



Scikit-learn strives to have a uniform interface across all methods, and we'll see examples of these below.

Given a scikit-learn *estimator* object named `model`, the following methods are available...



Scikit Learn

- Available in **all Estimators**
 - `model.fit()` : fit training data.
 - For supervised learning applications, this accepts two arguments: the data `X` and the labels `y` (e.g. `model.fit(X, y)`).
 - For unsupervised learning applications, this accepts only a single argument, the data `X` (e.g. `model.fit(X)`).



Available in **supervised estimators**

- `model.predict()` : given a trained model, predict the label of a new set of data. This method accepts one argument, the new data `X_new` (e.g. `model.predict(X_new)`), and returns the learned label for each object in the array.



Available in **supervised estimators**

- `model.predict_proba()` : For classification problems, some estimators also provide this method, which returns the probability that a new observation has each categorical label. In this case, the label with the highest probability is returned by `model.predict()`.



Available in **supervised estimators**

- `model.score()` : for classification or regression problems, most estimators implement a score method. Scores are between 0 and 1, with a larger score indicating a better fit.



Available in **unsupervised estimators**

- `model.predict()` : predict labels in clustering algorithms.



Available in **unsupervised estimators**

- `model.transform()` : given an unsupervised model, transform new data into the new basis. This also accepts one argument `X_new`, and returns the new representation of the data based on the unsupervised model.

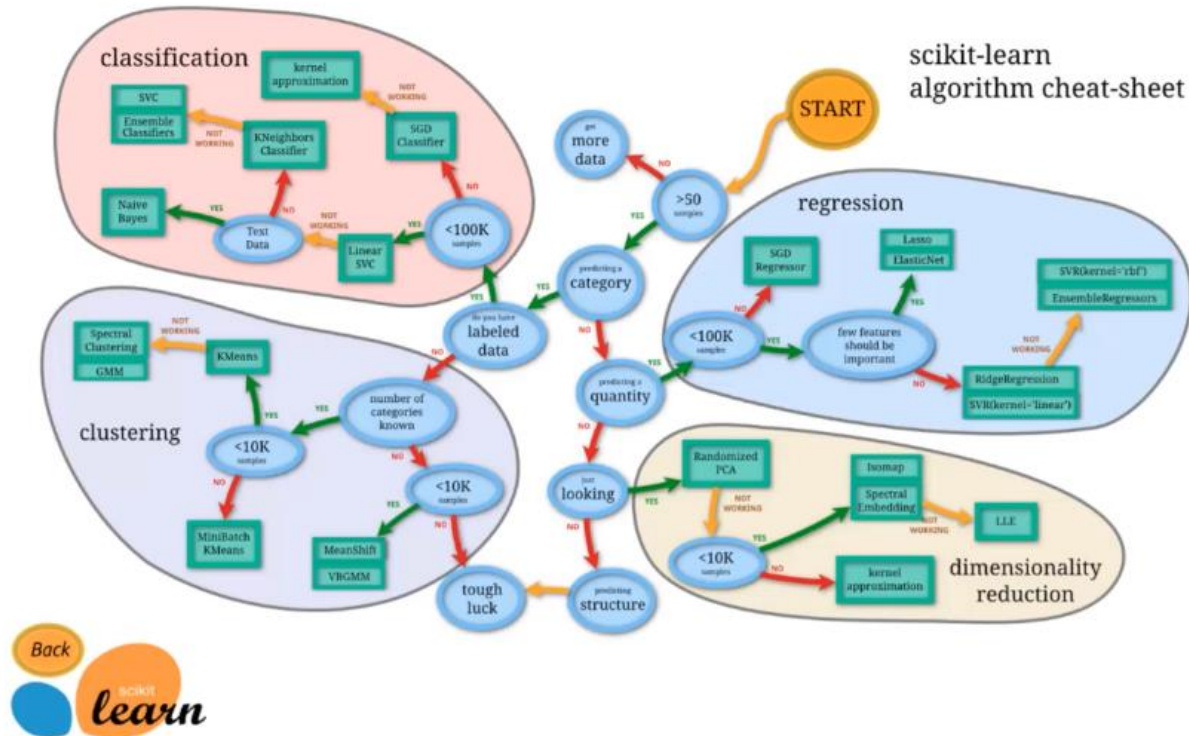


Available in **unsupervised estimators**

- `model.fit_transform()` : some estimators implement this method, which more efficiently performs a fit and a transform on the same input data.



Choosing an Algorithm



ML with Python