Preprocessing

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Why preprocess?

```
library(caret); library(kernlab); data(spam)
## Loading required package: lattice
## Loading required package: ggplot2
##
## Attaching package: 'kernlab'
## The following object is masked from 'package:ggplot2':
##
       alpha
##
inTrain <- createDataPartition(y=spam$type,</pre>
                                p=0.75, list=FALSE)
training <- spam[inTrain,]</pre>
testing <- spam[-inTrain,]</pre>
```

hist(training\$capitalAve,main="",xlab="ave. capital run

Why preprocess?

```
mean(training$capitalAve)

## [1] 4.910247

sd(training$capitalAve)

## [1] 28.30757
```

Standardizing

```
trainCapAve <- training$capitalAve</pre>
trainCapAveS <- (trainCapAve - mean(trainCapAve))/sd(train</pre>
mean(trainCapAveS)
## [1] -1.150422e-17
sd(trainCapAveS)
## [1] 1
```

Standardizing - test set

```
testCapAve <- testing$capitalAve</pre>
testCapAveS <- (testCapAve - mean(trainCapAve))/sd(trainCapAve)</pre>
mean(testCapAveS)
## [1] 0.03975327
sd(testCapAveS)
## [1] 1.423405
```

Standardizing - preProcess function

[1] 1

```
preObj <- preProcess(training[,-58],method=c("center","scale
trainCapAveS <- predict(preObj,training[,-58])$capitalAve
mean(trainCapAveS)

## [1] -1.150422e-17

sd(trainCapAveS)</pre>
```

Standardizing - preProcess function

```
testCapAveS <- predict(preObj,testing[,-58])$capitalAve
mean(testCapAveS)</pre>
```

```
## [1] 0.03975327
```

```
sd(testCapAveS)
```

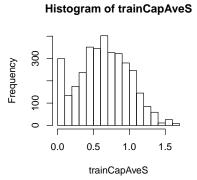
[1] 1.423405

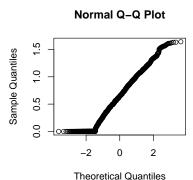
Standardizing - preProcess argument

```
set.seed(32343)
modelFit <- train(type ~.,data=training,
                 preProcess=c("center", "scale"), method="g"
## Warning: glm.fit: fitted probabilities numerically 0 or
```

Standardizing - Box-Cox transforms

preObj <- preProcess(training[,-58],method=c("BoxCox"))
trainCapAveS <- predict(preObj,training[,-58])\$capitalAve
par(mfrow=c(1,2)); hist(trainCapAveS); qqnorm(trainCapAveS)</pre>





Standardizing - Imputing data

```
set.seed(13343)
# Make some values NA
training$capAve <- training$capitalAve
selectNA <- rbinom(dim(training)[1],size=1,prob=0.05)==1</pre>
training$capAve[selectNA] <- NA
# Impute and standardize
preObj <- preProcess(training[,-58],method="knnImpute")</pre>
capAve <- predict(preObj,training[,-58])$capAve
# Standardize true values
capAveTruth <- training$capitalAve</pre>
capAveTruth <- (capAveTruth-mean(capAveTruth))/sd(capAveTru</pre>
```

Standardizing - Imputing data

```
quantile(capAve - capAveTruth)
                                               75%
##
            0% 25%
                            50%
## -2.816863e+00 -1.317386e-03 -1.415817e-05 5.857060e-04
quantile((capAve - capAveTruth)[selectNA])
                                               75%
##
            0% 25%
                                   50%
## -2.8168632080 -0.0169680286 0.0009679843 0.0218804267
quantile((capAve - capAveTruth)[!selectNA])
##
            0% 25%
                                   50%
                                               75%
```

-9.489725e-01 -1.232646e-03 -2.019562e-05 5.464626e-04

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Notes and further reading

- Training and test must be processed in the same way
- Test transformations will likely be imperfect
- Especially if the test/training sets collected at different times
- Careful when transforming factor variables!
- preprocessing with caret