In [54]: import numpy as np import pandas as pd import matplotlib.pyplot as plt import seaborn as sns #matplotlib inline from sklearn import datasets from sklearn.model selection import train test split from sklearn.linear_model import LinearRegression In [84]: Out[84]: array([[5.1, 3.5, 1.4, 0.2], [4.9, 3. , 1.4, 0.2], [4.7, 3.2, 1.3, 0.2], [4.6, 3.1, 1.5, 0.2], [5., 3.6, 1.4, 0.2], [5.4, 3.9, 1.7, 0.4], [4.6, 3.4, 1.4, 0.3], [5., 3.4, 1.5, 0.2], [4.4, 2.9, 1.4, 0.2], [4.9, 3.1, 1.5, 0.1], [5.4, 3.7, 1.5, 0.2], [4.8, 3.4, 1.6, 0.2], [4.8, 3., 1.4, 0.1], [4.3, 3., 1.1, 0.1], [5.8, 4., 1.2, 0.2], [5.7, 4.4, 1.5, 0.4], [5.4, 3.9, 1.3, 0.4], [5.1, 3.5, 1.4, 0.3], [5.7, 3.8, 1.7, 0.3], [5.1, 3.8, 1.5, 0.3], [5.4, 3.4, 1.7, 0.2], [5.1, 3.7, 1.5, 0.4], [4.6, 3.6, 1., 0.2], [5.1, 3.3, 1.7, 0.5], [4.8, 3.4, 1.9, 0.2], [5., 3., 1.6, 0.2], [5., 3.4, 1.6, 0.4], [5.2, 3.5, 1.5, 0.2], [5.2, 3.4, 1.4, 0.2], [4.7, 3.2, 1.6, 0.2], [4.8, 3.1, 1.6, 0.2], [5.4, 3.4, 1.5, 0.4], [5.2, 4.1, 1.5, 0.1], [5.5, 4.2, 1.4, 0.2], [4.9, 3.1, 1.5, 0.2], [5., 3.2, 1.2, 0.2], [5.5, 3.5, 1.3, 0.2], [4.9, 3.6, 1.4, 0.1], [4.4, 3., 1.3, 0.2], [5.1, 3.4, 1.5, 0.2], [5., 3.5, 1.3, 0.3], [4.5, 2.3, 1.3, 0.3], [4.4, 3.2, 1.3, 0.2], [5., 3.5, 1.6, 0.6], [5.1, 3.8, 1.9, 0.4], [4.8, 3., 1.4, 0.3], [5.1, 3.8, 1.6, 0.2], [4.6, 3.2, 1.4, 0.2], [5.3, 3.7, 1.5, 0.2], [5., 3.3, 1.4, 0.2], [7., 3.2, 4.7, 1.4], [6.4, 3.2, 4.5, 1.5], [6.9, 3.1, 4.9, 1.5], [5.5, 2.3, 4., 1.3], [6.5, 2.8, 4.6, 1.5], [5.7, 2.8, 4.5, 1.3], [6.3, 3.3, 4.7, 1.6], [4.9, 2.4, 3.3, 1.], [6.6, 2.9, 4.6, 1.3], [5.2, 2.7, 3.9, 1.4], [5., 2., 3.5, 1.], [5.9, 3., 4.2, 1.5], [6., 2.2, 4., 1.], [6.1, 2.9, 4.7, 1.4], [5.6, 2.9, 3.6, 1.3], [6.7, 3.1, 4.4, 1.4], [5.6, 3., 4.5, 1.5], [5.8, 2.7, 4.1, 1.], [6.2, 2.2, 4.5, 1.5], [5.6, 2.5, 3.9, 1.1], [5.9, 3.2, 4.8, 1.8], [6.1, 2.8, 4. , 1.3], [6.3, 2.5, 4.9, 1.5], [6.1, 2.8, 4.7, 1.2], [6.4, 2.9, 4.3, 1.3], [6.6, 3., 4.4, 1.4], [6.8, 2.8, 4.8, 1.4], [6.7, 3., 5., 1.7],[6., 2.9, 4.5, 1.5], [5.7, 2.6, 3.5, 1.], [5.5, 2.4, 3.8, 1.1], [5.5, 2.4, 3.7, 1.], [5.8, 2.7, 3.9, 1.2], [6., 2.7, 5.1, 1.6], [5.4, 3., 4.5, 1.5], [6., 3.4, 4.5, 1.6], [6.7, 3.1, 4.7, 1.5], [6.3, 2.3, 4.4, 1.3], [5.6, 3., 4.1, 1.3], [5.5, 2.5, 4., 1.3], [5.5, 2.6, 4.4, 1.2], [6.1, 3., 4.6, 1.4], [5.8, 2.6, 4., 1.2], [5., 2.3, 3.3, 1.], [5.6, 2.7, 4.2, 1.3], [5.7, 3., 4.2, 1.2], [5.7, 2.9, 4.2, 1.3], [6.2, 2.9, 4.3, 1.3], [5.1, 2.5, 3., 1.1], [5.7, 2.8, 4.1, 1.3], [6.3, 3.3, 6., 2.5], [5.8, 2.7, 5.1, 1.9], [7.1, 3., 5.9, 2.1], [6.3, 2.9, 5.6, 1.8], [6.5, 3., 5.8, 2.2], [7.6, 3., 6.6, 2.1], [4.9, 2.5, 4.5, 1.7], [7.3, 2.9, 6.3, 1.8], [6.7, 2.5, 5.8, 1.8], [7.2, 3.6, 6.1, 2.5], [6.5, 3.2, 5.1, 2.], [6.4, 2.7, 5.3, 1.9], [6.8, 3., 5.5, 2.1], [5.7, 2.5, 5., 2.], [5.8, 2.8, 5.1, 2.4], [6.4, 3.2, 5.3, 2.3], [6.5, 3., 5.5, 1.8], [7.7, 3.8, 6.7, 2.2], [7.7, 2.6, 6.9, 2.3], [6., 2.2, 5., 1.5], [6.9, 3.2, 5.7, 2.3], [5.6, 2.8, 4.9, 2.], [7.7, 2.8, 6.7, 2.], [6.3, 2.7, 4.9, 1.8], [6.7, 3.3, 5.7, 2.1], [7.2, 3.2, 6., 1.8], [6.2, 2.8, 4.8, 1.8], [6.1, 3., 4.9, 1.8], [6.4, 2.8, 5.6, 2.1], [7.2, 3., 5.8, 1.6], [7.4, 2.8, 6.1, 1.9], [7.9, 3.8, 6.4, 2.], [6.4, 2.8, 5.6, 2.2], [6.3, 2.8, 5.1, 1.5], [6.1, 2.6, 5.6, 1.4], [7.7, 3., 6.1, 2.3], [6.3, 3.4, 5.6, 2.4], [6.4, 3.1, 5.5, 1.8], [6., 3., 4.8, 1.8], [6.9, 3.1, 5.4, 2.1], [6.7, 3.1, 5.6, 2.4], [6.9, 3.1, 5.1, 2.3], [5.8, 2.7, 5.1, 1.9], [6.8, 3.2, 5.9, 2.3], [6.7, 3.3, 5.7, 2.5], [6.7, 3., 5.2, 2.3], [6.3, 2.5, 5., 1.9], [6.5, 3., 5.2, 2.], [6.2, 3.4, 5.4, 2.3], [5.9, 3., 5.1, 1.8]]) In [107... #convert to dataFrame df = pd.DataFrame(data=iris_dataset.data) df Out[107... 0 1 2 3 **0** 5.1 3.5 1.4 0.2 **1** 4.9 3.0 1.4 0.2 **2** 4.7 3.2 1.3 0.2 **3** 4.6 3.1 1.5 0.2 **4** 5.0 3.6 1.4 0.2 **145** 6.7 3.0 5.2 2.3 **146** 6.3 2.5 5.0 1.9 **147** 6.5 3.0 5.2 2.0 **148** 6.2 3.4 5.4 2.3 **149** 5.9 3.0 5.1 1.8 150 rows × 4 columns In [108... df.head() Out[108... 0 1 2 3 **0** 5.1 3.5 1.4 0.2 **1** 4.9 3.0 1.4 0.2 **2** 4.7 3.2 1.3 0.2 **3** 4.6 3.1 1.5 0.2 **4** 5.0 3.6 1.4 0.2 In [39]: df.tail() Out[39]: 0 1 2 3 **145** 6.7 3.0 5.2 2.3 **146** 6.3 2.5 5.0 1.9 **147** 6.5 3.0 5.2 2.0 **148** 6.2 3.4 5.4 2.3 **149** 5.9 3.0 5.1 1.8 df.columns = ['A', 'B', 'C', 'D'] Out[109... A B C D **0** 5.1 3.5 1.4 0.2 **1** 4.9 3.0 1.4 0.2 **2** 4.7 3.2 1.3 0.2 **3** 4.6 3.1 1.5 0.2 **4** 5.0 3.6 1.4 0.2 **145** 6.7 3.0 5.2 2.3 **146** 6.3 2.5 5.0 1.9 **147** 6.5 3.0 5.2 2.0 **148** 6.2 3.4 5.4 2.3 **149** 5.9 3.0 5.1 1.8 150 rows × 4 columns In [52]: df.index RangeIndex(start=0, stop=150, step=1) In [55]: sns.histplot(df) Out[55]: <AxesSubplot:ylabel='Count'> 80 ■ A **—** В 70 ___ c ___ D 60 50 40 Oonut 30 20 -10 -0 1 2 3 4 5 6 7 In [57]: sns.histplot(df.B) Out[57]: <AxesSubplot:xlabel='B', ylabel='Count'> 35 30 -25 y 20 · 15 -In [59]: plt.hist(df['C']) Out[59]: (array([37., 13., 0., 3., 8., 26., 29., 18., 11., 5.]), array([1. , 1.59, 2.18, 2.77, 3.36, 3.95, 4.54, 5.13, 5.72, 6.31, 6.9]), <BarContainer object of 10 artists>) 35 -30 25 20 15 10 In [60]: plt.boxplot(df.C) Out[60]: {'whiskers': [<matplotlib.lines.Line2D at 0x1d83ae88e20>, <matplotlib.lines.Line2D at 0x1d83969e850>], 'caps': [<matplotlib.lines.Line2D at 0x1d83969e340>, <matplotlib.lines.Line2D at 0x1d83a582f40>], 'boxes': [<matplotlib.lines.Line2D at 0x1d83ae88bb0>], 'medians': [<matplotlib.lines.Line2D at 0x1d83a582370>], 'fliers': [<matplotlib.lines.Line2D at 0x1d83a582c70>], 'means': []} In [66]: plt.figure(figsize=(10, 10)) sns.heatmap(df, annot=False) <AxesSubplot:> Out[66]: In [104... sns.scatterplot(x=df.A, y=df.B)<AxesSubplot:xlabel='A', ylabel='B'> 4.0 3.5 В 3.0 2.5 2.0 -5.5 6.0 6.5 7.0 5.0 In [116... X =df.drop(columns=['D']) y = df['D'] from sklearn.model_selection import train_test_split X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3) In [158... y_test 146 1.9 1.8 1.9 0.2 104 2.2 144 2.5 0.2 36 0.2 14 0.2 112 2.1 117 2.2 0.3 47 0.2 141 2.3 63 1.4 11 0.2 31 0.4 1.2 1.2 0.2 121 2.0 78 1.5 147 2.0 133 1.5 62 1.0 145 2.3 38 0.2 82 1.2 1.5 55 1.3 22 0.2 130 1.9 1.3 1.8 126 1.8 140 2.4 115 2.3 39 0.2 76 1.4 41 0.3 73 1.2 143 2.3 87 1.3 Name: D, dtype: float64 KNN Regressor $\textbf{from} \ \text{sklearn.neighbors} \ \textbf{import} \ \text{KNeighborsRegressor}$ knn = KNeighborsRegressor() knn.fit(X_train, y_train) KNeighborsRegressor() y_pred = knn.predict(X_test) y_pred.shape Out[170... In [171... y_train.shape y_test.shape Out[171... (45,) from sklearn.metrics import mean_absolute_error acrr = mean_absolute_error(y_test, y_pred) acrr 0.15511111111111111 from sklearn import datasets df = datasets.load_iris() df = pd.DataFrame(data=df.data) df.columns = ['a', 'b', 'c', 'd'] df.shape df.index df.columns Index(['a', 'b', 'c', 'd'], dtype='object') In [198... df.loc[: ,'a'] df Out[198... a b c d **0** 5.1 3.5 1.4 0.2 **1** 4.9 3.0 1.4 0.2 **2** 4.7 3.2 1.3 0.2 **3** 4.6 3.1 1.5 0.2 **4** 5.0 3.6 1.4 0.2 ••• **145** 6.7 3.0 5.2 2.3 **146** 6.3 2.5 5.0 1.9 **147** 6.5 3.0 5.2 2.0 **148** 6.2 3.4 5.4 2.3 **149** 5.9 3.0 5.1 1.8 150 rows × 4 columns In [207... x = df.iloc[:,:-1]y = df.iloc[:, -1]from sklearn.model_selection import train_test_split x_tr, x_te, y_tr, y_te = train_test_split(x, y) In [208... x_tr Out[208... a b c **90** 5.5 2.6 4.4 **34** 4.9 3.1 1.5 **106** 4.9 2.5 4.5 **55** 5.7 2.8 4.5 **96** 5.7 2.9 4.2 **32** 5.2 4.1 1.5 **75** 6.6 3.0 4.4 **35** 5.0 3.2 1.2 **9** 4.9 3.1 1.5 **125** 7.2 3.2 6.0 112 rows × 3 columns **Linear Regression** In [213... from sklearn.linear_model import LinearRegression lr = LinearRegression() lr.fit(x tr, y tr) y_pr=lr.predict(x_te) y_pr Out[213... array([1.75817343, 0.27559847, 1.02304746, 0.14855876, 1.64270853, 1.8088459 , 0.30094101, 2.04665424, 1.51746774, 1.84794356, 0.24157648, 0.31469613, 2.01226743, 1.35495737, 1.80378283, 1.82260102, 1.73645789, 1.11135631, 2.0647534 , 0.14819195, 1.37232688, 0.28030932, 0.30492221, 1.44979057, 0.21696557, 0.04358483, 0.29406444, 0.15398665, 1.49720087, 0.23398287, 1.32816615, 1.1298203 , 0.1662931 , 0.21841425, 1.2836532 , 0.18547611, 1.64705258, 0.29224895]) In [216... from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score mse = mean_absolute_error(y_te, y_pr) mse*****100 11.295066010419927 In [219... mae = mean_squared_error(y_te, y_pr) mae*****100 Out[219... 2.072481352442615 **SVR** from sklearn.svm import SVR svr = SVR()svr.fit(x_tr, y_tr) svr1= svr.predict(x_te) Out[222... array([1.81260934, 0.32631443, 1.0876026 , 0.24130479, 1.65439047, 1.89879133, 0.22706761, 2.11215326, 1.58169124, 1.8528243, 0.28376158, 0.24067514, 2.11452663, 1.31537502, 1.82745983, 1.86937525, 1.78745846, 1.13537161, 2.1731173 , 0.23622959, 1.3757722 , 0.22514937, 0.27920421, 1.40346811, 0.2358552 , 0.24633838, 0.24549112, 0.19210746, 1.53383503, 0.21549048, 1.3423918 , 1.05725079, 0.21676422, 0.23260624, 1.25357179, 0.21236111, 1.70371203, 0.34830935]) In [224... svr_mse = mean_absolute_error(y_te, svr1) svr_mse Out[224... 0.1208165623964546