Bagging

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Bootstrap aggregating (bagging)

Basic idea:

- 1. Resample cases and recalculate predictions
- 2. Average or majority vote

Notes:

- Similar bias
- Reduced variance
- More useful for non-linear functions

Ozone data

```
library(ElemStatLearn); data(ozone,package="ElemStatLearn"]
ozone <- ozone[order(ozone$ozone),]
head(ozone)</pre>
```

##		ozone	radiation	temperature	wind
##	17	1	8	59	9.7
##	19	4	25	61	9.7
##	14	6	78	57	18.4
##	45	7	48	80	14.3
##	106	7	49	69	10.3
##	7	8	19	61	20.1

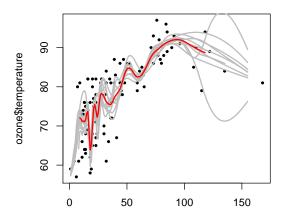
http://en.wikipedia.org/wiki/Bootstrap_aggregating

Bagged loess

```
11 <- matrix(NA,nrow=10,ncol=155)
for(i in 1:10){
   ss <- sample(1:dim(ozone)[1],replace=T)
   ozone0 <- ozone[ss,]; ozone0 <- ozone0[order(ozone0$ozone0
   loess0 <- loess(temperature ~ ozone,data=ozone0,span=0.20
   ll[i,] <- predict(loess0,newdata=data.frame(ozone=1:155))
}</pre>
```

Bagged loess

```
plot(ozone$ozone,ozone$temperature,pch=19,cex=0.5)
for(i in 1:10){lines(1:155,ll[i,],col="grey",lwd=2)}
lines(1:155,apply(ll,2,mean),col="red",lwd=2)
```



Bagging in caret

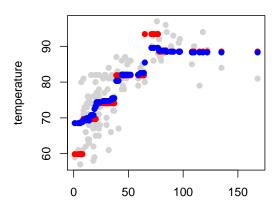
- Some models perform bagging for you, in train function consider method options
- bagEarth
- ▶ treebag
- ▶ bagFDA
- Alternatively you can bag any model you choose using the bag function

More bagging in caret

```
library(caret)
## Loading required package: lattice
## Loading required package: ggplot2
predictors = data.frame(ozone=ozone$ozone)
temperature = ozone$temperature
treebag <- bag(predictors, temperature, B = 10,
                bagControl = bagControl(fit = ctreeBag$fit
                                        predict = ctreeBags
                                        aggregate = ctreeBa
## Warning: executing %dopar% sequentially: no parallel back
http:
//www.inside-r.org/packages/cran/caret/docs/nbBag
```

Example of custom bagging (continued)

```
plot(ozone$ozone,temperature,col='lightgrey',pch=19)
points(ozone$ozone,predict(treebag$fits[[1]]$fit,predictors
points(ozone$ozone,predict(treebag,predictors),pch=19,col='
```



Parts of bagging

ctreeBag\$fit

```
## function (x, y, ...)
## {
## loadNamespace("party")
## data <- as.data.frame(x)
## data$y <- y
## party::ctree(y ~ ., data = data)
## }
## <environment: namespace:caret>
```

Parts of bagging

ctreeBag\$pred

```
## function (object, x)
## {
        if (!is.data.frame(x))
##
##
            x \leftarrow as.data.frame(x)
        obsLevels <- levels(object@data@get("response")[, 1]
##
##
        if (!is.null(obsLevels)) {
##
            rawProbs <- party::treeresponse(object, x)</pre>
            probMatrix <- matrix(unlist(rawProbs), ncol = le</pre>
##
##
                 byrow = TRUE)
##
            out <- data.frame(probMatrix)</pre>
##
            colnames(out) <- obsLevels
            rownames(out) <- NULL
##
        }
##
##
        else out <- unlist(party::treeresponse(object, x))</pre>
##
        out
                                         4□ > 4□ > 4□ > 4 = > 4 = > 9 < 0</p>
## }
```

Parts of bagging

ctreeBag\$aggregate

```
## function (x, type = "class")
## {
       if (is.matrix(x[[1]]) \mid is.data.frame(x[[1]]))  {
##
            pooled \leftarrow x[[1]] & NA
##
##
            classes <- colnames(pooled)</pre>
##
            for (i in 1:ncol(pooled)) {
##
                tmp <- lapply(x, function(y, col) y[, col],</pre>
                tmp <- do.call("rbind", tmp)</pre>
##
##
                pooled[, i] <- apply(tmp, 2, median)</pre>
            }
##
##
            if (type == "class") {
                out <- factor(classes[apply(pooled, 1, which
##
##
                     levels = classes)
##
##
            else out <- as.data.frame(pooled)
                                            →□→→=→=→== 990
##
```

Notes and further resources

Notes:

- Bagging is most useful for nonlinear models
- Often used with trees an extension is random forests
- Several models use bagging in caret's train function

Further resources:

- Bagging
- Bagging and boosting
- Elements of Statistical Learning