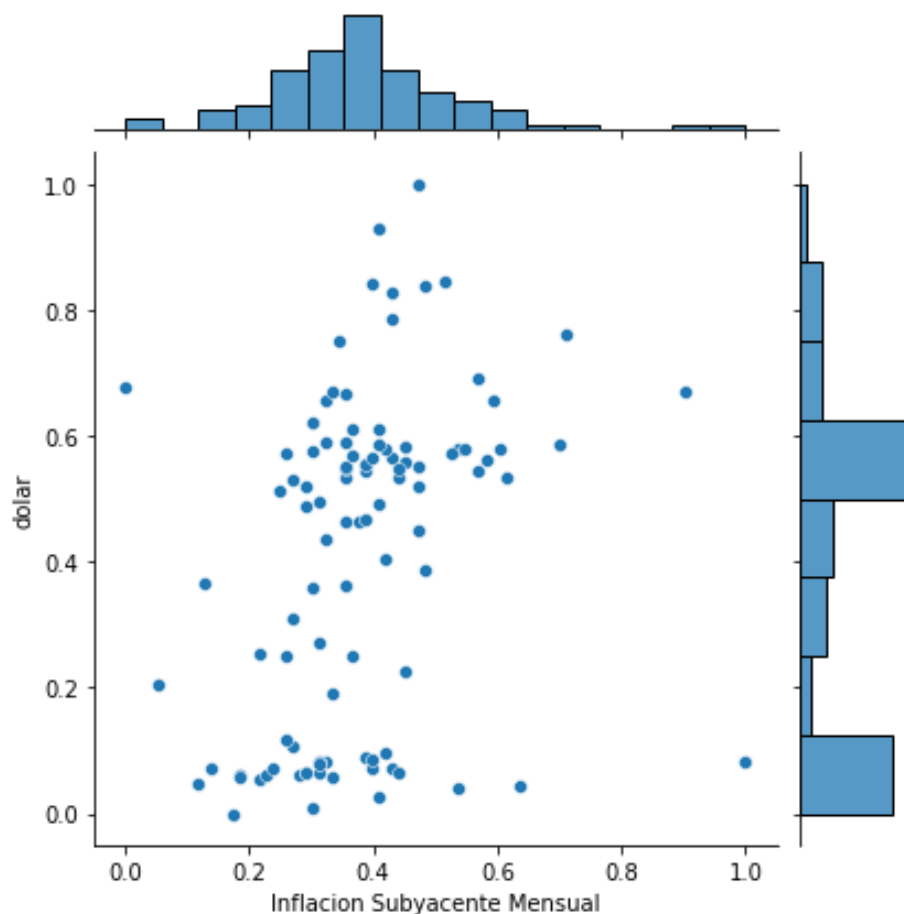


Variable: Inflacion Subyacente Mensual

Gráfica como función del tiempo

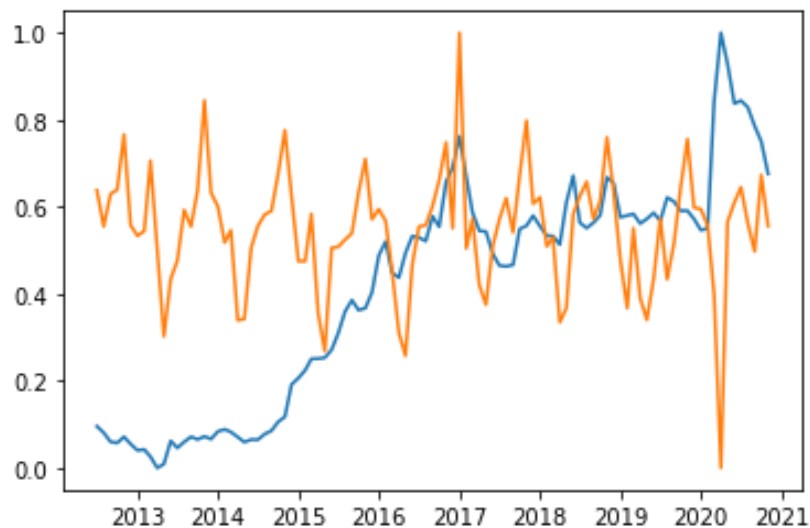


Grafica de variable vs dolar e histogramas



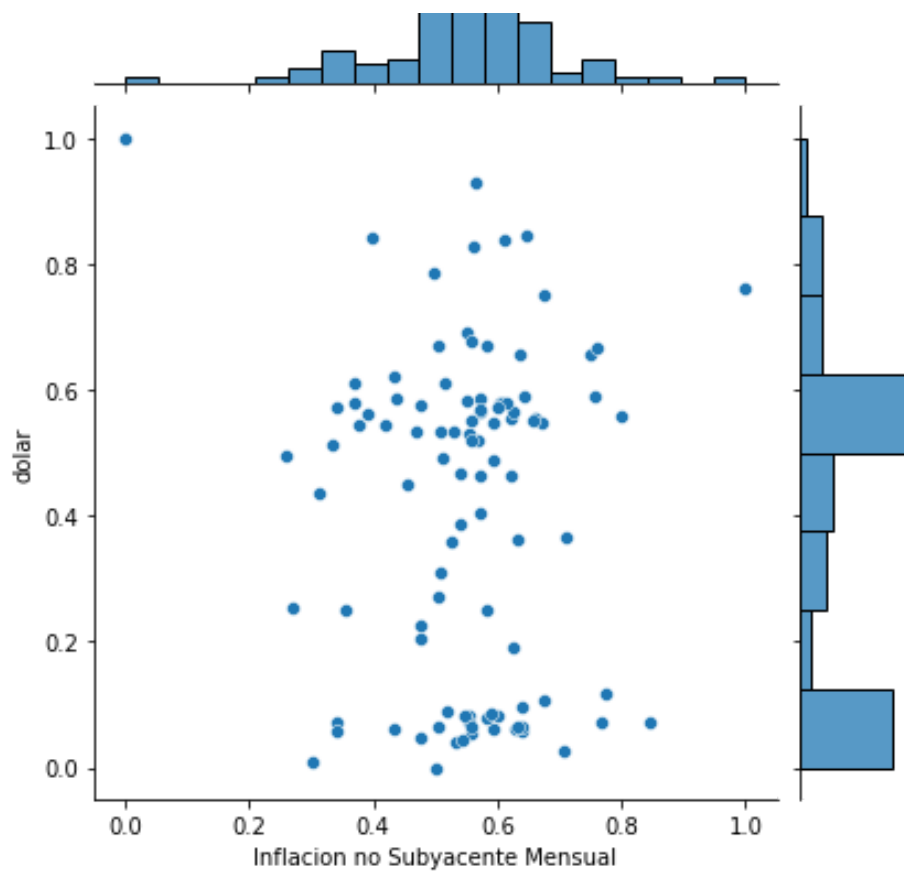
Variable: Inflacion no Subyacente Mensual

Gráfica como función del tiempo



Grafica de variable vs dolar e histogramas



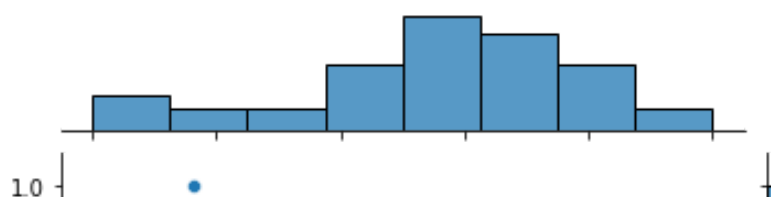


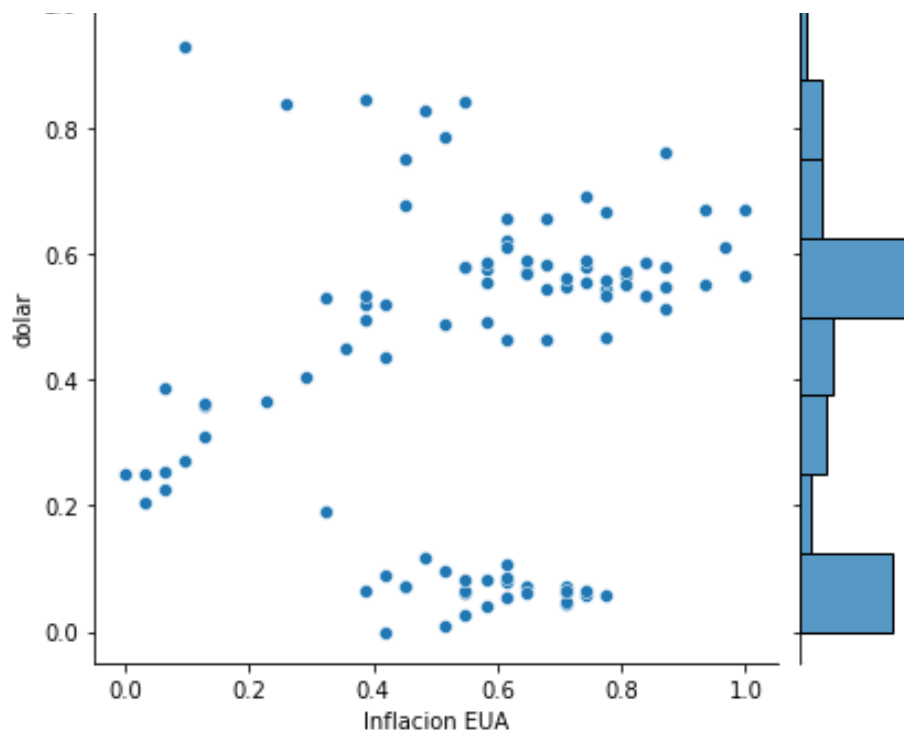
Variable: Inflacion EUA

Gráfica como función del tiempo



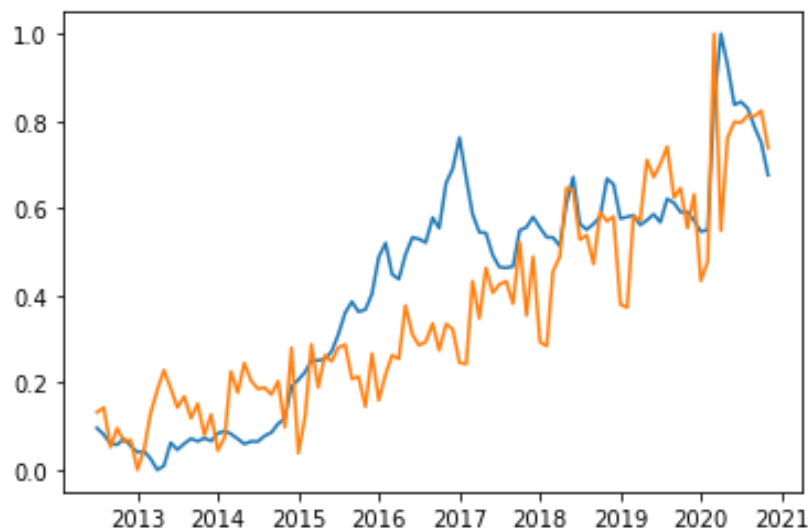
Grafica de variable vs dolar e histogramas



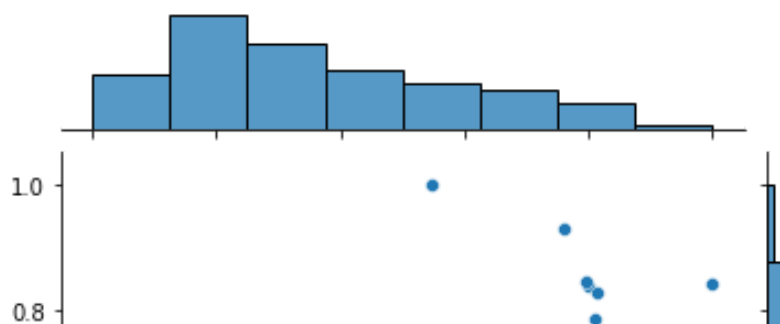


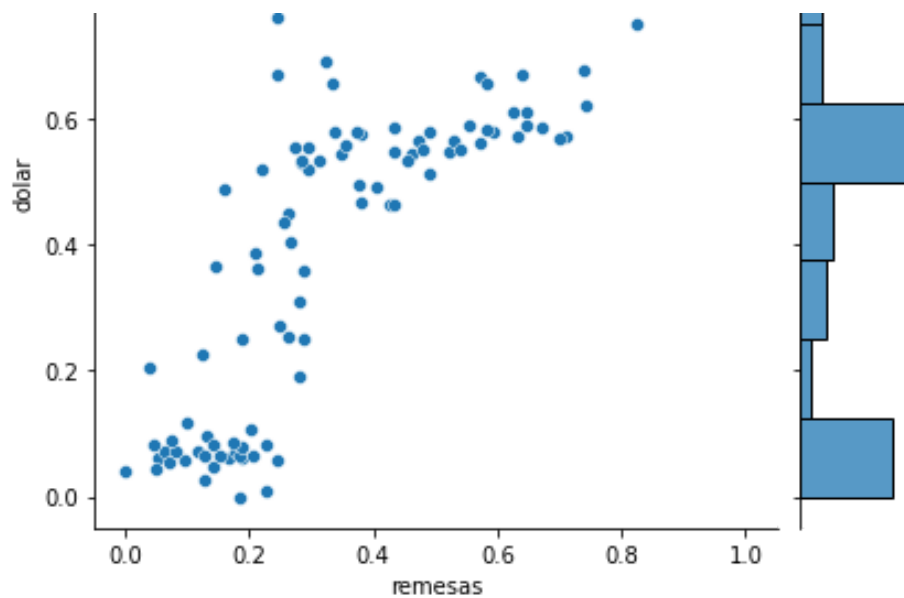
Variable: remesas

Gráfica como función del tiempo



Gráfica de variable vs dolar e histogramas



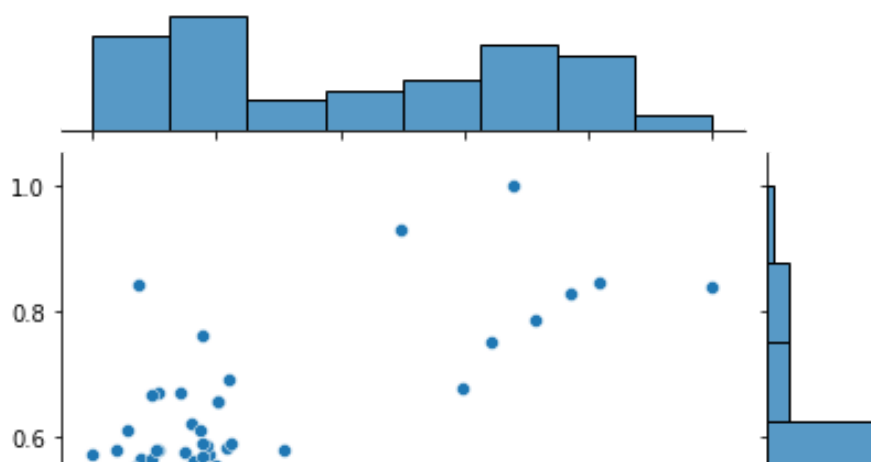


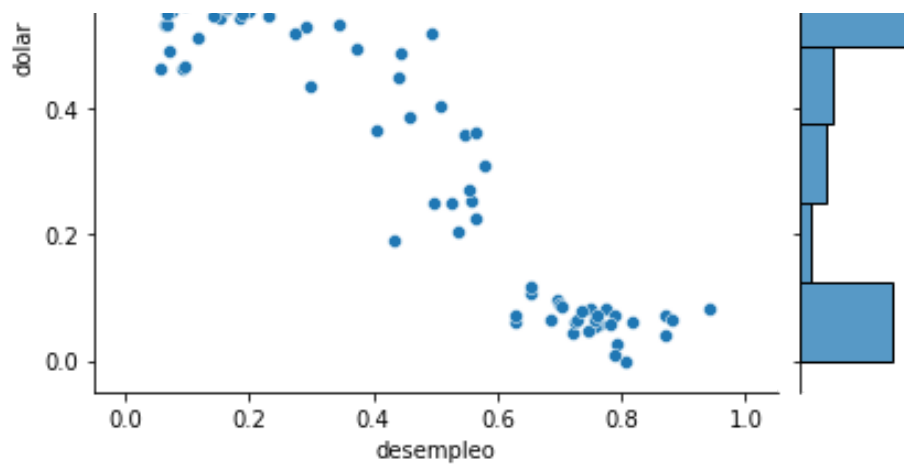
Variable: desempleo

Gráfica como función del tiempo



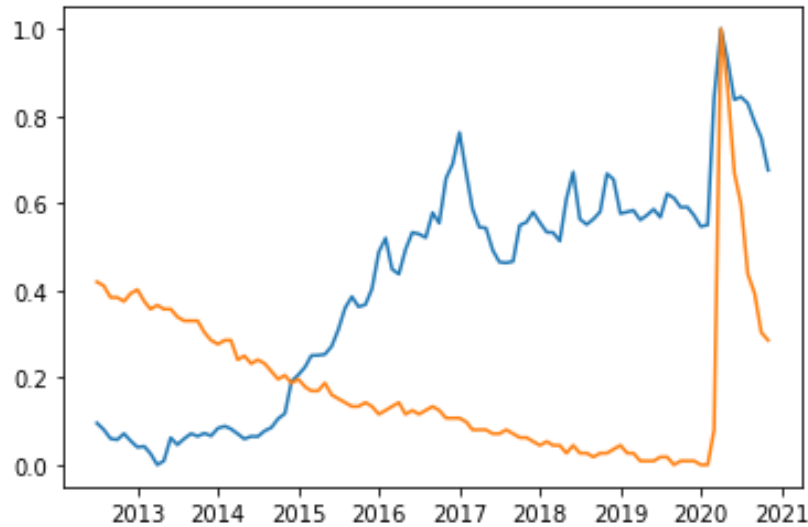
Grafica de variable vs dolar e histogramas



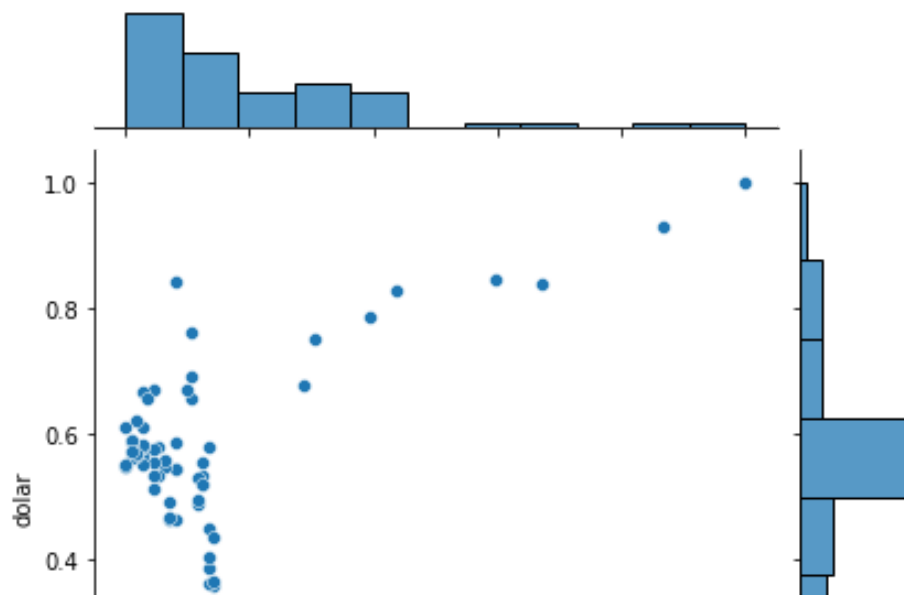


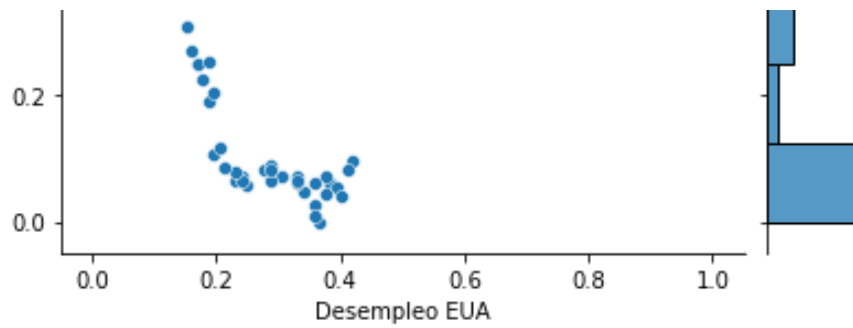
Variable: Desempleo EUA

Gráfica como función del tiempo



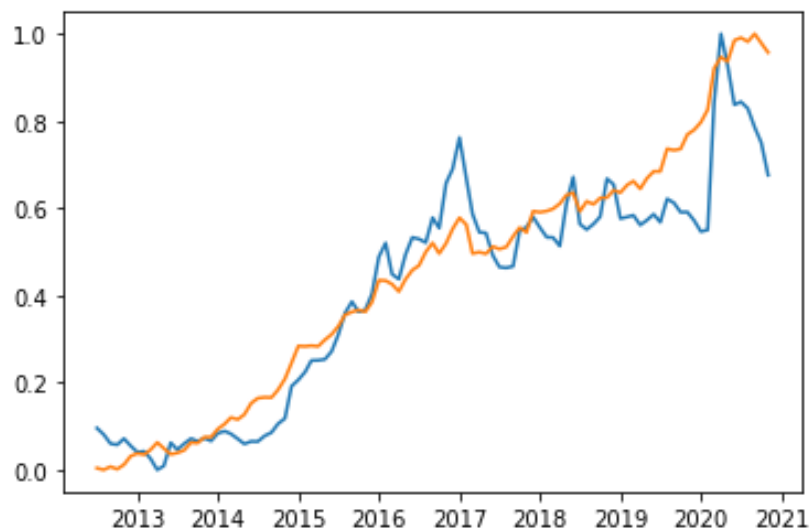
Grafica de variable vs dolar e histogramas



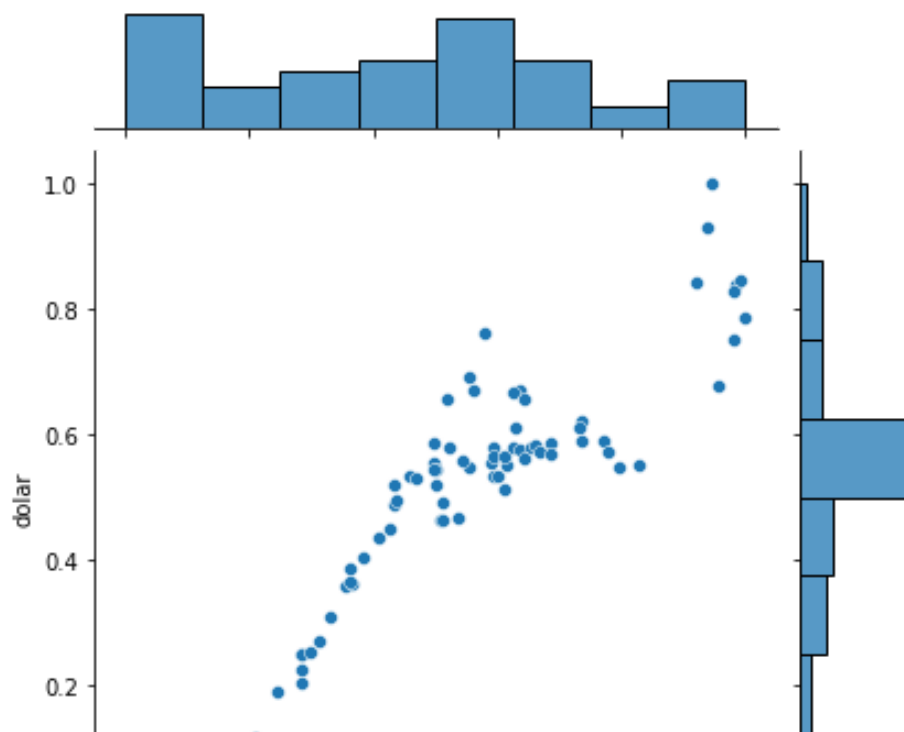


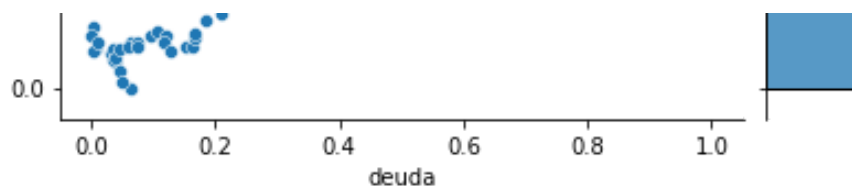
Variable: deuda

Gráfica como función del tiempo



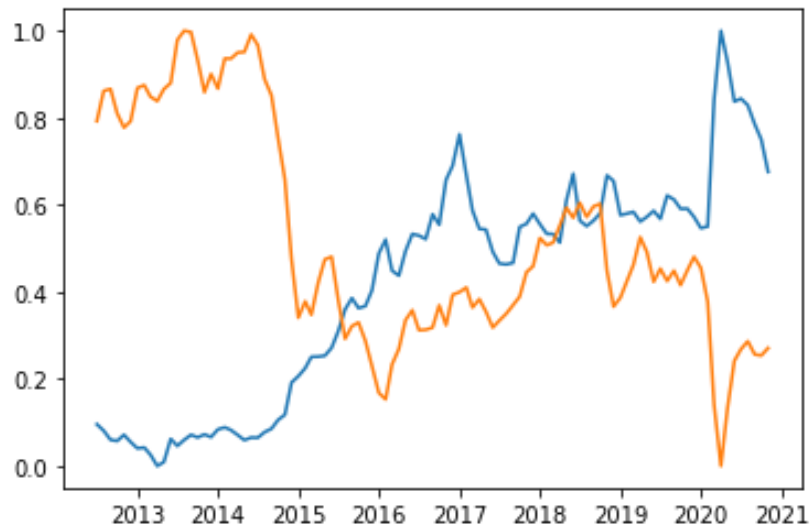
Grafica de variable vs dolar e histogramas



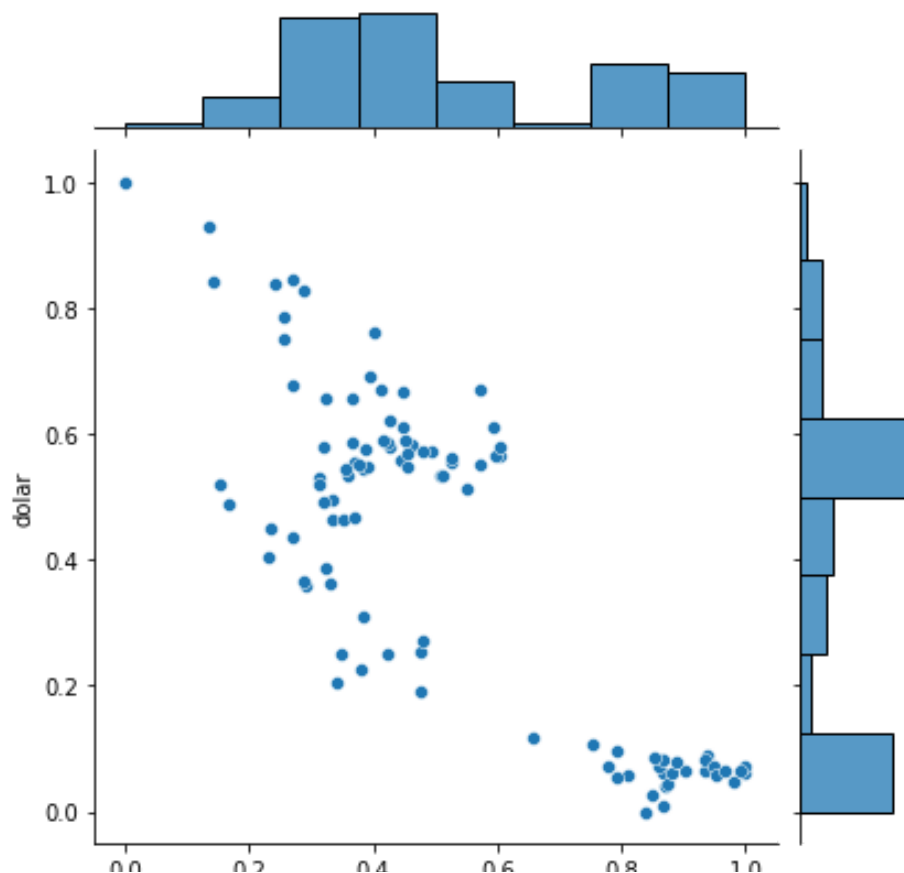


Variable: Crude Oil Price

Gráfica como función del tiempo



Grafica de variable vs dolar e histogramas



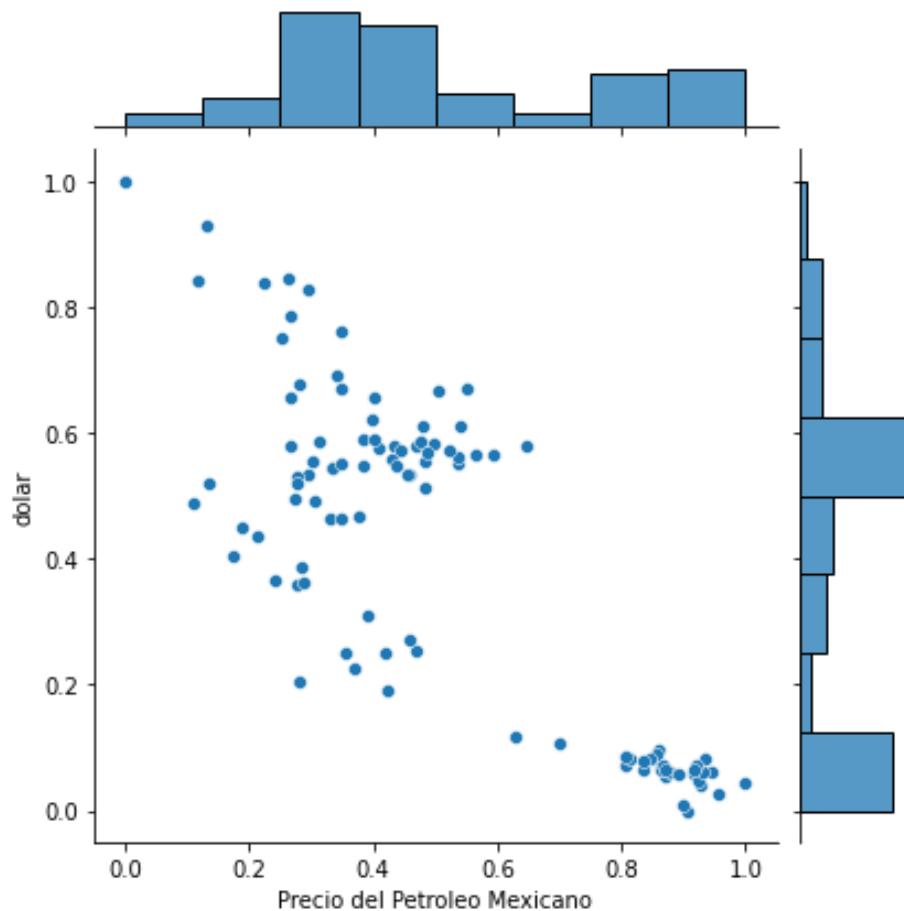
Crude Oil Price

Variable: Precio del Petroleo Mexicano

Gráfica como función del tiempo

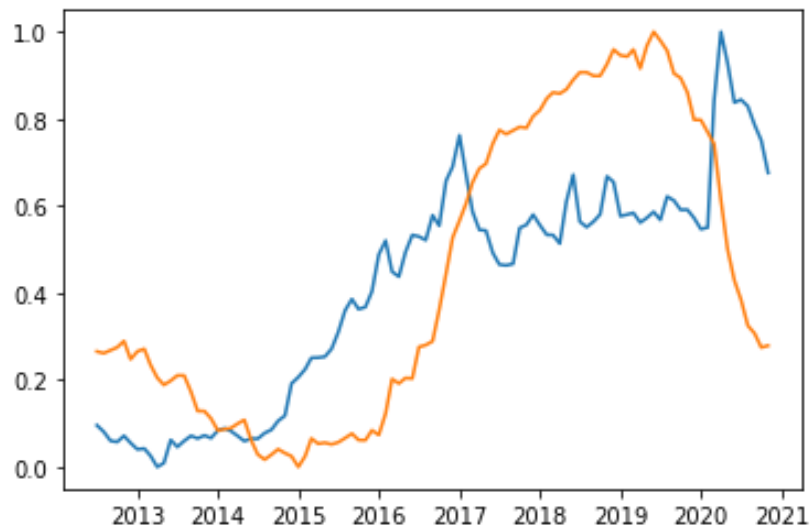


Grafica de variable vs dolar e histogramas

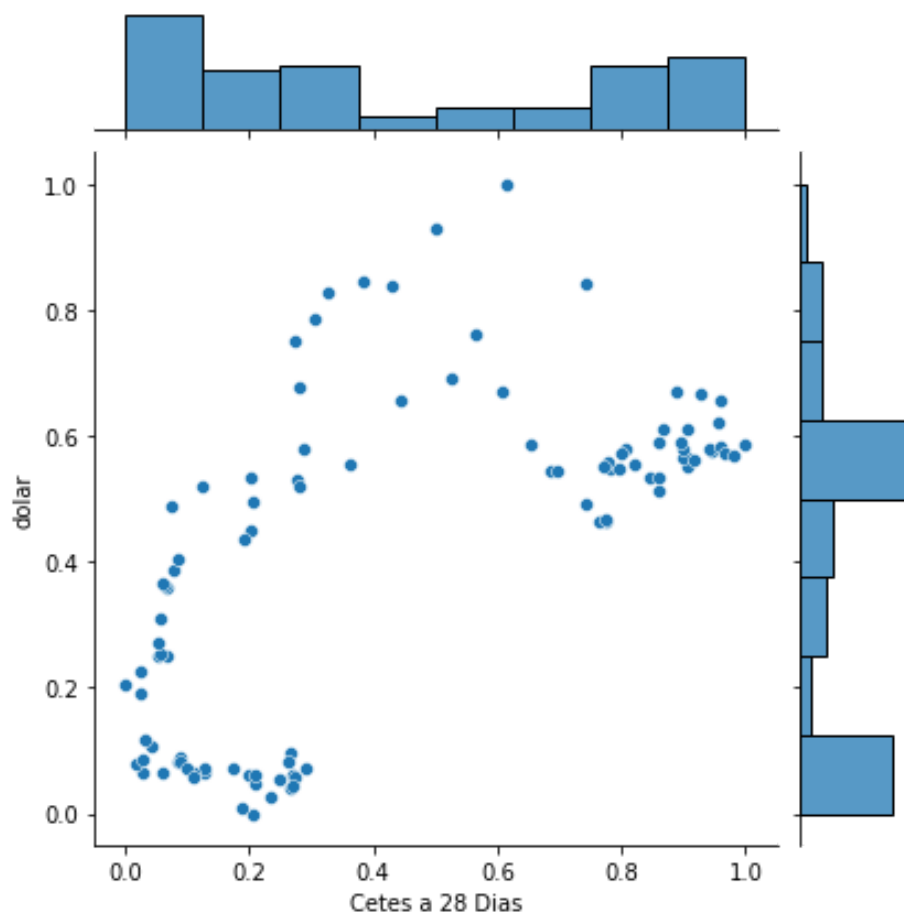


Variable: Cetes a 28 Dias

Gráfica como función del tiempo

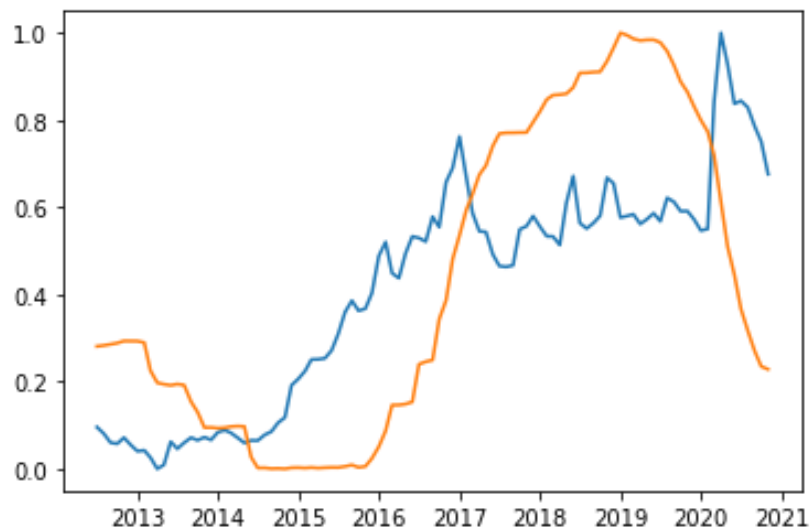


Grafica de variable vs dolar e histogramas

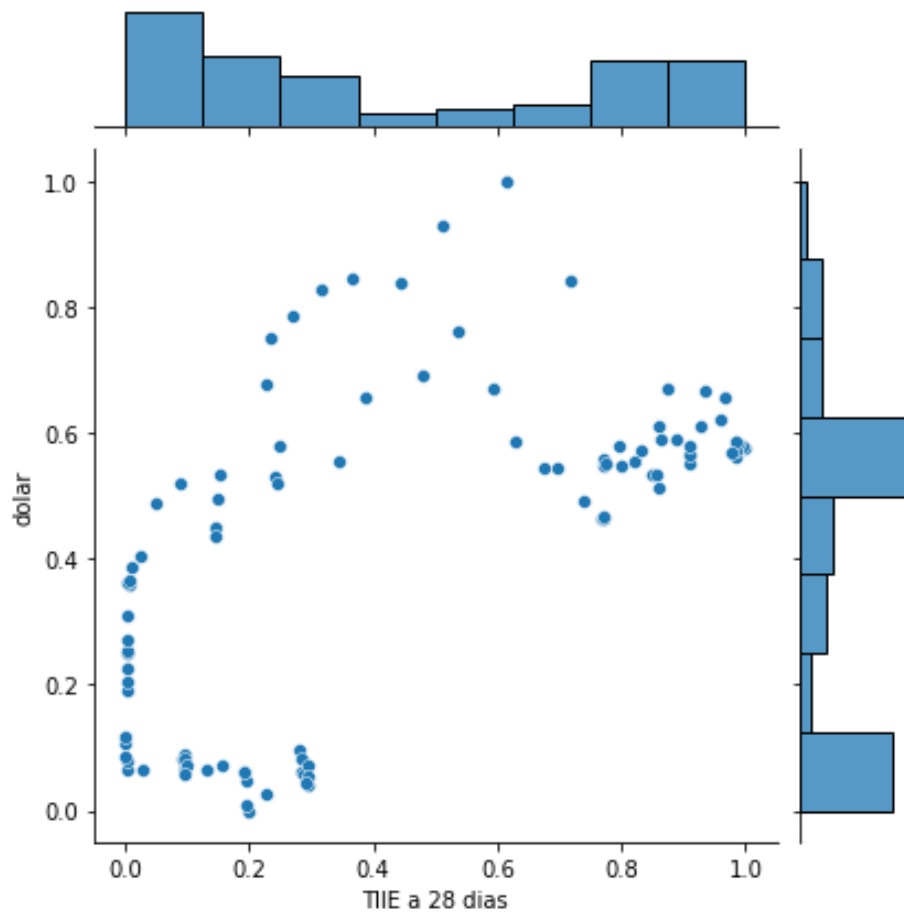


Variable: TIIIE a 28 dias

Gráfica como función del tiempo



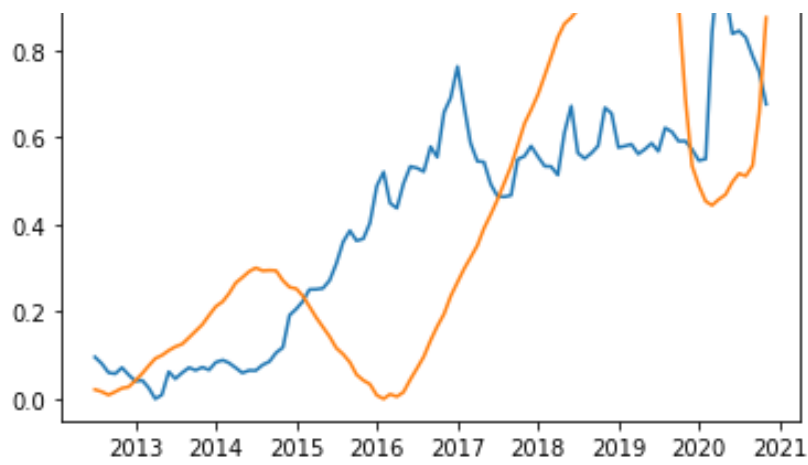
Grafica de variable vs dolar e histogramas



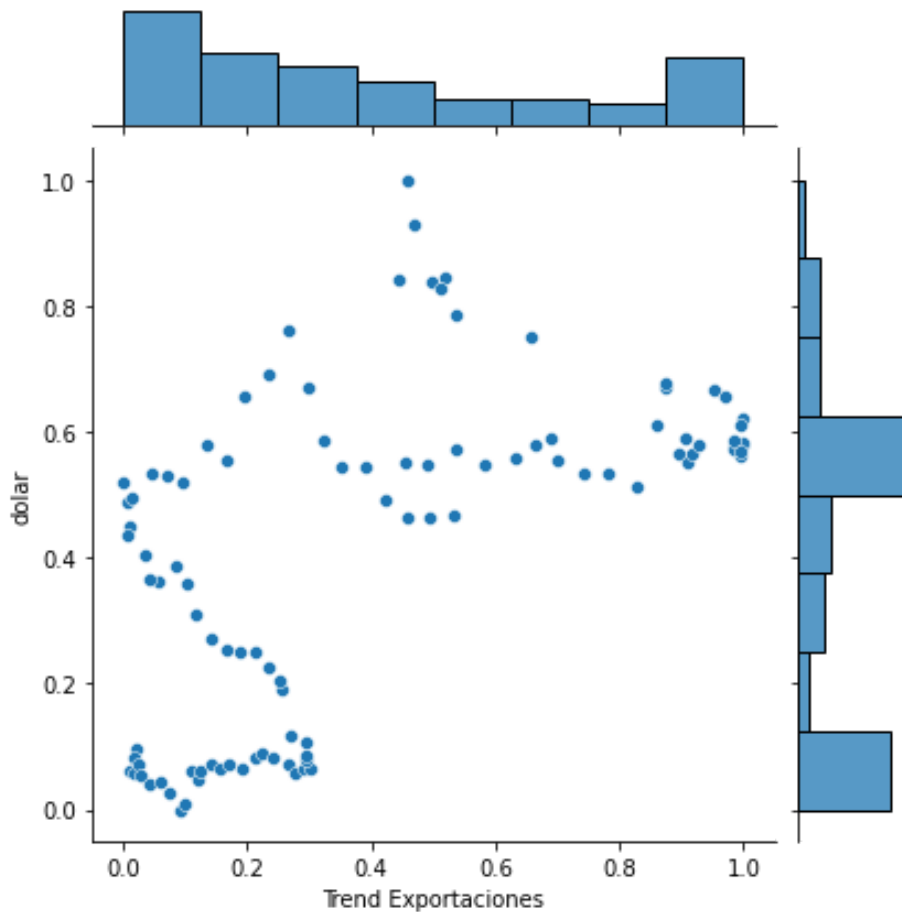
Variable: Trend Exportaciones

Gráfica como función del tiempo





Grafica de variable vs dolar e histogramas



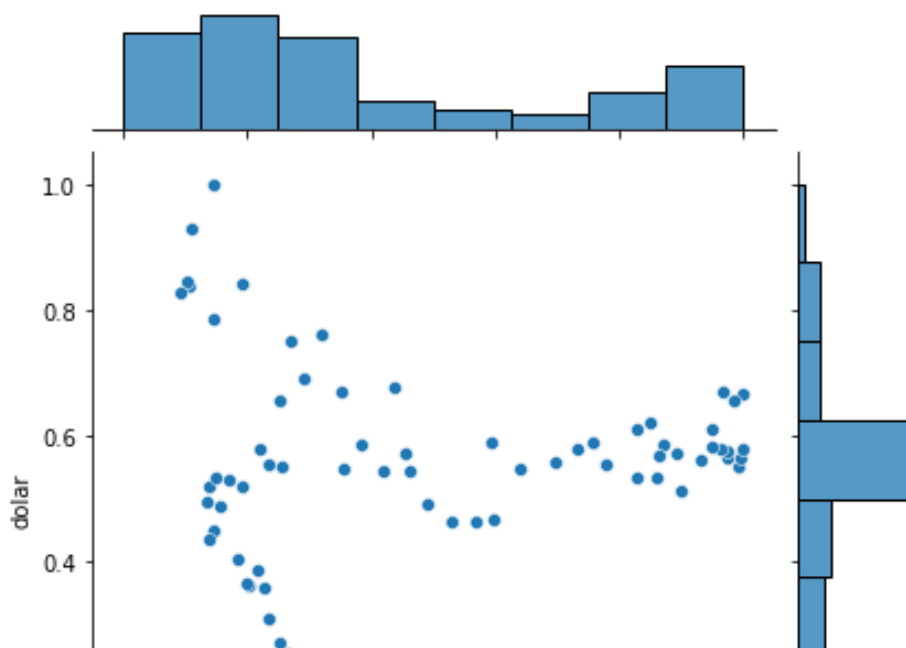
Variable: Trend Importaciones

Gráfica como función del tiempo





Grafica de variable vs dolar e histogramas

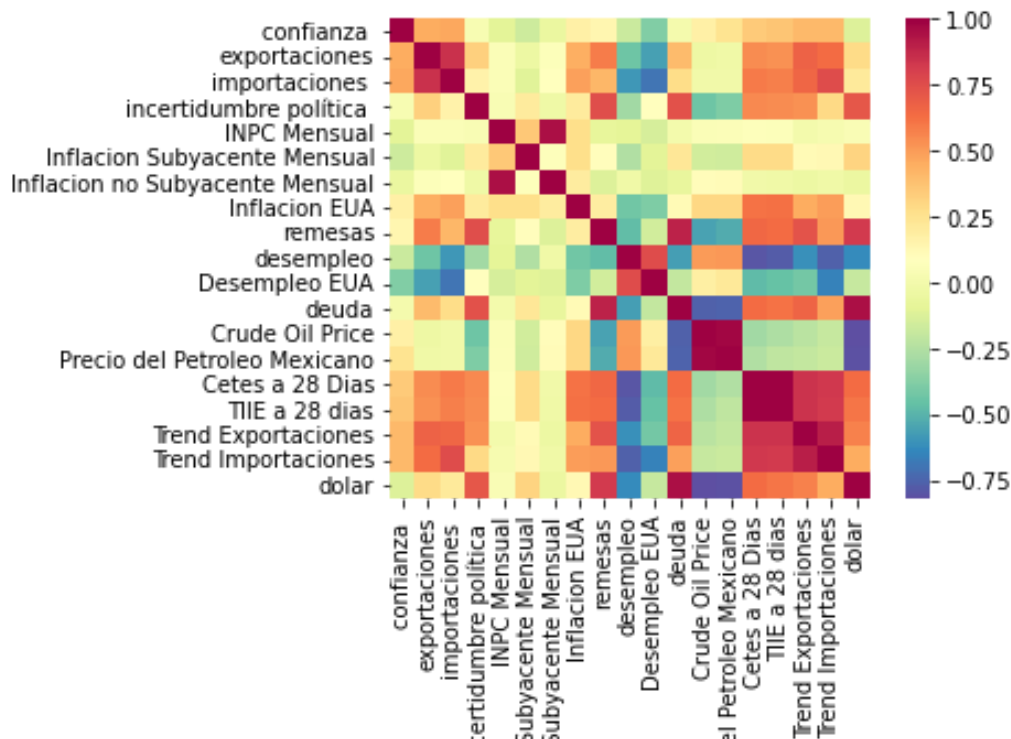


```
sns.pairplot(df)
```

```
# Correlaciones
```

```
corrmat = df.corr()
hm = sns.heatmap(corrmat,
                  cbar=True,
                  annot=False,
                  square=True,
                  fmt='.2f', 8
                  annot_kws={'size': 10},
                  yticklabels=df.columns,
                  xticklabels=df.columns,
                  cmap="Spectral_r")

plt.show()
```



```
## Test de Causalidad
```

```
from statsmodels.tsa.stattools import grangercausalitytests
```

```
def grangers_causation_matrix(data, variables, test='ssr_chi2test', verbose=False):
    maxlag=3
```

```
    """Check Granger Causality of all possible combinations of the Time series.
    The rows are the response variable, columns are predictors. The values in the table
    are the P-Values. P-Values lesser than the significance level (0.05), implies
    the Null Hypothesis that the coefficients of the corresponding past values is
    zero, that is, the X does not cause Y can be rejected.
```

```
    data      : pandas dataframe containing the time series variables
```

```
    variables : list containing names of the time series variables.
```

```
    """
```

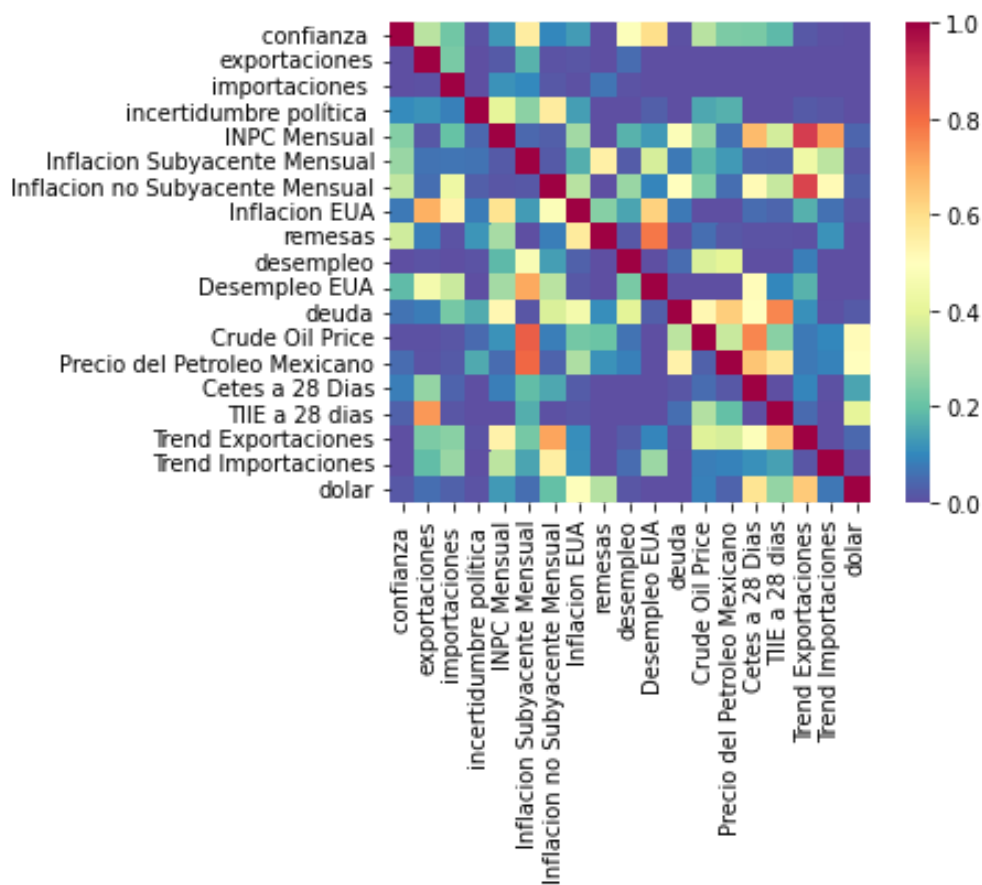
```
    df = pd.DataFrame(np.zeros((len(variables), len(variables))), columns=variables,
    for c in df.columns:
        for r in df.index:
            test_result = grangercausalitytests(data[[r, c]], maxlag=maxlag, verbose=False)
            p_values = [round(test_result[i+1][0][test][1],4) for i in range(maxlag)]
            if verbose: print(f'Y = {r}, X = {c}, P Values = {p_values}')
            min_p_value = np.min(p_values)
            df.loc[r, c] = min_p_value
    df.columns = [var + '_x' for var in variables]
    df.index = [var + '_y' for var in variables]
    return df
```

```
df_caus = grangers_causation_matrix(df, variables = df.columns)
```

```
# Causalidad
```

```
hm = sns.heatmap(df_caus,
                  cbar=True,
                  annot=False,
                  square=True,
                  fmt='.2f',
                  annot_kws={'size': 10},
                  yticklabels=df.columns,
                  xticklabels=df.columns,
                  cmap="Spectral_r")
```

```
plt.show()
```



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✓ 1 s se ejecutó 12:23

