Supplementary Material for Accurate directional inference in Gaussian graphical models

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1 Introduction

The current report reproduces the Figures and numerical results in Sections 4 and 5 of the main text. The outputs have been produced using R version 4.1.3 (R Core Team 2020). The code chunk below loads the necessary R packages.

```
library(matrixcalc)
library(Matrix)
library(mvtnorm)
library(ggm)
library(Rgraphviz)
library(gRbase)
```

We also provide code to reproduce all simulation results in the paper. The R scripts to carry out the simulation experiments, and the results from those, are provided in the DirInfGGM_code+results.zip archive. res_dir is the directory where the contents of the archive are and needs to be set appropriately.

```
res_dir <- "~/Dropbox/Directional/Supplementary/DirInfGGM_code+results"</pre>
```

2 Simulation studies

This section provides the R code that reproduces the outputs of Section 4 in the paper.

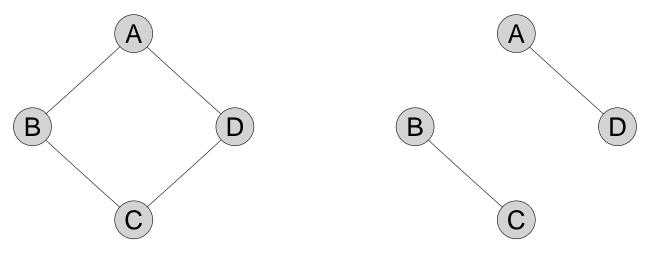
2.1 Scenario 1

The code chunk below reproduces Figure 1.

```
defAttrs <- getDefaultAttrs()
eAttrs <- list()
eAttrs$color <- c("A~B" = "white", "D~C" = "white", "C~B" = "black", "D~A" = "black")

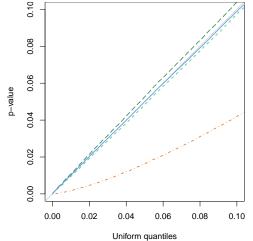
ug0 <- ug(~ A*B + D*C)
ug1 <- ug(~ A*B + D*C + C*B + D*A)

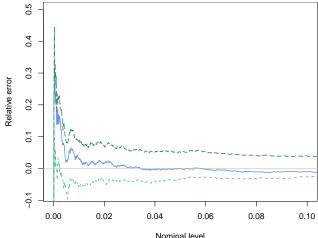
par(mfrow = c(1, 2))
plot(ug1, attrs = list(node = list(fillcolor = "lightgrey", fontsize = "10")))
plot(ug1, edgeAttrs = eAttrs, attrs = list(node = list(fillcolor = "lightgrey", fontsize = "10")))</pre>
```



The following code chunk uses the image dir_small_res.rda to reproduce Figure 2 of the main text. dir_small_res.rda results by running the script simulations1.R, available in the supplementary code archive.

```
setwd(res_dir)
load("dir_small_res.rda")
par(mfrow = c(1, 2), mai = c(1, 1, 0, 0.5))
old.pty <- par("pty")</pre>
par(pty = "s")
# 1st plot: empirical p-values
index <- seq(1, length(res_small$first.order), by = 10)</pre>
x <- sort(res_small$first.order)</pre>
x <- x[index]
plot(ppoints(x), x, type = "1",
     xlim=c(0, 0.1), lwd = 1.5, ylim=c(0, 0.1), xlab = "Uniform quantiles",
     ylab = "p-value", pty = "s", col = "chocolate2", lty = "dotdash", cex.lab = 1)
abline(0, 1, col ="grey")
x <- sort(res small$skovgaard1)</pre>
x <- x[index]
lines(ppoints(x), x, col = "seagreen3",
      lwd = 1.5, lty = "dashed")
x <- sort(res_small$skovgaard2)</pre>
x <- x[index]
lines(ppoints(x), x, col = "seagreen4",
      lty = "longdash", lwd = 1.5)
x <- sort(res_small$directional)</pre>
x <- x[index]
lines(ppoints(x), x, col = "cornflowerblue", lwd = 1.5)
par(pty = old.pty)
# 2nd plot: relative error for p-values lower than 0.10 (without w)
x <- sort(res_small$directional)</pre>
x <- x[index]
plot(ppoints(x), (x - ppoints(x))/ppoints(x), xlim = c(0, 0.1), lwd = 1.5,
     ylab ="Relative error", xlab = "Nominal level", type ="1", ylim = c(-0.08, 0.5),
     cex.lab = 1, col = "cornflowerblue")
abline(h = 0, col = "grey")
```





2.2 Scenario 2

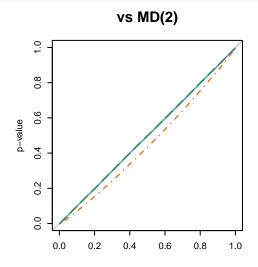
The code chunk below uses the image dir_MDq11_res.rda to reproduce Table 1 of the main text. Tables 2 and 3 are obtained similarly starting from the files dir_MDq30_res.rda and dir_MDq50_res.rda, respectively. All these images result by running the script simulations2.R, available in the supplementary code archive.

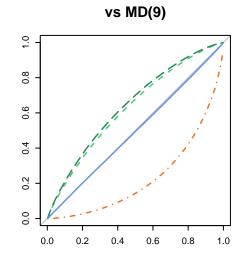
```
summary.sim <- function(out, levels = c(0.01, 0.025, 0.05, 0.10, 0.25, 0.5, 0.75,
                                            0.9, 0.95, 0.975, 0.99)) {
  fo <- sapply(levels, function(x) mean(out$first.order < x))</pre>
  sk1 <- sapply(levels, function(x) mean(out$skovgaard1 < x))</pre>
  sk2 <- sapply(levels, function(x) mean(out$skovgaard2 < x))</pre>
  dir2 <- sapply(levels, function(x) mean(out$directional < x))</pre>
  sim.se <- sqrt(levels * (1 - levels)/length(out$first.order))</pre>
  res <- rbind(levels, fo, sk1, sk2, dir2, sim.se)
  rownames(res) <- c("nominal", "first-order", "Skovgaard W*",</pre>
                       "Skovgaard W**", "directional", "sim-error")
  res
}
setwd(res_dir)
load("dir_MDq11_res.rda")
res9md <- round(summary.sim(resMD9), 3)</pre>
res6md <- round(summary.sim(resMD6), 3)</pre>
res3md <- round(summary.sim(resMD3), 3)</pre>
res2md <- round(summary.sim(resMD2), 3)</pre>
```

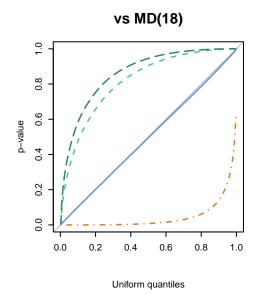
```
tab1 <- rbind(res2md[2:5,], res3md[2:5,],
             res6md[2:5,], res9md[2:5,])
round(tab1 * 100, digits = 1)
##
                 [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9] [,10] [,11]
## first-order
                 1.4 3.3 6.3 12.0 28.4 53.7 77.5 91.2 95.6 97.8 99.1
                  1.0 2.5 5.1 10.0 25.1 50.2 75.1 89.9 94.9 97.4 99.0
## Skovgaard W*
## Skovgaard W** 1.0 2.5 5.1 10.0 25.1 50.2 75.1 89.9 94.9 97.4 99.0
## directional
                  1.0 2.5 5.1 10.0 25.2 50.3 75.2 90.1 95.0 97.5 99.0
## first-order
                  1.8 3.9 7.2 13.5 30.4 56.1 79.3 92.0 96.0 98.1 99.2
## Skovgaard W*
                 1.1 2.6 5.0 10.0 24.6 49.6 74.6 89.6 94.7 97.3 98.9
## Skovgaard W** 1.0 2.5 5.0 9.9 24.5 49.5 74.5 89.5 94.7 97.2 98.9
                  1.0 2.6 5.1 10.1 25.0 50.3 75.4 90.2 95.0 97.5 99.0
## directional
## first-order
                  2.5 5.5 9.8 17.4 36.2 62.2 83.3 94.0 97.2 98.6 99.5
                 0.8 2.1 4.3 8.8 22.4 46.4 71.7 87.8 93.6 96.6 98.5
## Skovgaard W*
## Skovgaard W** 0.8 2.1 4.2 8.6 22.0 45.9 71.2 87.5 93.4 96.4 98.5
## directional
                  1.0 2.5 4.9 10.0 25.0 50.3 75.3 90.2 95.1 97.5 99.0
## first-order
                  3.3 6.9 12.0 20.6 40.8 66.2 85.9 95.2 97.8
                                                               99.0 99.6
                  0.7 1.8 3.7 7.8 20.7 43.7 69.1 86.3 92.6 96.1 98.2
## Skovgaard W*
                 0.7 1.8 3.6 7.5 20.1 42.8 68.2 85.7 92.2 95.8 98.1
## Skovgaard W**
                  1.0 2.4 4.9 9.9 25.2 50.0 75.0 90.1 95.1 97.5 99.0
## directional
The following code chunk uses the image dir_MDq30_res.rda to reproduce Figure 3 of the main text.
plot3.sim <- function(out, out1, out2, out3, step = 10) {</pre>
  index <- seq(1, length(out$first.order), by = step)</pre>
  par(mfrow = c(2, 2), cex = 0.6, mai = c(0.4, 0.8, 0.15, 0), mex = 0.8,
     pty = "s", oma = c(0.2, 0.2, 0.2, 0.2), mgp = c(3, 1, 0))
  x <- sort(out$first.order)</pre>
  x <- x[index]
  plot(ppoints(x), x, xlab = ' ', ylab = '', pch = ".", axes = FALSE, type ="l",
       xlim = c(0, 1), lwd = 1.5, ylim = c(0, 1), col = "chocolate2", lty = "dotdash")
  abline(0, 1, col ="grey")
  x <- sort(out$directional)</pre>
  x <- x[index]
  lines(ppoints(x), x, col = "cornflowerblue", lwd = 1.5)
  x <- sort(out$skovgaard1)</pre>
  x <- x[index]
  lines(ppoints(x), x, col = "seagreen3",
        lwd = 1.5, lty = "dashed")
  x <- sort(out$skovgaard2)
  x <- x[index]
  lines(ppoints(x), x, col = "seagreen4",
        lty = "longdash", lwd = 1.5)
  box()
  axis(2)
  axis(1)
  title("vs MD(2)", line = 2, cex.main = 1.5)
  title(ylab = "p-value", cex.lab = 1, line = 3)
  x <- sort(out1$first.order)</pre>
  x <- x[index]
  plot(ppoints(x), x, xlab = ' ', ylab = '', pch = ".", axes = FALSE, type = "l",
       xlim = c(0, 1), lwd = 1.5, ylim = c(0, 1), lty = "dotdash", col = "chocolate2")
  abline(0, 1, col ="grey")
```

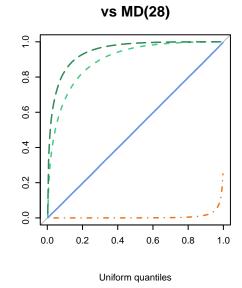
```
x <- sort(out1$skovgaard1)</pre>
x \leftarrow x[index]
lines(ppoints(x), x, col = "seagreen3",
      lwd = 1.5, lty = "dashed")
x <- sort(out1$skovgaard2)</pre>
x <- x[index]
lines(ppoints(x), x, col = "seagreen4",
      lty = "longdash", lwd = 1.5)
x <- sort(out1$directional)</pre>
x <- x[index]
lines(ppoints(x), x, col = "cornflowerblue", lwd = 1.5)
box()
axis(2)
axis(1)
title("vs MD(9)", line = 2, cex.main = 1.5)
x <- sort(out2$first.order)</pre>
x <- x[index]
plot(ppoints(x), x, xlab = ' ', ylab = '', pch = ".", axes = FALSE, type = "l",
     xlim = c(0, 1), lwd = 1.5, ylim = c(0, 1), lty = "dotdash", col = "chocolate2")
abline(0, 1, col ="grey")
x <- sort(out2$skovgaard1)</pre>
x <- x[index]
lines(ppoints(x), x, col = "seagreen3",
      lwd = 1.5, lty = "dashed")
x <- sort(out2$skovgaard2)</pre>
x <- x[index]
lines(ppoints(x), x, col = "seagreen4",
      lty = "longdash", lwd = 1.5)
x <- sort(out2$directional)</pre>
x \leftarrow x[index]
lines(ppoints(x), x, col = "cornflowerblue", lwd = 1.5)
box()
axis(2)
axis(1)
title(xlab = "Uniform quantiles", cex.lab = 1, line = 5)
title("vs MD(18)", line = 2, cex.main = 1.5)
title(ylab = "p-value", cex.lab = 1, line = 3)
x <- sort(out3$first.order)</pre>
x <- x[index]
plot(ppoints(x), x, xlab = ' ', ylab = '', pch = ".", axes = FALSE, type = "l",
     xlim = c(0, 1), lwd = 1.5, ylim = c(0, 1), lty = "dotdash", col = "chocolate2")
abline(0, 1, col ="grey")
x <- sort(out3$skovgaard1)</pre>
x <- x[index]
lines(ppoints(x), x, col = "seagreen3",
      lwd = 1.5, lty = "dashed")
x <- sort(out3$skovgaard2)</pre>
x \leftarrow x[index]
lines(ppoints(x), x, col = "seagreen4",
      lty = "longdash", lwd = 1.5)
x <- sort(out3$directional)</pre>
```

```
x <- x[index]
lines(ppoints(x), x, col = "cornflowerblue", lwd = 1.5)
box()
axis(2)
axis(1)
title(xlab = "Uniform quantiles", cex.lab = 1, line = 5)
title("vs MD(28)", line = 2, cex.main = 1.5)
}
setwd(res_dir)
load("dir_MDq30_res.rda")
plot3.sim(resMD2, resMD9, resMD18, resMD28, step = 500)</pre>
```









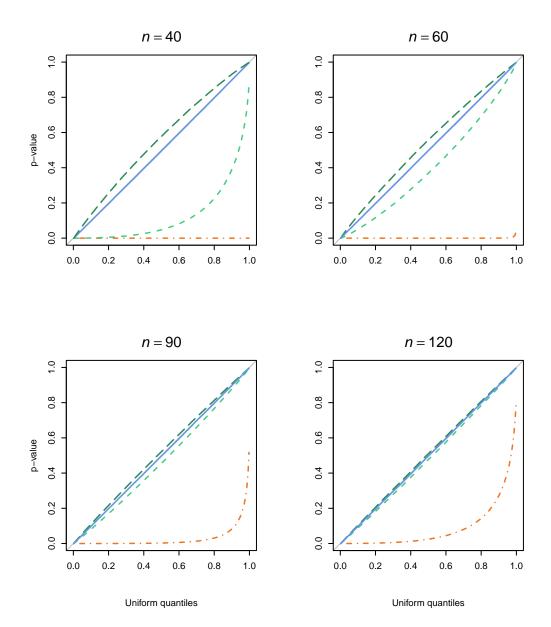
2.3 Scenario 3

The following two code chunks use the image dir_block_res.rda to reproduce Table 4 and Figure 4, respectively, of the main text. The file dir_block_res.rda results by running the script simulations3.R, available in the supplementary code archive.

```
setwd(res_dir)
load("dir block res.rda")
res40n <- round(summary.sim(res40n), 3)
res60n <- round(summary.sim(res60n), 3)</pre>
res90n <- round(summary.sim(res90n), 3)
res120n <- round(summary.sim(res120n), 3)</pre>
tab4 <- rbind(res40n[2:5,], res60n[2:5,],
              res90n[2:5,], res120n[2:5,])
round(tab4 * 100, digits = 1)
##
                        [,2]
                              [,3]
                                    [,4] [,5] [,6] [,7]
                                                             [,8] [,9] [,10] [,11]
                  [,1]
## first-order
                 100.0 100.0 100.0 100.0 100.0 100.0 100.0 100.0 100.0 100.0 100.0
                              50.6
                                    63.5 81.7
                                                93.7
                                                       98.5 99.7
## Skovgaard W*
                  27.1
                        39.1
                                                                   99.9 100.0 100.0
## Skovgaard W**
                   0.7
                         1.7
                               3.4
                                     7.2
                                          19.3
                                                 42.0
                                                       68.0
                                                             85.9
                                                                   92.5
                                                                         96.1
                                     10.1 25.2 50.2 75.2 90.0 94.9
## directional
                   1.0
                         2.5
                               5.0
                                                                        97.4
## first-order
                  98.4
                        99.3
                              99.7
                                     99.9 100.0 100.0 100.0 100.0 100.0 100.0 100.0
## Skovgaard W*
                   2.4
                         5.3
                               9.6
                                     17.3
                                           36.6
                                                62.9
                                                       84.2 94.6 97.6 98.9
## Skovgaard W**
                   0.6
                         1.7
                               3.5
                                     7.5
                                           20.4
                                                 43.9
                                                       70.3
                                                             87.4
                                                                   93.5
                                                                          96.7
## directional
                   1.0
                         2.5
                               5.0
                                     10.0
                                           25.1
                                                50.1
                                                       75.2 90.2 95.1
                                                                        97.6 99.0
## first-order
                        77.1
                              85.0
                                     91.5
                                           97.3
                                                 99.4
                                                       99.9 100.0 100.0 100.0 100.0
                  65.9
## Skovgaard W*
                   1.3
                         3.2
                               6.1
                                     12.0
                                           28.5
                                                 54.2
                                                       78.2
                                                             91.7
                                                                   96.0
                                                                         98.1
                                           23.0
                                                 47.6
                                                       73.2
                                                             89.0
## Skovgaard W**
                   0.8
                         2.1
                               4.3
                                      8.9
                                                                   94.5
                                                                          97.2
## directional
                                          25.0 50.1 75.1
                                                             90.1
                                                                   95.1 97.6 99.0
                   0.9
                         2.5
                               5.0
                                    10.1
## first-order
                  36.6
                        50.0
                              61.6
                                     73.6
                                           88.6
                                                 96.7
                                                       99.3
                                                             99.9 100.0 100.0 100.0
                                                 52.2
                                                       76.5
                                                             90.9
                                                                   95.5 97.8
## Skovgaard W*
                   1.1
                         2.9
                               5.6
                                     11.0
                                           26.8
## Skovgaard W**
                   0.9
                         2.3
                               4.6
                                      9.4
                                           24.0
                                                48.7
                                                       73.9
                                                             89.4
                                                                   94.6 97.3
                                                                                98.9
## directional
                               5.0
                                    10.1 25.1 50.1 75.0
                                                             90.0 95.0 97.5 99.0
                   1.0
                         2.5
plot4.sim <- function(out, out1, out2, out3, step = 10) {</pre>
  index <- seq(1, length(out$first.order), by = step)</pre>
  par(mfrow = c(2, 2), cex = 0.6, mai = c(0.4, 0.8, 0.15, 0), mex = 0.8,
      pty = "s", oma = c(0.2, 0.2, 0.2, 0.2), mgp = c(3, 1, 0))
  x <- sort(out$first.order)</pre>
  x <- x[index]
  plot(ppoints(x), x, xlab = ' ', ylab = '', pch = ".", axes = FALSE, type ="l",
       xlim = c(0, 1), lwd = 1.5, ylim = c(0, 1), col = "chocolate2", lty = "dotdash")
  abline(0, 1, col ="grey")
  x <- sort(out$directional)</pre>
  x \leftarrow x[index]
  lines(ppoints(x), x, col = "cornflowerblue", lwd = 1.5)
  x <- sort(out$skovgaard1)
  x <- x[index]
  lines(ppoints(x), x, col = "seagreen3",
        lwd = 1.5, lty = "dashed")
  x <- sort(out$skovgaard2)</pre>
  x \leftarrow x[index]
  lines(ppoints(x), x, col = "seagreen4",
```

```
lty = "longdash", lwd = 1.5)
box()
axis(2)
axis(1)
title(expression(italic(n) == 40), line = 2, cex.main = 1.5)
title(ylab = "p-value", cex.lab = 1, line = 3)
x <- sort(out1$first.order)</pre>
x <- x[index]
plot(ppoints(x), x, xlab = ' ', ylab = '', pch = ".", axes = FALSE, type = "l",
     xlim = c(0, 1), lwd = 1.5, ylim = c(0, 1), lty = "dotdash", col = "chocolate2")
abline(0, 1, col ="grey")
x <- sort(out1$skovgaard1)</pre>
x \leftarrow x[index]
lines(ppoints(x), x, col = "seagreen3",
      lwd = 1.5, lty = "dashed")
x <- sort(out1$skovgaard2)</pre>
x <- x[index]
lines(ppoints(x), x, col = "seagreen4",
      lty = "longdash", lwd = 1.5)
x <- sort(out1$directional)
x \leftarrow x[index]
lines(ppoints(x), x, col = "cornflowerblue", lwd = 1.5)
box()
axis(2)
axis(1)
title(expression(italic(n) == 60), line = 2, cex.main = 1.5)
x <- sort(out2$first.order)</pre>
x <- x[index]
plot(ppoints(x), x, xlab = ' ', ylab = '', pch = ".", axes = FALSE, type = "l",
     x = c(0, 1), lwd = 1.5, ylim = c(0, 1), lty = "dotdash", col = "chocolate2"
abline(0, 1, col ="grey")
x <- sort(out2$skovgaard1)</pre>
x <- x[index]
lines(ppoints(x), x, col = "seagreen3",
      lwd = 1.5, lty = "dashed")
x <- sort(out2$skovgaard2)</pre>
x \leftarrow x[index]
lines(ppoints(x), x, col = "seagreen4",
      lty = "longdash", lwd = 1.5)
x <- sort(out2$directional)
x \leftarrow x[index]
lines(ppoints(x), x, col = "cornflowerblue", lwd = 1.5)
box()
axis(2)
axis(1)
title(xlab = "Uniform quantiles", cex.lab = 1, line = 5)
title(expression(italic(n) == 90), line = 2, cex.main = 1.5)
title(ylab = "p-value", cex.lab = 1, line = 3)
x <- sort(out3$first.order)</pre>
x \leftarrow x[index]
```

```
plot(ppoints(x), x, xlab = ' ', ylab = '', pch = ".", axes = FALSE, type = "l",
       xlim = c(0, 1), lwd = 1.5, ylim = c(0, 1), lty = "dotdash", col = "chocolate2")
  abline(0, 1, col ="grey")
  x <- sort(out3$skovgaard1)</pre>
  x <- x[index]</pre>
  lines(ppoints(x), x, col = "seagreen3",
        lwd = 1.5, lty = "dashed")
  x <- sort(out3$skovgaard2)</pre>
  x <- x[index]</pre>
  lines(ppoints(x), x, col = "seagreen4",
        lty = "longdash", lwd = 1.5)
  x <- sort(out3$directional)</pre>
  x <- x[index]
  lines(ppoints(x), x, col = "cornflowerblue", lwd = 1.5)
  box()
  axis(2)
  axis(1)
 title(xlab = "Uniform quantiles", cex.lab = 1, line = 5)
 title(expression(italic(n) == 120), line = 2, cex.main = 1.5)
setwd(res_dir)
load("dir_block_res.rda")
plot4.sim(res40n, res60n, res90n, res120n, step = 500)
```



3 Applications

This section provides the R code that reproduces the outputs of Section 5 in the paper. The functions to be used in order to run the following analyses can be found in the script simulation_functions.R in the archive.

3.1 Cow data

The code chunk below loads the file cowdata.txt (Kenward 1987) from the archive and reproduces the numerical results regarding the veterinary trial on cattle reported in the main text.

```
setwd(res_dir)
source("simulation_functions.R")
```

##

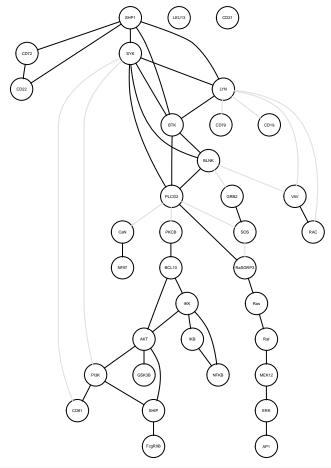
```
## Attaching package: 'plyr'
## The following object is masked from 'package:graph':
##
##
      join
## Loading required package: foreach
## Loading required package: iterators
## Loading required package: parallel
cow <- read.table("cowdata.txt", skip = 7, header = TRUE)</pre>
names(cow)[-(1:2)] <- paste("T", 1:11, sep = '')
cow.times \leftarrow cow[, -(1:2)]
cow.timesA \leftarrow cow[cow$G == "A", -(1:2)]
cow.timesB \leftarrow cow[cow$G == "B", -(1:2)]
## group A
cowsA <- cow.timesA</pre>
S <- cov(cowsA)
n <- nrow(cowsA)</pre>
# unsaturated model under H1
# MD(3)
T6*T7*T8*T9 + T7*T8*T9*T10 + T8*T9*T10*T11)
model1 <- fitConGraph(G1, S, n)</pre>
# model under HO
# MD(1)
T10*T11)
model0 <- fitConGraph(G0, S, n)</pre>
### tests
## W
lrt(model0, model1, n)
## $Woss
## [1] 28.38429
## $pvalue
## [1] 0.04064335
## w* and w**
Wstar(model0, model1, n, G1 = G1)
## $wstar1
           [,1]
## [1,] 22.97676
##
## $p.value1
##
            [,1]
## [1,] 0.1500088
##
```

```
## $wstar2
##
                                                                           [,1]
## [1,] 22.6913
##
## $p.value2
##
                                                                                     [,1]
## [1,] 0.1595728
## directional p-value
dir_p(model0, model1, n, G1 = G1)$pvalue
## [1] 0.1113944
## group B
cowsB <- cow.timesB</pre>
S <- cov(cowsB)
n <- nrow(cowsB)</pre>
# unsaturated model under H1
# MD(3)
\texttt{G1} \; \leftarrow \; \texttt{UG}(\; \sim \; \texttt{T1} \times \texttt{T2} \times \texttt{T3} \times \texttt{T4} \; + \; \texttt{T2} \times \texttt{T3} \times \texttt{T4} \times \texttt{T5} \; + \; \texttt{T3} \times \texttt{T4} \times \texttt{T5} \times \texttt{T6} \; + \; \texttt{T4} \times \texttt{T5} \times \texttt{T6} \times \texttt{T7} \; + \; \texttt{T5} \times \texttt{T6} \times \texttt{T7} \times \texttt{T8} \; + \; \texttt{T5} \times \texttt{T6} \times \texttt{T7} \times \texttt{T8} \; + \; \texttt{T5} \times \texttt{T6} \times \texttt{T7} \times \texttt{T8} \; + \; \texttt{T5} \times \texttt{T6} \times \texttt{T7} \times \texttt{T8} \; + \; \texttt{T5} \times \texttt{T6} \times \texttt{T7} \times \texttt{T8} \; + \; \texttt{T5} \times \texttt{T6} \times \texttt{T7} \times \texttt{T8} \; + \; \texttt{T5} \times \texttt{T6} \times \texttt{T7} \times \texttt{T8} \times \texttt{T9} \times \texttt{T9
                                                                           T6*T7*T8*T9 + T7*T8*T9*T10 + T8*T9*T10*T11)
model1 <- fitConGraph(G1, S, n)</pre>
# model under HO
 # MD(1)
GO <- UG(~ T1*T2 + T2*T3 + T3*T4 + T4*T5 + T5*T6 + T6*T7 + T7*T8 + T8*T9 + T9*T10 +
                                                                           T10*T11)
model0 <- fitConGraph(G0, S, n)</pre>
### tests
lrt(model0, model1, n)
## $Woss
## [1] 31.89548
##
## $pvalue
## [1] 0.01550424
## w* and w**
Wstar(model0, model1, n, G1 = G1)
## $wstar1
##
                                                                                   [,1]
## [1,] 30.05542
## $p.value1
                                                                                                 [,1]
## [1,] 0.02594814
## $wstar2
##
                                                                                   [,1]
## [1,] 30.02809
##
```

```
## $p.value2
## [,1]
## [1,] 0.02614321
## directional p-value
dir_p(model0, model1, n, G1 = G1)$pvalue
## [1] 0.02887994
```

3.2 Leukemia data

The next code chunks load from the archive the file BCRdata.rda, also accessible via the R package topologyGSA (Massa and Sales 2016), and reproduces Figure 5 and the numerical results regarding the genetic study on lymphocytic leukemia reported in the main text.



```
G1 <- as(ug_bcell, "matrix")</pre>
n \leftarrow nrow(y2)
S <- cov(y2)
# H1: unsaturated model
model1 <- fitConGraph(G1, S, n)</pre>
# HO: simpler BCR signaling pathway
GO <- G1
GO["SYK", "PI3K"] <- GO["SYK", "CD81"] <- GO["PI3K", "SYK"] <-
 GO["CD81", "SYK"] <- 0
GO["LYN", "CD79"] <- GO["LYN", "CD19"] <- GO["LYN", "VAV"] <- GO["LYN", "RAC"] <-
 GO[ "CD79", "LYN"] <- GO[ "CD19", "LYN"] <- GO["VAV", "LYN"] <-
  GO["RAC", "LYN"] <- 0
GO["BLNK", "VAV"] <- GO["BLNK", "GRB2"] <- GO["VAV", "BLNK"] <-
  GO["GRB2", "BLNK"] <- 0
GO["PLCG2", "CaN"] <- GO["PLCG2", "PKCB"] <- GO["PLCG2", "SOS"] <-
  GO["CaN", "PLCG2"] <- GO["PKCB", "PLCG2"] <- GO["SOS", "PLCG2"] <- 0
GO["RaSGRP3", "SOS"] <- GO["SOS", "RaSGRP3"] <- 0
model0 <- fitConGraph(G0, S, n)</pre>
## W
lrt(model0, model1, n)
```

```
## $Woss
## [1] 33.51951
##
## $pvalue
## [1] 0.0008027688
## w* and w**
Wstar(model0, model1, n, G1 = G1)
## $wstar1
##
            [,1]
## [1,] 32.17165
##
## $p.value1
##
               [,1]
## [1,] 0.001301832
##
## $wstar2
##
            [,1]
## [1,] 32.15782
##
## $p.value2
##
               [,1]
## [1,] 0.001308256
## directional p-value
dir_p(model0, model1, n, G1 = G1)$pvalue
## [1] 0.001394133
```

References

Kenward, Michael G. 1987. "A Method for Comparing Profiles of Repeated Measurements." J. R. Statist. Soc. C 36: 296–308.

Massa, Sofia, and Gabriele Sales. 2016. topologyGSA: Gene Set Analysis Exploiting Pathway Topology. https://CRAN.R-project.org/package=topologyGSA.

R Core Team. 2020. R: A Language and Environment for Statistical Computing. Vienna, Austria: R Foundation for Statistical Computing. https://www.R-project.org/.