PCA and LDA

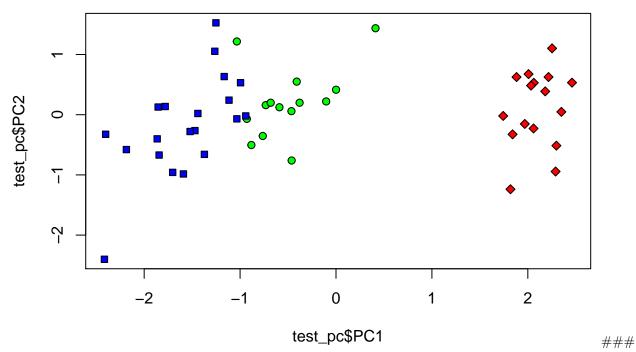
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Run PCA on the iris data

```
library(caret)
## Warning: package 'caret' was built under R version 3.4.3
## Loading required package: lattice
## Loading required package: ggplot2
data(iris)
i <- sample(1:150, 100, replace=FALSE)</pre>
train <- iris[i,]</pre>
test <- iris[-i,]</pre>
set.seed(1234)
pca_out <- preProcess(train[,1:4], method=c("center", "scale", "pca"))</pre>
pca_out
## Created from 100 samples and 4 variables
##
## Pre-processing:
## - centered (4)
   - ignored (0)
   - principal component signal extraction (4)
##
##
     - scaled (4)
##
## PCA needed 2 components to capture 95 percent of the variance
```

PCA plot

```
train_pc <- predict(pca_out, train[, 1:4])
test_pc <- predict(pca_out, test[,])
plot(test_pc$PC1, test_pc$PC2, pch=c(23,21,22)[unclass(test_pc$Species)], bg=c("red","green","blue")[unclass(test_pc$Species)]</pre>
```



PCA data in knn

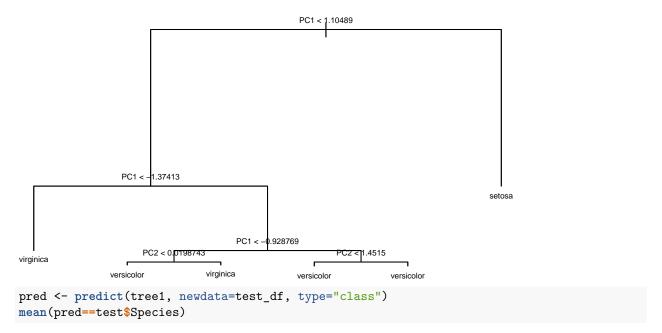
Now let's see if our two principal components can predict class.

```
train_df <- data.frame(train_pc$PC1, train_pc$PC2, train$Species)
test_df <- data.frame(test_pc$PC1, test_pc$PC2, test$Species)
library(class)
set.seed(1234)
pred <- knn(train=train_df[,1:2], test=test_df[,1:2], cl=train_df[,3], k=3)
mean(pred==test$Species)</pre>
```

[1] 0.9

The accuracy is .94 which is lower than if we used all 4 predictors.

```
library(tree)
colnames(train_df) <- c("PC1", "PC2", "Species")
colnames(test_df) <- c("PC1", "PC2", "Species")
set.seed(1234)
tree1 <- tree(Species~., data=train_df)
plot(tree1)
text(tree1, cex=0.5, pretty=0)</pre>
```



[1] 0.92

With the decison tree we got a little lower accuracy.

LDA

```
library(MASS)
lda1 <- lda(Species~., data=train)</pre>
lda1$means
              Sepal.Length Sepal.Width Petal.Length Petal.Width
##
                   5.044118
                               3.491176
                                             1.432353
## setosa
                                                         0.2470588
## versicolor
                   5.930556
                               2.713889
                                             4.247222
                                                         1.2972222
## virginica
                   6.683333
                               2.986667
                                             5.653333
                                                         2.0600000
```

predict on test

```
lda_pred <- predict(lda1, newdata=test, type="class")</pre>
lda_pred$class
   [1] setosa
                  setosa
                             setosa
                                        setosa
                                                   setosa
                                                              setosa
  [7] setosa
                  setosa
                             setosa
                                        setosa
                                                   setosa
                                                              setosa
## [13] setosa
                  setosa
                             setosa
                                        setosa
                                                   versicolor versicolor
## [19] versicolor versicolor versicolor versicolor versicolor virginica
## [25] versicolor versicolor versicolor versicolor versicolor
## [31] virginica virginica virginica virginica virginica virginica
## [37] virginica virginica virginica virginica virginica virginica
## [43] virginica versicolor virginica virginica virginica virginica
## [49] virginica virginica
## Levels: setosa versicolor virginica
mean(lda_pred$class==test$Species)
```

plot

plot(lda_pred\$x[,1], lda_pred\$x[,2], pch=c(23,21,22)[unclass(lda_pred\$class)], bg=c("red", "green", "blue

