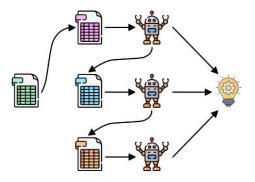


Module 10- Part II Boosting Models AdaBoost, GBM, XGBoost



Prof. Pedram Jahangiry

Boosting



Sequential





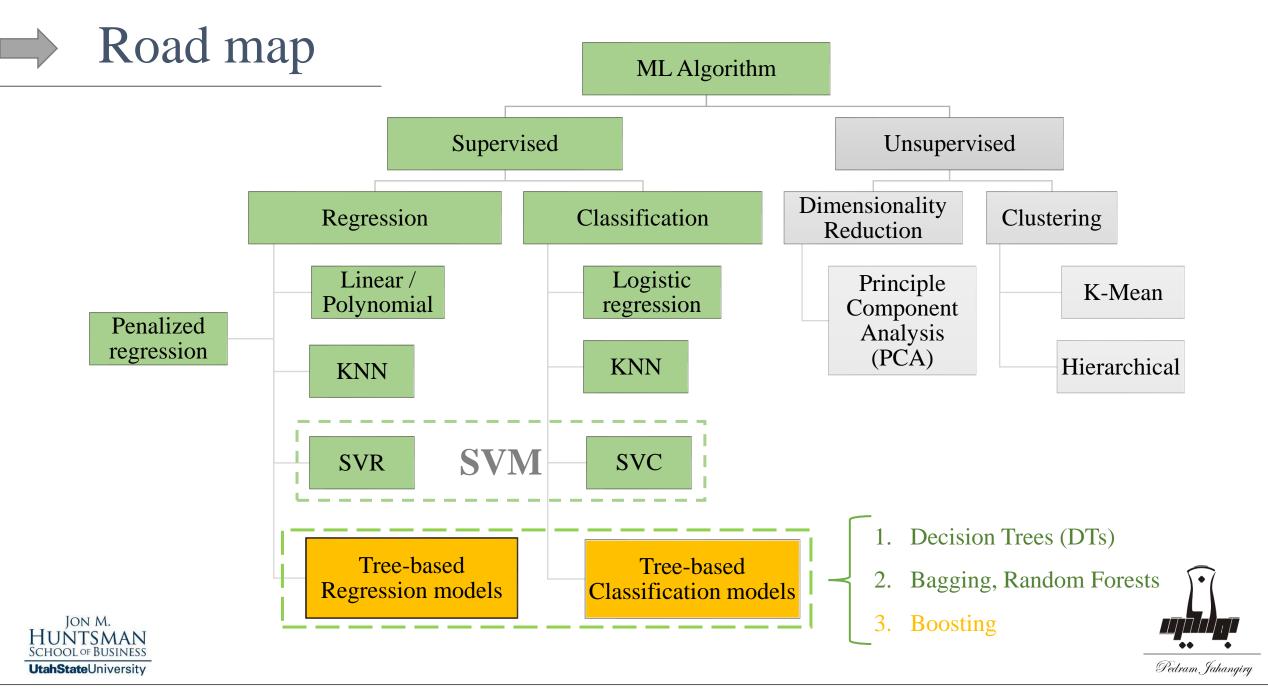


Class Modules

- Module 1- Introduction to Deep Learning
- Module 2- Setting up Machine Learning Environment
- Module 3- Linear Regression (Econometrics approach)
- Module 4- Machine Learning Fundamentals
- Module 5- Linear Regression (Machine Learning approach)
- Module 6- Penalized Regression (Ridge, LASSO, Elastic Net)
- Module 7- Logistic Regression
- Module 8- K-Nearest Neighbors (KNN)
- Module 9- Classification and Regression Trees (CART)
- Module 10- Bagging and Boosting
- Module 11- Dimensionality Reduction (PCA)
- Module 12- Clustering (KMeans Hierarchical)









Topics

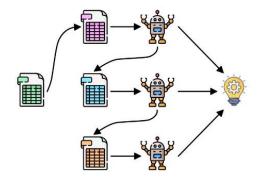
Part I

- 1. Bagging vs Boosting
- 2. AdaBoost
- 3. Gradient Boosting Machine (GBM)
- 4. XGBoost

Part II

Pros and Cons

Boosting



Sequential





Part I

- 1. Bagging vs Boosting
- 2. AdaBoost
- 3. Gradient Boosting Machine (GBM)
- 4. XGBoost





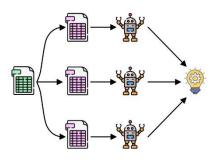


Bagging vs Boosting

- Bagging consists of creating many "copies" of the training data (each copy is slightly different from another) and then apply the weak learner to each copy to obtain multiple weak models and then combine them.
- In bagging, the bootstrapped trees are independent from each other.
- Boosting consists of using the "original" training data and iteratively creating multiple models by using a weak learner. Each new model would be different from the previous ones in the sense that the weak learner, by building each new model tries to "fix" the errors which previous models make.
- In boosting, each tree is grown using information from previous tree.

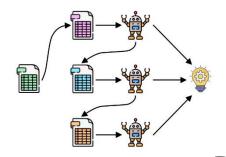


Bagging



Parallel

Boosting



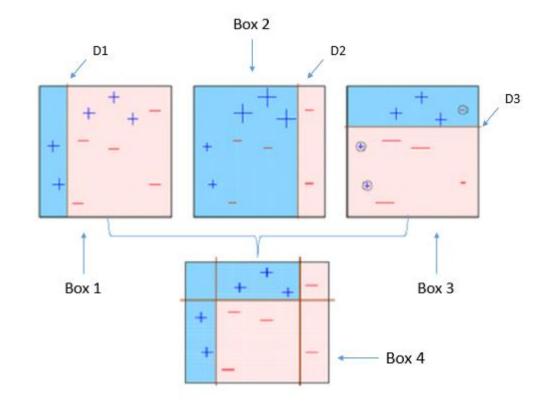
Sequential





AdaBoost (Adaptive Boosting)

- Forest of weak learners (trees with only 1 feature; stumps).
- Each tree (stump) depends on the previous tree's errors rather than being independent.
- 1) Starting with usual splitting criteria!
- 2) Each tree (stump) gets different weight based on its prediction accuracy.
- 3) Each observation gets a weight inversely related to its predicted outcome. (ex, misclassified ones get more weight).
- 4) Aggregation is done based on each weak learner's weight.



Source: Towards data science





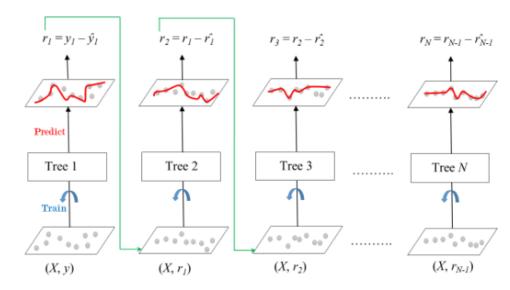


UtahStateUniversity

Gradient Boosting Machine (GBM)

Source: <u>Geeksforgeeks</u>

- In gradient boosting, each weak learner corrects its predecessor's error.
- Unlike AdaBoost, the weights of the training instances are not tweaked, instead, each predictor is trained using the residual errors of predecessor as labels.
- Unlike AdaBoost, each tree can be larger than a stump. However, the trees are still small. By fitting a small tree to the residuals, the GBM slowly improve \hat{f} in areas where it does not perform well.

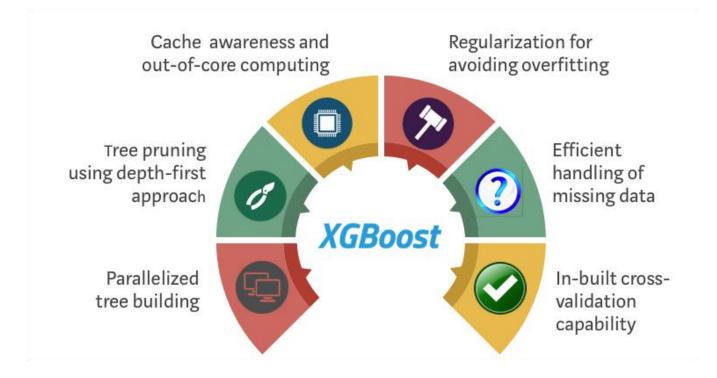


- Learning rate shrinks the contribution of each tree. There is a trade-off between learning rate and number of trees. Learning rate slows down the process even further, allowing for more and different shaped trees to attack the residuals.
- Aggregation is done by adding the first tree predictions and a scaled (shrunk) version of the following ION Mtrees.



Extreme Gradient Boosting (XGBoost)

- XGBoost is a refined and customized version of a gradient boosting decision tree system, created with performance and speed in mind.
- Extreme refers to the fact that the algorithms and methods have been customized to push the limit of what is possible for gradient boosting algorithms.

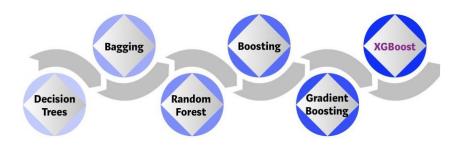






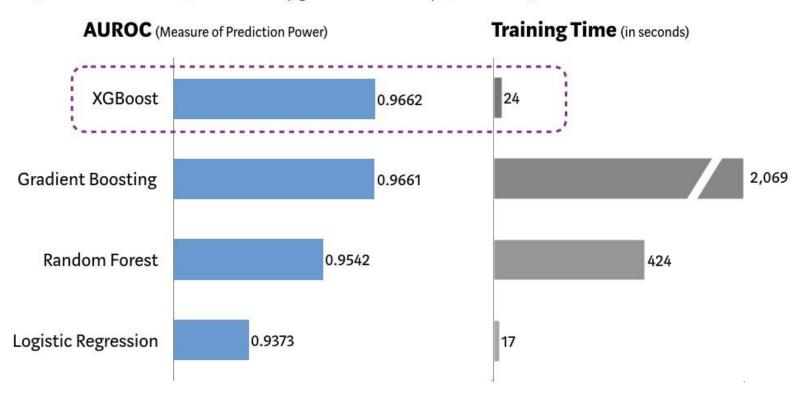


Put it all together!



Performance Comparison using SKLearn's 'Make_Classification' Dataset

(5 Fold Cross Validation, 1MM randomly generated data sample, 20 features)







Part II Pros and Cons

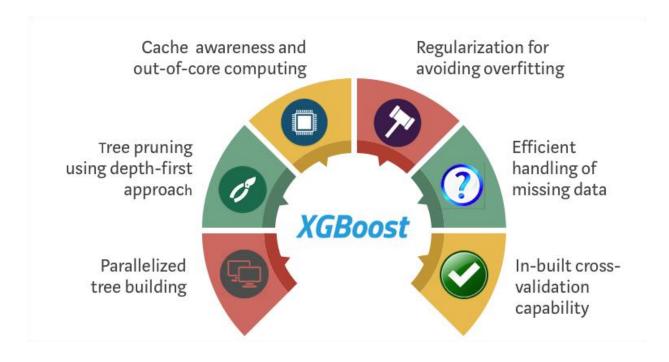






XGBoost's Pros and Cons

Pros:



Cons:

• XGBoost is more difficult to understand, visualize and to tune compared to AdaBoost and random forests. There is a multitude of hyperparameters that can be tuned to increase performance.







Class Modules

- ✓ Module 1- Introduction to Deep Learning
- ✓ Module 2- Setting up Machine Learning Environment
- ✓ Module 3- Linear Regression (Econometrics approach)
- ✓ Module 4- Machine Learning Fundamentals
- ✓ Module 5- Linear Regression (Machine Learning approach)
- ✓ Module 6- Penalized Regression (Ridge, LASSO, Elastic Net)
- ✓ Module 7- Logistic Regression
- ✓ Module 8- K-Nearest Neighbors (KNN)
- ✓ Module 9- Classification and Regression Trees (CART)
- ✓ Module 10- Bagging and Boosting
- Module 11- Dimensionality Reduction (PCA)
- Module 12- Clustering (KMeans Hierarchical)



