## Problem 3

```
set.seed(123)
(K \leftarrow cbind(c(10,7,7,0),c(7,20,0,7),c(7,0,30,7),c(0,7,7,40)))
##
         [,1] [,2] [,3] [,4]
## [1,]
           10
                  7
## [2,]
             7
                 20
                        0
                              7
             7
                              7
## [3,]
                       30
## [4,]
                             40
\texttt{data} \leftarrow \texttt{as.data.frame(mvrnorm(n=10000,mu=c(0,0,0,0),Sigma=solve(K)))}
colnames(data) <- c("X1","X2","X3","X4")</pre>
```

## Conditional independency

It represents following independencies:

$$X_1 \perp \!\!\! \perp X_4 | X_2, X_3$$
 and  $X_2 \perp \!\!\! \perp X_3 | X_1, X_4$ 

The corresponding graph

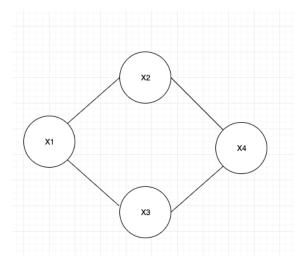


Figure 1: fig 3-1

Fit with OLS

## Residuals:

```
lmodel = lm(X1 ~ X4 + X2 + X3, data=data)
summary(lmodel)

##
## Call:
## lm(formula = X1 ~ X4 + X2 + X3, data = data)
##
```

```
Median
                 1Q
## -1.36729 -0.21127 0.00304 0.21389 1.20994
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
                                     0.616
## (Intercept) 0.001934
                          0.003141
                                              0.538
                                              0.692
## X4
               0.007927
                           0.020037
                                     0.396
## X2
               -0.682729
                           0.012203 -55.950
                                              <2e-16 ***
                          0.015540 -44.741
## X3
              -0.695282
                                              <2e-16 ***
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.3141 on 9996 degrees of freedom
## Multiple R-squared: 0.4564, Adjusted R-squared: 0.4563
## F-statistic: 2798 on 3 and 9996 DF, p-value: < 2.2e-16
```

X4 is not significant while X2 and X3 are. This means X4 and X1 is independent given X2 and X3.

```
lmodel = lm(X2 ~ X3 + X1 + X4, data=data)
summary(lmodel)
```

```
##
## Call:
## lm(formula = X2 ~ X3 + X1 + X4, data = data)
## Residuals:
       Min
                  10
                      Median
                                    30
## -0.90282 -0.15318  0.00188  0.15342  0.85952
##
## Coefficients:
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) 0.001141
                           0.002247
                                      0.508
                                               0.612
## X3
               0.012316
                           0.012177
                                      1.011
                                               0.312
## X1
              -0.349303
                           0.006243 -55.950
                                              <2e-16 ***
## X4
              -0.352810
                           0.013891 -25.398
                                              <2e-16 ***
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 0.2246 on 9996 degrees of freedom
## Multiple R-squared: 0.3841, Adjusted R-squared: 0.3839
## F-statistic: 2078 on 3 and 9996 DF, p-value: < 2.2e-16
```

X3 is not significant while X1 and X4 are. This means X2 and X3 is independent given X1 and X4.

Fit with gRim

cannot install package, remember to do it later

```
glist <- list( 'X1', 'X2', 'X3', 'X4' )</pre>
ddd <- cov.wt(data, method="ML")</pre>
fit <- ggmfit(ddd$cov, ddd$n.obs, glist) # Estimate parameters using IPF
fit$K
```

## Х1 X2 ХЗ Х4

```
## X1 5.513255 0.00000 0.00000 0.00000
## X2 0.000000 12.21077 0.00000 0.00000
## X3 0.000000 0.00000 20.54787 0.00000
## X4 0.000000 0.00000 0.00000 33.73434
```

It did not work. K has more elements equal to zero than the original one.

## Problem 4

**a**)

Correlation represented by graph

 $X_1 \perp \!\!\! \perp X_3 \ X_1 \perp \!\!\! \perp X_4 \ X_2 \perp \!\!\! \perp X_4$  and they are not independent given  $X_2$ 

Correlation Matrix

```
solve(Sig)
```

```
## [,1] [,2] [,3] [,4]

## [1,] 0.5427350 0.4487179 -0.2094017 -0.2094017

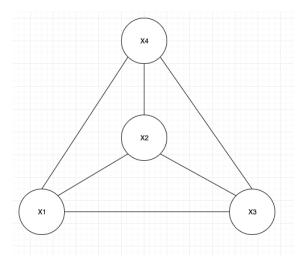
## [2,] 0.4487179 0.9615385 -0.4487179 -0.4487179

## [3,] -0.2094017 -0.4487179 0.5427350 0.2094017

## [4,] -0.2094017 -0.4487179 0.2094017 0.5427350
```

b)

The moralized graph looks like



Every element of the precision matrix is not equal to 0 because every vertex is adjacent to another one. It does not imply the correlation suggested in (a)

**c**)

```
glist <- list( 'X1', 'X2', 'X3', 'X4' )
ddd <- cov.wt(data, method="ML")
fit <- ggmfit(ddd$cov, ddd$n.obs, glist) # Estimate parameters using IPF
solve(fit$K)

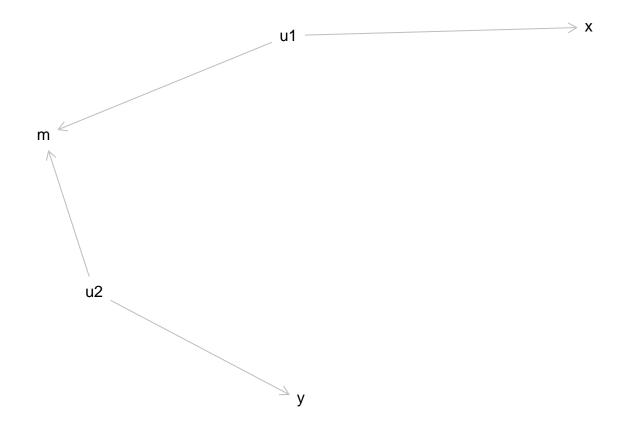
## X1 X2 X3 X4
## X1 2.991722 0.000000 0.000000 0.000000
## X2 0.000000 2.959982 0.000000 0.000000
## X3 0.000000 0.000000 2.966525 0.000000
## X4 0.000000 0.000000 0.000000 3.077011</pre>
```

It is different from original covariance matrix as the elements on the diagonal are not the same.

## Problem 5

```
g <- dagitty( "dag{ x <- u1; u1 -> m <- u2; u2 -> y }" ) df = simulateSEM(g, N = 1000, standardized = TRUE) plot(g)
```

## Plot coordinates for graph not supplied! Generating coordinates, see ?coordinates for how to set you



```
reg = lm(y \sim x + m, data = df)
summary(reg)
##
## Call:
## lm(formula = y \sim x + m, data = df)
## Residuals:
                 1Q Median
## -2.89042 -0.68302 -0.03076 0.67329 3.15642
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.02718
                          0.03113 -0.873 0.3829
## x
              -0.06637
                          0.03043 -2.181
                                            0.0294 *
## m
               0.17789
                          0.03119
                                   5.703 1.55e-08 ***
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.9843 on 997 degrees of freedom
## Multiple R-squared: 0.03459, Adjusted R-squared: 0.03265
## F-statistic: 17.86 on 2 and 997 DF, p-value: 2.398e-08
reg = lm(y \sim u2, data = df)
summary(reg)
```

```
##
## Call:
## lm(formula = y \sim u2, data = df)
##
## Residuals:
##
      \mathtt{Min}
               1Q Median
                                3Q
                                      Max
## -2.6370 -0.5855 -0.0396 0.6020 3.1702
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.01729
                           0.02861 -0.604
                                             0.546
                           0.02941 14.999 <2e-16 ***
               0.44110
## u2
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 0.9045 on 998 degrees of freedom
## Multiple R-squared: 0.184, Adjusted R-squared: 0.1831
## F-statistic: 225 on 1 and 998 DF, p-value: < 2.2e-16
if (!require("BiocManager", quietly = TRUE))
    install.packages("BiocManager")
BiocManager::install("graph")
## Bioconductor version 3.15 (BiocManager 1.30.18), R 4.2.1 (2022-06-23)
## Warning: package(s) not installed when version(s) same as current; use 'force = TRUE' to
   re-install: 'graph'
```