

Scikit-Learn Python Cheat Sheet

by Manasa via cheatography.com/121399/cs/22207/

Machine Learning Supervised Unsupervised learning Learning

The model maps No training is given to input to an the model and it has to output based on discover the features of input by self training input-output mechanism.

Scikit learn can be used in Classi¬fic¬ation, Regres¬sion, Cluste¬ring, Dimens¬ion¬ality reduct¬ion¬,Model Selection and prepro¬cessing by supervised and unsupe¬rvised training models.

Basic Commands

- >>> from sklearn import neighbors, datasets, preprocessing
- >>> from sklearn.model_selection import train_test_split
- >>> from sklearn.metrics import accuracy_-score
- >>> iris = datasets.load_iris()
- >>> X, y = iris.data[:, :2], iris.target
- >>> X_train, X_test, y_train, y_test = train_-test_split(X, y, random_state=33)
- >>> scaler = preprocessing.StandardScale-r().fit(X_train)
- >>> X_train = scaler.transform(X_train)
- >>> X_test = scaler.transform(X_test)
- >>> knn = neighbors.KNeighborsClassifier-(n_neighbors=5)
- >>> knn.fit(X_train, y_train)
- >>> y_pred = knn.predict(X_test)
- >>> accuracy_score(y_test, y_pred)

Loading Data example

- >>> import numpy as np
- >>> X = np.random.random((20,2))
- >>> y = np.array(['A','B','C','D','E','F','G','-A','C','A','B'])
- >>> X[X < 0.7] = 0

The data being loaded should be numeric and has to be stored as NumPy arrays or SciPy sparse matrices.

Processing Loaded Data

Standardi- zation	Normal- ization	Binarization
>>> from sklearn.prep- rocessing import StandardS- caler	>>> from sklearn.p- reproc- essing import Normalizer	>>> from sklearn.p- reproc- essing import Binarizer
>>> scaler = StandardS- caler().fit(X_t- rain)	>>> scaler = Normalize- r().fit(Xtrain)	<pre>>>> binarizer = Binarizer- (threshol- d=0.0).fit(X)</pre>
>>> standa- rdized_X = scaler.trans- form(X_train)	>>> normal- ized_X = scaler.tr- ansform(X- _train)	<pre>>>> binary_X = binarizer.tr- ansform(X)</pre>
<pre>>>> standa- rdized_X_test = scaler.tr- ansform(X-</pre>	>>> normalized_X_test = scaler.transform(X_test)	

Training And Test Data

test)

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

Creating Model		
Supervised Learn	ing Estimato	rs
Linear Regression	Support Vector Machines (SVM)	Naive Bayes
>>> from sklearn.line- ar_model import Linear- Regression	>>> from sklear- n.svm import SVC	>>> from sklearn.naiv- e_bayes import GaussianNB
>>> Ir = Linear- Regression(n- ormalize=True)	>>> svc = SVC(ke- rnel='lin- ear')	>>> gnb = Gaussi- anNB()

Creating Model

Unsupervised Learning Estimators		
Principal Component Analysis (PCA)	K Means	
>>> from sklear- n.decomposition import PCA	>>> from sklearn.c- luster import KMeans	
>>> pca = PCA(n_components=0.95)	>>> k_means = KMeans(n_clusters=3, random_state=0)	

Model Fitting	
Supervised	Unsupervised learning
Learning	

>>> knn.fit(X- >>> pca_model = pca.fi-_train, y_train) t_transform(X_train)

>>> k_means.fit(X_train)

>>> svc.fit(X_train, y_train)

>>> Ir.fit(X, y)

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Predicting output		
Supervised Estimators	Unsupervised Estimators	
>>> y_pred = svc.predict(np.ran- dom.random((2,5)))	>>> y_pred = k_means.p- redict(X_test)	
>>> y_pred = Ir.predict(X_test)		
>>> y_pred = knn.predict_proba(X_test))		

Classification Metrics Model Performance			
Accuracy Score	Classification Report	Confusion Matrix	
>>> knn.score- (X_test, y_test)	>>> from sklearn.m- etrics import classi- fication_report	>>> from sklear- n.metrics import confusion_matrix	
>>> from sklear- n.metrics import accuracy_score	<pre>>>> print(classific- ation_report(y_test, y_pred)))</pre>	<pre>>>> print(confus- ion_matrix(y_test, y_pred)))</pre>	
>>> accuracy_score(y_test, y_pred)			

>>>	accuracy_	_score(y_	_test, y_	prea)

Clustering Metrics Model Performance			
Adjusted Rand Index	Homogeneity	Cross-Validation	
<pre>>>> from sklear- n.metrics import adjusted_ran- d_score</pre>	<pre>>>> from sklear- n.metrics import homogeneity score</pre>	<pre>>>> print(cross val_score(knn, X_train, y_train, cv=4))</pre>	
<pre>>>> adjusted_ran- d_score(y_true, y_pred))</pre>	<pre>>>> homogenei- ty_score(y_true, y_pred))</pre>	<pre>>>> print(cross val_score(Ir, X, y, cv=2))</pre>	



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