class 08

Breast Cancer Project

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

	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_mean	compactness_mean	concavity_mean	concave.points_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	0.2419	0.07871	1.0950	0.9053	8.589
842517	0.1812	0.05667	0.5435	0.7339	3.398
84300903	0.2069	0.05999	0.7456	0.7869	4.585
84348301	0.2597	0.09744	0.4956	1.1560	3.445
84358402	0.1809	0.05883	0.7572	0.7813	5.438
843786	0.2087	0.07613	0.3345	0.8902	2.217

	area_se	smoothness_se	compactness_se	concavity_se	concave.points_se
842302	153.40	0.006399	0.04904	0.05373	0.01587
842517	74.08	0.005225	0.01308	0.01860	0.01340
84300903	94.03	0.006150	0.04006	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	0.006193	25.38	17.33
842517	0.01389	0.003532	24.99	23.41
84300903	0.02250	0.004571	23.57	25.53
84348301	0.05963	0.009208	14.91	26.50
84358402	0.01756	0.005115	22.54	16.67
843786	0.02165	0.005082	15.47	23.75

	perimeter_worst	area_worst	smoothness_worst	compactness_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_worst	concave.points_worst	symmetry_worst
842302	0.7119	0.2654	0.4601
842517	0.2416	0.1860	0.2750
84300903	0.4504	0.2430	0.3613
84348301	0.6869	0.2575	0.6638
84358402	0.4000	0.1625	0.2364
843786	0.5355	0.1741	0.3985

	fractal_dimension_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?

```r
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.

```r
diagnosis <- as.factor(wisc.data$diagnosis)
#diagnosis
```

Now exclude the diagnosis column from the data

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

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

```r
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

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

radius_mean	texture_mean	perimeter_mean
14	19	92
area_mean	smoothness_mean	compactness_mean
655	0	0
concavity_mean	concave.points_mean	symmetry_mean
0	0	0
fractal_dimension_mean	radius_se	texture_se
0	0	1

perimeter_se	area_se	smoothness_se
3	40	0
compactness_se	concavity_se	concave.points_se
0	0	0
symmetry_se	fractal_dimension_se	radius_worst
0	0	16
texture_worst	perimeter_worst	area_worst
26	107	881
smoothness_worst	compactness_worst	concavity_worst
0	0	0
concave.points_worst	symmetry_worst	fractal_dimension_worst
0	0	0

```r
pca <- prcomp(wisc, scale=TRUE)
summary(pca)
```

```
Importance of components:
                          PC1     PC2     PC3     PC4     PC5     PC6     PC7
Standard deviation     3.6444  2.3857 1.67867 1.40735 1.28403 1.09880 0.82172
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
                          PC8     PC9    PC10    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    PC20   PC21
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
```

```r
attributes(pca)
```

```
$names
[1] "sdev"     "rotation" "center"   "scale"    "x"

$class
[1] "prcomp"
```
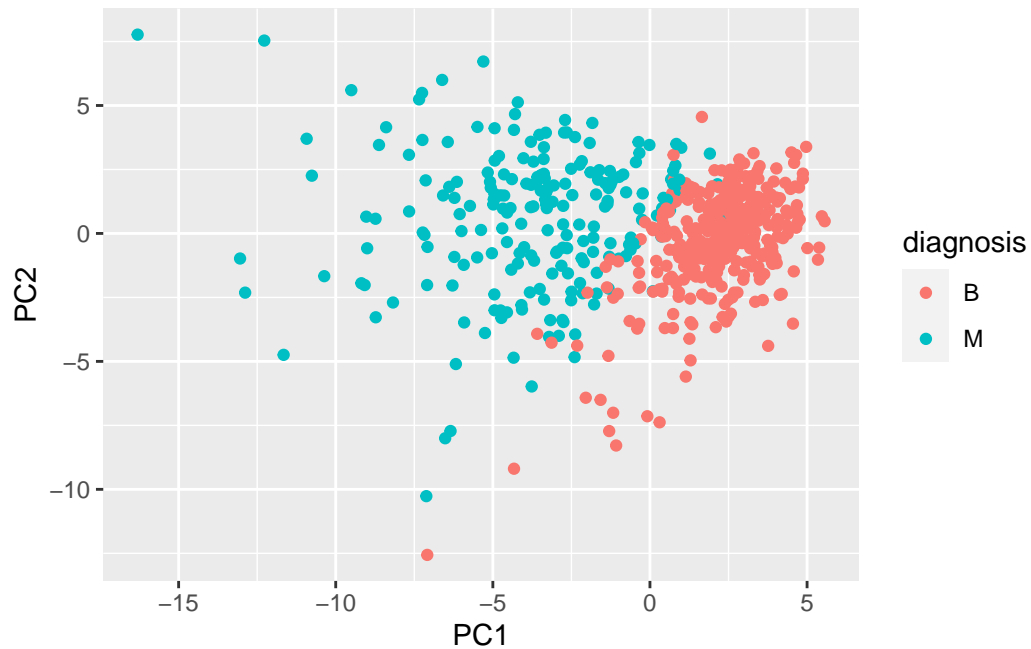
```r
plot(pca$x[,1], pca$x[,2], col=diagnosis)
```
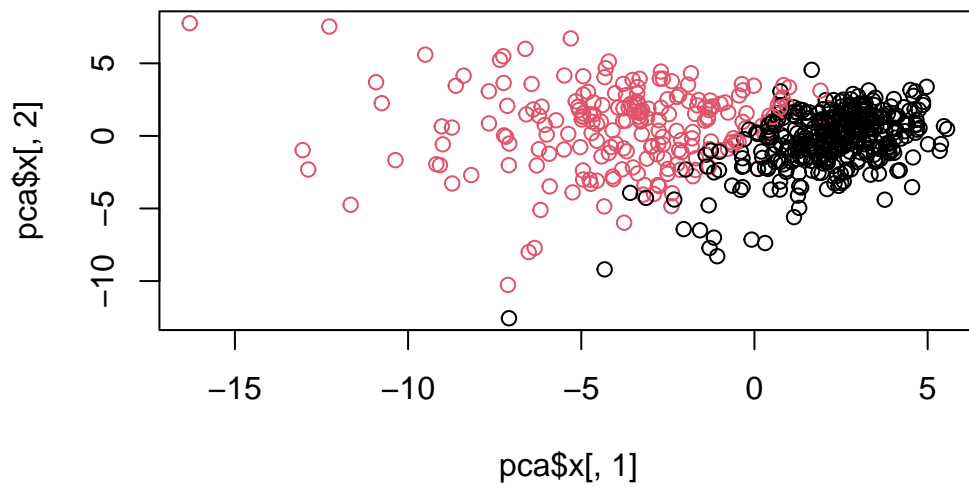


```r
library(ggplot2)

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

ggplot(x) +
  aes(PC1, PC2, col=diagnosis) +
  geom_point()
```

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)
```

Cluster Dendrogram
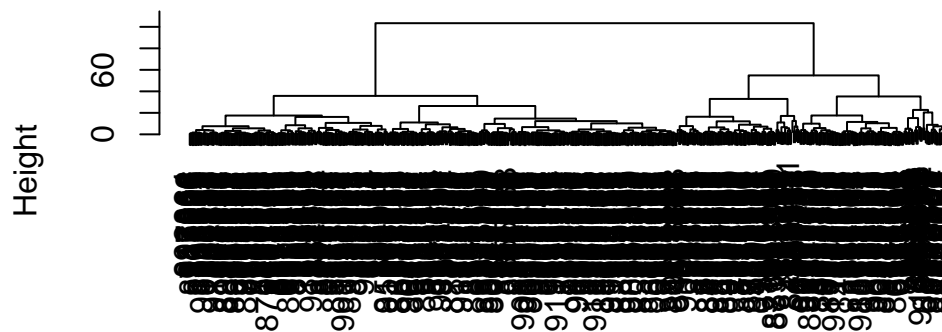
d
hclust (*, "ward.D2")

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
```

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
```

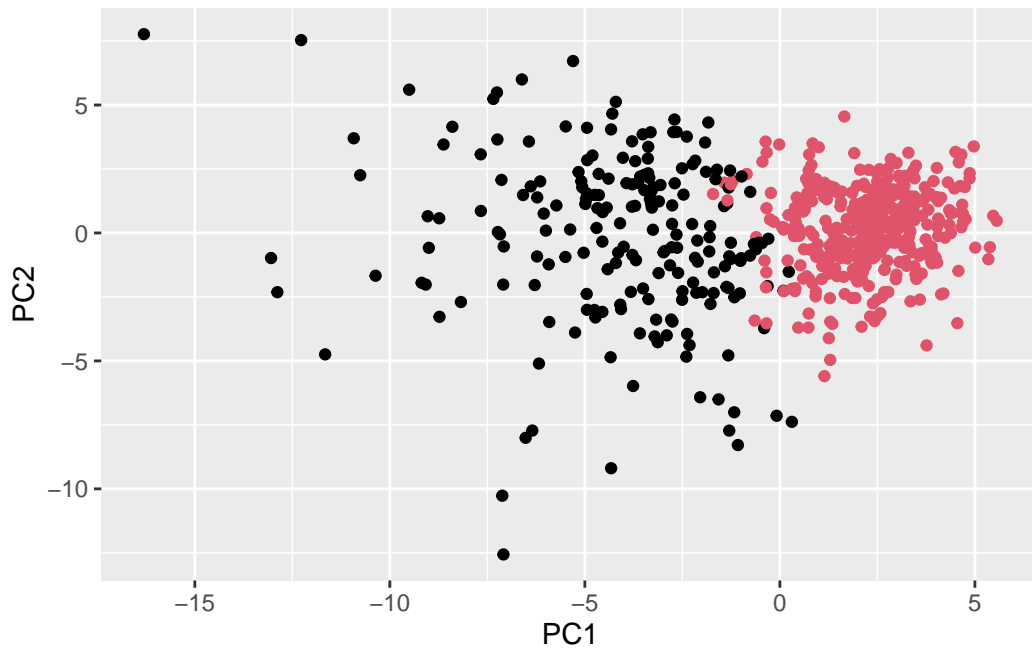We can also plot our results using our clustering vector `grps`

```
plot(pca$x[,1], pca$x[,2], col=grps)
```
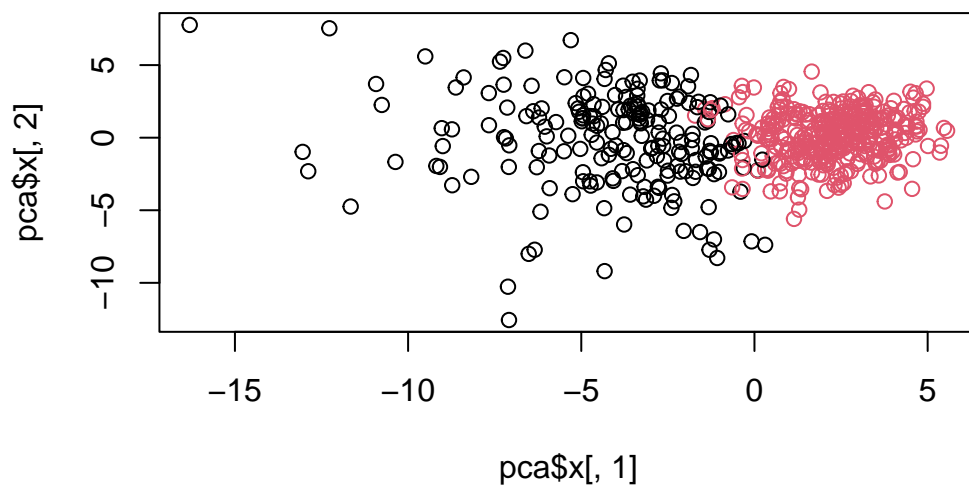
```
library(ggplot2)

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

ggplot(x) +
  aes(PC1, PC2) +
  geom_point(col=grps)
```


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.