MACHINE LEARNING

Ensemble Methods – Bagging and Boosting



ENSEMBLE METHODS

- Ensemble methods are meta-algorithms which combine several machine learning models together and creating a single predictive model out of it.
- The main goal of using ensemble methods are : to reduce variance (bagging), to reduce bias (boosting) and to improve the predictions.
- Stacking is also another method under ensemble methods but most widely used techniques are Bagging and Boosting.
- Bagging and boosting algorithms can be used in both regression problems and classification problems.
- Random Forest is one of the prominent bagging algorithm & XGBoosting (Extreme Gradient Boosting) is one of the most widely used boosting algorithm. However Adaboost and Gradient Boosting are also used.



BAGGING – RANDOM FOREST

- Bagging is the short for Bootstrap Aggregation.
- Random Forest is an extension to the decision tree algorithm.
- Constructing multiple decision trees creates the forest.
- Step 1: Create various bootstrapped datasets from the existing training data.
- Step 2: Create decision tree with the randomly selected bootstrapped data.
- Step 3: Repeat step 1 for "n" number of times and create "n" decision trees
- Step 4: Calculate the performance metrics for the dataset.



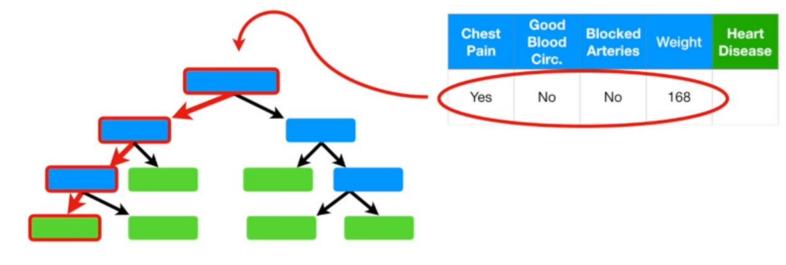


...and now we want to know if they have heart disease or not.



Image courtesy: Random Forest by Statquest

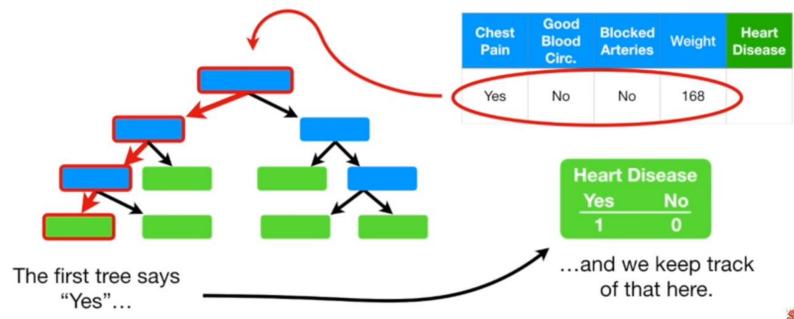




The first tree says "Yes"...

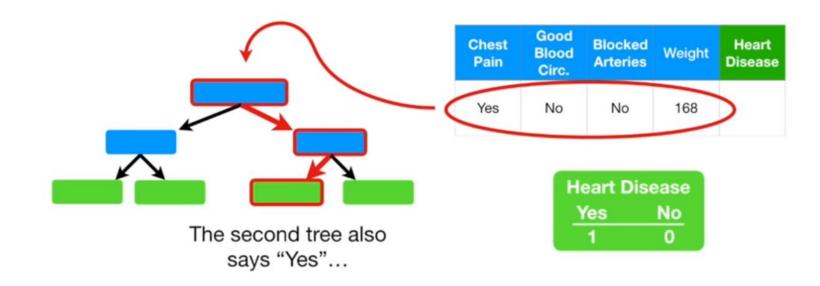






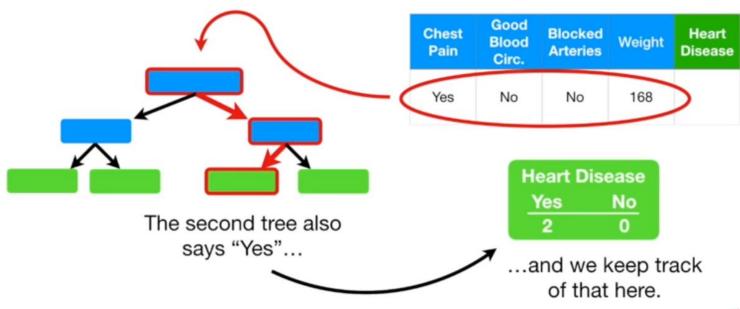






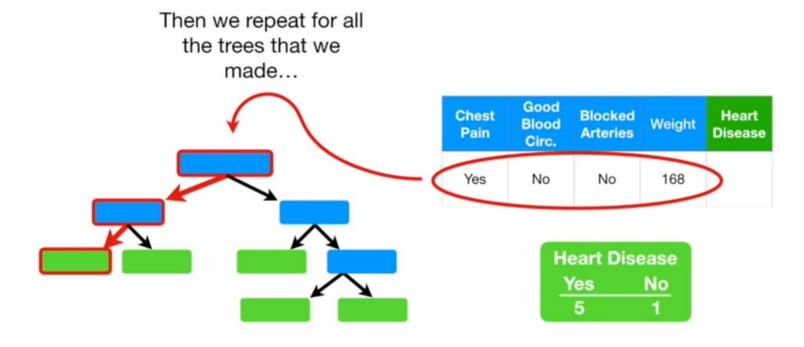


















After running the data down all of the trees in the random forest, we see which option received more votes.







SUMMARY

- More the number of trees, more will be the accuracy. Having exceedingly higher number of trees can lead to overfitting.
- For every stage of boostrapped data, we will have few data left behind. They are called as Out-of-bag samples (OOB).
- The trees generated will be functioning in parallel.

