## PRODUCT SALES ANALYSIS **Problem Definition** Consider incorporating machine learning algorithms to predict future sales trends or customer behaviors Data visualization: Data visualization plays a crucial role in data analysis, enabling the extraction of meaningful insights from complex datasets. This abstract outlines the process of creating informative and visually appealing data visualizations using Python's Matplotlib library within the Jupyter Anaconda environment. The dataset chosen for this project is various parameter used for analysing water quality for domestic purpose, and the goal is to effectively communicate key trends, patterns, and relationships within the data. Matplotlib, a versatile and widely used data visualization library, offers a wide range of tools for creating various types of plots, such as line charts, bar graphs, scatter plots, heatmaps, and more. In this project, we will walk through the step-by-step process of data visualization, including data pre-processing, selecting appropriate plot types, customizing aesthetics (colours, labels, legends, etc.), and adding context to the visualizations with titles, annotations, and captions. We will also explore how to handle common data visualization challenges, such as handling missing data, creating subplots, and generating interactive visualizations. import pandas as pd In [3]: import numpy as ny from statsmodels.tsa.arima.model import ARIMA In [4]: df=pd.read\_csv('ste.csv') df Out[4]: Unnamed: 0 Date Q-P1 Q-P2 Q-P3 Q-P4 S-P1 S-P2 S-P3 S-P4 0 13-06-2010 5422 3725 576 907 17187.74 23616.50 3121.92 6466.91 1 14-06-2010 7047 779 3578 1574 22338.99 4938.86 19392.76 11222.62 2 2 15-06-2010 1572 2082 595 1145 4983.24 13199.88 3224.90 8163.85 3 16-06-2010 5657 2399 3140 1672 17932.69 15209.66 17018.80 11921.36 4 4 17-06-2010 3668 3207 2184 708 11627.56 20332.38 11837.28 5048.04 4595 4595 30-01-2023 2476 3419 525 1359 7848.92 21676.46 2845.50 9689.67 4596 4596 31-01-2023 7446 841 4825 1311 23603.82 5331.94 26151.50 9347.43 4597 4597 01-02-2023 6289 3143 3588 474 19936.13 19926.62 19446.96 3379.62 4598 4598 02-02-2023 3122 1188 5899 517 9896.74 7531.92 31972.58 3686.21 4599 4599 03-02-2023 1234 3854 2321 406 3911.78 24434.36 12579.82 2894.78 4600 rows × 10 columns In [5]: from sklearn.linear\_model import LinearRegression lin\_model=LinearRegression() from sklearn.ensemble import RandomForestRegressor model=RandomForestRegressor(n\_estimators=100, max\_features=3, random\_state=1) import numpy as np x1, x2, x3, y=df['Q-P1'], df['Q-P2'], df['Q-P3'], df['Q-P4'] x1, x2, x3, y=np.array(x1), np.array(x2), np.array(x3), np.array(y)x1, x2, x3, y=x1.reshape(-1,1), x2.reshape(-1,1), x3.reshape(-1,1), y.reshape(-1,1)final\_x=np.concatenate((x1, x2, x3), axis=1) print(final\_x) [[5422 3725 576] [7047 779 3578] [1572 2082 595] [6289 3143 3588] [3122 1188 5899] [1234 3854 2321]] In [8]: X\_train, X\_test, y\_train, y\_test=final\_x[:-30], final\_x[-30:], y[:-30], y[-30:] model.fit(X\_train,y\_train) In [9]: lin\_model.fit(X\_train,y\_train) C:\Users\Dell\anaconda3\envs\fre\lib\site-packages\sklearn\base.py:1151: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape o f y to (n\_samples,), for example using ravel(). return fit\_method(estimator, \*args, \*\*kwargs) Out[9]: LinearRegression LinearRegression() In [10]: pred=model.predict(X\_test) import matplotlib.pyplot as plt plt.rcParams["figure.figsize"] = (12,8) plt.plot(pred, label='Random\_Forest\_Predictions') plt.plot(y\_test, label='Actual Product') plt.legend(loc="upper left") plt.show() Random\_Forest\_Predictions 2000 **Actual Product** 1750 1500 1250 1000 750 500 5 10 15 20 25 30 In [11]: lin\_pred=lin\_model.predict(X\_test) import matplotlib.pyplot as plt plt.rcParams["figure.figsize"] = (11,6) plt.plot(lin\_pred, label='Linear\_Regression\_Predictions') plt.plot(y\_test, label='Actual Products') plt.legend(loc="upper left") plt.show() 2000 Linear\_Regression\_Predictions Actual Products 1750 1500 1250 1000 750 500 15 20 25 0 5 10 30 In [12]: **import** numpy **as** np x1, x2, x3, y=df['S-P1'], df['S-P2'], df['S-P3'], df['S-P4'] x1, x2, x3, y=np.array(x1), np.array(x2), np.array(x3), np.array(y)x1, x2, x3, y=x1.reshape(-1,1), x2.reshape(-1,1), x3.reshape(-1,1), y.reshape(-1,1)final\_x=np.concatenate((x1,x2,x3),axis=1) print(final\_x) [[17187.74 23616.5 3121.92] [22338.99 4938.86 19392.76] [ 4983.24 13199.88 3224.9 ] [19936.13 19926.62 19446.96] [ 9896.74 7531.92 31972.58] [ 3911.78 24434.36 12579.82]] In [13]: X\_train, X\_test, y\_train, y\_test=final\_x[:-90], final\_x[-90:], y[:-90], y[-90:] In [14]: model.fit(X\_train,y\_train) lin\_model.fit(X\_train,y\_train) C:\Users\Dell\anaconda3\envs\fre\lib\site-packages\sklearn\base.py:1151: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape o f y to (n\_samples,), for example using ravel(). return fit\_method(estimator, \*args, \*\*kwargs) Out[14]: ▼ LinearRegression LinearRegression() In [15]: pred=model.predict(X\_test) import matplotlib.pyplot as plt plt.rcParams["figure.figsize"] = (12,8) plt.plot(pred, label='Random\_Forest\_Predictions') plt.plot(y\_test, label='Actual Sales') plt.legend(loc="upper left") plt.show() Random\_Forest\_Predictions 14000 -Actual Sales 12000 10000 8000 6000 4000 2000 0 20 40 60 80 In [32]: from sklearn.metrics import mean\_squared\_error from math import sqrt rmse\_rf=sqrt(mean\_squared\_error(pred,y\_test)) rmse\_lr=sqrt(mean\_squared\_error(lin\_pred,y\_test)) In [19]: lin\_pred=lin\_model.predict(X\_test) import matplotlib.pyplot as plt plt.rcParams["figure.figsize"] = (11,6) plt.plot(lin\_pred, label='Linear\_Regression\_Predictions') plt.plot(y\_test,label='Actual Sales') plt.legend(loc="upper left") plt.show() Linear\_Regression\_Predictions 14000 Actual Sales 12000 10000 8000 6000 4000

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