

```
In [ ]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import plotly.express as px
import seaborn as sns

In [ ]: df_ti=pd.read_excel('Consulta_20230622-185234029.xlsx')
df_pib=pd.read_excel('Indicadores20230623152744.xlsx')
df_mxn_x=pd.read_csv('MXN=X (2).csv')
df_indicadores=pd.read_excel('Indicadores20230622190157.xls')
df_fed=pd.read_csv('FEDFUNDS (1).csv')
df_reservas=pd.read_excel('Consulta_20230623-125314125.xlsx')
df_gdp=pd.read_csv('GDP.csv')
df_desempleo_usa=pd.read_csv('UNRATE.csv')
df_cpi=pd.read_csv('CPIAUCSL.csv')

c:\Users\Pablo\anaconda3\lib\site-packages\openpyxl\styles\stylesheet.py:226: UserWarning: Workbook contains no default style, apply openpyxl's default
warn("Workbook contains no default style, apply openpyxl's default")

In [ ]: df_reservas=df_reservas[['Banco de México','Unnamed: 1']]
df_reservas.rename(columns={'Banco de México':'Periodos',
                             'Unnamed: 1':'Reserva Extranjera'},inplace=True)
df_reservas.drop(df_reservas.index[0:11],inplace=True)
df_reservas['Periodos']=pd.to_datetime(df_reservas['Periodos'])
trimestres=df_reservas['Periodos'].dt.month.isin([3, 6, 9, 12])
df_reservas=df_reservas[trimestres]
df_reservas['Periodos'] = df_reservas['Periodos'] + pd.offsets.QuarterEnd()

In [ ]: periodos=df_reservas['Periodos']
periodos=pd.DataFrame({'Periodos':periodos}).reset_index(drop=True)

In [ ]: df_pib.rename(columns={'Instituto Nacional de Estadística y Geografía (INEGI)':'Periodos',
                             'Unnamed: 1':'PIB_pct'},inplace=True)
df_pib.drop(df_pib.index[0:4],inplace=True)
df_pib.dropna(inplace=True)
df_pib.drop(['Periodos'],axis=1,inplace=True)
df_pib.reset_index(drop=True,inplace=True)
df_pib=pd.concat([periodos,df_pib],axis=1)
df_pib['Periodos'] = df_pib['Periodos'] + pd.offsets.QuarterEnd()

In [ ]: df_gdp.rename(columns={'DATE':'Periodos',
                             'GDP_PCH':'GDP_pct'},inplace=True)
df_gdp.drop(['Periodos'],axis=1,inplace=True)
df_gdp=pd.concat([periodos,df_gdp],axis=1)
df_gdp['Periodos'] = df_gdp['Periodos'] + pd.offsets.QuarterEnd()

In [ ]: df_cpi.rename(columns={'DATE':'Periodos', 'CPIAUCSL':'CPI'}, inplace=True)
df_cpi['Periodos']=pd.to_datetime(df_cpi['Periodos'])
trimestres=df_cpi['Periodos'].dt.month.isin([3,6,9,12])
df_cpi=df_cpi[trimestres]
df_cpi['Periodos'] = df_cpi['Periodos'] + pd.offsets.QuarterEnd()

In [ ]: df_desempleo_usa.rename(columns={'DATE':'Periodos','FEDFUNDS':'Tasa Desempleo USA'}, inplace=True)
df_desempleo_usa['Periodos']=pd.to_datetime(df_desempleo_usa['Periodos'])
trimestres=df_desempleo_usa['Periodos'].dt.month.isin([3,6,9,12])
df_desempleo_usa=df_desempleo_usa[trimestres]
df_desempleo_usa['Periodos'] = df_desempleo_usa['Periodos'] + pd.offsets.QuarterEnd()

In [ ]: df_ti=df_ti.drop(df_ti.iloc[:,1:-1],axis=1)
df_ti.rename(columns={'Banco de México':'Periodos', 'Unnamed: 7':'Tasa de interes'},inplace=True)
df_ti.drop(df_ti.index[0:17], inplace=True)
df_ti=df_ti.loc[-df_ti['Tasa de interes'].isin(['N/E'])]

In [ ]: df_mxn_x=df_mxn_x.drop(df_mxn_x[['Open','High','Low','Adj Close','Volume']],axis=1)
df_mxn_x.rename(columns={'Date':'Periodos',
                           'Close':'MX=X'},inplace=True)

In [ ]: df_mxn_x.rename(columns={'Date':'Periodos',
                              'Close':'MX=X'},inplace=True)
df_indicadores.loc[3].values
df_indicadores.rename(columns={'Instituto Nacional de Estadística y Geografía (INEGI)':'Periodos',
                              'Unnamed: 1':'Tasa de ocupacion',
                              'Unnamed: 2':'INPC',
                              'Unnamed: 3':'Exportaciones (FOB)',
                              'Unnamed: 4':'Importaciones (FOB)',
                              'Unnamed: 5':'Balanza Comercial',
                              'Unnamed: 6':'Ingresos y egresos brutos'},inplace=True)
df_indicadores=df_indicadores.drop(['Exportaciones (FOB)','Importaciones (FOB)',axis=1)
df_indicadores.drop(df_indicadores.index[0:4], inplace=True)
df_indicadores.drop(['Ingresos y egresos brutos'],axis=1, inplace=True)
df_indicadores=df_indicadores.dropna()
df_indicadores['Periodos']=pd.to_datetime(df_indicadores['Periodos'])
trimestres=df_indicadores['Periodos'].dt.month.isin([3, 6, 9, 12])
df_indicadores=df_indicadores[trimestres]
df_indicadores['Periodos'] = df_indicadores['Periodos'] + pd.offsets.QuarterEnd()

In [ ]: df_mxn_x['Periodos']=pd.to_datetime(df_mxn_x['Periodos'])
trimestres=df_mxn_x['Periodos'].dt.month.isin([3,6,9,12])
df_mxn_x=df_mxn_x[trimestres]
df_mxn_x['Periodos'] = df_mxn_x['Periodos'] + pd.offsets.QuarterEnd()

In [ ]: df_ti['Periodos']=pd.to_datetime(df_ti['Periodos'])
df_ti=df_ti.resample('Q', on='Periodos').last().reset_index()

In [ ]: df_fed.rename(columns={'DATE':'Periodos', 'FEDFUNDS':'Tasa Fed'}, inplace=True)
df_fed['Periodos']=pd.to_datetime(df_fed['Periodos'])
trimestres=df_fed['Periodos'].dt.month.isin([3,6,9,12])
df_fed=df_fed[trimestres]
df_fed['Periodos'] = df_fed['Periodos'] + pd.offsets.QuarterEnd()
fecha=pd.to_datetime('2023-03-31')
dato=4.33
datos={'Periodos':fecha,'Tasa Fed':dato}
df_fed=df_fed.append(datos,ignore_index=True)

C:\Users\Pablo\AppData\Local\Temp\ipykernel_52716\1699872842.py:9: FutureWarning: The frame.append method is deprecated and will be removed from pandas in a future version. Use pandas.concat instead.
df_fed=df_fed.append(datos,ignore_index=True)

In [ ]: def unir(df1,df2,columna,variable,join='inner'):
globals()[variable]=pd.merge(df1,df2,how=join,on=columna)

In [ ]: unir(df_ti, df_mxn_x, 'Periodos', 'df_total', 'right')
unir(df_total,df_indicadores,'Periodos','df_total','right')
unir(df_total,df_fed,'Periodos','df_total','right')
unir(df_total,df_cpi,'Periodos','df_total','right')
unir(df_total,df_reservas,'Periodos','df_total','right')
unir(df_total,df_pib,'Periodos','df_total','right')
unir(df_total,df_gdp,'Periodos','df_total','right')
unir(df_total,df_desempleo_usa,'Periodos','df_total','right')

In [ ]: df_total.dropna(inplace=True)
df_total=df_total.set_index('Periodos')

In [ ]: def grafico_lineal(data):
for column in data:
fig=px.line(x=data.index,y=data[column],
            title=f'{column}',height=800,width=1200)
fig.show()

In [ ]: grafico_lineal(df_total)

In [ ]: df_total.info()

<class 'pandas.core.frame.DataFrame'>
DatetimeIndex: 61 entries, 2008-03-31 to 2023-03-31
Data columns (total 11 columns):
 #   Column                Non-Null Count  Dtype
---  -
 0   Tasa de interes        61 non-null    float64
 1   MX=X                   61 non-null    float64
 2   Tasa de ocupacion      61 non-null    object
 3   INPC                   61 non-null    object
 4   Balanza Comercial      61 non-null    object
 5   Tasa Fed               61 non-null    float64
 6   CPI                   61 non-null    float64
 7   Reserva Extranjera     61 non-null    object
 8   PIB_pct               61 non-null    object
 9   GDP_pct               61 non-null    float64
10  UNRATE                 61 non-null    float64
dtypes: float64(6), object(5)
memory usage: 5.7+ KB

In [ ]: def convertir_valores(data):
if any(data.dtypes != 'float64'):
data = data.apply(pd.to_numeric, errors='coerce')
return data

In [ ]: df_total=convertir_valores(df_total)

In [ ]: df_total.describe().to_excel('descripcion_estadistica.xlsx', index=True)
df_total.describe()

Out[ ]:

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	Tasa de interes	MX=X	Tasa de ocupacion	INPC	Balanza Comercial	Tasa Fed	CPI	Reserva Extranjera	PIB_pct	GDP_pct	UN
count	61.000000	61.000000	61.000000	61.000000	61.000000	61.000000	61.000000	61.000000	61.000000	61.000000	61.0
mean	5.553279	16.321623	59.468337	91.248804	338.231954	0.751475	243.059475	165329.596148	1.55059	0.980166	6.1
std	2.018206	3.584161	1.265571	16.878043	1809.909573	0.19266	22.794233	37827.128831	4.61974	1.841201	2.1
min	3.000000	10.299000	53.111205	66.019891	-3971.164000	0.070000	211.398000	81476.217000	-18.61400	-8.827640	3.5
25%	4.250000	13.027310	59.243066	77.792385	-518.037000	0.100000	227.223000	149208.695000	1.17900	0.617950	4.3
50%	4.500000	16.913639	59.711229	87.752419	147.159000	0.180000	237.657000	178008.699000	2.62100	1.060470	5.6
75%	7.250000	19.639999	60.053019	103.476000	1001.516000	1.210000	255.159000	193331.800000	3.39100	1.430080	7.9
max	11.250000	23.833000	61.471040	128.389000	6274.687000	4.330000	301.808000	212002.533000	19.61200	8.788110	11.0

```
In [ ]: y=df_total['MX=X']

In [ ]: x=df_total.drop(['MX=X'],axis=1)

In [ ]: def correlacion(x, y):
correlacion = x.corrwith(y).abs().sort_values(ascending=False)
return correlacion

In [ ]: correlacion(x,y).to_excel('correlacion.xlsx', index=True)
correlacion(x,y)

Out[ ]:

```

INPC		0.859404
CPI		0.775686
Reserva Extranjera		0.696736
UNRATE		0.584170
Tasa de ocupacion		0.449930
Tasa de interes		0.329998
Tasa Fed		0.259633
Balanza Comercial		0.198155
PIB_pct		0.159689
GDP_pct		0.065667

```
In [ ]: def tabla_correlacion(x, y):
corr_matrix = correlacion(x, y)
plt.figure(figsize=(10, 8))
sns.heatmap(corr_matrix.to_frame(), annot=True, cmap='coolwarm', linewidths=0.5)
plt.title(f'Correlación respecto a {y.name}')
plt.xlabel('Correlacion')
plt.ylabel('Variable')
plt.show()

In [ ]: tabla_correlacion(x,y)

Correlación respecto a MX=X

```

Variable	INPC	CPI	Reserva Extranjera	UNRATE	Tasa de ocupacion	Tasa de interes	Tasa Fed	Balanza Comercial	PIB_pct	GDP_pct
	0.86	0.78	0.7	0.58	0.45	0.33	0.26	0.2	0.16	0.066

```
In [ ]: def dispersion(x,y):
corr_matrix=correlacion(x,y)
for column, correlation in corr_matrix.iteritems():
fig = px.scatter(x=x[column], y=y, trendline='ols', title=f'Dispersión entre {column} y {y.name}',height=400)
fig.show()

In [ ]: dispersion(x,y)

C:\Users\Pablo\AppData\Local\Temp\ipykernel_52716\3129074370.py:3: FutureWarning:
iteritems is deprecated and will be removed in a future version. Use .items instead.

In [ ]: from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.compose import ColumnTransformer
from sklearn.ensemble import RandomForestRegressor
from sklearn.linear_model import LinearRegression
from sklearn.pipeline import Pipeline,make_pipeline
from sklearn.experimental import enable_iterative_imputer
from sklearn.metrics import mean_squared_error, r2_score, mean_absolute_error

In [ ]: X_train,X_test,y_train,y_test=train_test_split(x[['INPC','CPI']],y,test_size=0.2,random_state=42)

In [ ]: num_col=[k for k, v in x[['INPC','CPI']].dtypes!= 'object'].to_dict().items() if v]

In [ ]: variables=Pipeline(
steps=[
('imputer',IterativeImputer()),
('scaler',StandardScaler())
])
preprocesado=ColumnTransformer(
transformers=[
('num',variables,num_col)
])
regresion=Pipeline(
steps=[
('preprocesado',preprocesado),
('regresion',RandomForestRegressor())
])

In [ ]: regresion.fit(X_train,y_train)
pipe=make_pipeline(preprocesado,regresion)
pipe

Out[ ]:

```

```

> Pipeline
> columntransformer: ColumnTransformer
> num
> IterativeImputer
> StandardScaler
> pipeline: Pipeline
> preprocesado: ColumnTransformer
> num
> IterativeImputer
> StandardScaler
> RandomForestRegressor

```

```
In [ ]: y_pred=regresion.predict(X_test)
r2_score(y_test,y_pred)

Out[ ]: 0.9501157692698131

In [ ]: importancia= regresion.named_steps['regresion'].feature_importances_
importancia=pd.DataFrame({'Variables':X_train.columns,
                           'Importancia':importancia}).sort_values(by='Importancia',ascending=False)
importancia.to_excel('importancia.xlsx',index=True)
importancia

Out[ ]:

```

	Variables	Importancia
0	INPC	0.688316
1	CPI	0.311684

```
In [ ]: def EvaluationMetric(Xt,yt,yp,disp="on"):
MSE=round(mean_squared_error(y_true=yt,y_pred=yp),4)
RMSE=(np.sqrt(MSE))
R2=(r2_score(y_true=yt,y_pred=yp))
Adjusted_R2=(1-(1-R2_score(yt,yp))*((Xt.shape[0]-1)/(Xt.shape[0]-Xt.shape[1]-1)))
if disp=="on":
print("MSE: ",MSE,"RMSE: ",RMSE)
print("R2 :",R2,"ADJUSTED R2 :",Adjusted_R2)

plt.figure(figsize=(18,6))
plt.plot((yp)[:100])
plt.plot((np.array(yt)[:100]))
plt.legend(["Prediccion","Real"])
plt.title("Precio Real y Prediccion del modelo")

return (MSE,RMSE,R2,Adjusted_R2)

In [ ]: y_predt=regresion.predict(X_train)

In [ ]: EvaluationMetric(X_train,y_train,y_predt)

MSE: 0.2208 RMSE: 0.4698936049788292
R2 : 0.9815251860996601 ADJUSTED R2 : 0.980704083259645
(0.2208, 0.4698936049788292, 0.9815251860996601, 0.980704083259645)

Out[ ]:

```

Precio Real y Prediccion del modelo

```
In [ ]: predict_df=regresion.predict(X_test)
predict_df=pd.DataFrame(np.ravel(predict), columns=['Predicciones'])
predict_df.index = y_test.index # Asignar Los mismos índices que y_test al DataFrame predict_df
comparasion_df=pd.concat([y_test, predict_df], axis=1)
comparasion_df.columns = ['MX=X', 'Predicciones']
comparasion_df.round(2)
comparasion_df.to_excel('Comparasion.xlsx',index=True)
comparasion_df

Out[ ]:

```

	MX=X	Predicciones
Periodos		
2008-03-31	10.635500	11.207503
2009-06-30	13.179000	13.764792
2019-09-30	19.678699	19.552014
2015-12-31	17.269760	16.696498
2011-06-30	11.711200	13.296581
2021-12-31	20.459101	20.238773
2016-09-30	19.368931	19.072759
2020-06-30	23.056570	22.474935
2011-03-31	11.891000	13.243269
2018-03-31	18.146400	19.586182
2016-06-30	18.322950	17.665513
2022-12-31	19.467100	19.546153
2012-03-31	12.797500	13.495377

Prueba unicamente para ver si podía predecir con ARIMA las variables y así generar una prediccion con esas mismas.