

## **sklearn.base** (version 1.1.3)

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Base classes for all estimators.

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class **BaseEstimator**([builtins.object](#))

Base class for all estimators in scikit-learn.

Notes

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All estimators should specify all the parameters that can be set at the class level in their ``\_\_init\_\_`` as explicit keyword arguments (no ``\*args`` or ``\*\*kwargs``).

Methods defined here:

**\_\_getstate\_\_**(self)

**\_\_repr\_\_**(self, N\_CHAR\_MAX=700)  
Return repr(self).

**\_\_setstate\_\_**(self, state)

**get\_params**(self, deep=True)  
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 : dict

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Parameter names mapped to their values.

**set\_params**(self, \*\*params)

Set the parameters of this estimator.

The method works on simple estimators as well as on nested objects (such as :class:`~sklearn.pipeline.Pipeline`). 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 : estimator instance  
Estimator instance.

---

Data descriptors defined here:

**\_\_dict\_\_**

dictionary for instance variables (if defined)

**\_\_weakref\_\_**

list of weak references to the object (if defined)

class **BiclusterMixin**(builtins.object)

Mixin class for all bicluster estimators in scikit-learn.

Methods defined here:

**get\_indices**(self, i)

Row and column indices of the `i`'th bicluster.

Only works if ``rows`` and ``columns`` attributes exist.

Parameters

-----

i : int  
The index of the cluster.

Returns

-----

row\_ind : ndarray, dtype=np.intp  
Indices of rows in the dataset that belong to the bicluster.  
col\_ind : ndarray, dtype=np.intp  
Indices of columns in the dataset that belong to the bicluster.

**get\_shape**(self, i)

Shape of the `i`'th bicluster.

Parameters

-----

i : int  
The index of the cluster.

Returns

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```
-----
n_rows : int
    Number of rows in the bicluster.

n_cols : int
    Number of columns in the bicluster.

get_submatrix(self, i, data)
    Return the submatrix corresponding to bicluster `i`.

    Parameters
    -----
    i : int
        The index of the cluster.
    data : array-like of shape (n_samples, n_features)
        The data.

    Returns
    -----
    submatrix : ndarray of shape (n_rows, n_cols)
        The submatrix corresponding to bicluster `i`.

    Notes
    -----
    Works with sparse matrices. Only works if ``rows_`` and
    ``columns_`` attributes exist.
```

---

Readonly properties defined here:

**biclusters\_**  
Convenient way to get row and column indicators together.

Returns the ``rows\_`` and ``columns\_`` members.

---

Data descriptors defined here:

**\_\_dict\_\_**  
dictionary for instance variables (if defined)

**\_\_weakref\_\_**  
list of weak references to the object (if defined)

class **ClassifierMixin**([builtins.object](#))

Mixin class for all classifiers in scikit-learn.

Methods defined here:

**score**(self, X, y, sample\_weight=None)  
Return the mean accuracy on the given test data and labels.

In multi-label classification, this is the subset accuracy which is a harsh metric since you require for each sample that each label set be correctly predicted.

Parameters  
-----  
X : array-like of shape (n\_samples, n\_features)  
 Test samples.

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`y` : array-like of shape `(n_samples,)` or `(n_samples, n_outputs)`  
True labels for ``X``.

`sample_weight` : array-like of shape `(n_samples,)`, default=None  
Sample weights.

Returns

-----

`score` : float  
Mean accuracy of ``self.predict(X)`` wrt. ``y``.

---

Data descriptors defined here:

`__dict__`

dictionary for instance variables (if defined)

`__weakref__`

list of weak references to the object (if defined)

class **ClusterMixin**(builtins.object)

Mixin class for all cluster estimators in scikit-learn.

Methods defined here:

**fit\_predict**(self, X, y=None)

Perform clustering on ``X`` and returns cluster labels.

Parameters

-----

`X` : array-like of shape `(n_samples, n_features)`  
Input data.

`y` : Ignored  
Not used, present for API consistency by convention.

Returns

-----

`labels` : ndarray of shape `(n_samples,)`, dtype=np.int64  
Cluster labels.

---

Data descriptors defined here:

`__dict__`

dictionary for instance variables (if defined)

`__weakref__`

list of weak references to the object (if defined)

class **DensityMixin**(builtins.object)

Mixin class for all density estimators in scikit-learn.

Methods defined here:

**score**(self, X, y=None)

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Return the score of the model on the data `X`.

Parameters

-----

X : array-like of shape (n\_samples, n\_features)  
Test samples.

y : Ignored

Not used, present for API consistency by convention.

Returns

-----

score : float

---

Data descriptors defined here:

**\_\_dict\_\_**

dictionary for instance variables (if defined)

**\_\_weakref\_\_**

list of weak references to the object (if defined)

class **MetaEstimatorMixin**([builtins.object](#))

Data descriptors defined here:

**\_\_dict\_\_**

dictionary for instance variables (if defined)

**\_\_weakref\_\_**

list of weak references to the object (if defined)

class **MultiOutputMixin**([builtins.object](#))

Mixin to mark estimators that support multioutput.

Data descriptors defined here:

**\_\_dict\_\_**

dictionary for instance variables (if defined)

**\_\_weakref\_\_**

list of weak references to the object (if defined)

class **OutlierMixin**([builtins.object](#))

Mixin class for all outlier detection estimators in scikit-learn.

Methods defined here:

**fit\_predict**(self, X, y=None)

Perform fit on X and returns labels for X.

Returns -1 for outliers and 1 for inliers.

Parameters

-----

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X : {array-like, sparse matrix} of shape (n\_samples, n\_features)  
The input samples.

y : Ignored  
Not used, present for API consistency by convention.

Returns

-----

y : ndarray of shape (n\_samples,)  
1 for inliers, -1 for outliers.

---

Data descriptors defined here:

**\_\_dict\_\_**

dictionary for instance variables (if defined)

**\_\_weakref\_\_**

list of weak references to the object (if defined)

class **RegressorMixin**([builtins.object](#))

Mixin class for all regression estimators in scikit-learn.

Methods defined here:

**score**(self, X, y, sample\_weight=None)

Return the coefficient of determination of the prediction.

The coefficient of determination  $R^2$  is defined as  $1 - \frac{u}{v}$ , where  $u$  is the residual sum of squares  $((y_{\text{true}} - y_{\text{pred}})^2).sum()$  and  $v$  is the total sum of squares  $((y_{\text{true}} - y_{\text{true.mean()}})^2).sum()$ . The best possible score is 1.0 and it can be negative (because the model can be arbitrarily worse). A constant model that always predicts the expected value of  $y$ , disregarding the input features, would get a  $R^2$  score of 0.0.

Parameters

-----

X : array-like of shape (n\_samples, n\_features)  
Test samples. For some estimators this may be a precomputed kernel matrix or a list of generic objects instead with shape  $((n_{\text{samples}}, n_{\text{samples\_fitted}}))$ , where  $n_{\text{samples\_fitted}}$  is the number of samples used in the fitting for the estimator.

y : array-like of shape (n\_samples,) or (n\_samples, n\_outputs)  
True values for  $X$ .

sample\_weight : array-like of shape (n\_samples,), default=None  
Sample weights.

Returns

-----

score : float  
 $R^2$  of `self.predict(X)` wrt.  $y$ .

Notes

-----

The  $R^2$  score used when calling `score` on a regressor uses `multioutput='uniform_average'` from version 0.23 to keep consistent

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```
with default value of :func:`~sklearn.metrics.r2_score`.
This influences the ``score`` method of all the multioutput
regressors (except for
:class:`~sklearn.multioutput.MultiOutputRegressor`).
```

---

Data descriptors defined here:

```
__dict__
    dictionary for instance variables (if defined)

__weakref__
    list of weak references to the object (if defined)
```

class **TransformerMixin**([builtins.object](#))

Mixin class for all transformers in scikit-learn.

Methods defined here:

```
fit_transform(self, X, y=None, **fit_params)
    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 : array-like of shape (n_samples, n_features)
        Input samples.

    y : array-like of shape (n_samples,) or (n_samples, n_outputs),
        Target values (None for unsupervised transformations).
                                                default=None

    **fit_params : dict
        Additional fit parameters.

    Returns
    -----
    X_new : ndarray array of shape (n_samples, n_features_new)
        Transformed array.
```

---

Data descriptors defined here:

```
__dict__
    dictionary for instance variables (if defined)

__weakref__
    list of weak references to the object (if defined)
```

## Functions

**clone**(estimator, \*, safe=True)  
Construct a new unfitted estimator with the same parameters.

Clone does a deep copy of the model in an estimator without actually copying attached data. It returns a new estimator with the same parameters that has not been fitted on any data.

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### Parameters

-----

estimator : {list, tuple, set} of estimator instance or a single estimator instance  
The estimator or group of estimators to be cloned.  
safe : bool, default=True  
If safe is False, clone will fall back to a deep copy on objects that are not estimators.

### Returns

-----

estimator : object  
The deep copy of the input, an estimator if input is an estimator.

### Notes

-----

If the estimator's `random_state` parameter is an integer (or if the estimator doesn't have a `random_state` parameter), an *\*exact clone\** is returned: the clone and the original estimator will give the exact same results. Otherwise, *\*statistical clone\** is returned: the clone might return different results from the original estimator. More details can be found in :ref:`randomness`.

### **is\_classifier(estimator)**

Return True if the given estimator is (probably) a classifier.

### Parameters

-----

estimator : object  
Estimator object to test.

### Returns

-----

out : bool  
True if estimator is a classifier and False otherwise.

### **is\_outlier\_detector(estimator)**

Return True if the given estimator is (probably) an outlier detector.

### Parameters

-----

estimator : estimator instance  
Estimator object to test.

### Returns

-----

out : bool  
True if estimator is an outlier detector and False otherwise.

### **is\_regressor(estimator)**

Return True if the given estimator is (probably) a regressor.

### Parameters

-----

estimator : estimator instance  
Estimator object to test.

### Returns

-----

out : bool  
True if estimator is a regressor and False otherwise.