

Project 1

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Problem 3: Learning Bayesian networks from protein data

Preparation to run the code

Setting seed for reproducibility and loading packages.

```
library("GGally")
library("BiDAG")
library("igraph")
```

Loading data

```
data <- read.table("2005_sachs_2_cd3cd28icam2_log_std.csv", sep=",", header=TRUE)
```

Variables and observations

```
num_variables <- ncol(data)
num_observations <- nrow(data)
```

Visualisation of transformed data

```
png(file="plot.png", width=650, height=1000)
ggpairs(data, progress=FALSE)
dev.off()
```

```
## pdf
## 2
```

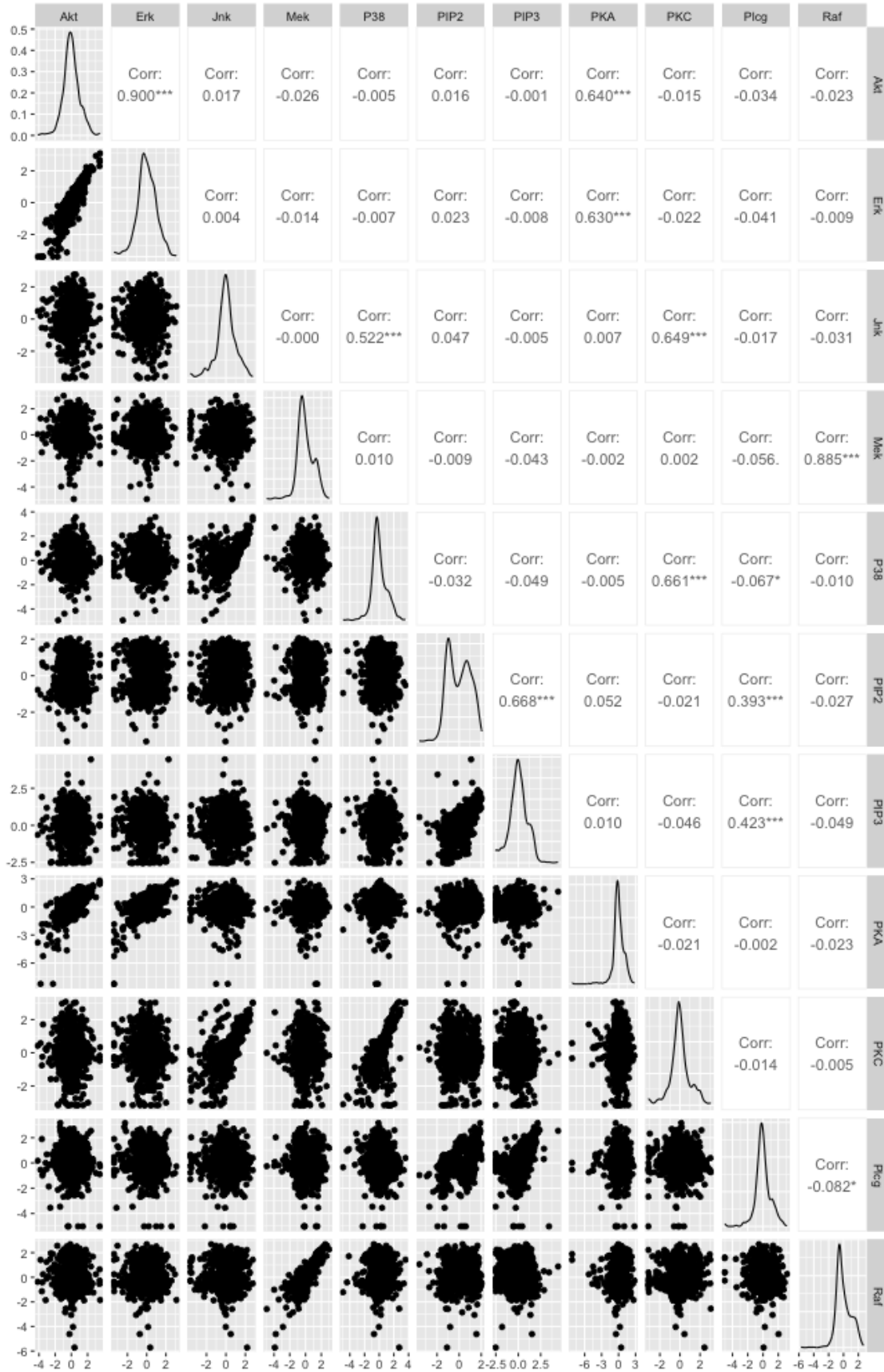


Figure 1: Plot of transformed data

Defining functions

```
splitting_data <- function(data) {  
  # Shuffling data  
  indices <- 1:nrow(data)  
  indices <- sample(length(indices))  
  
  # Splitting data  
  train_size <- ceiling(nrow(data)*0.8)  
  train_indices <- indices[1:train_size]  
  test_indices <- indices[(train_size+1):(length(indices))]  
  
  # Checking if there is no overlap  
  if(length(unique(c(train_indices, test_indices))) != length(c(train_indices, test_indices))) {  
    print("Overlap!")  
  }  
  
  train_data <- data[row.names(data) %in% train_indices, ]  
  test_data <- data[row.names(data) %in% test_indices, ]  
  
  split_data <- list("train_data"=train_data, "test_data"=test_data)  
  
  return(split_data)  
}  
  
training_BN <- function(data, bgepar) {  
  library("BiDAG")  
  init_score_par <- scoreparameters("bge", data$train_data, bgepar)  
  learnt_BN <- iterativeMCMC(init_score_par, verbose=FALSE)  
  return(learnt_BN)  
}  
  
testing_BN <- function(data, BN, bgepar) {  
  library("BiDAG")  
  test_score_par <- scoreparameters("bge", data$test_data, bgepar)  
  test_score <- scoreagainstDAG(test_score_par, BN$DAG)  
  return(mean(test_score))  
}  
  
plot_DAG <- function(BN) {  
  library("igraph")  
  g <- graph_from_adjacency_matrix(BN$DAG)  
  plot(g)  
}  
  
get_number_of_edges <- function(BN) {  
  library("igraph")  
  return(length(E(graph_from_adjacency_matrix(BN$DAG))))  
}
```

Default parameters

```

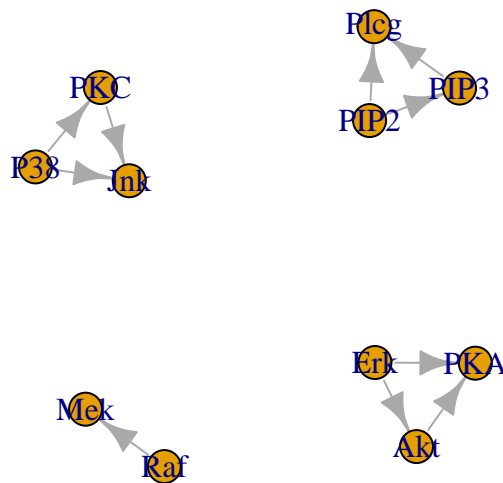
bgepar <- list(am=1, aw=NULL, edgepf=1)
split_data <- splitting_data(data)
learnt_BN <- training_BN(split_data, bgepar)

mean_test_score <- testing_BN(split_data, learnt_BN, bgepar)
print(paste0("Average BGe score on testing data: ", mean_test_score))

## [1] "Average BGe score on testing data: -12.4139751176187"

plot_DAG(learnt_BN)

```



Different parameters

```

library(parallel)
library(doParallel)

# Set the number of cores to use
num_cores <- detectCores()

# Register parallel backend
cl <- makeCluster(num_cores)
registerDoParallel(cl)

ams <- c(10−5, 10−3, 10−1, 10, 102)

res <- foreach(am = ams, .combine=c) %dopar% {

```

```

set.seed(42)
bgepar <- list(am=am, aw=NULL, edgepf=1)
numbers_of_edges <- c()
mean_test_scores <- c()
for(i in 1:10) {
  RNGkind("L'Ecuyer-CMRG")
  split_data <- splitting_data(data)
  learnt_BN <- training_BN(split_data, bgepar)
  number_of_edges <- get_number_of_edges(learnt_BN)
  numbers_of_edges <- append(numbers_of_edges, number_of_edges)
  mean_test_score <- testing_BN(split_data, learnt_BN, bgepar)
  mean_test_scores <- append(mean_test_scores, mean_test_score)
}
return(c(am, mean(numbers_of_edges), mean(mean_test_scores)))
}

stopCluster(cl)

res_m <- t(matrix(data=res, nrow=5, ncol=3, byrow=TRUE))
rownames(res_m) <- c("Parameter am", "Average number of edges", "Average BGe score of the test data")
print(res_m)

```

```

##                                [,1]      [,2]      [,3]      [,4]      [,5]
## Parameter am                  0.00001    0.00100    0.10000    10.00000   100.00000
## Average number of edges       7.00000    7.00000    9.10000    13.40000   16.70000
## Average BGe score of the test data -12.73416 -12.73416 -12.70972 -12.80457 -13.58552

```

Retraining the best BN

The best BN (based on the highest BGe score) was the one with $\alpha_m = 10^{-1}$.

```

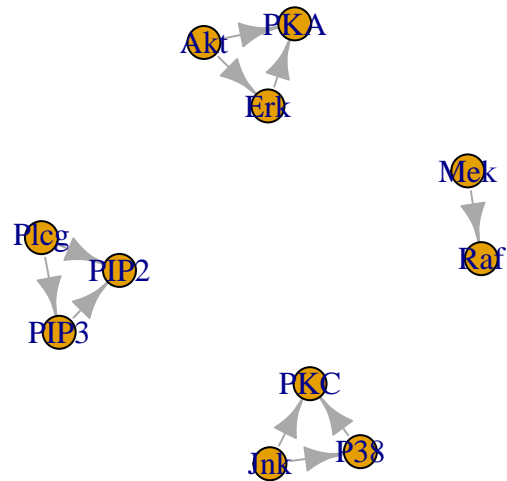
set.seed(42)
bgepar <- list(am=10^(-1), aw=NULL, edgepf=1)
split_data <- splitting_data(data)
learnt_BN <- training_BN(split_data, bgepar)

mean_test_score <- testing_BN(split_data, learnt_BN, bgepar)
print(paste0("Average BGe score on testing data: ", mean_test_score))

## [1] "Average BGe score on testing data: -12.1209831813562"

plot_DAG(learnt_BN)

```



Render this .rmd into a pdf

```
library(rmarkdown)
render("1.Rmd", pdf_document(TRUE), "1.pdf") # TRUE adds table of content
```