

Escherichia Coli Network

Example for GeneNet 1.2.13 (August 2015) or later

This note reproduces the “Escherichia coli” network example from J. Schäfer and K. Strimmer. 2005. *A shrinkage approach to large-scale covariance estimation and implications for functional genomics*. Statist. Appl. Genet. Mol. Biol. 4: 32 (<http://dx.doi.org/10.2202/1544-6115.1175>)

Load GeneNet package

```
library("GeneNet")
```

```
## Loading required package: corpcor
```

```
## Loading required package: longitudinal
```

```
## Loading required package: fdrtool
```

E. Coli data set (9 time points for 102 genes):

```
data(ecoli)
dim(ecoli)
```

```
## [1] 9 102
```

Estimation of partial correlations

Estimate matrix of partial correlation using a shrinkage estimator:

```
pc = ggm.estimate.pcor(ecoli)
```

```
## Estimating optimal shrinkage intensity lambda (correlation matrix): 0.1804
```

```
dim(pc)
```

```
## [1] 102 102
```

Assign p-values, q-values and empirical posterior probabilities to all 5151 potential edges in the network:

```
ecoli.edges = network.test.edges(pc, direct=TRUE, fdr=TRUE)
```

```
## Estimate (local) false discovery rates (partial correlations):
## Step 1... determine cutoff point
## Step 2... estimate parameters of null distribution and eta0
## Step 3... compute p-values and estimate empirical PDF/CDF
## Step 4... compute q-values and local fdr
## Step 5... prepare for plotting
```

```
##
## Estimate (local) false discovery rates (log ratio of spvars):
## Step 1... determine cutoff point
## Step 2... estimate parameters of null distribution and eta0
## Step 3... compute p-values and estimate empirical PDF/CDF
## Step 4... compute q-values and local fdr
## Step 5... prepare for plotting
```

```
dim(ecoli.edges)
```

```
## [1] 5151 10
```

The table lists all edges in the order strength of partial correlations:

```
ecoli.edges[1:5,]
```

```
##          pcor node1 node2          pval          qval          prob          log.spvar
## 1  0.2318566    51    53 2.220446e-16 3.612205e-13 1.0000000 -0.043537019
## 2  0.2240555    52    53 2.220446e-16 3.612205e-13 1.0000000 -0.040249854
## 3  0.2150782    51    52 2.220446e-16 3.612205e-13 1.0000000 -0.003287165
## 4  0.1732886     7    93 3.108624e-15 3.792816e-12 0.9999945 -0.025293430
## 5 -0.1341889    29    86 1.120812e-09 1.093997e-06 0.9999516  0.022305368
##          pval.dir  qval.dir  prob.dir
## 1 0.3803869 0.7557272 1.110223e-15
## 2 0.4173922 0.7724561 1.110223e-15
## 3 0.9471949 0.8851073 1.110223e-15
## 4 0.6103234 0.8323249 1.110223e-15
## 5 0.6531371 0.8415749 1.110223e-15
```

Decide which edges to include in the network

To obtain a graph you need to select top ranking edges according to a suitable criterion. Here are some suggestions:

1. Use local fdr cutoff 0.2, i.e. include all edges with posterior probability of at least 0.8.

```
ecoli.net = extract.network(ecoli.edges)
```

```
##
## Significant edges: 125
##   Corresponding to 2.43 % of possible edges
##
## Significant directions: 377
##   Corresponding to 7.32 % of possible directions
## Significant directions in the network: 17
##   Corresponding to 13.6 % of possible directions in the network
```

```
dim(ecoli.net)
```

```
## [1] 125 11
```

2. Use local fdr cutoff 0.1, i.e. include all edges with posterior probability of at least 0.9.

```
ecoli.net = extract.network(ecoli.edges, cutoff.ggm=0.9, cutoff.dir=0.9)
```

```
##
## Significant edges: 65
##   Corresponding to 1.26 % of possible edges
##
## Significant directions: 269
##   Corresponding to 5.22 % of possible directions
## Significant directions in the network: 6
##   Corresponding to 9.23 % of possible directions in the network
```

```
dim(ecoli.net)
```

```
## [1] 65 11
```

3. Include a fixed number of edges, say the 70 strongest edges

```
ecoli.net = extract.network(ecoli.edges, method.ggm="number", cutoff.ggm=70)
```

```
##
## Significant edges: 70
##   Corresponding to 1.36 % of possible edges
##
## Significant directions: 377
##   Corresponding to 7.32 % of possible directions
## Significant directions in the network: 9
##   Corresponding to 12.86 % of possible directions in the network
```

```
dim(ecoli.net)
```

```
## [1] 70 11
```

Plot network

For plotting we use the graph and Rgraphviz packages from Bioconductor.

```
library("Rgraphviz")
```

```
## Loading required package: graph
```

```
## Loading required package: BiocGenerics
```

```
##
```

```
## Attaching package: 'BiocGenerics'
```

```
## The following objects are masked from 'package:stats':
```

```
##
```

```
## IQR, mad, sd, var, xtabs
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
## anyDuplicated, aperm, append, as.data.frame, basename, cbind,  
## colnames, dirname, do.call, duplicated, eval, evalq, Filter, Find,  
## get, grep, grepl, intersect, is.unsorted, lapply, Map, mapply,  
## match, mget, order, paste, pmax, pmax.int, pmin, pmin.int,  
## Position, rank, rbind, Reduce, rownames, sapply, setdiff, table,  
## tapply, union, unique, unsplit, which.max, which.min
```

```
## Loading required package: grid
```

Create graph object from the list of edges:

```
node.labels = colnames(ecoli)  
gr = network.make.graph(ecoli.net, node.labels, drop.singles=TRUE)  
table( edge.info(gr)$dir )
```

```
##
```

```
## forward none
```

```
## 9 61
```

```
sort( node.degree(gr), decreasing=TRUE)
```

```
## sucA cspG fixC yheI lacA lacY lacZ asnA eutG yceP yedE ygcE pspA  
## 11 8 7 7 6 6 6 5 5 5 5 5 4  
## atpD b1191 b1583 cspA icdA mopB pspB tnaA yaeM ycgX yfaD dnaG dnaK  
## 3 3 3 3 3 3 3 3 3 3 3 2 2  
## hupB ibpB yfiA aceB atpG b1963 cchB dnaJ flgD folK ftsJ gltA lpdA  
## 2 2 2 1 1 1 1 1 1 1 1 1 1  
## nmpC nuoM sucD yecO ygbD yhdM yjbO  
## 1 1 1 1 1 1 1
```

Set node and edge attributes for more beautiful graph plotting:

```

globalAttrs = list()
globalAttrs$edge = list(color = "black", lty = "solid", lwd = 1, arrowsize=1)
globalAttrs$node = list(fillcolor = "lightblue", shape = "ellipse", fixedsize = FALSE)

nodeAttrs = list()
nodeAttrs$fillcolor = c('sucA' = "yellow")

edi = edge.info(gr)
edgeAttrs = list()
edgeAttrs$dir = edi$dir # set edge directions
edgeAttrs$lty = ifelse(edi$weight < 0, "dotted", "solid") # negative correlation -> dotted
edgeAttrs$color = ifelse(edi$dir == "none", "black", "red")
edgeAttrs$label = round(edi$weight, 2) # use partial correlation as edge labels

plot(gr, attrs = globalAttrs, nodeAttrs = nodeAttrs, edgeAttrs = edgeAttrs, "fdp")

```

