# Regression and Classification Trees

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## Regression trees

As an example, we use household-level private label share data aggregated at the year level.

```
load("./Data/PL_shares_annual.RData")
```

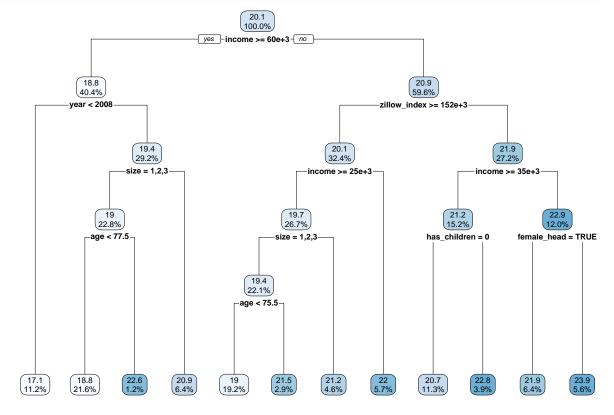
Convert all categorical variables to factors. The rpart package will then understand that these variables are categorical.

```
categorical_vars = c("unemployed", "has_children", "female_head", "size")
PL_shares_DT[, (categorical_vars) := lapply(.SD, as.factor), .SDcols = categorical_vars]
```

Now build a regression tree (method = "anova").

The cost-complexity parameter cp is key: If the algorithm does not split, make the parameter smaller. minsplit determines the minimum number of observations in a node such that a split is attempted, maxdepth determines the depth (length) of a tree.

```
N_terminal_nodes = length(unique(tree$where))
rpart.plot(tree, tweak = 0.8, digits = 3)
```



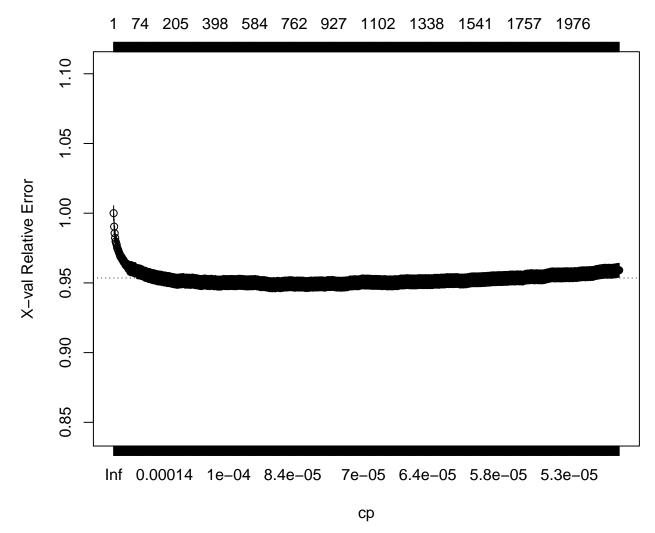
#### Cost-complexity pruning

Grow a large tree:

Visualize the cross-validation results:

```
plotcp(tree)
```

## size of tree



Find the optimal complexity parameter:

```
index_optimal = which.min(tree$cptable[, "xerror"])
cp_optimal = tree$cptable[index_optimal, "CP"]
size_optimal = tree$cptable[index_optimal, "nsplit"] + 1
```

Prune the tree:

```
pruned_tree = prune(tree, cp = cp_optimal)
# rpart.plot(pruned_tree)
```

## Classification trees

Use the RFM data to predict if a customer buys.

```
load("./Data/RFM-Data.RData")
```

Use the method = "class" option for classification.

