## R In-Class Lab: Principal Component Analysis with mtcars

## Using PCA to Reduce Dimensionality

Using the functions discussed above, we will now show how PCA can be used to reduce the dimension of the mtcars dataset, which is a readily available dataset in base R. See https://www.datacamp.com/community/t utorials/pca-analysis-r for more information about the mtcars dataset.

```
head(mtcars)
```

```
qsec vs am gear carb
##
                       mpg cyl disp hp drat
                                                 wt
## Mazda RX4
                                160 110 3.90 2.620 16.46
## Mazda RX4 Wag
                             6
                                160 110 3.90 2.875 17.02
                                                            0
                                                                    4
                                                                          4
                      21.0
                                     93 3.85 2.320 18.61
                                                                          1
## Datsun 710
                      22.8
## Hornet 4 Drive
                      21.4
                             6
                                258 110 3.08 3.215 19.44
                                                                    3
                                                                          1
                                                                    3
                                                                          2
## Hornet Sportabout 18.7
                             8
                                360 175 3.15 3.440 17.02
## Valiant
                      18.1
                             6
                                225 105 2.76 3.460 20.22
                                                                    3
                                                                          1
```

There are two categorical features, namely 'vs' and 'am', which will be excluded, since PCA should only be used on numerical features. Also, we partition the mtcars dataset into two separate datasets, so we can have some "new" data to show how new observations are handled.

```
mtcars <- mtcars[, c(1:7, 10:11)]
indices <- sample(1:32, size = 27)

mtcars_newdata <- mtcars[-indices, ]
mtcars <- mtcars[indices, ]</pre>
```

Let's try reducing the 9 numerical features to a smaller set using the principal components. First, as above, we use the prcomp function to calculate the principal components. Here, we both center and scale the features, since they are measured in different units.

```
pca_mtcars <- prcomp(mtcars, center = TRUE, scale = TRUE)
summary(pca_mtcars)</pre>
```

```
## Importance of components:
                                    PC2
                                             PC3
                                                     PC4
                                                             PC5
                                                                     PC6
                                                                              PC7
                             PC1
                          2.3817 1.4679 0.67436 0.52508 0.38499 0.35156 0.30408
## Standard deviation
## Proportion of Variance 0.6303 0.2394 0.05053 0.03063 0.01647 0.01373 0.01027
                          0.6303 0.8697 0.92024 0.95088 0.96734 0.98108 0.99135
## Cumulative Proportion
##
                              PC8
## Standard deviation
                          0.24809 0.12766
## Proportion of Variance 0.00684 0.00181
## Cumulative Proportion 0.99819 1.00000
```

From these results, we see that the first 3 principal components explain approximately 92% of the variation in the mtcars dataset. Thus, we can try reducing the 9 original features by projecting onto the first 3 principal components. To do this, we first center and scale the mtcars dataset, since we applied both of these actions when we calculated the principal components. Then, we rotate the data, which gives us the projection of the (centered and scaled) mtcars dataset onto the 9 principal components. Only the first 3 resulting columns are kept, which correspond to the projection onto the first 3 principal components.

```
PC1
                                        PC2
                                                   PC3
##
## Datsun 710
                      -2.316827 -0.1740157
                                            0.2337354
                       1.476906 2.9695487 0.2222096
## Ford Pantera L
                      -3.440017 -0.2035849 0.1191914
## Fiat 128
## Merc 230
                      -2.291891 -1.1093910 -1.8467126
## Lincoln Continental 3.580864 -0.9567075 -0.8468006
                       2.468255 0.7046585 0.3867386
## Camaro Z28
```

Now, we have 3 new features called PC1, PC2, and PC3, since they are the projections of the data onto the first 3 principal components. To handle new observations, we perform the same steps as above:

It is important to notice that we use the center and scale, as well as rotation, based on the mtcars dataset and not the new data!