

# Class10: MacineLearningProject\_take2

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## Get our input data

Our data for today come from the Wisconsin Breast Cancer Diagnostic Data Set

```
wisc.df <- read.csv("WisconsinCancer.csv")
head(wisc.df)
```

```
##      id diagnosis radius_mean texture_mean perimeter_mean area_mean
## 1   842302      M      17.99      10.38      122.80      1001.0
## 2   842517      M      20.57      17.77      132.90      1326.0
## 3  84300903      M      19.69      21.25      130.00      1203.0
## 4  84348301      M      11.42      20.38      77.58      386.1
## 5  84358402      M      20.29      14.34      135.10      1297.0
## 6   843786      M      12.45      15.70      82.57      477.1
## smoothness_mean compactness_mean concavity_mean concave.points_mean
## 1      0.11840      0.27760      0.3001      0.14710
## 2      0.08474      0.07864      0.0869      0.07017
## 3      0.10960      0.15990      0.1974      0.12790
## 4      0.14250      0.28390      0.2414      0.10520
## 5      0.10030      0.13280      0.1980      0.10430
## 6      0.12780      0.17000      0.1578      0.08089
## symmetry_mean fractal_dimension_mean radius_se texture_se perimeter_se
## 1      0.2419      0.07871      1.0950      0.9053      8.589
## 2      0.1812      0.05667      0.5435      0.7339      3.398
## 3      0.2069      0.05999      0.7456      0.7869      4.585
## 4      0.2597      0.09744      0.4956      1.1560      3.445
## 5      0.1809      0.05883      0.7572      0.7813      5.438
## 6      0.2087      0.07613      0.3345      0.8902      2.217
## area_se smoothness_se compactness_se concavity_se concave.points_se
## 1   153.40      0.006399      0.04904      0.05373      0.01587
## 2    74.08      0.005225      0.01308      0.01860      0.01340
## 3    94.03      0.006150      0.04006      0.03832      0.02058
## 4    27.23      0.009110      0.07458      0.05661      0.01867
## 5    94.44      0.011490      0.02461      0.05688      0.01885
## 6    27.19      0.007510      0.03345      0.03672      0.01137
## symmetry_se fractal_dimension_se radius_worst texture_worst perimeter_worst
## 1      0.03003      0.006193      25.38      17.33      184.60
## 2      0.01389      0.003532      24.99      23.41      158.80
## 3      0.02250      0.004571      23.57      25.53      152.50
## 4      0.05963      0.009208      14.91      26.50      98.87
## 5      0.01756      0.005115      22.54      16.67      152.20
## 6      0.02165      0.005082      15.47      23.75      103.40
## area_worst smoothness_worst compactness_worst concavity_worst
```

```
## 1      2019.0      0.1622      0.6656      0.7119
## 2      1956.0      0.1238      0.1866      0.2416
## 3      1709.0      0.1444      0.4245      0.4504
## 4       567.7      0.2098      0.8663      0.6869
## 5      1575.0      0.1374      0.2050      0.4000
## 6       741.6      0.1791      0.5249      0.5355
##      concave.points_worst symmetry_worst fractal_dimension_worst X
## 1      0.2654      0.4601      0.11890 NA
## 2      0.1860      0.2750      0.08902 NA
## 3      0.2430      0.3613      0.08758 NA
## 4      0.2575      0.6638      0.17300 NA
## 5      0.1625      0.2364      0.07678 NA
## 6      0.1741      0.3985      0.12440 NA
```

```
wisc.data <- as.matrix(wisc.df[,3:32])
head(wisc.data)
```

```
##      radius_mean texture_mean perimeter_mean area_mean smoothness_mean
## [1,]      17.99      10.38      122.80      1001.0      0.11840
## [2,]      20.57      17.77      132.90      1326.0      0.08474
## [3,]      19.69      21.25      130.00      1203.0      0.10960
## [4,]      11.42      20.38       77.58      386.1      0.14250
## [5,]      20.29      14.34      135.10      1297.0      0.10030
## [6,]      12.45      15.70      82.57      477.1      0.12780
##      compactness_mean concavity_mean concave.points_mean symmetry_mean
## [1,]      0.27760      0.3001      0.14710      0.2419
## [2,]      0.07864      0.0869      0.07017      0.1812
## [3,]      0.15990      0.1974      0.12790      0.2069
## [4,]      0.28390      0.2414      0.10520      0.2597
## [5,]      0.13280      0.1980      0.10430      0.1809
## [6,]      0.17000      0.1578      0.08089      0.2087
##      fractal_dimension_mean radius_se texture_se perimeter_se area_se
## [1,]      0.07871      1.0950      0.9053      8.589      153.40
## [2,]      0.05667      0.5435      0.7339      3.398      74.08
## [3,]      0.05999      0.7456      0.7869      4.585      94.03
## [4,]      0.09744      0.4956      1.1560      3.445      27.23
## [5,]      0.05883      0.7572      0.7813      5.438      94.44
## [6,]      0.07613      0.3345      0.8902      2.217      27.19
##      smoothness_se compactness_se concavity_se concave.points_se symmetry_se
## [1,]      0.006399      0.04904      0.05373      0.01587      0.03003
## [2,]      0.005225      0.01308      0.01860      0.01340      0.01389
## [3,]      0.006150      0.04006      0.03832      0.02058      0.02250
## [4,]      0.009110      0.07458      0.05661      0.01867      0.05963
## [5,]      0.011490      0.02461      0.05688      0.01885      0.01756
## [6,]      0.007510      0.03345      0.03672      0.01137      0.02165
##      fractal_dimension_se radius_worst texture_worst perimeter_worst area_worst
## [1,]      0.006193      25.38      17.33      184.60      2019.0
## [2,]      0.003532      24.99      23.41      158.80      1956.0
## [3,]      0.004571      23.57      25.53      152.50      1709.0
## [4,]      0.009208      14.91      26.50      98.87      567.7
## [5,]      0.005115      22.54      16.67      152.20      1575.0
## [6,]      0.005082      15.47      23.75      103.40      741.6
##      smoothness_worst compactness_worst concavity_worst concave.points_worst
## [1,]      0.1622      0.6656      0.7119      0.2654
## [2,]      0.1238      0.1866      0.2416      0.1860
```

```
## [3,]          0.1444          0.4245          0.4504          0.2430
## [4,]          0.2098          0.8663          0.6869          0.2575
## [5,]          0.1374          0.2050          0.4000          0.1625
## [6,]          0.1791          0.5249          0.5355          0.1741
##      symmetry_worst fractal_dimension_worst
## [1,]          0.4601          0.11890
## [2,]          0.2750          0.08902
## [3,]          0.3613          0.08758
## [4,]          0.6638          0.17300
## [5,]          0.2364          0.07678
## [6,]          0.3985          0.12440
```

Q. How many patients are there in this dataset?

```
nrow(wisc.df)
```

```
## [1] 569
```

Q. How many cancer and non-cancer patients are there?

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

```
##
##      B      M
## 357 212
```

```
sum(wisc.df$diagnosis == "M")
```

```
## [1] 212
```

Q. How many columns are "\_mean" values

```
colnames(wisc.df)
```

```
## [1] "id"                "diagnosis"
## [3] "radius_mean"       "texture_mean"
## [5] "perimeter_mean"    "area_mean"
## [7] "smoothness_mean"   "compactness_mean"
## [9] "concavity_mean"    "concave.points_mean"
## [11] "symmetry_mean"     "fractal_dimension_mean"
## [13] "radius_se"         "texture_se"
## [15] "perimeter_se"      "area_se"
## [17] "smoothness_se"     "compactness_se"
## [19] "concavity_se"      "concave.points_se"
## [21] "symmetry_se"       "fractal_dimension_se"
## [23] "radius_worst"      "texture_worst"
## [25] "perimeter_worst"   "area_worst"
## [27] "smoothness_worst"  "compactness_worst"
## [29] "concavity_worst"   "concave.points_worst"
## [31] "symmetry_worst"    "fractal_dimension_worst"
## [33] "X"
```

We can use the `grep()` function to see this

```
grep("_mean", colnames(wisc.data), value=TRUE)
```

```
## [1] "radius_mean"       "texture_mean"       "perimeter_mean"
## [4] "area_mean"         "smoothness_mean"    "compactness_mean"
## [7] "concavity_mean"    "concave.points_mean" "symmetry_mean"
## [10] "fractal_dimension_mean"
```

We can take the `length()` of this to find how many matches there are

```
length(grep("_mean", colnames(wisc.data)))
```

```
## [1] 10
```

```
#View(wisc.data)
```

## Enter Principal Component Analysis

First we need to check whether our input data should be scaled. Lets check the `sd()` and `mean()` of all our columns in `wisc.data`

```
round(apply(wisc.data, 2, sd), 2)
```

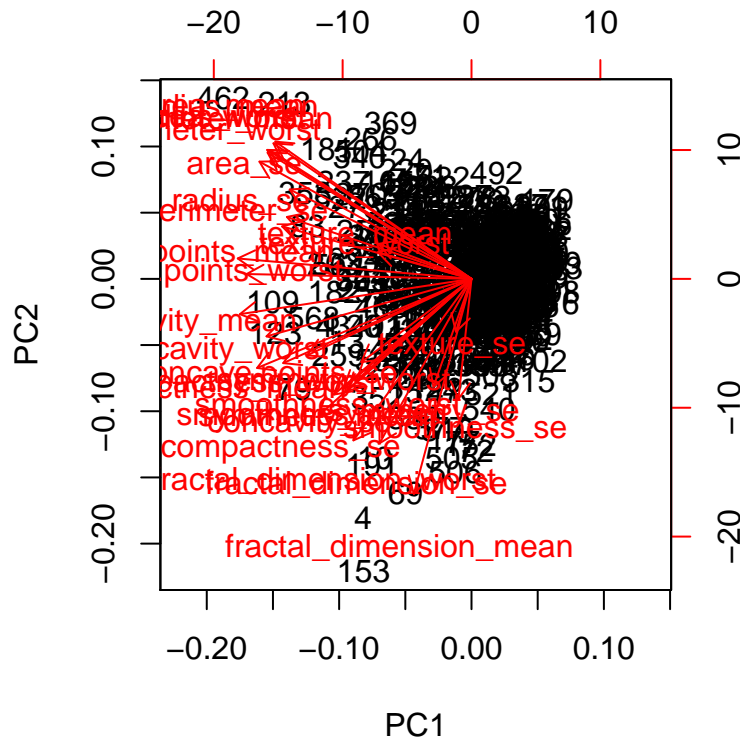
```
##           radius_mean      texture_mean      perimeter_mean
##           3.52         4.30         24.30
##           area_mean      smoothness_mean      compactness_mean
##           351.91         0.01         0.05
##           concavity_mean      concave.points_mean      symmetry_mean
##           0.08         0.04         0.03
## fractal_dimension_mean      radius_se      texture_se
##           0.01         0.28         0.55
##           perimeter_se      area_se      smoothness_se
##           2.02         45.49         0.00
##           compactness_se      concavity_se      concave.points_se
##           0.02         0.03         0.01
##           symmetry_se      fractal_dimension_se      radius_worst
##           0.01         0.00         4.83
##           texture_worst      perimeter_worst      area_worst
##           6.15         33.60         569.36
##           smoothness_worst      compactness_worst      concavity_worst
##           0.02         0.16         0.21
##           concave.points_worst      symmetry_worst      fractal_dimension_worst
##           0.07         0.06         0.02
```

after class break

```
wisc.pr <- prcomp( wisc.data, scale=TRUE )
summary(wisc.pr)
```

```
## 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
```



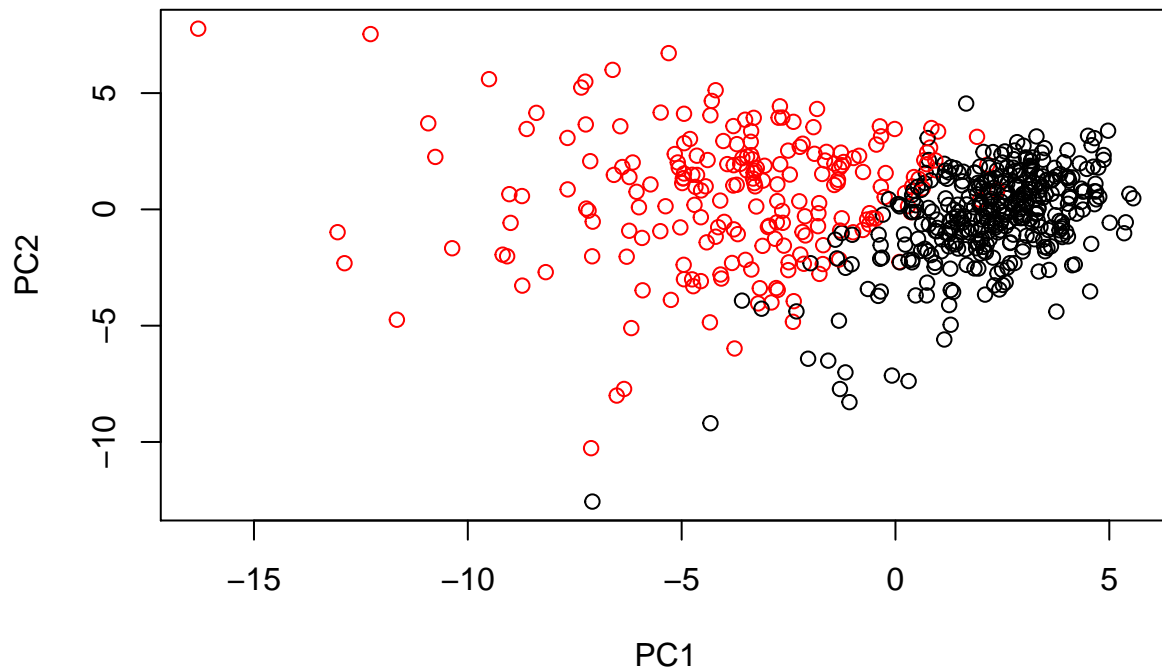
Biplot just doesn't cut it, time to look at this using PCA. First need to access the results within the `wisc.pr` object.

```
attributes(wisc.pr)
```

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

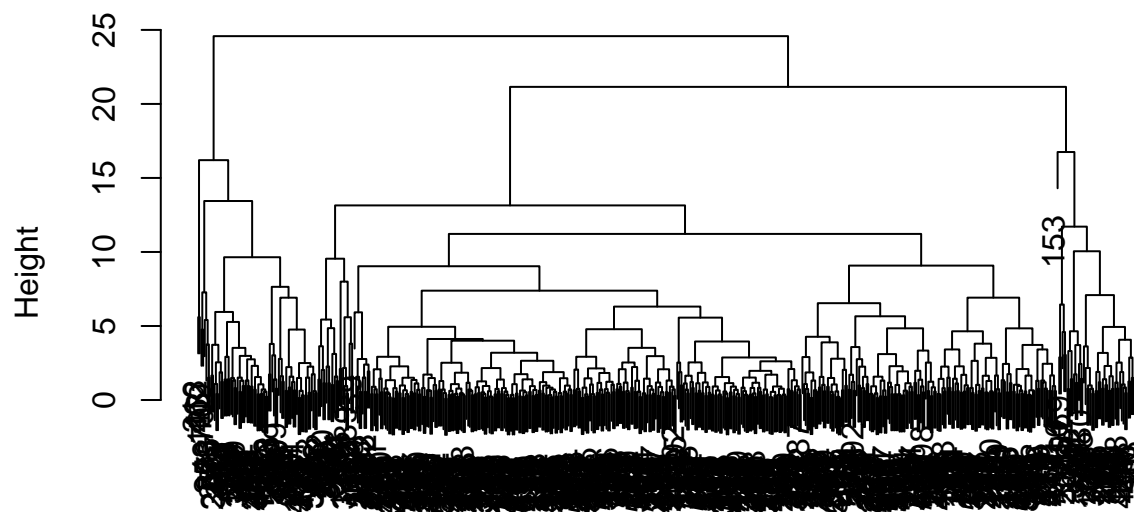
We want the \$x\$ component to make our PCA plot!

```
plot(wisc.pr$x[,1:2], col=wisc.df$diagnosis)
```



```
PCA.wisc.pr<- wisc.pr$x[,1:3]
hclust.wisc.pr<- hclust(dist(PCA.wisc.pr))
#View(hclust.wisc.pr)
plot(hclust.wisc.pr)
```

### Cluster Dendrogram

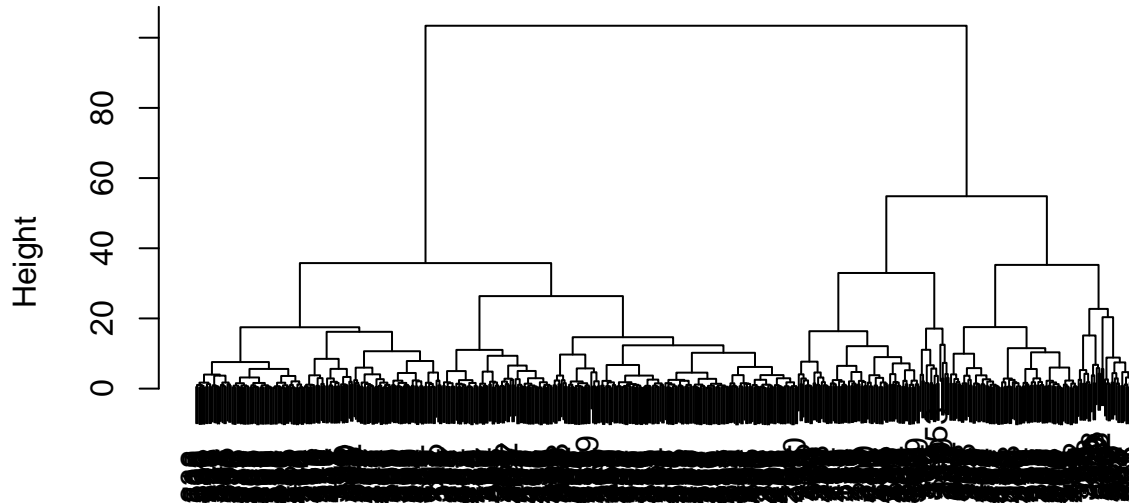


```
dist(PCA.wisc.pr)
hclust (*, "complete")
```

I don't know where it is good to *cut* a tree like this...

```
wisc.pr.hc<- hclust( dist(wisc.pr$x[,1:3]), method="ward.D2")
plot(wisc.pr.hc)
```

## Cluster Dendrogram



```
dist(wisc.pr$x[, 1:3])
hclust (*, "ward.D2")
```

```
grps <- cutree(wisc.pr.hc, k=2)
table(grps)
```

```
## grps
## 1 2
## 203 366
```

```
table(grps, wisc.df$diagnosis)
```

```
##
## grps B M
## 1 24 179
## 2 333 33
```

```
kmeans(wisc.data, centers = 2)
```

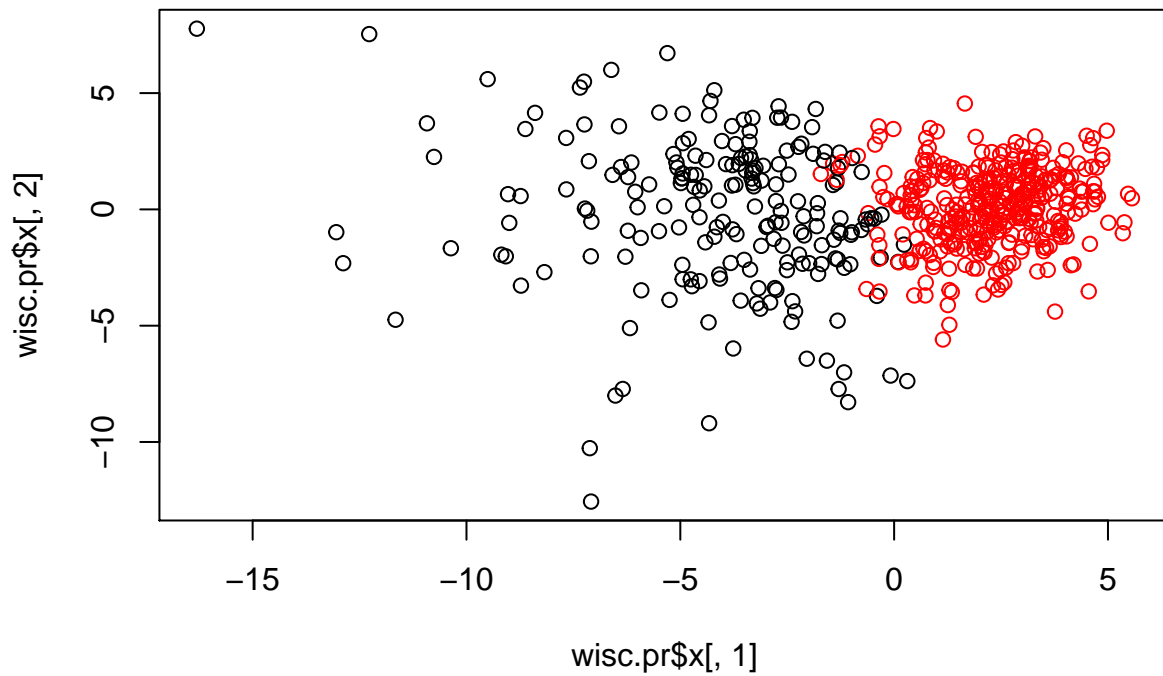
```
## K-means clustering with 2 clusters of sizes 438, 131
##
## Cluster means:
## radius_mean texture_mean perimeter_mean area_mean smoothness_mean
## 1 12.55630 18.57037 81.12347 496.0619 0.0948845
## 2 19.37992 21.69458 128.23130 1185.9298 0.1012946
## compactness_mean concavity_mean concave.points_mean symmetry_mean
## 1 0.09109982 0.06243776 0.03343254 0.1780580
## 2 0.14861298 0.17693947 0.10069878 0.1915397
## fractal_dimension_mean radius_se texture_se perimeter_se area_se
## 1 0.06345402 0.3041909 1.215153 2.152881 23.78529
```

```

## 2          0.06060290 0.7428038  1.222538    5.250580 95.67817
## smoothness_se compactness_se concavity_se concave.points_se symmetry_se
## 1  0.007173263    0.02347469  0.02874551    0.01063632 0.02061358
## 2  0.006598687    0.03217669  0.04241977    0.01567398 0.02030397
## fractal_dimension_se radius_worst texture_worst perimeter_worst area_worst
## 1      0.003747503    14.04390    24.70954    91.93751  619.6479
## 2      0.003953389    23.70947    28.91267    158.49618 1753.0229
## smoothness_worst compactness_worst concavity_worst concave.points_worst
## 1      0.1299591      0.2233118    0.2192149    0.09132984
## 2      0.1404247      0.3577577    0.4493061    0.19243107
## symmetry_worst fractal_dimension_worst
## 1      0.2835537      0.08328194
## 2      0.3118817      0.08616550
##
## Clustering vector:
## [1] 2 2 2 1 2 1 2 1 1 1 1 2 2 1 1 1 1 2 2 1 1 1 1 2 2 1 2 2 2 1 2 2 2 2 1
## [38] 1 1 1 1 1 2 1 1 2 1 1 1 1 1 1 1 2 1 1 2 1 1 1 1 1 1 1 1 1 1 1 1 1 2 1 2 1
## [75] 1 2 1 2 2 1 1 1 2 2 1 2 1 2 1 1 1 1 1 1 2 1 1 1 1 1 1 1 1 1 1 1 1 1 2 1 1
## [112] 1 1 1 1 1 1 1 1 2 2 1 2 2 1 1 1 1 2 1 2 1 1 1 1 2 1 1 1 1 1 1 2 1 1 1 1 1
## [149] 1 1 1 1 1 1 1 1 2 1 1 1 1 2 2 1 2 1 1 2 2 1 1 1 1 1 1 1 1 1 1 1 1 2 2 2 1 1
## [186] 1 2 1 1 1 1 1 1 1 1 1 1 1 2 2 1 1 2 2 1 1 1 1 2 1 1 2 1 1 2 1 1 1 1 2 2 1 1
## [223] 1 1 1 1 1 1 1 1 2 1 1 1 2 1 1 2 2 1 2 1 1 1 1 2 1 1 1 1 1 2 1 2 2 2 1 2 1 2
## [260] 1 2 2 2 1 2 2 1 1 1 1 1 1 2 1 2 1 1 2 1 1 2 1 1 2 1 2 1 1 1 1 1 1 1 1 1 1 1
## [297] 1 1 1 1 2 1 2 1 1 1 1 1 1 1 1 1 1 1 1 1 1 2 1 1 1 2 1 2 1 1 1 1 1 1 1 1 1 1
## [334] 1 1 2 1 2 1 2 1 1 1 2 1 1 1 1 1 1 1 1 1 2 1 1 1 1 1 1 1 1 1 1 1 1 2 2 1 2 2
## [371] 1 1 2 2 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 2 1 1 2 2 1 1 1 1 1 1 2 1 1 1 1 1 1
## [408] 1 2 1 1 1 1 1 1 1 1 2 1 1 1 1 1 1 1 1 1 1 1 1 2 2 1 1 1 1 1 1 1 1 1 2 1 1
## [445] 2 1 2 1 1 2 1 2 1 1 1 1 1 1 1 1 1 2 2 1 1 1 1 1 2 1 1 1 1 1 1 1 1 1 1 1 1
## [482] 1 1 1 1 1 1 2 1 1 1 2 2 1 1 1 1 1 2 2 1 1 1 2 1 1 1 1 1 1 1 1 1 1 1 1 2 2
## [519] 1 1 1 2 1 1 1 1 1 1 1 1 1 1 1 1 2 1 2 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## [556] 1 1 1 1 1 1 1 1 2 2 2 1 2 1
##
## Within cluster sum of squares by cluster:
## [1] 28559677 49383423
## (between_SS / total_SS =  69.6 %)
##
## Available components:
##
## [1] "cluster"      "centers"      "totss"        "withinss"     "tot.withinss"
## [6] "betweenss"    "size"         "iter"         "ifault"
plot(wisc.pr$x[,1], wisc.pr$x[,2], col=grps)

```

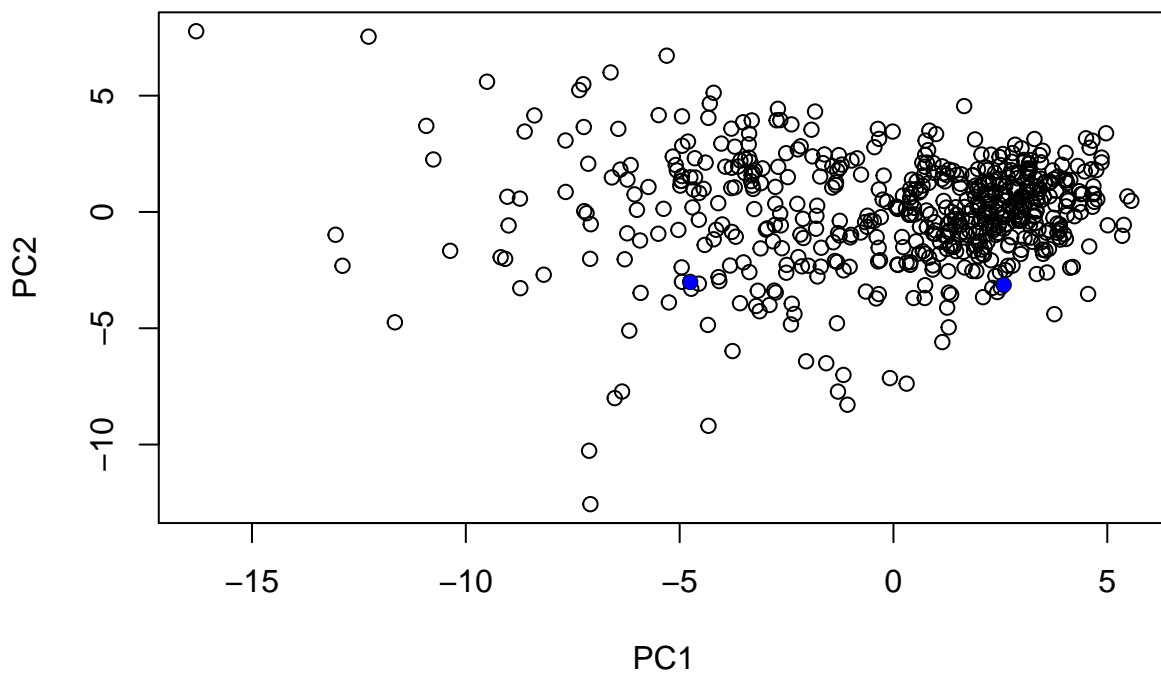




one last hoorah

Predicting Malignancy Of New samples

```
url <- "https://tinyurl.com/new-samples-CSV"
new <- read.csv(url)
npc <- predict(wisc.pr, newdata=new)
plot(wisc.pr$x[,1:2])
points(npc[,1], npc[,2], col="blue", pch=16)
```



```
plot(wisc.pr$x[,1], wisc.pr$x[,2], col=grps)
points(npc[,1], npc[,2], col="blue", pch=16, cex=3)
text(npc[,1], npc[,2], labels=c(1,2), col="white")
```

