

Data Mining and Machine Learning in Bioinformatics

Exercise Series 8

Group members (Name, Student ID, E-Mail):

- Baldomero Valdez, Valenzuela, 2905175, baldmer.w@gmail.com
 - Omar Trinidad Gutierrez Mendez, 2850441, omar.vpa@gmail.com
 - Shinho Kang, 2890169, wis.shinho.kang@gmail.com
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Task 7-2:

A) GMM - cluster patients in one of two categories on the basis of expressions from 25 genes.

i) Degrees of freedom:

$n(\pi_k) = 25$, $n(\mu_k) = 25$, $n(\Sigma) = 25 \times 25$, where $n(p)$ is number of adjustable parameter p
totally, $25+25+(25 \times 25) = \mathbf{675}$

ii) with diagonal covariance matrix

$n(\pi_k) = 25$, $n(\mu_k) = 25$, $n(\Sigma) = 25$ totally, $25+25+25 = \mathbf{75}$

iii) no correlation, the variation for each gene is same for each Gaussian.

$n(\pi_k) = 25$, $n(\mu_k) = 25$ totally, $25+25 = \mathbf{50}$

B) GMM clustering

```
#install.packages("mclust")
library(mclust)
library(colonCA)

data(colonCA)
colon.ds <- log(exprs(colonCA))

pvalues <- apply(colon.ds, 1, function(x) {
  return (t.test(x[colonCA$class=='t'], x[colonCA$class=='n'])$p.value)
})
alpha = 0.0001
colon.signif = colon.ds[pvalues <= alpha,]
```

i) patients based on their gene expression profiles

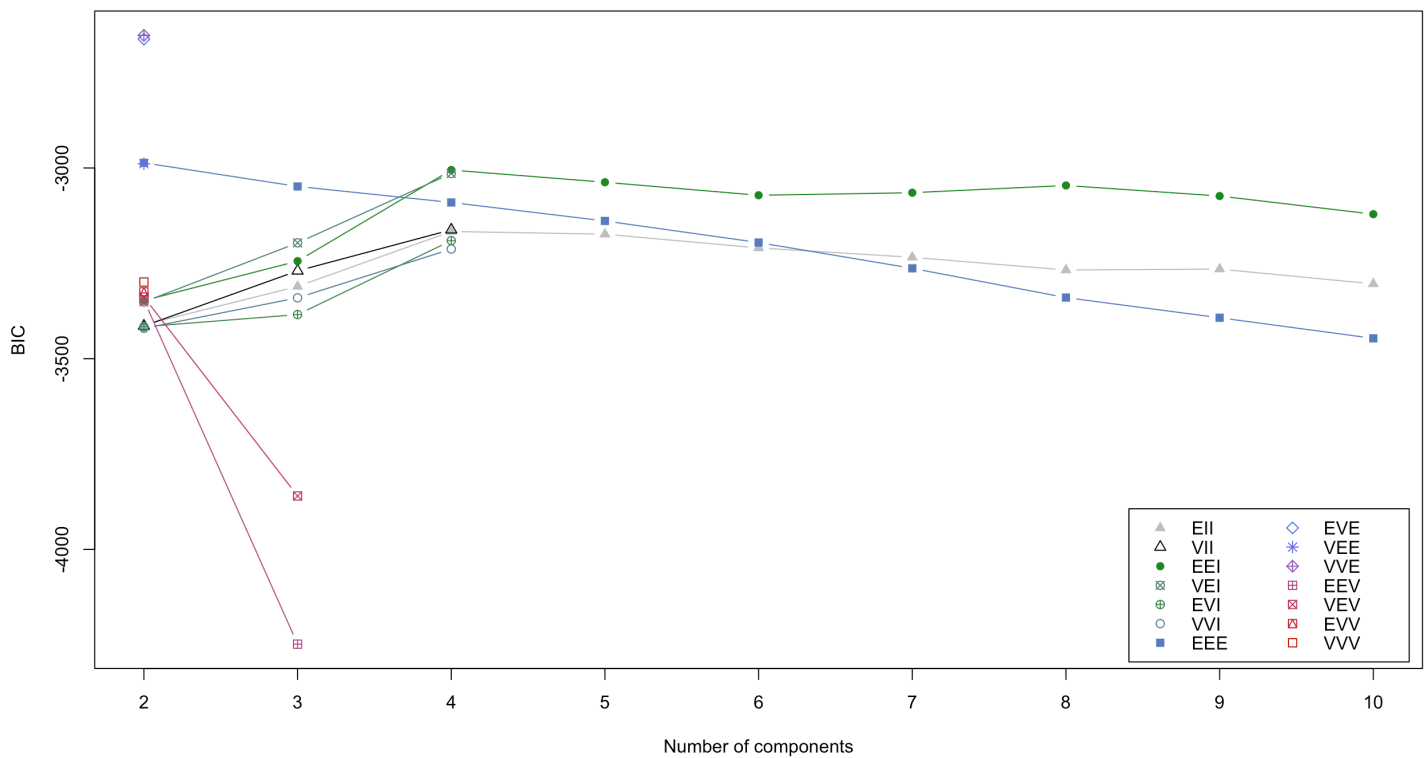
```
c11 = Mclust(t(colon.signif), G=2:10)
summary(c11)
plot(c11, what = "BIC")
```

ii) differentially expressed genes based on their profiles across patients

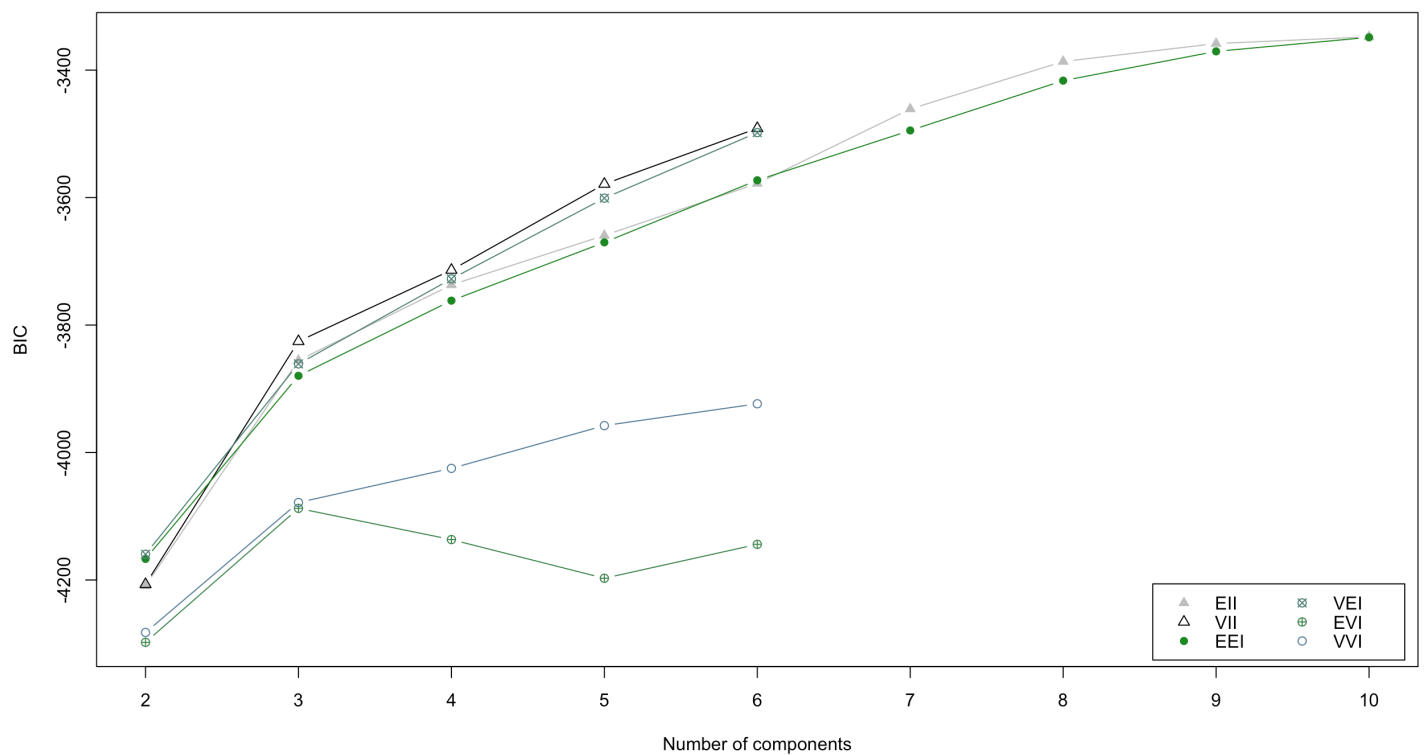
```
c12 = Mclust(colon.signif, G=2:10)
summary(c12)
plot(c12, what = "BIC")
```

RESULT

- GMM for patients: Mclust VVE (ellipsoidal, equal orientation) model with **2 components**, with BIC **-2652.285**



- GMM for genes: Mclust EII (spherical, equal volume) model with **10 components**, with BIC **-3347.762**



C) GMM clustering - Standardized gene expressions

```
scaled.colon.signif <- scale(colon.signif, center=TRUE, scale=TRUE)
```

i) GMM for patients

ii) GMM for genes

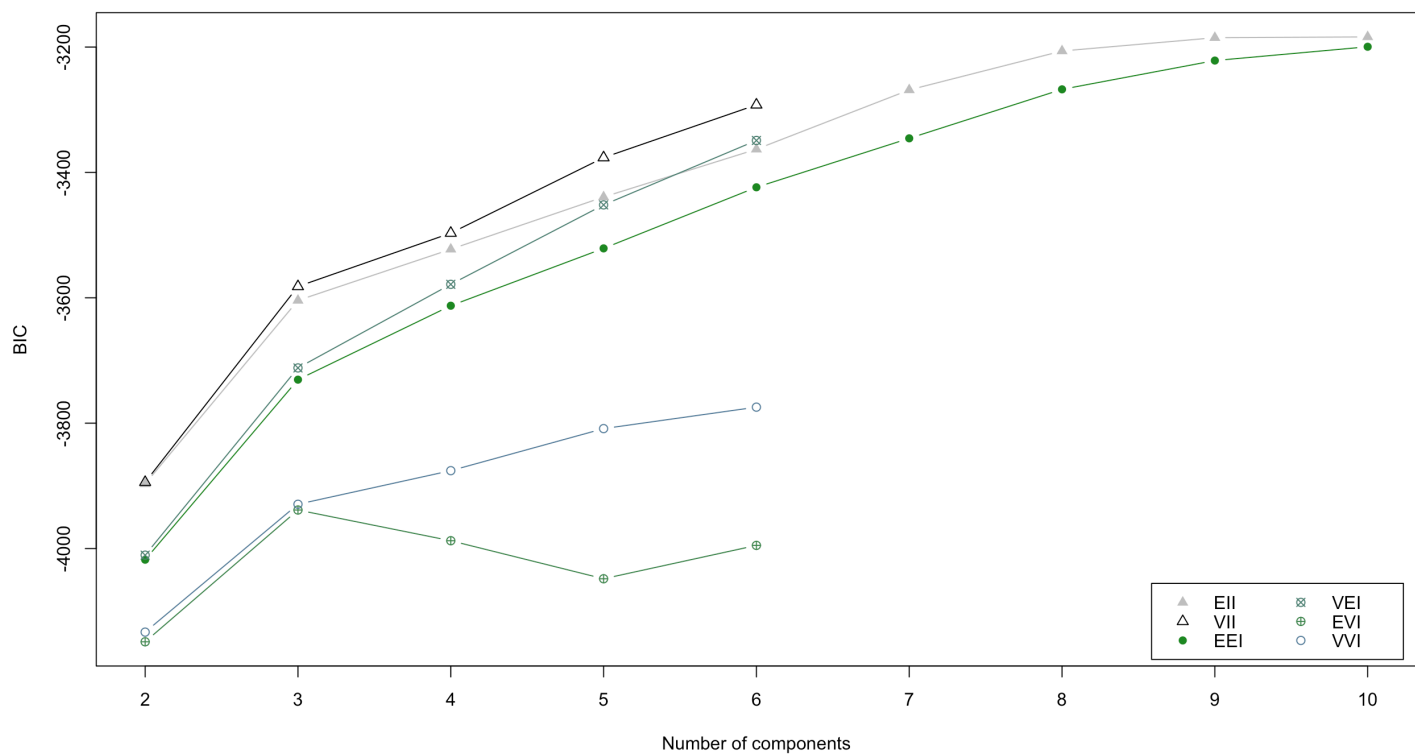
RESULT

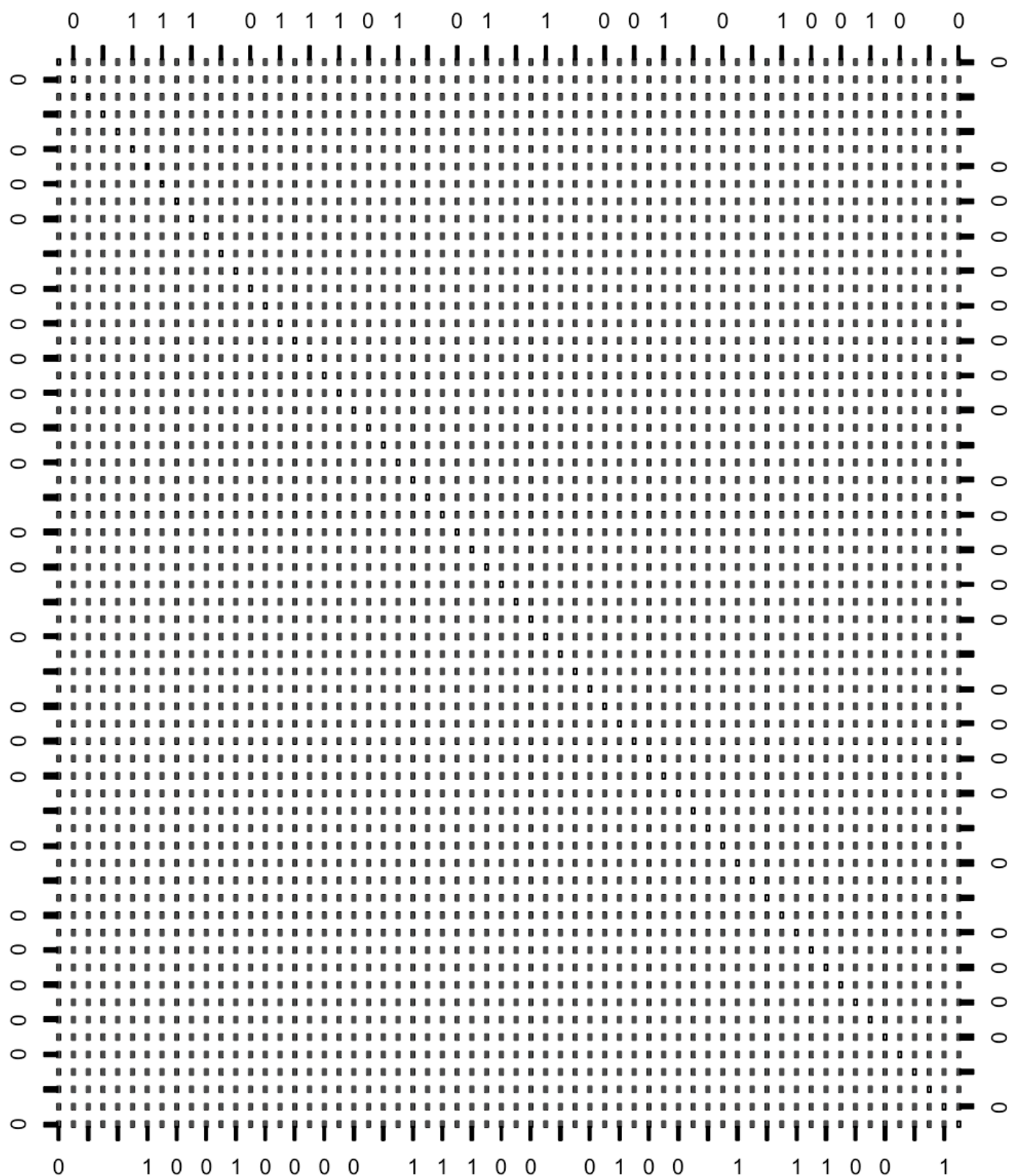
-
- Figure 1 is a line graph showing the Bayesian Information Criterion (BIC) for 15 different models across 2 to 10 components. The y-axis represents BIC, ranging from -2800 to -2200. The x-axis represents the Number of components, ranging from 2 to 10. The models are: EII (grey triangles), VII (black triangles), EEI (green circles), VEI (teal crosses), EVI (green plus signs), VVI (blue circles), EEE (blue squares), EVE (blue diamonds), VEE (purple asterisks), VVE (purple diamonds), EEV (green plus signs), VEV (red crosses), EVV (red squares), and VVV (red squares). The graph shows that BIC generally decreases as the number of components increases, with some models showing a sharp drop at 8 components.



24 variables

- GMM for genes: Mclust EII (spherical, equal volume) model with **10 components**, with BIC **-3183.784**





62 variables

D) Comparing above clusterings

- GMM for patients
 - b) **2 components**, with BIC **-2652.285**
 - c) **3 components**, with BIC **-2107.218**

- GMM for genes
 - b) **10 components**, with BIC **-3347.762**
 - c) **10 components**, with BIC **-3183.784**

For genes, it is difficult to say that using standardized data makes big difference. It probably over-fitted with 10 components.

For patients, standardizing made a bigger number of clusters than the clustering without standardizing.

Standardizing data is recommended because otherwise the range of values in each feature will act as a weight when determining how to cluster data, which is typically undesired.

If one of the features has a range of values much larger than the others, clustering will be completely dominated by that one feature.

E) silhouettes