

sklearn.model_selection.train_test_split

`sklearn.model_selection.train_test_split(*arrays, **options)`

[\[source\]](#)

Split arrays or matrices into random train and test subsets

Quick utility that wraps input validation and `next(ShuffleSplit().split(X, y))` and application to input data into a single call for splitting (and optionally subsampling) data in a oneliner.

Read more in the [User Guide](#).

Parameters

****arrayssequence of indexables with same length / shape[0]***

Allowed inputs are lists, numpy arrays, scipy-sparse matrices or pandas dataframes.

test_sizefloat or int, default=None

If float, should be between 0.0 and 1.0 and represent the proportion of the dataset to include in the test split. If int, represents the absolute number of test samples. If None, the value is set to the complement of the train size.

If `train_size` is also None, it will be set to 0.25.

train_sizefloat or int, default=None

If float, should be between 0.0 and 1.0 and represent the proportion of the dataset to include in the train split. If int, represents the absolute number of train samples. If None, the value is automatically set to the complement of the test size.

random_stateint or RandomState instance, default=None

Controls the shuffling applied to the data before applying the split. Pass an int for reproducible output across multiple function calls. See [Glossary](#).

shufflebool, default=True

Whether or not to shuffle the data before splitting. If `shuffle=False` then `stratify` must be None.

stratifyarray-like, default=None

If not None, data is split in a stratified fashion, using this as the class labels.

Returns

splittinglist, length=2 * len(arrays)

List containing train-test split of inputs.

New in version 0.16: If the input is sparse, the output will be a `scipy.sparse.csr_matrix`. Else, output type is the same as the input type.

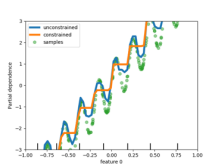
Examples

```
>>> import numpy as np
>>> from sklearn.model_selection import train_test_split
>>> X, y = np.arange(10).reshape((5, 2)), range(5)
>>> X
array([[0, 1],
       [2, 3],
       [4, 5],
       [6, 7],
       [8, 9]])
>>> list(y)
[0, 1, 2, 3, 4]
```

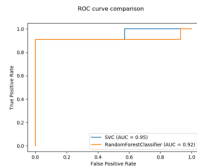
```
>>> X_train, X_test, y_train, y_test = train_test_split(
...     X, y, test_size=0.33, random_state=42)
...
>>> X_train
array([[4, 5],
       [0, 1],
       [6, 7]])
>>> y_train
[2, 0, 3]
>>> X_test
array([[2, 3],
       [8, 9]])
>>> y_test
[1, 4]
```

```
>>> train_test_split(y, shuffle=False)
[[0, 1, 2], [3, 4]]
```

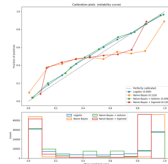
Examples using `sklearn.model_selection.train_test_split`



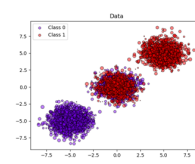
[Release Highlights for scikit-learn 0.23](#)



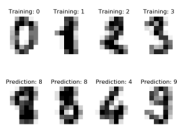
[Release Highlights for scikit-learn 0.22](#)



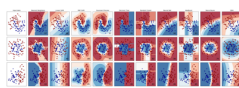
[Probability Calibration curves](#)



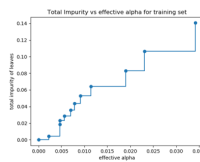
[Probability calibration of classifiers](#)



[Recognizing hand-written digits](#)



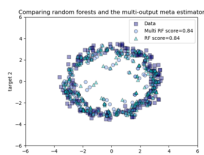
[Classifier comparison](#)



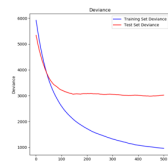
[Post pruning decision trees with cost complexity pruning](#)



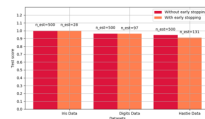
[Understanding the decision tree structure](#)



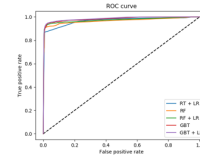
[Comparing random forests and the multi-output meta estimator](#)



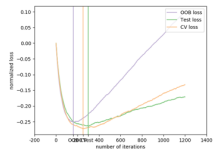
[Gradient Boosting regression](#)



[Early stopping of Gradient Boosting](#)



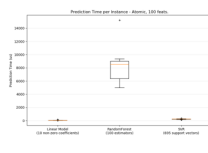
[Feature transformations with ensembles of trees](#)



[Gradient Boosting Out-of-Bag estimates](#)



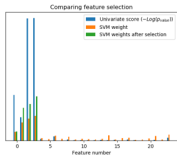
[Faces recognition example using eigenfaces and SVMs](#)



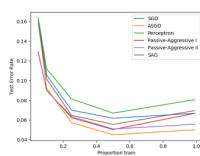
[Prediction Latency](#)



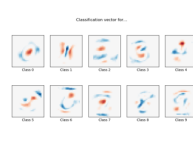
[Pipeline Anova SVM](#)



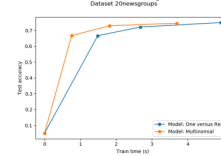
[Univariate Feature Selection](#)



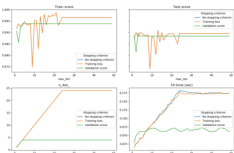
[Comparing various on-line solvers](#)



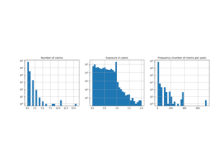
[MNIST classification using multinomial logistic + L1](#)



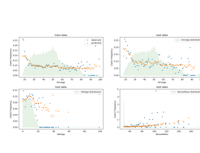
[Multiclass sparse logistic regression on 20newsgroups](#)



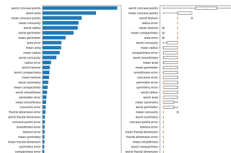
[Early stopping of Stochastic Gradient Descent](#)



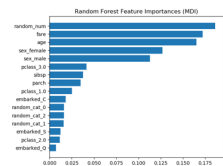
[Poisson regression and non-normal loss](#)



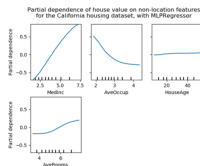
[Tweedie regression on insurance claims](#)



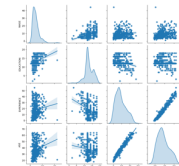
[Permutation Importance with Multicollinear or Correlated Features](#)



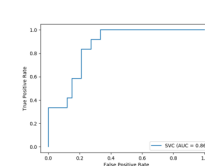
[Permutation Importance vs Random Forest Feature Importance \(MDI\)](#)



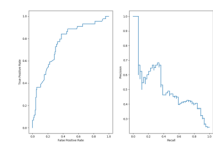
[Partial Dependence Plots](#)



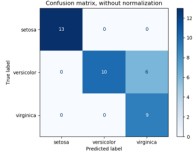
[Common pitfalls in interpretation of coefficients of linear models](#)



[ROC Curve with Visualization API](#)



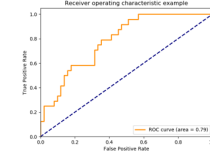
[Visualizations with Display Objects](#)



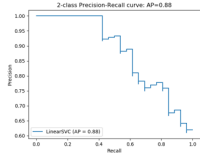
[Confusion matrix](#)



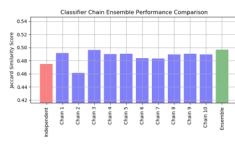
[Parameter estimation using grid search with cross-validation](#)



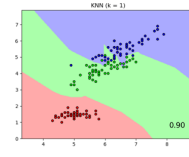
[Receiver Operating Characteristic \(ROC\)](#)



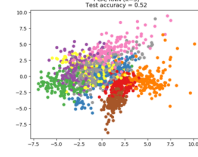
[Precision-Recall](#)



[Classifier Chain](#)



[Comparing Nearest Neighbors with and without Neighborhood Components Analysis](#)

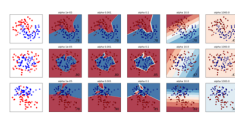


[Dimensionality Reduction with Neighborhood Components Analysis](#)

100 components extracted by RBM



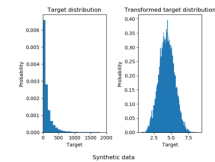
[Restricted Boltzmann Machine features for digit classification](#)



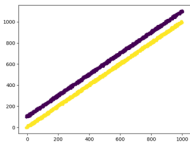
[Varying regularization in Multi-layer Perceptron](#)



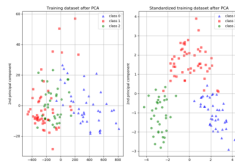
[Column Transformer with Mixed Types](#)



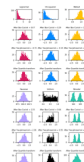
[Effect of transforming the targets in regression model](#)



[Using FunctionTransformer to select columns](#)



[Importance of Feature Scaling](#)



[Map data to a normal distribution](#)



[Feature discretization](#)