

# 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',
delimite=',')
```

##### • Using Pandas:

```
>>> import pandas as pd
>>> df = pd.read_csv('file_name.csv', header=0)
```

### 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)
```

##### • Normalization

```
>>> from sklearn.preprocessing import
Normalizer
>>> pd.read_csv("File_name.csv")
>>> x_array = np.array(df[0:Column10])
# 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 Bayes:

```
>>> from sklearn.naive_bayes import
GaussianNB
>>> new_gnb = GaussianNB()
```

##### • KNN:

```
>>> from sklearn import neighbors
>>>
knn = neighbors.KNeighborsClassifier(n_neighbors=1)
```

### 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
>>> params = {"n_neighbors": np.arange(1,3), "metric":
["euclidean", "cityblock"]}
>>> 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. R<sup>2</sup> 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))
>>>
print(cross_val_score(new_
lr, X, y, cv=2))
```



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