

Open-source Python packages for Feature Selection

Open-source for Feature engineering



Feature-engine

Fit – transform functionality



- `fit()` → finds important features
- `transform()` → transforms data
 - Removes unwanted features

• Pipeline



: *# we stack all the selection methods inside a pipeline*

```
pipe = Pipeline([
    ('constant', DropConstantFeatures(tol=0.998)),
    ('duplicated', DropDuplicateFeatures()),
    ('correlation', SmartCorrelatedSelection(selection_method='variance')),
])

pipe.fit(X_train)
```

Pipeline



train pipeline

```
price_pipe.fit(X_train, y_train)
```

transform data

```
price_pipe.transform(X_train)
```

```
price_pipe.transform(X_test)
```

Please [cite us](#) if you use the software.

1.13. Feature selection

1.13.1. Removing features with low variance

1.13.2. Univariate feature selection

1.13.3. Recursive feature elimination

1.13.4. Feature selection using SelectFromModel

1.13.5. Sequential Feature Selection

1.13.6. Feature selection as part of a pipeline

1.13. Feature selection

The classes in the `sklearn.feature_selection` module can be used for feature selection/dimensionality reduction on sample sets, either to improve estimators' accuracy scores or to boost their performance on very high-dimensional datasets.

1.13.1. Removing features with low variance

VarianceThreshold is a simple baseline approach to feature selection. It removes all features whose variance doesn't meet some threshold. By default, it removes all zero-variance features, i.e. features that have the same value in all samples.

As an example, suppose that we have a dataset with boolean features, and we want to remove all features that are either one or zero (on or off) in more than 80% of the samples. Boolean features are Bernoulli random variables, and the variance of such variables is given by

$$\text{Var}[X] = p(1 - p)$$

so we can select using the threshold `.8 * (1 - .8)`:

```
>>> from sklearn.feature_selection import VarianceThreshold
>>> X = [[0, 0, 1], [0, 1, 0], [1, 0, 0], [0, 1, 1], [0, 1, 0], [0, 1, 1]]
>>> sel = VarianceThreshold(threshold=(.8 * (1 - .8)))
>>> sel.fit_transform(X)
array([[0, 1],
       [1, 0],
       [0, 0],
       [1, 1],
       [1, 0],
       [1, 1]])
```

As expected, `VarianceThreshold` has removed the first column, which has a probability $p = 5/6 > .8$ of containing a zero.

Welcome to mlxtend's
documentation!

Links

Examples

License

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Welcome to mlxtend's documentation!

Mlxtend (machine learning extensions) is a Python library of useful tools for the day-to-day data science tasks.

JOSS [10.21105/joss.00638](#) pypi package [0.18.0](#) python [3.6](#) python [3.7](#) license [BSD](#) discuss [google group](#)

Links

- Documentation: <http://rasbt.github.io/mlxtend>
- Source code repository: <https://github.com/rasbt/mlxtend>
- PyPI: <https://pypi.python.org/pypi/mlxtend>
- Questions? Check out the [Google Groups mailing list](#)

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documentation!

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- mlxtend.classifier
- mlxtend.cluster
- mlxtend.data
- mlxtend.evaluate
- mlxtend.feature extraction
- mlxtend.feature selection
- mlxtend.file io
- mlxtend.frequent patterns
- mlxtend.image
- mlxtend.plotting
- mlxtend.preprocessing
- mlxtend.regressor
- mlxtend.text
- mlxtend.utils



mlxtend's documentation!

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8 pypi package 0.18.0 python 3.6 python 3.7 license BSD discuss google group

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Feature-engine



- <https://www.trainindata.com/feature-engine>
- <https://feature-engine.readthedocs.io/en/latest/>
- https://github.com/solegalli/feature_engine

```
pip install feature-engine
```

```
conda install -c conda-forge feature_engine
```



1.0.0

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LEARNING RESOURCES

Feature-engine: A Python library for Feature Engineering for Machine Learning



Feature-engine rocks!

Feature-engine is a Python library with multiple transformers to engineer features for use in machine learning models. Feature-engine preserves Scikit-learn functionality with methods `fit()` and `transform()` to learn parameters from and then transform the data.

Feature-engine includes transformers for:

- Missing data imputation
- Categorical variable encoding
- Discretisation
- Variable transformation

LEARNING RESOURCES**CONTRIBUTE**

Feature Selection

Feature-engine's feature selection transformers are used to drop subsets of variables. Or in other words to select subsets of variables.

- [DropFeatures](#)
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Thank you

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