# sklearn.pipeline.make\_pipeline

sklearn.pipeline.make\_pipeline(\*steps, memory=None, verbose=False)

[source]

Construct a **Pipeline** from the given estimators.

This is a shorthand for the <u>Pipeline</u> constructor; it does not require, and does not permit, naming the estimators. Instead, their names will be set to the lowercase of their types automatically.

#### Parameters:

#### \*steps: list of Estimator objects

List of the scikit-learn estimators that are chained together.

## memory: str or object with the joblib. Memory interface, default=None

Used to cache the fitted transformers of the pipeline. By default, no caching is performed. If a string is given, it is the path to the caching directory. Enabling caching triggers a clone of the transformers before fitting. Therefore, the transformer instance given to the pipeline cannot be inspected directly. Use the attribute named\_steps or steps to inspect estimators within the pipeline. Caching the transformers is advantageous when fitting is time consuming.

### verbose: bool, default=False

If True, the time elapsed while fitting each step will be printed as it is completed.

#### **Returns:**

# p: Pipeline

Returns a scikit-learn **Pipeline** object.

#### See also:

**Pipeline** 

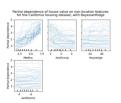
Class for creating a pipeline of transforms with a final estimator.

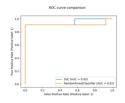
# **Examples**

# Examples using sklearn.pipeline.make\_pipeline



2 a months and a month and a m







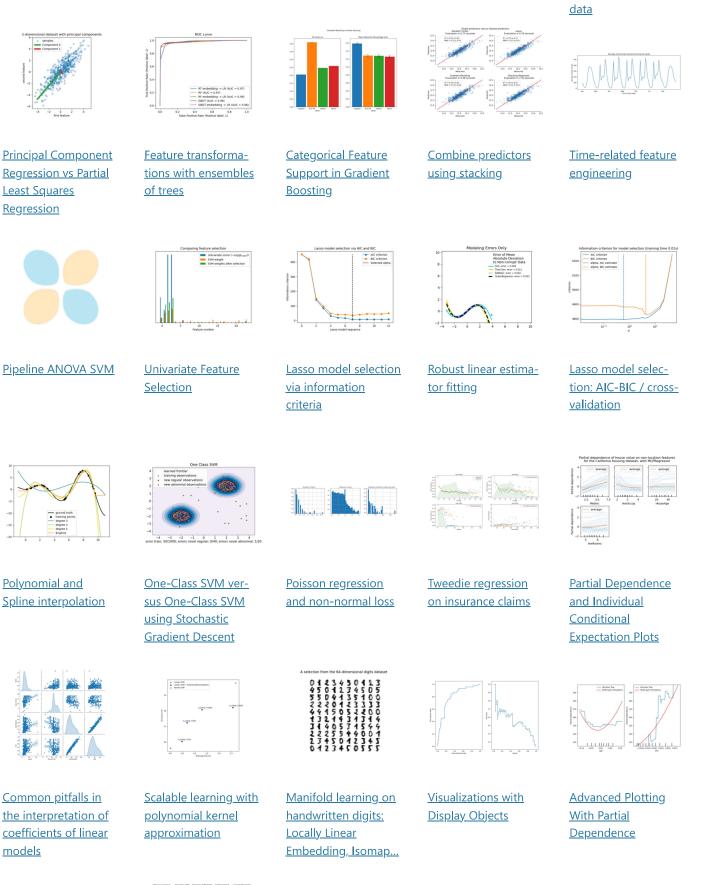
Release Highlights for scikit-learn 1.0

Release Highlights for scikit-learn 0.23

Release Highlights for scikit-learn 0.24

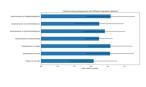
Release Highlights for scikit-learn 0.22

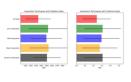
A demo of K-Means clustering on the handwritten digits

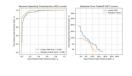












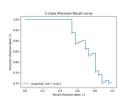
<u>Displaying Pipelines</u> <u>Comparing</u> detection

Comparing anomaly detection algorithms for outlier detection

Imputing missing values with variants of IterativeImputer Imputing missing values before building an estimator

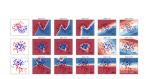
Detection error tradeoff (DET) curve

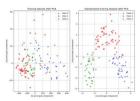
# on toy datasets











Precision-Recall

Dimensionality
Reduction with
Neighborhood
Components Analysis



<u>Varying regularization</u> <u>in Multi-layer</u> <u>Perceptron</u>

Importance of Feature Scaling





Feature discretization

<u>Clustering text docu-</u> <u>ments using k-means</u>

© 2007 - 2022, scikit-learn developers (BSD License). Show this page source