

[sklearn.feature_selection](#).SelectKBest

```
class sklearn.feature_selection.SelectKBest(score_func=<function f_classif>, k=10)
```

[\[source\]](#)

Select features according to the k highest scores.

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

Parameters:

score_func : callable

Function taking two arrays X and y, and returning a pair of arrays (scores, pvalues) or a single array with scores. Default is f_classif (see below "See also"). The default function only works with classification tasks.

k : int or "all", optional, default=10

Number of top features to select. The "all" option bypasses selection, for use in a parameter search.

Attributes:

scores_ : array-like of shape (n_features,)

Scores of features.

pvalues_ : array-like of shape (n_features,)

p-values of feature scores, None if score_func returned only scores.

See also:

[f_classif](#)

ANOVA F-value between label/feature for classification tasks.

[mutual_info_classif](#)

Mutual information for a discrete target.

[chi2](#)

Chi-squared stats of non-negative features for classification tasks.

[f_regression](#)

F-value between label/feature for regression tasks.

[mutual_info_regression](#)

Mutual information for a continuous target.

[SelectPercentile](#)

Select features based on percentile of the highest scores.

[SelectFpr](#)

Select features based on a false positive rate test.

[SelectFdr](#)

Select features based on an estimated false discovery rate.

[SelectFwe](#)

Select features based on family-wise error rate.

[GenericUnivariateSelect](#)

Univariate feature selector with configurable mode.

Notes

Ties between features with equal scores will be broken in an unspecified way.

Examples

```
>>> from sklearn.datasets import load_digits
>>> from sklearn.feature_selection import SelectKBest, chi2
>>> X, y = load_digits(return_X_y=True)
>>> X.shape
(1797, 64)
>>> X_new = SelectKBest(chi2, k=20).fit_transform(X, y)
>>> X_new.shape
(1797, 20)
```

Methods

[fit\(self, X, y\)](#)

Run score function on (X, y) and get the appropriate features.

Toggle Menu

[transform\(self, X\[, y\]\)](#)

Fit to data, then transform it.

<code>get_params(self, deep)</code>	Get parameters for this estimator.
<code>get_support(self, indices)</code>	Get a mask, or integer index, of the features selected
<code>inverse_transform(self, X)</code>	Reverse the transformation operation
<code>set_params(self, **params)</code>	Set the parameters of this estimator.
<code>transform(self, X)</code>	Reduce X to the selected features.

```
__init__(self, score_func=<function f_classif at 0x7f4d964b6320>, k=10)
```

[\[source\]](#)

Initialize self. See help(type(self)) for accurate signature.

```
fit(self, X, y)
```

[\[source\]](#)

Run score function on (X, y) and get the appropriate features.

Parameters:

X : array-like of shape (n_samples, n_features)

The training input samples.

y : array-like of shape (n_samples,)

The target values (class labels in classification, real numbers in regression).

Returns:

self : object

```
fit_transform(self, X, y=None, **fit_params)
```

[\[source\]](#)

Fit to data, then transform it.

Fits transformer to X and y with optional parameters fit_params and returns a transformed version of X.

Parameters:

X : numpy array of shape [n_samples, n_features]

Training set.

y : numpy array of shape [n_samples]

Target values.

****fit_params : dict**

Additional fit parameters.

Returns:

X_new : numpy array of shape [n_samples, n_features_new]

Transformed array.

```
get_params(self, deep=True)
```

[\[source\]](#)

Get parameters for this estimator.

Parameters:

deep : bool, default=True

If True, will return the parameters for this estimator and contained subobjects that are estimators.

Returns:

params : mapping of string to any

Parameter names mapped to their values.

```
get_support(self, indices=False)
```

[\[source\]](#)

Get a mask, or integer index, of the features selected

Parameters:

indices : boolean (default False)

If True, the return value will be an array of integers, rather than a boolean mask.

Toggle Menu

Returns:**support : array**

An index that selects the retained features from a feature vector. If `indices` is `False`, this is a boolean array of shape `[# input features]`, in which an element is `True` iff its corresponding feature is selected for retention. If `indices` is `True`, this is an integer array of shape `[# output features]` whose values are indices into the input feature vector.

`inverse_transform(self, X)`[\[source\]](#)

Reverse the transformation operation

Parameters:**X : array of shape `[n_samples, n_selected_features]`**

The input samples.

Returns:**X_r : array of shape `[n_samples, n_original_features]`**

x with columns of zeros inserted where features would have been removed by [transform](#).

`set_params(self, **params)`[\[source\]](#)

Set the parameters of this estimator.

The method works on simple estimators as well as on nested objects (such as pipelines). The latter have parameters of the form `<component>__<parameter>` so that it's possible to update each component of a nested object.

Parameters:****params : dict**

Estimator parameters.

Returns:**self : object**

Estimator instance.

`transform(self, X)`[\[source\]](#)

Reduce X to the selected features.

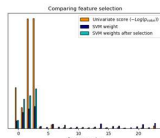
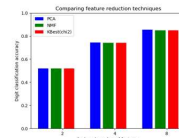
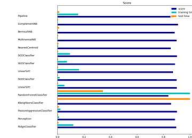
Parameters:**X : array of shape `[n_samples, n_features]`**

The input samples.

Returns:**X_r : array of shape `[n_samples, n_selected_features]`**

The input samples with only the selected features.

Examples using `sklearn.feature_selection.SelectKBest`

[Pipeline Anova SVM](#)[Univariate Feature Selection](#)[Concatenating multiple feature extraction methods](#)[Selecting dimensionality reduction with Pipeline and GridSearchCV](#)[Classification of text documents using sparse features](#)