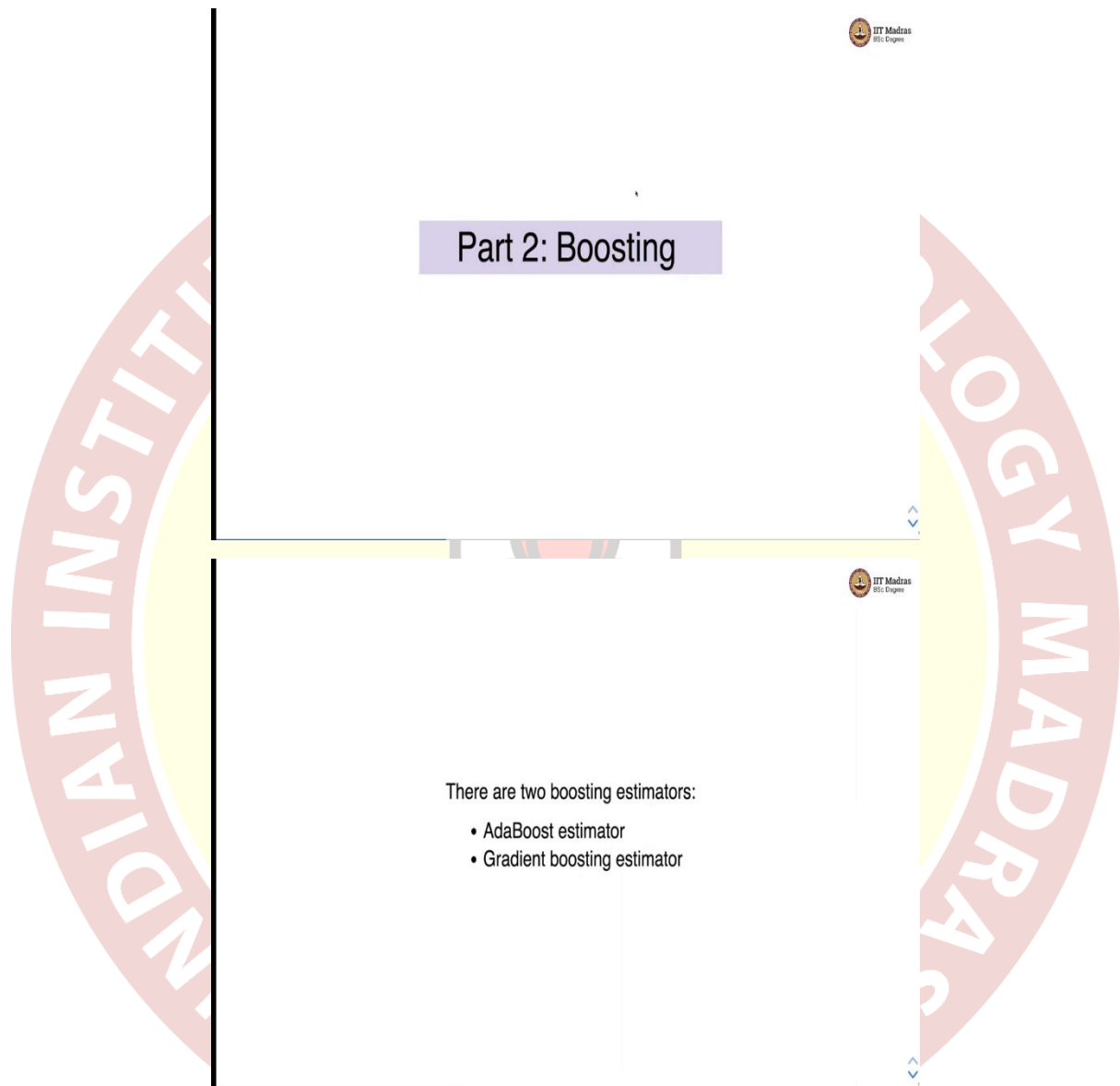


IIT Madras

ONLINE DEGREE

Machine Learning Practice
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Boosting: AdaBoost, GradientBoost

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Namaste! Welcome to the next video of Machine Learning Practice Course. In this video, we will discuss about boosting techniques as implemented in sklearn. There are 2 boosting estimators AdaBoost estimator and Gradient boosting estimator that are of our interest as far as this course is concerned.

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AdaBoost estimator

Class: `sklearn.ensemble.AdaBoostClassifier`

Class: `sklearn.ensemble.AdaBoostRegressor`

Class: `sklearn.ensemble.AdaBoostClassifier`

<code>base_estimator</code>	<ul style="list-style-type: none">• Default estimator is <code>DecisionTreeClassifier</code> with <code>depth = 1</code>.
<code>n_estimators</code>	<ul style="list-style-type: none">• Maximum number of estimators where boosting is terminated. The default value is 50.
<code>learning_rate</code>	<ul style="list-style-type: none">• Weight applied to each classifier during boosting.• Higher value here would increase contribution of individual classifiers.• There is a trade-off between <code>n_estimators</code> and <code>learning_rate</code>.

AdaBoost estimators there are 2 of them, one is `AdaBoostClassifier`, and `AdaBoostRegressor` and they are implemented as part of `sklearn.ensemble` model. `AdaBoostClassifier` takes `base_estimator` parameter that tells us what kind of estimator we will use to start the boosting procedure. The default estimator is `DecisionTreeClassifier` with `depth = 1`, then there is a parameter called `n_estimators` that provides the maximum number of estimators where boosting is terminated, the default value is 50.

Then there is a `learning_rate`, which specifies the weight to be applied to each classifier during the boosting. Higher value for `learning_rate` would increase the contribution of individual classifiers. There is a trade-off between `n_estimators` parameter and `learning_rate` parameter.

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Class: `sklearn.ensemble.AdaBoostRegressor`

- | | |
|-----------------------------|--|
| <code>base_estimator</code> | <ul style="list-style-type: none">• Default estimator is <code>DecisionTreeRegressor</code> with <code>depth = 3</code>. |
| <code>n_estimators</code> | <ul style="list-style-type: none">• Maximum number of estimators where boosting is terminated. The default value is 50. |
| <code>learning_rate</code> | <ul style="list-style-type: none">• Weight applied to each regressor at each boosting iteration.• Higher value here would increase contribution of individual regressor.• There is a trade-off between <code>n_estimators</code> and <code>learning_rate</code>. |

Let us look at `AdaBoostRegressor`. It also has a `base_estimator`. And the default `base_estimator` is `DecisionTreeRegressor` with `depth = 3`. Number of estimators are set to default value of 50 and they specify the maximum number of estimators where the boosting will be terminated. And `learning_rate`, which is weight applied to each regressor, higher value here would increase contribution of individual regressor and there is a trade-off between `n_estimators` and `learning_rate`.

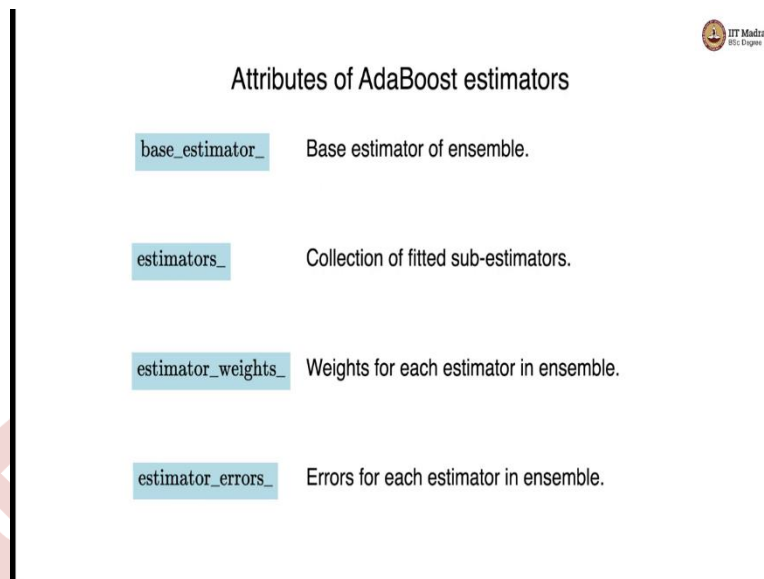
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The main parameters to tune to obtain good results are

- `n_estimators` and
- Complexity of the base estimators (e.g. its depth `max_depth` or `min_samples_split`).

There are a couple of parameters that we can tune to obtain good results in case of `AdaBoost` estimators. One is the number of estimators and second is complexity of `base_estimator`, where the complexity is defined in terms of the depth of the tree, and minimum samples required to split a node.

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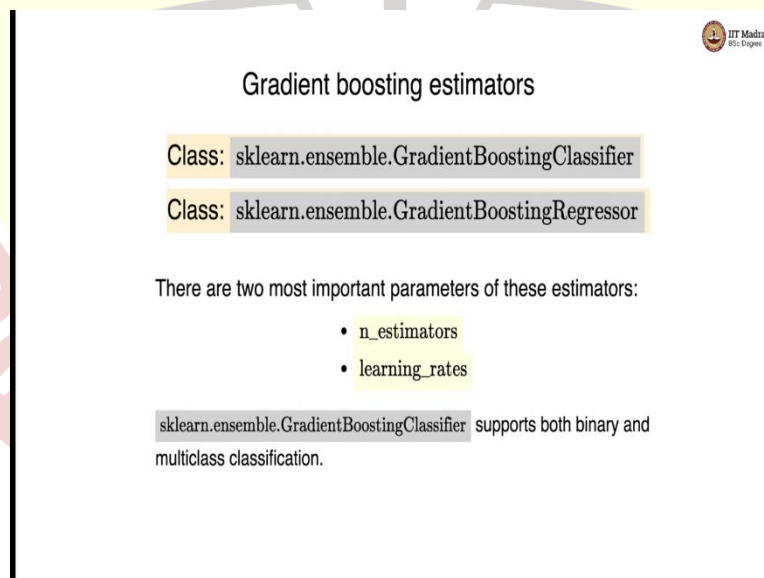


Attributes of AdaBoost estimators

- `base_estimator_` Base estimator of ensemble.
- `estimators_` Collection of fitted sub-estimators.
- `estimator_weights_` Weights for each estimator in ensemble.
- `estimator_errors_` Errors for each estimator in ensemble.

Let us look at attributes of AdaBoost estimators. So, there is an attribute which is `base_estimator_` that provides basic estimator of the ensemble. `Estimator_` is a collection of fitted sub-estimators. An `estimator_weights_` provides the weights of each estimator in ensemble. An `estimator_errors_` provides errors of each estimator in ensemble.

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Gradient boosting estimators

- Class: `sklearn.ensemble.GradientBoostingClassifier`
- Class: `sklearn.ensemble.GradientBoostingRegressor`

There are two most important parameters of these estimators:

- `n_estimators`
- `learning_rate`

`sklearn.ensemble.GradientBoostingClassifier` supports both binary and multiclass classification.

Let us look at gradient boosting estimators, there are there is a `GradientBoostingClassifier` and `GradientBoostingRegressor`. And there are 2 important parameters in these estimators again, which is number of estimators where the boosting will be terminated under learning `_rate`. `GradientBoostingClassifier` supports both binary and multiclass classification.

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We will directly demonstrate XGBoost through colab demonstration.

And finally, there is one more boosting method which is XGboost which we will demonstrate directly through colab. So, in this video, we look at various boosting estimators in sklearn. So, AdaBoost and gradient boosting is implemented in sklearn, whereas XGBoost is not directly implemented in sklearn. And we will be implementing an XGboost with some other library. And we will see that through the colab demonstration.

