

# More on population substructure

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## Recalling previous code

```
url="http://www.biostat.umn.edu/~cavanr/FMS_data.txt"
fms <- read.delim("url", header=T, sep="\t")

# Necessary code from Example 3.9:
attach(fms)
NamesAkt1Snps <- names(fms)[substr(names(fms),1,4)==
  "akt1"]
FMSgeno <- fms[,is.element(names(fms),NamesAkt1Snps)]
FMSgenoNum <- data.matrix(FMSgeno)
FMSgenoNum[is.na(FMSgenoNum)] <- 4
DistFmsGeno <- as.matrix(dist(FMSgenoNum))
```

## Multidimensional scaling

Next, one just uses the figure to set up indicator variables

```
mds=cmdscale(DistFmsGeno)
ind1=c(mds[,1]>0 & mds[,2]<4)
ind2=c(mds[,1]>0 & mds[,2]>4)
ind3=c(mds[,1]<0 & mds[,2]>4)

m1=glm(Met_syn~nr3c1_rs4582314+ind1+ind2+ind3,
       family=binomial)
```

# Principal components analysis

Same set up as first slide

```
PCFMS <- prcomp(FMSgenoNum)
```

```
sind1=c(PCFMS$x[,1]<0 & PCFMS$x[,2]<4)
```

```
sind2=c(PCFMS$x[,1]<0 & PCFMS$x[,2]>4)
```

```
sind3=c(PCFMS$x[,1]>0 & PCFMS$x[,2]>4)
```

# Principal components analysis

We can see that these are the same indicator variables

```
> table(ind1,sind1)
      sind1
ind1  FALSE TRUE
FALSE  870   0
TRUE   0   527
```

## Cluster analysis

We can also use a clustering algorithm on this data set.

```
library(mclust)
```

```
m1=Mclust(FMSgenoNum)
```

```
summary(m1)
```

```
-----  
Gaussian finite mixture model fitted by EM algorithm  
-----
```

```
Mclust VEV (ellipsoidal, equal shape) model with 2
```

```
components:
```

log.likelihood	n	df	BIC	ICL
1676.917	1397	626	-1179.709	-1179.709

# Cluster analysis

```
> table(m1$class)
```

```
 1  2  
774 623
```

```
> table(m1$class, ind1)
```

```
  ind1  
  FALSE TRUE  
1   774   0  
2   96  527
```

# Cluster analysis

```
> table(m1$class,ind2)
```

```
  ind2
```

```
  FALSE TRUE
```

```
1    774    0
```

```
2    527    96
```

```
> table(m1$class,ind3)
```

```
  ind3
```

```
  FALSE TRUE
```

```
1    737    37
```

```
2    623    0
```