

Arabidopsis Thaliana Network

Example for GeneNet 1.2.13 (August 2015) or later

This note reproduces the “Arabidopsis thaliana” network example from R. Opgen-Rhein and K. Strimmer. 2007. *From correlation to causation networks: a simple approximate learning algorithm and its application to high-dimensional plant gene expression data*. BMC Syst. Biol. **1**: 37. (<http://dx.doi.org/10.1186/1752-0509-1-37>)

The original source of the data is Smith et al. 2004. *Diurnal changes in the transcriptom encoding enzymes of starch metabolism provide evidence for both transcriptional and posttranscriptional regulation of starch metabolism in Arabidopsis leaves*. Plant Physiol. **136**: 2687-2699.

This example was suggested by Papapit Ingkasuwan, Division of Biotechnology, School of Bioresources and Technology, King Mongkut’s University of Technology Thonburi, Bangkok, Thailand.

Inspect Data

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

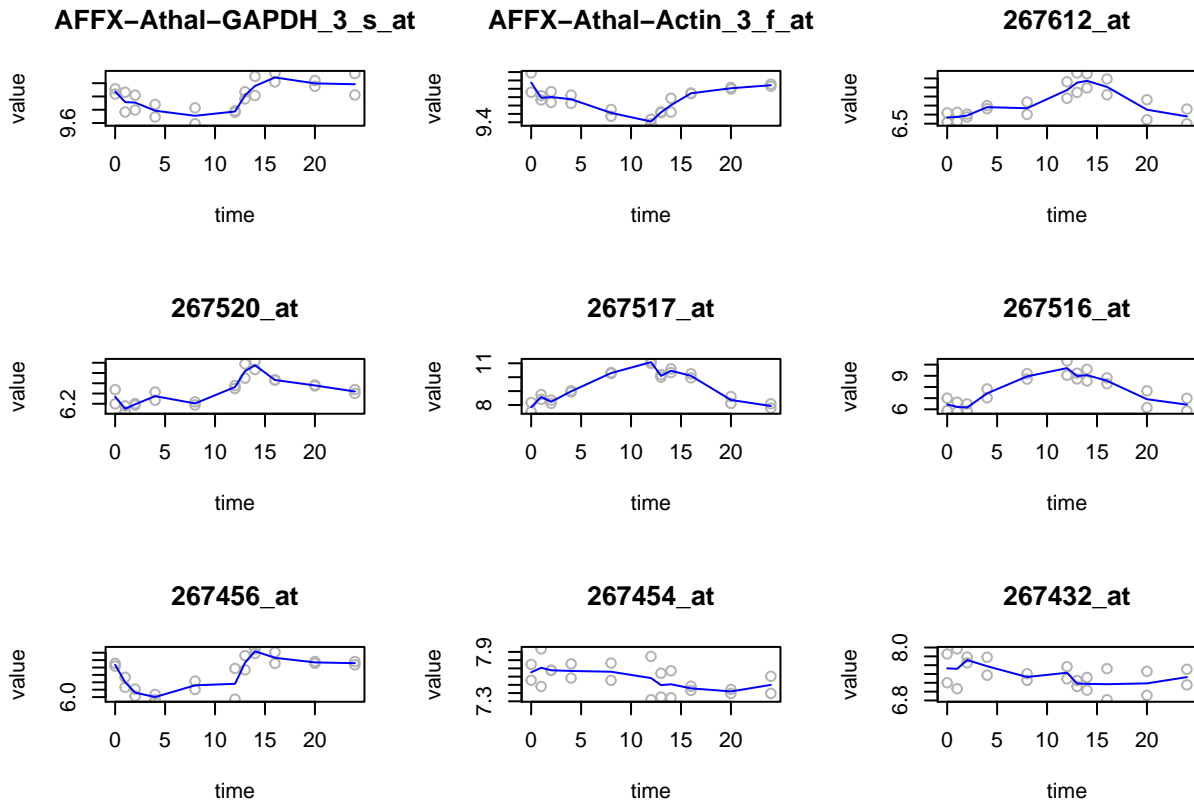
```
## Loading required package: corpcor  
## Loading required package: longitudinal  
## Loading required package: fdrtool
```

```
data("arth800")  
summary(arth800.expr)
```

```
## Longitudinal data:  
## 800 variables measured at 11 different time points  
## Total number of measurements per variable: 22  
## Repeated measurements: yes  
##  
## To obtain the measurement design call 'get.time.repeats()'.
```

Plot time series:

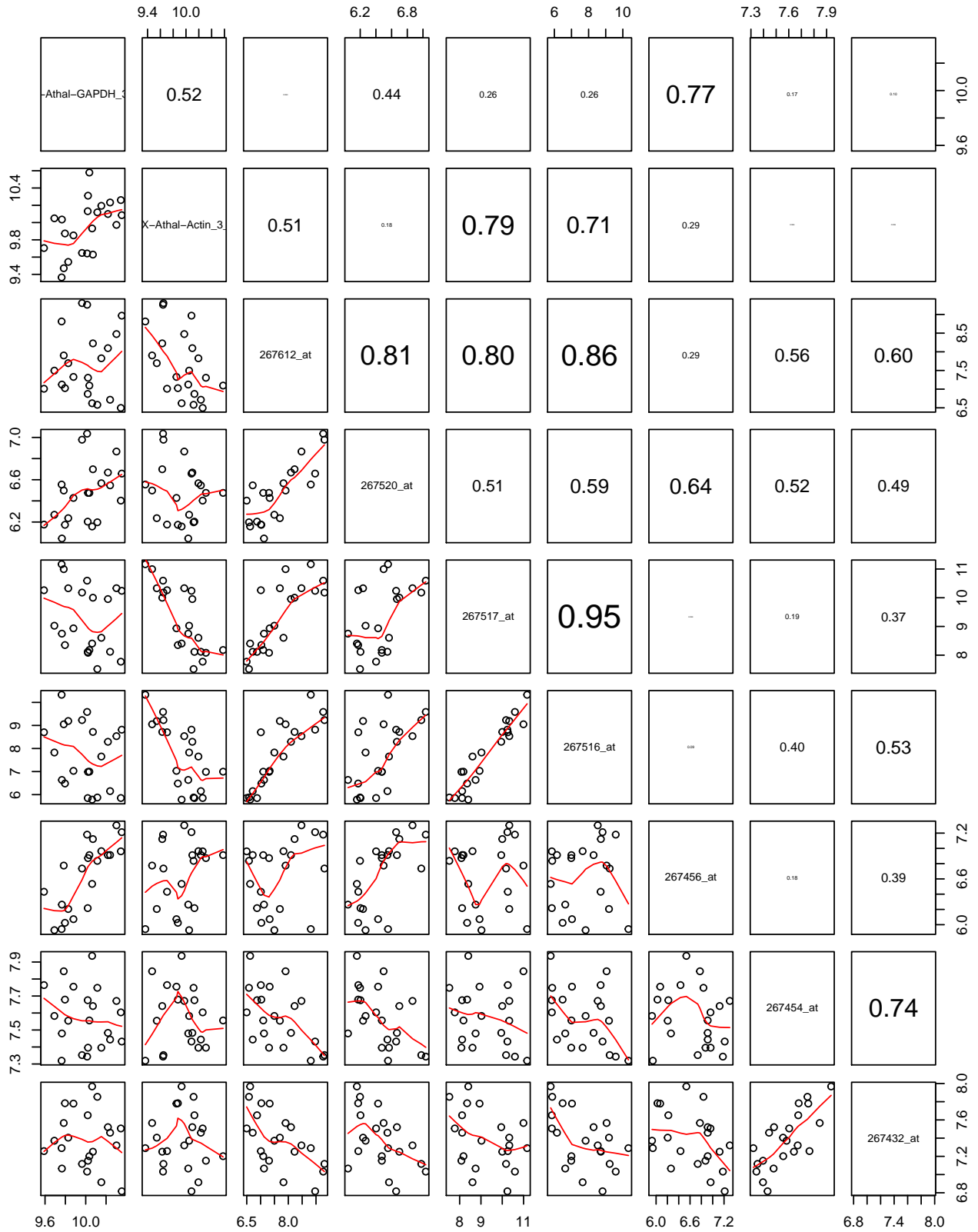
```
plot(arth800.expr, 1:9)
```



Inspect pairwise scatter plots:

```
panel.cor = function(x, y, digits=2, prefix="", cex.cor)
{
  usr = par("usr"); on.exit(par(usr))
  par(usr = c(0, 1, 0, 1))
  r = abs(cor(x, y))
  txt = format(c(r, 0.123456789), digits=digits)[1]
  txt = paste(prefix, txt, sep="")
  if(missing(cex.cor)) cex = 0.8/strwidth(txt)
  text(0.5, 0.5, txt, cex = cex * r)
}
```

```
pairs(arth800.expr[,1:9], lower.panel=panel.smooth, upper.panel=panel.cor)
```



Compute Partial Correlations and Select Relevant Edges

```
pcor.dyn = ggm.estimate.pcor(arth800.expr, method = "dynamic")
```

```
## Estimating optimal shrinkage intensity lambda (correlation matrix): 0.185
```

```
arth.edges = network.test.edges(pcor.dyn,direct=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(arth.edges)
```

```
## [1] 319600    10
```

We use the strongest 150 edges:

```
arth.net = extract.network(arth.edges, method.ggm="number", cutoff.ggm=150)
```

```
##  
## Significant edges: 150  
##   Corresponding to 0.05 % of possible edges  
##  
## Significant directions: 10516  
##   Corresponding to 3.29 % of possible directions  
## Significant directions in the network: 55  
##   Corresponding to 36.67 % of possible directions in the network
```

Construct Graph

```
library("graph")
```

```
node.labels = as.character(1:ncol(arth800.expr))  
gr = network.make.graph( arth.net, node.labels, drop.singles=TRUE)
```

Some information about the graph Number of nodes:

```
num.nodes(gr)
```

```
## [1] 107
```

Correlations:

```
edge.info(gr)$weight
```

```
##      8~209    13~331    13~422    20~573    20~781    26~781    47~793    47~422
## -0.03088  0.03289  0.03416 -0.03221  0.03300 -0.03751  0.03101  0.03302
##      47~736    47~714    47~414    47~331    47~629    47~452    47~479    61~480
##  0.03460  0.03553  0.03559  0.03897  0.04035  0.04036  0.04235  0.03112
##      61~111    63~198    63~738    63~444    78~726    81~377    81~712    81~198
##  0.03414 -0.03085 -0.03953 -0.04345 -0.03184  0.03139  0.03157  0.03184
##      81~368    81~108    81~211    81~61    81~666    81~636    81~665    81~793
##  0.03244  0.03311  0.03314  0.03334 -0.03342  0.03367 -0.03382  0.03402
##      81~248    81~622    81~144    81~111    81~570    81~767    86~738    86~181
## -0.03423 -0.03501 -0.03593  0.03912  0.04281 -0.05957  0.03068  0.03268
##     100~296   100~245   100~412   101~443   108~603   108~272   111~537   111~496
##  0.03181  0.03278  0.03401  0.03449  0.03091 -0.03117 -0.03102  0.03405
##     126~783   155~679   181~783   198~779   198~686   209~738   226~573   269~783
## -0.03384  0.03345  0.04190 -0.03128 -0.03631 -0.03711 -0.03118  0.03479
##     272~289   272~603   272~414   272~560   272~452   272~726   289~414   289~452
##  0.03364 -0.03782  0.03797  0.04137  0.04530 -0.05022  0.03132  0.03479
##     289~786   289~726   299~444   328~519   331~714   331~479   331~414   331~452
## -0.03598 -0.03933 -0.03135  0.03108  0.03194  0.03278  0.03404  0.03628
##     331~422   414~786   414~452   414~726   422~479   422~736   422~714   422~629
##  0.03881 -0.03464  0.03520 -0.03543  0.03104  0.03277  0.03573  0.03663
##     422~627   443~565   443~573   443~600   444~738   452~714   452~726   452~603
## -0.03972  0.03092 -0.03130 -0.03159  0.03340  0.03216 -0.03400 -0.03526
##     479~677   479~714   479~793   479~629   480~738   539~360   539~758   539~778
##  0.03261  0.03656  0.03837  0.04523 -0.04365  0.03080 -0.03104  0.03219
##     539~596   539~93    539~585   539~197   539~4    540~738   558~783   558~342
##  0.03269 -0.03381 -0.03483  0.03614  0.04977  0.04087  0.03174 -0.03211
##     558~96    558~269   558~363   558~161   558~739   558~126   558~256   560~793
##  0.03223  0.03271  0.03387 -0.03596 -0.03647 -0.03727 -0.04139 -0.03093
##     560~627   560~726   570~623   570~767   570~256   570~234   570~460   570~219
##  0.03807 -0.04152  0.03122 -0.03127 -0.03187  0.03225 -0.03322  0.03364
##     570~378   570~135   570~699   570~585   570~576   570~422   570~38    570~554
## -0.03373  0.03376  0.03565  0.03722  0.03794 -0.03815 -0.03951 -0.03989
##     570~651   570~61    570~598   570~111   570~464   600~699   603~781   627~738
##  0.04060  0.04262  0.04284  0.04411  0.04751  0.03330  0.03092 -0.03246
##     627~661   627~281   629~714   629~793   636~738   677~714   679~783   726~786
##  0.03495  0.03570  0.03806  0.04579  0.03234  0.03134  0.03705  0.03638
##     779~798   781~783   783~640   783~187   783~547   783~454
##  0.03063  0.03146 -0.03264 -0.03306  0.03761 -0.04094
```

Number of directed (“forward”) and undirected (“none”) edges:

```
table( edge.info(gr)$dir )
```

```
##
## forward    none
##      55      95
```

Well-Connected Nodes

Nodes connected with many edges:

```
sort(node.degree(gr), decreasing=TRUE)[1:10]
```

```
## 570 81 783 47 422 558 452 539 738 272  
## 20 17 10 9 9 9 8 8 8 7
```

```
arth800.descr[570]
```

```
## [1] "AP2 transcription factor - like protein"
```

```
arth800.descr[81]
```

```
## [1] "ATRPAC43; DNA binding / DNA-directed RNA polymerase; DNA-directed RNA polymerase, putative, identical to ATRPAC43"
```

```
arth800.descr[558]
```

```
## [1] "structural constituent of ribosome; 60S ribosomal protein L35 (RPL35C), various ribosomal L35 proteins"
```

```
arth800.descr[539]
```

```
## [1] "DNA binding / transcription factor; basic helix-loop-helix (bHLH) family protein, contains Pfam domain PF00022"
```

```
arth800.descr[783]
```

```
## [1] "RNA binding / RNA methyltransferase; tRNA/rRNA methyltransferase (SpoU) family protein, similar to SpoU"
```

Plot Network

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

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

For a more beautiful plot of the network set node and edge parameters: Set global node and edge attributes:

```
globalAttrs = list()  
globalAttrs$edge = list(color = "black", lty = "solid", lwd = 1, arrowsize=1)  
globalAttrs$node = list(fillcolor = gray(.95), shape = "ellipse", fixedsize = FALSE)
```

Set attributes of some particular nodes:

```
nodeAttrs = list()
nodeAttrs$fillcolor = c('570' = "red", "81" = "red") # highlight hub nodes
```

Set edge attributes:

```
edi = edge.info(gr) # edge directions and correlations
edgeAttrs = list()
edgeAttrs$dir = edi$dir # set edge directions
cutoff = quantile(abs(edi$weight), c(0.2, 0.8)) # thresholds for line width / coloring
edgeAttrs$lty = ifelse(edi$weight < 0, "dotted", "solid") # negative correlation
edgeAttrs$color = ifelse(abs(edi$weight) <= cutoff[1], "grey", "black") # lower 20% quantile
edgeAttrs$lwd = ifelse(abs(edi$weight) >= cutoff[2], 2, 1) # upper 20% quantile
```

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

