Semi-supervised learning

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.. currentmodule:: sklearn.semi_supervised
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Semi-supervised learning is a situation in which in your training data some of the samples are not labeled. The semi-supervised estimators in mod: sklearn.semi_supervised are able to make use of this additional unlabeled data to better capture the shape of the underlying data distribution and generalize better to new samples. These algorithms can perform well when we have a very small amount of labeled points and a large amount of unlabeled points.

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Unlabeled entries in y

It is important to assign an identifier to unlabeled points along with the labeled data when training the model with the fit method. The identifier that this implementation uses is the integer value -1. Note that for string labels, the dtype of y should be object so that it can contain both strings and integers.

Note

Semi-supervised algorithms need to make assumptions about the distribution of the dataset in order to achieve performance gains. See here for more details.

Self Training

This self-training implementation is based on Yarowsky's [1] algorithm. Using this algorithm, a given supervised classifier can function as a semi-supervised classifier, allowing it to learn from unlabeled data.

class: SelfTrainingClassifier` can be called with any classifier that implements *predict_proba*, passed as the parameter base_classifier. In each iteration, the base_classifier predicts labels for the unlabeled samples and adds a subset of these labels to the labeled dataset.

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The choice of this subset is determined by the selection criterion. This selection can be done using a *threshold* on the prediction probabilities, or by choosing the k_best samples according to the prediction probabilities.

The labels used for the final fit as well as the iteration in which each sample was labeled are available as attributes. The optional max_iter parameter specifies how many times the loop is executed at most.

The max_iter parameter may be set to None, causing the algorithm to iterate until all samples have labels or no new samples are selected in that iteration.

Note

When using the self-training classifier, the ref calibration calibration of the classifier is important.

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Examples

ref. sphx glr auto examples semi supervised plot self training varying threshold.py

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• ref. sphx glr auto examples semi supervised plot semi supervised versus svm iris.py

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References

[1] :doi:\"Unsupervised word sense disambiguation rivaling supervised methods" <10.3115/981658.981684>\text{\text}\) David Yarowsky, Proceedings of the 33rd annual meeting on Association for Computational Linguistics (ACL '95). Association for Computational Linguistics, Stroudsburg, PA, USA, 189-196.

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Label Propagation

Label propagation denotes a few variations of semi-supervised graph inference algorithms.

A few features available in this model:

- Used for classification tasks
- Kernel methods to project data into alternate dimensional spaces

scikit-learn provides two label propagation models: :class: LabelPropagation` and :class: LabelSpreading`. Both work by constructing a similarity graph over all items in the input dataset.

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An illustration of label-propagation: the structure of unlabeled observations is consistent with the class structure, and thus the class label can be propagated to the unlabeled observations of the training set.

class: LabelPropagation` and class: LabelSpreading` differ in modifications to the similarity matrix that graph and the clamping effect on the label distributions. Clamping allows the algorithm to change the weight of the true ground labeled data to some degree. The class: LabelPropagation` algorithm performs hard clamping of input labels, which means $\alpha = 0$. This clamping factor can be relaxed, to say $\alpha = 0.2$, which means that we will always retain 80 percent of our original label distribution, but the algorithm gets to change its confidence of the distribution within 20 percent.

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xclass: LabelPropagation` uses the raw similarity matrix constructed from the data with no modifications. In contrast, xclass: LabelSpreading` minimizes a loss function that has regularization properties, as such it is often more robust to noise. The algorithm iterates on a modified version of the original graph and normalizes the edge weights by computing the normalized graph Laplacian matrix. This procedure is also used in ref: spectral clustering`.

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Label propagation models have two built-in kernel methods. Choice of kernel effects both scalability and performance of the algorithms. The following are available:

- rbf (exp($-\gamma |x-y|^2$), $\gamma > 0$). γ is specified by keyword gamma.
- $knn(1[x' \in kNN(x)])$. k is specified by keyword n neighbors.

The RBF kernel will produce a fully connected graph which is represented in memory by a dense matrix. This matrix may be very large and combined with the cost of performing a full matrix multiplication calculation for each iteration of the algorithm can lead to prohibitively long running times. On the other hand, the KNN kernel will produce a much more memory-friendly sparse matrix which can drastically reduce running times.

Examples

ref. sphx_glr_auto_examples_semi_supervised_plot_semi_supervised_versus_svm_iris.py

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• ref: sphx_glr_auto_examples_semi_supervised_plot_label_propagation_structure.py

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ref. sphx glr auto examples semi supervised plot label propagation digits.py

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• ref. sphx glr auto examples semi supervised plot label propagation digits active learning.py

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References

- [2] Yoshua Bengio, Olivier Delalleau, Nicolas Le Roux. In Semi-Supervised Learning (2006), pp. 193-216
- [3] Olivier Delalleau, Yoshua Bengio, Nicolas Le Roux. Efficient Non-Parametric Function Induction in Semi-Supervised Learning. AISTAT 2005 https://research.microsoft.com/en-us/people/nicolasl/efficient_ssl.pdf