```
In [1]: ▶ import pandas as pd
          import numpy as np
df.head()
   Out[2]:
               ΑT
                   V
                         AP
                              RH
                                   PΕ
           0 14.96 41.76 1024.07 73.17 463.26
           1 25.18 62.96 1020.04 59.08 444.37
           2 5.11 39.40 1012.16 92.14 488.56
           3 20.86 57.32 1010.24 76.64 446.48
           4 10.82 37.50 1009.23 96.62 473.90
Out[3]: AT
               0
          AΡ
               0
          RH
               0
          PΕ
               0
          dtype: int64
       Multiple Linear Regression
y = df.iloc[:, -1]
In [5]: ▶ from sklearn.model selection import train test split
          X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0)
In [6]: ▶ from sklearn.linear_model import LinearRegression
          lm = LinearRegression()
          lm.fit(X\_train, y\_train)
          lm_predictions = lm.predict(X_test)
In [7]: 🔰 from sklearn.metrics import r2_score, mean_absolute_error, mean_squared_error
           print(f"r2_score: {r2_score(y_test, lm_predictions)}")
          print(f"mean_absolute_error: {mean_absolute_error(y_test, lm_predictions)}")
          print(f"mean_squared_error: {mean_squared_error(y_test, lm_predictions)}")
          print(f"Root Mean Square Error: {np.sqrt(mean_squared_error(y_test, lm_predictions))}")
          r2_score: 0.9325315554761302
          mean_squared_error: 19.73369930349765
          Root Mean Square Error: 4.442262858442491
       Polynomial Regression
y = df.iloc[:, -1]
from sklearn.model_selection import train_test_split
          poly_feature = PolynomialFeatures(degree=4)
          X = poly_feature.fit_transform(X)
          X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0)
poly_lm = LinearRegression()
          poly_lm.fit(X_train, y_train)
          poly_predictions = poly_lm.predict(X_test)
In [11]: M from sklearn.metrics import r2_score, mean_absolute_error, mean_squared_error
```

```
print(f"r2_score: {r2_score(y_test, poly_predictions)}")
                     print(f"mean\_absolute\_error: \{mean\_absolute\_error(y\_test, poly\_predictions)\}")
                     print(f"mean_squared_error: {mean_squared_error(y_test, poly_predictions)}")
                     print(f"Root Mean Square Error: {np.sqrt(mean_squared_error(y_test, poly_predictions))}")
                     r2_score: 0.9458193390165572
                     mean_absolute_error: 3.1360275695485944
                     mean_squared_error: 15.847184257134275
                     Root Mean Square Error: 3.9808522023725366
               Support Vector Regression
y = df.iloc[:, -1].to_frame()
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0)
In [14]: ▶ from sklearn.preprocessing import StandardScaler
                     sc X = StandardScaler()
                     sc_y = StandardScaler()
                                                                                       Scale for X and
                     X_train = sc_X.fit_transform(X_train)
                     y_train = sc_y.fit_transform(y_train)
                     X_test = sc_X.transform(X_test)
                    y_test = sc_y.transform(y_test)
In [15]: ► from sklearn.svm import SVR
                     svr = SVR() # kernel = 'rbf'
                     svr.fit(X_train, y_train)
                     svr_predictions = svr.predict(X_test)
                     C:\Users\Perry\anaconda3\lib\site-packages\sklearn\utils\validation.py:63: DataConversionWarning:
                     A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samp
                     les, ), for example using ravel().
                     return f(*args, **kwargs)
In [16]: M from sklearn.metrics import r2_score, mean_absolute_error, mean_squared_error
                     print(f"r2_score: {r2_score(y_test, svr_predictions)}")
                     print(f"mean_absolute_error: {mean_absolute_error(y_test, svr_predictions)}")
                     print(f"mean_squared_error: {mean_squared_error(y_test, svr_predictions)}")
                     print(f"Root Mean Square Error: {np.sqrt(mean_squared_error(y_test, svr_predictions))}")
                     r2_score: 0.9480784049986266
                     mean_absolute_error: 0.1756034994155729
                     mean_squared_error: 0.05220075426721192
                     Root Mean Square Error: 0.22847484383890476
               2
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0)
In [19]: ▶ from sklearn.preprocessing import StandardScaler
                                                                                    Scale just only for X (Right way, no need
                     sc_X = StandardScaler()
                     sc_y = StandardScaler()
                     X train = sc X.fit transform(X train)
                                                                                    scale for label)
                    y_train = sc_y.fit_transform(y_train)
In [20]: ► from sklearn.svm import SVR
                     svr = SVR() # kernel = 'rbf'
                     svr.fit(X_train, y_train)
                     svr_predictions = svr.predict(sc_X.transform(X_test))
                     \verb|C:\Users\le \Perry\anaconda3\lib\site-packages\sklearn\utils\validation.py:63: DataConversionWarning: | Perry\anaconda3\lib\site-packages\sklearn\utils\validation.py:63: DataConversionWarning: | Perry\anaconda3\lib\site-packages\sklearn\utils\svalidation.py:63: DataConversionWarning: | Perry\anaconda3\lib\site-packages\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\sklearn\utils\
                     A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samp
                     les, ), for example using ravel().
                       return f(*args, **kwargs)
In [21]: M from sklearn.metrics import r2_score, mean_absolute_error, mean_squared_error
```

```
print(f"r2\_score: \{r2\_score(y\_test, sc\_y.inverse\_transform(svr\_predictions))\}")
            print( \textbf{f"mean\_squared\_error}: \\ \{ \textbf{mean\_squared\_error}(\textbf{y\_test}, \ \textbf{sc\_y\_inverse\_transform}(\textbf{svr\_predictions})) \}") \} \\ ") \\ "
            print(f"Root Mean Square Error: {np.sqrt(mean_squared_error(y_test, sc_y.inverse_transform(svr_pred
            r2_score: 0.9480784049986266
            mean_absolute_error: 2.9951783924512863
            mean_squared_error: 15.186434937781799
            Root Mean Square Error: 3.896977667087893
        Decision Tree Regression
y = df.iloc[:, -1]
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0)
In [24]: ► from sklearn.tree import DecisionTreeRegressor
            dtr = DecisionTreeRegressor(random_state=0)
            {\tt dtr.fit}({\tt X\_train},\ {\tt y\_train})
            dtr_predictions = dtr.predict(X_test)
In [25]: ┡ from sklearn.metrics import r2_score, mean_absolute_error, mean_squared_error
            print(f"r2_score: {r2_score(y_test, dtr_predictions)}")
            print(f"mean_absolute_error: {mean_absolute_error(y_test, dtr_predictions)}")
            print(f"mean_squared_error: {mean_squared_error(y_test, dtr_predictions)}")
            print(f"Root Mean Square Error: {np.sqrt(mean_squared_error(y_test, dtr_predictions))}")
            r2_score: 0.922905874177941
            mean_absolute_error: 3.103923719958203
            mean_squared_error: 22.549093991640547
            Root Mean Square Error: 4.748588631545224
        Random Forest Regression
In [26]: M = df.iloc[:, :-1]
           y = df.iloc[:, -1]
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0)
In [28]: ▶ from sklearn.ensemble import RandomForestRegressor
            {\tt rfr = RandomForestRegressor(n\_estimators=10, random\_state=0)}
            {\tt rfr.fit}({\tt X\_train,\ y\_train})
            rfr_predictions = rfr.predict(X_test)
```

In [29]: N from sklearn.metrics import r2_score, mean_absolute_error, mean_squared_error

print(f"mean_absolute_error: {mean_absolute_error(y_test, rfr_predictions)}")
print(f"mean_squared_error: {mean_squared_error(y_test, rfr_predictions)}")

print(f"Root Mean Square Error: {np.sqrt(mean_squared_error(y_test, rfr_predictions))}")

print(f"r2_score: {r2_score(y_test, rfr_predictions)}")

r2_score: 0.9615908334363876

mean_absolute_error: 2.452366771159876 mean_squared_error: 11.234213991640557 Root Mean Square Error: 3.3517479009675766