## 14 Adaptive Resampling

Models can benefit significantly from tuning but the optimal values are rarely known beforehand. train can be used to define a grid of possible points and resampling can be used to generate good estimates of performance for each tuning parameter combination. However, in the nominal resampling process, all the tuning parameter combinations are computed for all the resamples before a choice is made about which parameters are good and which are poor.

caret contains the ability to adaptively resample the tuning parameter grid in a way that concentrates on values that are the in the neighborhood of the optimal settings. See this paper for the details.

To illustrate, we will use the Sonar data from one of the previous pages.

```
library(mlbench)

data(Sonar)

library(caret)

set.seed(998)

inTraining <- createDataPartition(Sonar$Class, p = .75, list = FAI

training <- Sonar[ inTraining,]

testing <- Sonar[-inTraining,]</pre>
```

We will tune a support vector machine model using the same tuning strategy as before but with random search:

Using this method, the optimal tuning parameters were a RBF kernel parameter of 0.0174 and a cost value of 459.4251397. To use the adaptive procedure, the trainControl option needs some additional arguments:

- min is the minimum number of resamples that will be used for each tuning parameter. The default value is 5 and increasing it will decrease the speed-up generated by adaptive resampling but should also increase the likelihood of finding a good model.
- alpha is a confidence level that is used to remove parameter settings. To date, this value has not shown much of an effect.

- method is either "gls" for a linear model or "BT" for a
  Bradley-Terry model. The latter may be more useful when you
  expect the model to do very well (e.g. an area under the ROC
  curve near 1) or when there are a large number of tuning
  parameter settings.
- complete is a logical value that specifies whether train should generate the <u>full resampling set</u> if it finds an optimal solution before the end of resampling. If you want to know the optimal parameter settings and don't care much for the estimated performance value, a value of FALSE would be appropriate here.

The new code is below. Recall that setting the random number seed just prior to the model fit will ensure the same resamples as well as the same random grid.

```
adaptControl <- trainControl(method = "adaptive cv",</pre>
                              number = 10, repeats = 10,
                              adaptive = list(min = 5, alpha = 0.0
                                               method = "gls", comp!
                              classProbs = TRUE,
                              summaryFunction = twoClassSummary,
                              search = "random")
set.seed(825)
svmAdapt <- train(Class ~ ., data = training,</pre>
                   method = "svmRadial",
                   trControl = adaptControl,
                   preProc = c("center", "scale"),
                   metric = "ROC",
                   tuneLength = 15)
```

The search finalized the tuning parameters on the 13th iteration of resampling and was 2.1-fold faster than the original analysis. Here, the optimal tuning parameters were a RBF kernel parameter of 0.0158 and a cost value of 20.3636946. These are close to the previous settings and result in a difference in the area under the ROC curve of 0.002 and the adaptive approach used 1308 fewer models.

Remember that this methodology is experimental, so please send any questions or bug reports to the package maintainer.