Train and Test Data

from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X,y,random_state=0)

Data Preprocessing

Standardization

Normalization

Imputing Missing Values

Linear Regression

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 + \dots + \beta_n x_n + \varepsilon$$

Linear Regression Using Sklearn

from sklearn.linear_model import LinearRegression
 model = LinearRegression()

Linear Regression Using statmodels

import statsmodels.api as sm
lmodel = sm.OLS(y_train, X_train)

Model Fitting

model.fit(X_train, y_train)

Prediction

y_pred = model.predict(X_test)

Feature Selection

Forward Selection

from mlxtend.feature_selection import SequentialFeatureSelector as sfs linreq forward = sfs(estimator = model, k features, forward = True)

Backward Elimination

from mlxtend.feature_selection import SequentialFeatureSelector as sfs linreg_backward = sfs(estimator = model, k_features, forward = False)

Recursive Forward Elimination

from sklearn.feature_selection import RFE
rfe model = RFE(estimator = model, n features to select)

Cross Validation

Cross Val Score

from sklearn.model_selection import cross_val_score

K-Fold CV

from sklearn.model_selection import KFold
 kf = KFold(n_splits)

Leave One Out CV

from sklearn.model_selection import LeaveOneOut
 loocv = LeaveOneOut()

Hyperparameter Tuning

GridSearchCV

from sklearn.model_selection import GridSearchCV
qrid model = GridSearchCV(estimator, param qrid , cv)

Gradient Descent

Stochastic Gradient Descent

from sklearn.linear_model import SGDRegressor
sqd = SGDRegressor()

Regularization

Ridge Regularization

from sklearn.linear_model import Ridge
 ridge = Ridge(alpha)

Lasso Regularization

Elastic-Net Regularization

from sklearn.linear_model import ElasticNet
 enet = ElasticNet(alpha, 11 ratio)

Model Evaluation

R-Squared

from sklearn.metrics import r2_score
 r_sq = r2_score(y_true, y_pred)

Adjusted R-Squared

 $\label{eq:norm} \begin{array}{ll} n \, = \, number \, \, of \, \, observations \\ k \, = \, number \, \, of \, \, columns \, \, \, (including \, \, intercept) \end{array}$

 $adj_r_squared = 1 - (((1 - r_sq) * (n - 1)) / (n - k - 1))$

Evaluating Model Performance

Mean Squared Error

from sklearn.metrics import mean_squared_error
MSE = mean_squared_error(y_test, y_pred)

Root Mean Squared Error

RMSE = np.sqrt(MSE)

Mean Absolute Error

from sklearn.metrics import mean_absolute_error
 MAE = mean_absolute_error(y_true, y_pred)

Mean Absolute Percentage Error

MAPE = np.mean(np.abs((y_test - y_pred) / y_test)) * 100