class 08

Breast Cancer Project

```
wisc.data <- read.csv("WisconsinCancer.csv", row.names = 1)
head(wisc.data)</pre>
```

| | diagnosis | radius_mean | texture_mean | perimeter_mean | area_mean | | |
|--|------------|--------------|----------------|-----------------|------------|----------|--|
| 842302 | M | 17.99 | 10.38 | 122.80 | 1001.0 | | |
| 842517 | M | 20.57 | 17.77 | 132.90 | 1326.0 | | |
| 84300903 | M | 19.69 | 21.25 | 130.00 | 1203.0 | | |
| 84348301 | M | 11.42 | 20.38 | 77.58 | 386.1 | | |
| 84358402 | M | 20.29 | 14.34 | 135.10 | 1297.0 | | |
| 843786 | M | 12.45 | 15.70 | 82.57 | 477.1 | | |
| | smoothness | s_mean compa | ctness_mean co | ncavity_mean co | oncave.poi | nts_mean | |
| 842302 | 0 | .11840 | 0.27760 | 0.3001 | | 0.14710 | |
| 842517 | 0 | .08474 | 0.07864 | 0.0869 | | 0.07017 | |
| 84300903 | 0 | .10960 | 0.15990 | 0.1974 | | 0.12790 | |
| 84348301 | 0 | .14250 | 0.28390 | 0.2414 | | 0.10520 | |
| 84358402 | 0 | .10030 | 0.13280 | 0.1980 | | 0.10430 | |
| 843786 | 0 | .12780 | 0.17000 | 0.1578 | | 0.08089 | |
| symmetry_mean fractal_dimension_mean radius_se texture_se perimeter_se | | | | | | | |
| 842302 | 0.5 | 2419 | 0.0787 | 1 1.0950 | 0.9053 | 8.589 | |
| 842517 | 0.3 | 1812 | 0.0566 | 7 0.5435 | 0.7339 | 3.398 | |
| 84300903 | 0.5 | 2069 | 0.0599 | 9 0.7456 | 0.7869 | 4.585 | |
| 84348301 | 0.5 | 2597 | 0.0974 | 4 0.4956 | 1.1560 | 3.445 | |
| 84358402 | 0.3 | 1809 | 0.0588 | 3 0.7572 | 0.7813 | 5.438 | |
| 843786 | 0.5 | 2087 | 0.0761 | 3 0.3345 | 0.8902 | 2.217 | |
| | area_se si | moothness_se | compactness_s | e concavity_se | concave.p | oints_se | |
| 842302 | 153.40 | 0.006399 | 0.0490 | 4 0.05373 | | 0.01587 | |
| 842517 | 74.08 | 0.005225 | 0.0130 | 8 0.01860 | | 0.01340 | |
| 84300903 | 94.03 | 0.006150 | 0.0400 | 6 0.03832 | | 0.02058 | |

| 84348301 | 27.23 | 0.009110 | 0.07458 | 0.05661 | 0.01867 | | | |
|---|---------------|---------------|------------|------------------|------------|--|--|--|
| 84358402 | 94.44 | 0.011490 | 0.02461 | 0.05688 | 0.01885 | | | |
| 843786 | 27.19 | 0.007510 | 0.03345 | 0.03672 | 0.01137 | | | |
| symmetry_se fractal_dimension_se radius_worst texture_worst | | | | | | | | |
| 842302 | 0.03003 | 0.0 | 006193 | 25.38 | 17.33 | | | |
| 842517 | 0.01389 | 0.0 | 003532 | 24.99 | 23.41 | | | |
| 84300903 | 0.02250 | 0.0 | 004571 | 23.57 | 25.53 | | | |
| 84348301 | 0.05963 | 0.0 | 009208 | 14.91 | 26.50 | | | |
| 84358402 | 0.01756 | 0.0 | 005115 | 22.54 | 16.67 | | | |
| 843786 | 0.02165 | 0.0 | 005082 | 15.47 | 23.75 | | | |
| | perimeter_wor | st area_worst | smoothness | s_worst compact: | ness_worst | | | |
| 842302 | 184. | 60 2019.0 | | 0.1622 | 0.6656 | | | |
| 842517 | 158. | 80 1956.0 | | 0.1238 | 0.1866 | | | |
| 84300903 | 152. | 50 1709.0 | | 0.1444 | 0.4245 | | | |
| 84348301 | 98. | 87 567.7 | | 0.2098 | 0.8663 | | | |
| 84358402 | 152. | 20 1575.0 | | 0.1374 | 0.2050 | | | |
| 843786 | 103. | 40 741.6 | | 0.1791 | 0.5249 | | | |
| | concavity_wor | st concave.po | ints_worst | symmetry_worst | | | | |
| 842302 | 0.71 | 19 | 0.2654 | 0.4601 | | | | |
| 842517 | 0.24 | 16 | 0.1860 | 0.2750 | | | | |
| 84300903 | 0.45 | 04 | 0.2430 | 0.3613 | | | | |
| 84348301 | 0.68 | 869 | 0.2575 | 0.6638 | | | | |
| 84358402 | 0.40 | 000 | 0.1625 | 0.2364 | | | | |
| 843786 | 0.53 | 555 | 0.1741 | 0.3985 | | | | |
| | fractal_dimen | sion_worst | | | | | | |
| 842302 | | 0.11890 | | | | | | |
| 842517 | | 0.08902 | | | | | | |
| 84300903 | | 0.08758 | | | | | | |
| 84348301 | | 0.17300 | | | | | | |
| 84358402 | | 0.07678 | | | | | | |
| 843786 | | 0.12440 | | | | | | |
| | | | | | | | | |

Q. How many patient samples are in this dataset?

nrow(wisc.data)

[1] 569

There are 569 patients in this dataset.

Q. How many cancer (M) and non cancer (B) samples are there?

```
table(wisc.data$diagnosis)
```

```
B M
357 212
```

Save the diagnosis for later use as a reference to compare how weel we do with PCA etc.

```
diagnosis <- as.factor(wisc.data$diagnosis)
#diagnosis</pre>
```

Now exclude the diagnosis column from the data

```
wisc <- wisc.data[,-1]

Q How many "dimensions", "variables", "columns" are there in this dataset?

ncol(wisc)
```

[1] 30

Principal Component Analysis (PCA)

To perform PCA in R we can use the prcomp() function, it takes as input a numeric dataset and optional scale=FALSE/TRUE argument

We generally always want to scale=TRUE but let's make sure by checking if the mean and standard deviation values are different across 30 columns

```
round( colMeans(wisc) )
```

| radius_mean 14 | texture_mean 19 | perimeter_mean 92 |
|-------------------------------------|----------------------------------|----------------------|
| area_mean 655 | ${\tt smoothness_mean} \\ 0$ | compactness_mean 0 |
| concavity_mean 0 | <pre>concave.points_mean 0</pre> | symmetry_mean O |
| <pre>fractal_dimension_mean 0</pre> | radius_se 0 | texture_se |

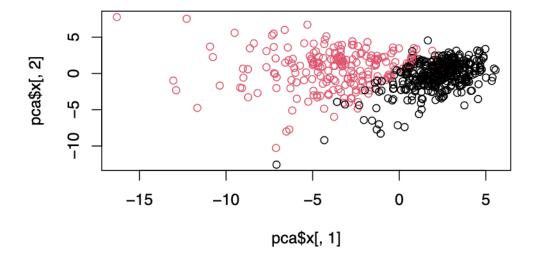
```
perimeter_se
                                                          smoothness_se
                                       area_se
                    3
                                            40
                                                                       0
      compactness_se
                                  concavity_se
                                                      concave.points_se
                                                                       0
                    0
         symmetry_se
                         fractal dimension se
                                                           radius worst
                    0
       texture worst
                              perimeter_worst
                                                             area worst
                   26
                                           107
                                                                     881
                                                        concavity_worst
    smoothness_worst
                            compactness_worst
concave.points_worst
                                symmetry_worst fractal_dimension_worst
                    0
                                             0
```

pca <- prcomp(wisc, scale=TRUE)
summary(pca)</pre>

Importance of components:

```
PC1
                                 PC2
                                         PC3
                                                 PC4
                                                          PC5
                                                                  PC6
                                                                          PC7
                       3.6444 2.3857 1.67867 1.40735 1.28403 1.09880 0.82172
Standard deviation
Proportion of Variance 0.4427 0.1897 0.09393 0.06602 0.05496 0.04025 0.02251
Cumulative Proportion 0.4427 0.6324 0.72636 0.79239 0.84734 0.88759 0.91010
                                         PC10
                           PC8
                                  PC9
                                                PC11
                                                        PC12
                                                                 PC13
                                                                         PC14
Standard deviation
                       0.69037 0.6457 0.59219 0.5421 0.51104 0.49128 0.39624
Proportion of Variance 0.01589 0.0139 0.01169 0.0098 0.00871 0.00805 0.00523
Cumulative Proportion 0.92598 0.9399 0.95157 0.9614 0.97007 0.97812 0.98335
                          PC15
                                  PC16
                                          PC17
                                                  PC18
                                                           PC19
Standard deviation
                       0.30681 0.28260 0.24372 0.22939 0.22244 0.17652 0.1731
Proportion of Variance 0.00314 0.00266 0.00198 0.00175 0.00165 0.00104 0.0010
Cumulative Proportion 0.98649 0.98915 0.99113 0.99288 0.99453 0.99557 0.9966
                          PC22
                                  PC23
                                         PC24
                                                 PC25
                                                          PC26
                                                                  PC27
                                                                          PC28
Standard deviation
                       0.16565 0.15602 0.1344 0.12442 0.09043 0.08307 0.03987
Proportion of Variance 0.00091 0.00081 0.0006 0.00052 0.00027 0.00023 0.00005
Cumulative Proportion 0.99749 0.99830 0.9989 0.99942 0.99969 0.99992 0.99997
                          PC29
                                  PC30
Standard deviation
                       0.02736 0.01153
Proportion of Variance 0.00002 0.00000
Cumulative Proportion 1.00000 1.00000
```

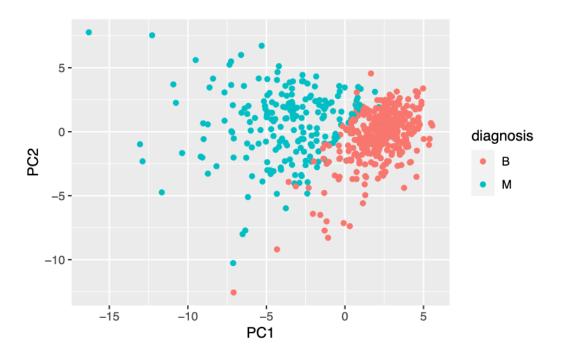
attributes(pca)



```
library(ggplot2)

x <- as.data.frame(pca$x)

ggplot(x) +
  aes(PC1, PC2, col=diagnosis) +
  geom_point()</pre>
```



Q. How much variance is captured in the top 3 PCs?

They capture 72.6% of the total variance.

Q9. For the first principal component, what is the component of the loading vector (i.e. wisc.pr\$rotation[,1]) for the feature concave.points_mean? This tells us how much this original feature contributes to the first PC.

```
pca$rotation["concave.points_mean", 1]
```

[1] -0.2608538

attributes(pca)

\$names

[1] "sdev" "rotation" "center" "scale" "x"

\$class

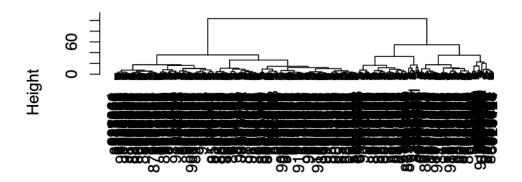
[1] "prcomp"

Combine PCA results with clustering

We can use our new PCA variables (i.e. the scores along the PCs contained in t pca\$x) as input for other methods such as clustering.

```
# Hclust needs a distance matrix as input
d <- dist( pca$x[,1:3] )
hc <- hclust(d, method="ward.D2")
plot(hc)</pre>
```

Cluster Dendrogram



d hclust (*, "ward.D2")

To get our cluster membership vector we can use the cutree() function and specify a height: (h) or number of groups (k).

```
grps <- cutree(hc, h=80)
table(grps)

grps
    1      2
203 366</pre>
```

I want to find out how many diagnosis "M" and "B" are in each group?

```
table(diagnosis)
```

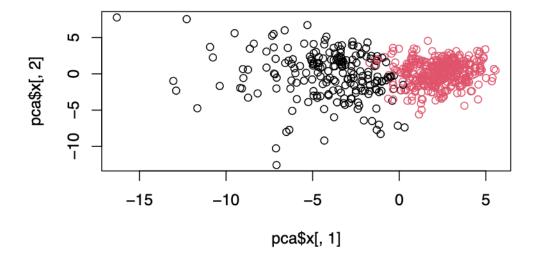
diagnosis B M 357 212

table(diagnosis, grps)

grps diagnosis 1 2 B 24 333 M 179 33

We can also plot our results using our clustering vector grps

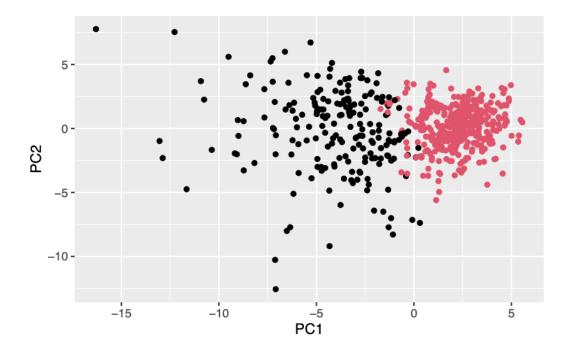
```
plot(pca$x[,1], pca$x[,2], col=grps)
```



```
library(ggplot2)

x <- as.data.frame(pca$x)

ggplot(x) +
   aes(PC1, PC2) +
   geom_point(col=grps)</pre>
```



Q15. What is the specificity and sensitivity of our current results?

```
#Sensitivity
179/(179+33)
```

[1] 0.8443396

#Specificity 333/(333+24)

[1] 0.9327731

Q16. Which of these new patients should we prioritize for follow up based on your results?

We should prioritize patient 2 because it is most similar to our previous results.