CLASSIFICATION II -BIS BEYOND DECISION TREE

Dr. Lablanche Pierre-Yves African Institute for Mathematical Sciences

REMINDER

Building a full (until complete purity) single decision tree has several drawbacks. Constraining the tree construction (complexity) allows to:

- limit the resources used
- Reduce the overfitting issue (does not suppress it)

But does not solve the local optimal vs global optimal problem!

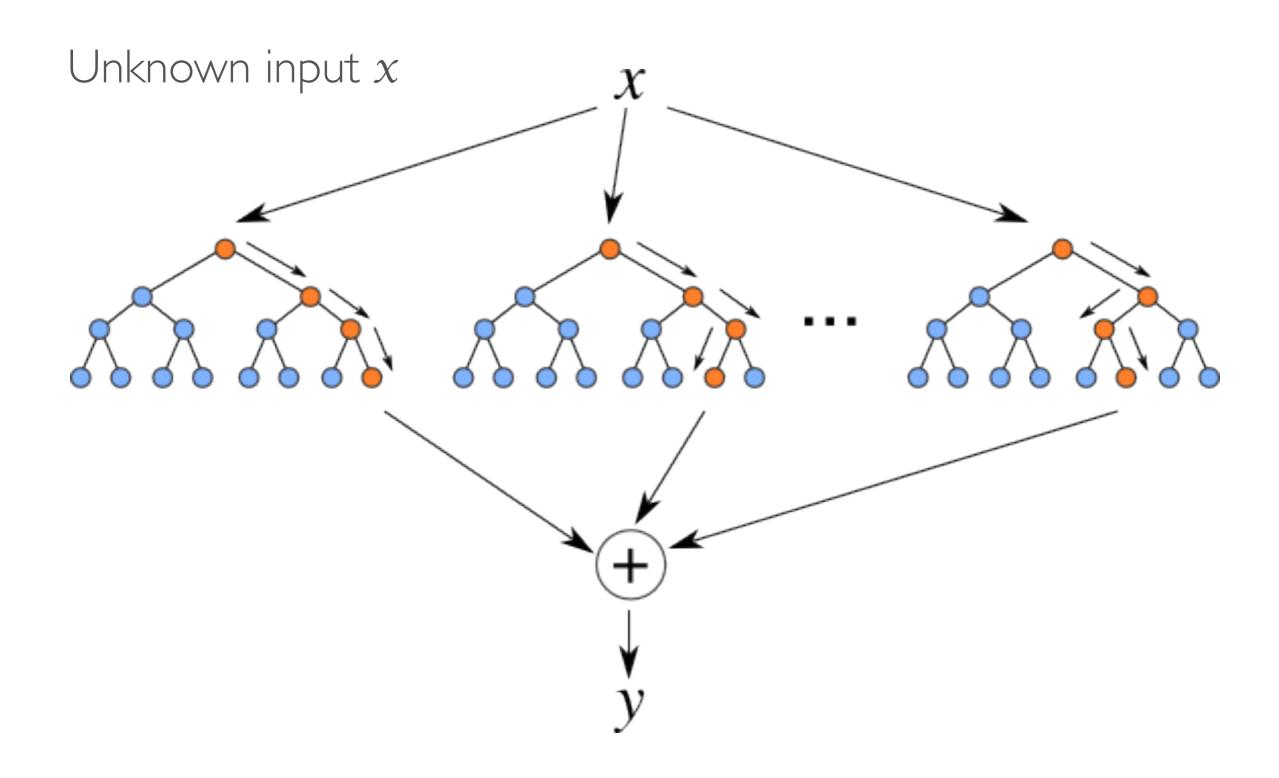
Solution: Ensemble Method

BAGGING

Bagging (also called **B**ootstrap **Agg**regat**ing**) in three simple steps : from data set \mathcal{D} size n

- Chose m new sub-training set of size n' from the original training set. (using sampling with replacement)
- Build a decision tree for each sub-set (*m* trees total)
- Combine each tree prediction by voting (averaging for regression)

BAGGING



FROM TREE TO FOREST

Ensemble = more than one

More than one tree = Forest

Thus:

Bootstrap Aggregation + Decision Tree= **Decision Forest**

Where does the "Random" come from?

CLEVER BAGGING

Bagging does not prevent decision trees from being correlated.

If decision trees are correlated, bagging is just a waste of resource.

Very complicated to predict how similar the trees will be.

Solution: Force Randomness at decision node

RANDOM FOREST

RANDOM FOREST

For each tree, limit the number of features to consider for the bet split at a decision node - Random draw of k' features out of original k features.

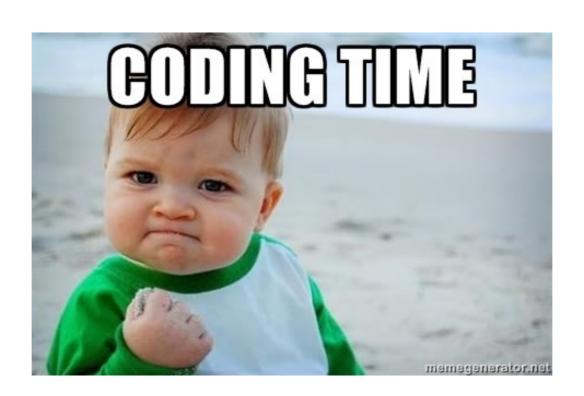
Best split threshold still computed by the machine.

Finally:

Bagging + Random Feature Selection Decision Tree= Random Forest

CODINGTIME (I)

- Use RandomTreeClassifier to classify the iris dataset.
- Play with the different parameters!



MORE RANDOMNESS !!!

Possible to push randomisation a step further.

Randomly choose k' features AND threshold by feature

Keep the best randomly generated feature/threshold pair

Bagging + Random Feature and Threshold Decision Tree = Extra (Randomised) Trees

...does not necessarily perform better than Random Forests.

CODINGTIME (2)

- Use ExtraTreesClassifier to classify the iris dataset.
- Recall DecisionTreeClassifier() and
 RandomTreeClassifier() to classify the iris dataset.
- Compare the three classifiers!

