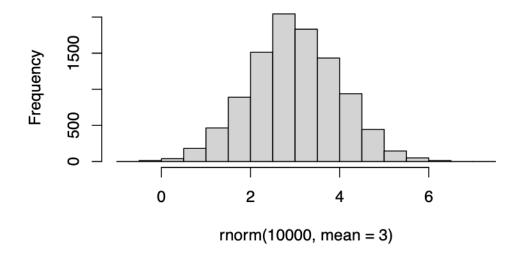
Class07

```
rnorm(10)

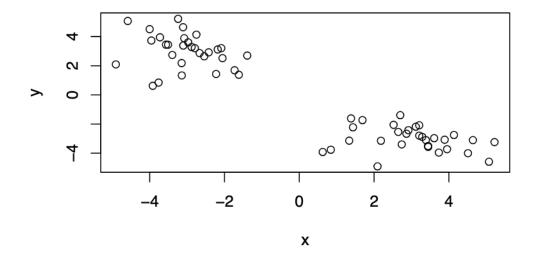
[1] -0.7041508 -0.3406981  1.3403915 -2.0842758  1.7103341  1.0190302
[7] -1.0318180  0.1484586  1.2454737 -1.3385844

hist(rnorm(10000,mean=3))
```

Histogram of rnorm(10000, mean = 3)



```
tmp<-c(rnorm(30,3),rnorm(30,-3))
x<- cbind(x=tmp,y=rev(tmp))</pre>
```



```
#k<- kmeans(x, centers,iter.max=10L...)
k<-kmeans(x,centers=2,nstart=20)
k</pre>
```

K-means clustering with 2 clusters of sizes 30, 30

Cluster means:

x y 1 2.994745 -3.014638 2 -3.014638 2.994745

Clustering vector:

Within cluster sum of squares by cluster:

[1] 61.44498 61.44498 (between_SS / total_SS = 89.8 %)

Available components:

- [1] "cluster" "centers" "totss" "withinss" "tot.withinss"
- [6] "betweenss" "size" "iter" "ifault"
 - Q. How many points are in each cluster

k\$size

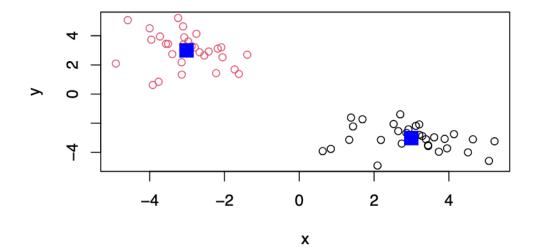
[1] 30 30

<Q2. The clustering result i.e. membership vector?

k\$cluster

Q4. Make a plot of our data colored by clustering results with optionally the cluster centers shown

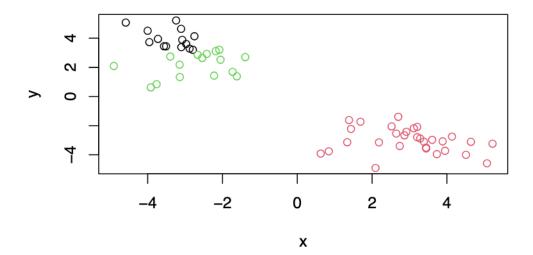
plot(x,col=k\$cluster)
points(k\$centers,col="blue",pch=15,cex=2)



Q5. Run Kmeans again but cluster into 3 groups and plot the results

```
k<-kmeans(x,centers=3,nstart=20)
k$cluster</pre>
```

```
plot(x,col=k$cluster)
```

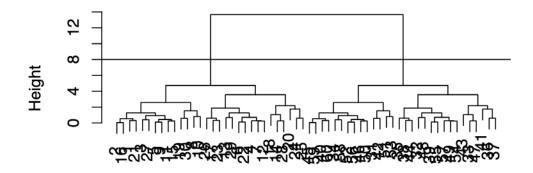


#Hierarchical Clustering

The main function in baseR is 'hclust()'

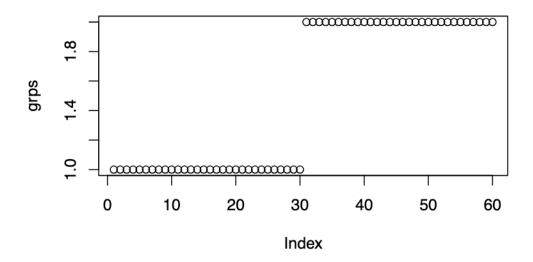
```
hc<-hclust(dist(x))
plot(hc)
abline(h=8)</pre>
```

Cluster Dendrogram



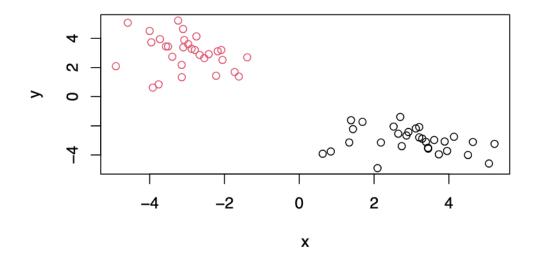
dist(x) hclust (*, "complete")

grps<-cutree(hc,h=8)
plot(grps)</pre>



Q. Plot our hclust rsults in terms of our data colored by cluster membership

```
plot(x,col=grps)
```



Lab portion

```
url <- "https://tinyurl.com/UK-foods"
x <- read.csv(url, row.names=1)
head(x)</pre>
```

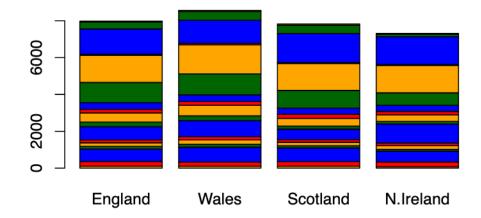
	England	Wales	Scotland	N.Ireland
Cheese	105	103	103	66
Carcass_meat	245	227	242	267
Other_meat	685	803	750	586
Fish	147	160	122	93
Fats_and_oils	193	235	184	209
Sugars	156	175	147	139

dim(x)

[1] 17 4

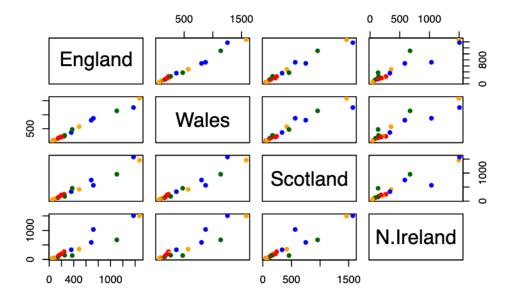
Q1> 5 columns and 17 rows, but if you do row.names=1, there are 17 rows and 4 columns Q2> I prefer to import the df using row.names=1 so I know I am always working with the same df in following lines '

```
colors= (c("orange","red","blue","dark green"))
barplot(as.matrix(x), beside=F, col=colors)
```



Q5> horiz! I"f FALSE, the bars are drawn vertically with the first bar to the left. If TRUE, the bars are drawn horizontally with the first at the bottom."

```
pairs(x, col=colors, pch=16)
```



 $\mathrm{Q6}{>}$ There is a stronger outlier in the N. Ireland dataset that may affect the coorelation value

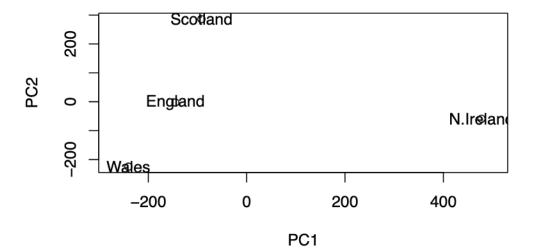
t(x) #transpose x

	Cheese	Carcass	_meat	Other_	meat	Fish	Fats_and	oils	Sugars
England	105		245		685	147		193	156
Wales	103		227		803	160		235	175
Scotland	103		242		750	122		184	147
N.Ireland	66		267		586	93		209	139
	Fresh_	otatoes	Fres	h_Veg	Other	_Veg	Processed	d_potat	toes
England		720)	253		488			198
Wales		874	1	265		570			203
Scotland		566	3	171		418			220
N.Ireland		1033	3	143		355			187
	Process	sed_Veg	Fresh	_fruit	Cere	als	Beverages	Soft_d	drinks
England		360		1102	2	1472	57		1374
Wales		365		1137	7	1582	73		1256
Scotland		337		957	•	1462	53		1572
N.Ireland		334		674	<u>.</u> :	1494	47		1506
	Alcohol	lic_drinl	ks Co	nfectio	nery				
England		3	375		54				

Importance of components:

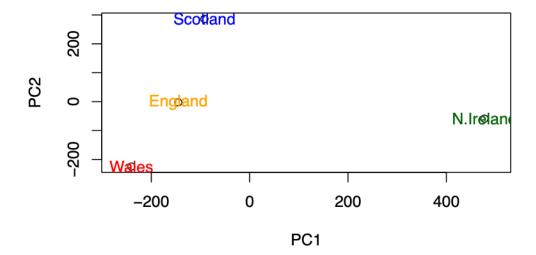
```
PC1 PC2 PC3 PC4
Standard deviation 324.1502 212.7478 73.87622 3.176e-14
Proportion of Variance 0.6744 0.2905 0.03503 0.000e+00
Cumulative Proportion 0.6744 0.9650 1.00000 1.000e+00
```

```
# Plot PC1 vs PC2
plot(pca$x[,1], pca$x[,2], xlab="PC1", ylab="PC2", xlim=c(-270,500))
text(pca$x[,1], pca$x[,2], colnames(x))
```



Q8> Customize your plot so that the colors of the country names match the colors in our UK and Ireland map and table at start of this document.

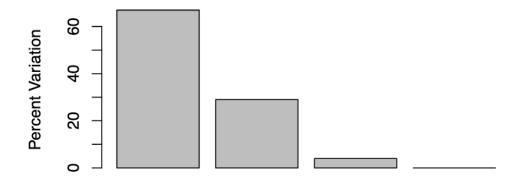
```
# Plot PC1 vs PC2
plot(pca$x[,1], pca$x[,2], xlab="PC1", ylab="PC2", xlim=c(-270,500))
text(pca$x[,1], pca$x[,2], colnames(x),col=colors)
```



```
#variance from diff PC
v <- round( pca$sdev^2/sum(pca$sdev^2) * 100 )
v</pre>
[1] 67 29 4 0
```

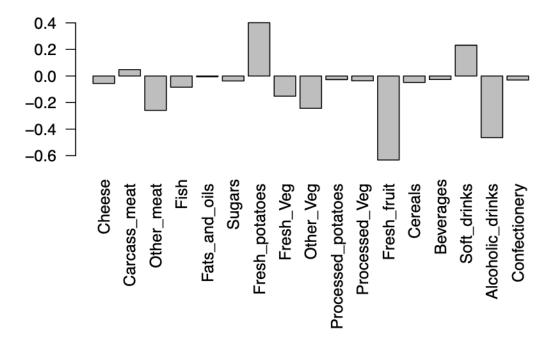
```
## or the second row here...
z <- summary(pca)
z$importance</pre>
```

PC1 PC2 PC3 PC4
Standard deviation 324.15019 212.74780 73.87622 3.175833e-14
Proportion of Variance 0.67444 0.29052 0.03503 0.000000e+00
Cumulative Proportion 0.67444 0.96497 1.00000 1.000000e+00



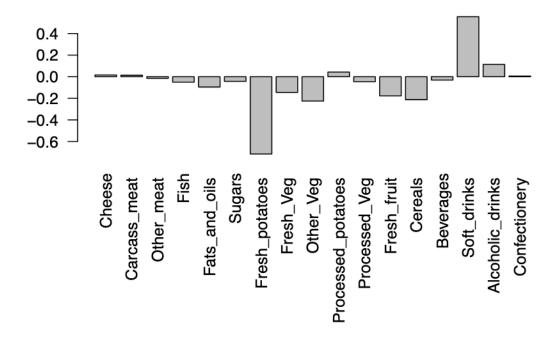
Principal Component

```
## Lets focus on PC1 as it accounts for > 90% of variance
par(mar=c(10, 3, 0.35, 0))
barplot( pca$rotation[,1], las=2 )
```



Q9: Generate a similar 'loadings plot' for PC2. What two food groups feature prominantely and what does PC2 maninly tell us about?

```
##PC2
par(mar=c(10, 3, 0.35, 0))
barplot( pca$rotation[,2], las=2 )
```



It tells us about Fresh_potatoes and soft_drinks. wales eats more fresh potatoes than the rest and drinks fewer soft drinks.

Q10> How many genes and samples are in this data set?

```
url2 <- "https://tinyurl.com/expression-CSV"</pre>
  rna.data <- read.csv(url2, row.names=1)</pre>
  head(rna.data)
       wt1 wt2
                      wt4 wt5 ko1 ko2 ko3 ko4 ko5
                 wt3
       439 458
                 408
                      429 420
                                90
                                    88
                                        86
                                            90
                                                 93
gene1
       219 200
                 204
                      210 187 427 423 434 433 426
gene3 1006 989 1030 1017 973 252 237 238 226 210
gene4
       783 792
                 829
                      856 760 849 856 835 885 894
       181 249
                 204
                      244 225 277 305 272 270 279
gene5
gene6
       460 502
                 491
                      491 493 612 594 577 618 638
  dim(rna.data)
```

[1] 100 10