PYTHON FOR DATA SCIENCE CHEAT SHEET

Python Scikit-Learn

Introduction

Scikit-learn: 'sklearn' is a machine learning library for the Python programming language. Simple and efficient tool for data mining, Data analysis and Machine Learning.

Importing Convention - import sklearn

Preprocessing

Data Loading

- Using NumPy:
- >>>import numpy as np
- >>>a=np.array([(1,2,3,4),(7,8,9,10)],dtype=int)
- >>>data = np.loadtxt('file_name.csv', delimiter=',')
- Using Pandas:
- >>>import pandas as pd
- >>>df=pd.read_csv('file_name.csv',header=o)

Train-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 Preparation

Standardization

- >>>from sklearn.preprocessing import
 StandardScaler
- >>>get names = df.columns
- >>>scaler =
- preprocessing.StandardScaler()
- >>>scaled df = scaler.fit transform(df)
- >>>scaled df =
- pd.DataFrame(scaled_df, columns=get names)m

Normalization

- >>>from sklearn.preprocessing import
 Normalizer
- >>>pd.read csv("File name.csv")
- >>>x array = np.array(df['Column1'])
- #Normalize Column1
- >>>normalized_X = preprocessing.normalize([x array])

Working On Model

Model Choosing

Supervised Learning Estimator:

- · Linear Regression:
- >>> from sklearn.linear_model import LinearRegression
- >>> new lr =
- LinearRegression(normalize=True)
- · Support Vector Machine:
- >>> from sklearn.svm import SVC
- >>> new_svc = SVC(kernel='linear')

Naive Baves:

- >>> from sklearn.naive_bayes import GaussianNB
- >>> new gnb = GaussianNB()
- KNN:
- >>> from sklearn import neighbors
- >>>
- knn=neighbors.KNeighborsClassifier(n_neighbors=1)

Unsupervised Learning Estimator:

- Principal Component Analysis (PCA):
- >>> from sklearn.decomposition import PCA
- >>> new_pca= PCA(n_components=0.95)
- K Means:
- >>> from sklearn.cluster import KMeans
- >>> k_means = KMeans(n_clusters=5, random state=0)

Train-Test Data

Supervised:

- >>>new_lr.fit(X, y)
- >>> knn.fit(X train, y train)
- >>>new svc.fit(X train, y train)

Unsupervised:

- >>> k means.fit(X train)
- >>> pca model fit =
- new_pca.fit_transform(X_train)

Post-Processing

Prediction

Supervised:

- >>> y predict =
- new svc.predict(np.random.random((3,5)))
- >>> y predict = new lr.predict(X test)
- >>> y_predict = knn.predict_proba(X_test)

Unsupervised:

>>> y_pred = k_means.predict(X_test)

Model Tuning

Grid Search:

- >>> from sklearn.grid search import GridSearchCV
- >>> grid = GridSearchCV(estimator=knn,
- param_grid=params)
- >>> grid.fit(X_train, y_train)
- >>> print(grid.best_score_)
- >>> print(grid.best_estimator_.n_neighbors)

Randomized Parameter Optimization:

- >>> from sklearn.grid search import RandomizedSearchCV
- >>> params = {"n neighbors": range(1,5), "weights":
- ["uniform", "distance"]}
- >>> rsearch = RandomizedSearchCV(estimator=knn,
- param_distributions=params, cv=4, n_iter=8, random_state=5)
- >>> rsearch.fit(X_train, y_train)
 >>> print(rsearch.best_score)
- Evaluate Performance

Classification:

1. Confusion Matrix:

- >>> from sklearn.metrics import confusion_matrix
- >>> print(confusion_matrix(y_test, y_pred))
- 2. Accuracy Score:
- >>> knn.score(X_test, y_test)
- >>> from sklearn.metrics import
 accuracy_score
 >>> accuracy score(y test, y pred)

Regression:

1. Mean Absolute Error:

- >>> from sklearn.metrics import mean_absolute_error
- >>> y_true = [3, -0.5, 2]
- >>> mean_absolute_error(y_true, y_predict)
- 2. Mean Squared Error:
- >>> from sklearn.metrics import mean_squared_error
 >>> mean_squared_error(y test, y predict)
- 3. R² Score:
- >>> from sklearn.metrics import r2_score
- >>> r2_score(y_true, y_predict)

Clustering:

1. Homogeneity:

- >>> from sklearn.metrics import homogeneity score
- >>> homogeneity_score(y_true,
- y_predict)
 2. V-measure:
- >>> from sklearn.metrics import
- v_measure_score
- >>> metrics.v_measure_score(y_true, y predict)

Cross-validation:

- >>> from sklearn.cross validation
- import cross_val_score
 >>>
- print(cross_val_score(knn, X train, y train, cv=4))
- x_train, y_ti
- print(cross_val_score(new_ lr, X, y, cv=2))

FURTHERMORE:

