

## RANDOM FORESTS IN STATA

For anybody keen to try some out-of-sample prediction models in Stata, there is a great package for running a random forest algorithm in Stata called `rforest`.

Random forest models are decision tree based, so are great for contexts where you don't think your data can be modelled as having a linear relationship.

While decision trees are unlikely to perform well in terms of out of sample prediction because they will over-fit the data, random forest models average predictions over many individual trees and use bootstrap aggregating to reduce overfitting, yielding better prediction accuracy.

The Stata syntax for `rforest` is:

```
rforest depvar indepvars [if] [in] , [ options ]  
predict newvar | varlist | stub* [if] [in] , [ pr ]
```

It can be used for either classification (categorical outcome variables) or regression (continuous outcome variables). You first need to split your sample into a training and a testing sample to use this comment. If you want to use a training dataset with a binary dependent variable to classify the observations in a new dataset into the same categories, you can use the option `type(class)` at the end.

Say for example you were trying to predict whether a village would experience flooding in a given year and had a set of appropriate explanatory variables. You could use the code:

```
rforest flood indepvars if dataset==training, type(class)  
predict flood if dataset==testing
```

If instead you are trying to predict a continuous variable, use the option `type(regression)`.

You can set the number of trees it will iterate over using the option `iterations(x)`. If you don't specify, the default will be 100. Sometimes it will be a trade-off between prediction accuracy and model running time – if you have a very large sample, `rforest` can take a while to run. You can also adjust the minimum leaf size, the depth of the tree and the number of variables to randomly investigate.