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# **brew Documentation**

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## brew: A Multiple Classifier Systems API

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This project was started in 2014 by *Dayvid Victor* and *Thyago Porpino* for the Multiple Classifier Systems class at Federal University of Pernambuco.

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The aim of this project is to provide an easy API for Ensembling, Stacking, Blending, Ensemble Generation, Ensemble Pruning, Dynamic Classifier Selection, and Dynamic Ensemble Selection.

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## Example

```
import numpy as np
import matplotlib.pyplot as plt
import matplotlib.gridspec as gridspec
import itertools

import sklearn

from sklearn.linear_model import LogisticRegression
from sklearn.svm import SVC
from sklearn.ensemble import RandomForestClassifier

from brew.base import Ensemble, EnsembleClassifier
from brew.stacking.stacker import EnsembleStack, EnsembleStackClassifier
from brew.combination.combiner import Combiner

from mlxtend.data import iris_data
from mlxtend.evaluate import plot_decision_regions

# Initializing Classifiers
clf1 = LogisticRegression(random_state=0)
clf2 = RandomForestClassifier(random_state=0)
clf3 = SVC(random_state=0, probability=True)

# Creating Ensemble
ensemble = Ensemble([clf1, clf2, clf3])
eclf = EnsembleClassifier(ensemble=ensemble, combiner=Combiner('mean'))

# Creating Stacking
layer_1 = Ensemble([clf1, clf2, clf3])
layer_2 = Ensemble([sklearn.clone(clf1)])

stack = EnsembleStack(cv=3)

stack.add_layer(layer_1)
stack.add_layer(layer_2)

sclf = EnsembleStackClassifier(stack)

clf_list = [clf1, clf2, clf3, eclf, sclf]
lbl_list = ['Logistic Regression', 'Random Forest', 'RBF kernel SVM', 'Ensemble', 'Stacking']

# Loading some example data
X, y = iris_data()
```

```
X = X[:, [0, 2]]

# Plotting Decision Regions
gs = gridspec.GridSpec(2, 3)
fig = plt.figure(figsize=(10, 8))

itt = itertools.product([0, 1, 2], repeat=2)

for clf, lab, grd in zip(clf_list, lbl_list, itt):
    clf.fit(X, y)
    ax = plt.subplot(gs[grd[0], grd[1]])
    fig = plot_decision_regions(X=X, y=y, clf=clf, legend=2)
    plt.title(lab)
plt.show()
```

## 1.1 Features

- Ensembling, Blending and Stacking.
- Dynamic Classifier Selection: OLA, LCA, A Priori, A Posteriori.
- Dynamic Ensemble Selection: KNORA E and KNORA U.
- Ensemble Combination Rules: majority vote, min, max, mean and median.
- Ensemble Diversity Metrics: Entropy Measure E, Kohavi Wolpert Variance, I Q Statistics, Correlation Coefficient p, Disagreement Measure, Agreement Measure, Double Fault Measure.
- Ensemble Classifier Generators: Bagging, Random Subspace, SMOTEBagging, ICS-Bagging, SMOTE-ICS-Bagging.
- Ensemble Pruning: EPIC.
- Oversampling: SMOTE.

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## Important References

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- Kuncheva, Ludmila I. Combining pattern classifiers: methods and algorithms. John Wiley & Sons, 2014.
- Zhou, Zhi-Hua. Ensemble methods: foundations and algorithms. CRC Press, 2012.



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## Dependencies

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- Python 2.6+
- scikit-learn  $\geq$  0.14.1
- Numpy  $\geq$  1.3
- SciPy  $\geq$  0.7
- Matplotlib  $\geq$  0.99.1 (examples, only)
- mlxtend (examples, only)



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**Contents:**

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## **4.1 brew package**

### **4.1.1 Subpackages**

**brew.combination package**

Submodules

**brew.combination.combiner module**

**brew.combination.rules module**

Module contents

**brew.generation package**

Submodules

**brew.generation.bagging module**

**brew.generation.base module**

**brew.generation.random\_subspace module**

**brew.generation.smote\_bagging module**

Module contents

**brew.metrics package**

Subpackages

**brew.metrics.diversity package**

**Submodules**

**brew.metrics.diversity.base module**

**brew.metrics.diversity.non\_paired module**

**brew.metrics.diversity.paired module**

**Module contents**

**Submodules**

**brew.metrics.evaluation module**

**Module contents**

**brew.preprocessing package**

**Submodules**

**brew.preprocessing.smote module**

**Module contents**

**brew.selection package**

**Subpackages**

**brew.selection.dynamic package**

**Submodules**

**brew.selection.dynamic.base module**

**brew.selection.dynamic.knora module**

**brew.selection.dynamic.lca module**

**brew.selection.dynamic.ola module**

**Module contents**

**brew.selection.pruning package**

## Submodules

**brew.selection.pruning.epic module**

### Module contents

### Module contents

**brew.utils package**

### Submodules

**brew.utils.data module**

### Module contents

## 4.1.2 Submodules

## 4.1.3 brew.base module

## 4.1.4 Module contents

# 4.2 Installation

At the command line either via `easy_install` or `pip`:

```
$ easy_install brew
$ pip install brew
```

Or, if you have `virtualenvwrapper` installed:

```
$ mkvirtualenv brew
$ pip install brew
```

# 4.3 Usage

To use brew in a project:

```
import brew
from brew.base import Ensemble
from brew.base import EnsembleClassifier
from brew.combination import Combiner

# here, clf1 and clf2 are sklearn classifiers or brew ensemble classifiers
# already trained. Keep in mind that brew requires your labels = [0,1,2,...]
# numerical with no skips.
clfs = [clf1, clf2]
ens = Ensemble(classifiers = clfs)

# create your Combiner
```

```
# the rules can be 'majority_vote', 'max', 'min', 'mean' or 'median'
comb = Combiner(rule='mean')

# now create your ensemble classifier
ensemble_clf = EnsembleClassifier(ensemble=ens, combiner=comb)
y_pred = ensemble_clf.predict(X)

# there you go, y_pred is your prediction.
```

## 4.4 Contributing

Contributions are welcome, and they are greatly appreciated! Every little bit helps, and credit will always be given.

You can contribute in many ways:

### 4.4.1 Types of Contributions

#### Report Bugs

Report bugs at <https://github.com/viisar/brew/issues>.

If you are reporting a bug, please include:

- Your operating system name and version.
- Any details about your local setup that might be helpful in troubleshooting.
- Detailed steps to reproduce the bug.

#### Fix Bugs

Look through the GitHub issues for bugs. Anything tagged with “bug” is open to whoever wants to implement it.

#### Implement Features

Look through the GitHub issues for features. Anything tagged with “feature” is open to whoever wants to implement it.

#### Write Documentation

brew could always use more documentation, whether as part of the official brew docs, in docstrings, or even on the web in blog posts, articles, and such.

#### Submit Feedback

The best way to send feedback is to file an issue at <https://github.com/viisar/brew/issues>.

If you are proposing a feature:

- Explain in detail how it would work.
- Keep the scope as narrow as possible, to make it easier to implement.

- Remember that this is a volunteer-driven project, and that contributions are welcome :)

## 4.4.2 Get Started!

Ready to contribute? Here's how to set up *brew* for local development.

1. **Fork** the *brew* repo on GitHub.
2. Clone your fork locally:

```
$ git clone git@github.com:your_name_here/brew.git
```

3. Create a branch for local development:

```
$ git checkout -b name-of-your-bugfix-or-feature
```

Now you can make your changes locally.

4. When you're done making changes, check that your changes pass style and unit tests, including testing other Python versions with `tox`:

```
$ tox
```

To get `tox`, just `pip` install it.

5. Commit your changes and push your branch to GitHub:

```
$ git add .  
$ git commit -m "Your detailed description of your changes."  
$ git push origin name-of-your-bugfix-or-feature
```

6. Submit a pull request through the GitHub website.

## 4.4.3 Pull Request Guidelines

Before you submit a pull request, check that it meets these guidelines:

1. The pull request should include tests.
2. If the pull request adds functionality, the docs should be updated. Put your new functionality into a function with a docstring, and add the feature to the list in `README.rst`.
3. The pull request should work for Python 2.6, 2.7, and 3.3, and for PyPy. Check <https://travis-ci.org/viisar/brew> under pull requests for active pull requests or run the `tox` command and make sure that the tests pass for all supported Python versions.

## 4.4.4 Tips

To run a subset of tests:

```
$ py.test test/test_brew.py
```

## 4.5 Credits

### 4.5.1 Development Lead

- Dayvid Victor <victor.dvro@gmail.com>
- Thyago Porpino <thyago.porpino@gmail.com>

### 4.5.2 Contributors

None yet. Why not be the first?

## 4.6 History

### 4.6.1 0.1.0 (2014-11-12)

- First release on PyPI.

## Feedback

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If you have any suggestions or questions about **brew** feel free to email me at [victor.dvro@gmail.com](mailto:victor.dvro@gmail.com).

If you encounter any errors or problems with **brew**, please let me know! Open an Issue at the GitHub <http://github.com/dvro/brew> main repository.



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