Basic R tutorial

February 22, 2024

Contents

roblem 3: Learning Bayesian networks from protein data	1
Preparation to run the code	1
Loading data	1
Variables and observations	1
Visualisation of transformed data	1
Defining functions	2
Default parameters	
Different parameters	
Render this .rmd into a pdf	4

Problem 3: Learning Bayesian networks from protein data

Preparation to run the code

Setting seed for reproducibility and loading packages.

```
set.seed(42)
library("GGally")
library("BiDAG")
library("igraph")
```

Loading data

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

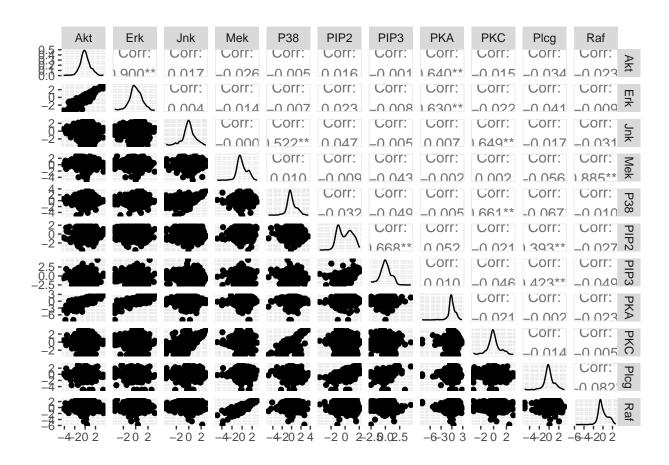
Variables and observations

```
num_variables <- ncol(data)
num_observations <- nrow(data)</pre>
```

Visualisation of transformed data

```
TODO: must resize the plot
```

```
pair_plot <- ggpairs(data, progress=FALSE)
pair_plot</pre>
```



Defining functions

```
splitting_data <- function(data) {</pre>
    # Shuffling data
    indices <- 1:nrow(data)</pre>
    indices <- sample(length(indices))</pre>
    # Splitting data
    train_size <- ceiling(nrow(data)*0.8)</pre>
    train_indices <- indices[1:train_size]</pre>
    test_indices <- indices[(train_size+1):(length(indices))]</pre>
    # 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, ]</pre>
    test_data <- data[row.names(data) %in% test_indices, ]</pre>
    split_data <- list("train_data"=train_data, "test_data"=test_data)</pre>
    return(split_data)
}
```

```
training_BN <- function(data, bgepar) {
    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) {
    test_score_par <- scoreparameters("bge", data$test_data, bgepar)
    test_score <- scoreagainstDAG(test_score_par, BN$DAG)
    return(mean(test_score))
}

get_number_of_edges <- function(BN) {
    return(length(E(graph_from_adjacency_matrix(BN$DAG))))
}</pre>
```

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))</pre>
```

[1] "Average BGe score on testing data: -12.7214725878423"

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)</pre>
```

TODO: compute average number of edges

TODO: make pretty-printing of results

```
ams <- c(10^(-5), 10^(-3), 10^(-1), 10, 10^2)

foreach(am = ams) %dopar% {
   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")
      library("BiDAG")
      library("igraph")
      split_data <- splitting_data(data)</pre>
```

```
learnt_BN <- training_BN(split_data, bgepar)</pre>
        number_of_edges <- get_number_of_edges(learnt_BN)</pre>
        numbers_of_edges <- append(numbers_of_edges, number_of_edges)</pre>
        mean_test_score <- testing_BN(split_data, learnt_BN, bgepar)</pre>
        mean_test_scores <- append(mean_test_scores, mean_test_score)</pre>
    print(paste(am, mean(mean_test_scores), mean(numbers_of_edges)))
}
## [[1]]
## [1] "1e-05 -12.7577737541497 7"
##
## [[2]]
## [1] "0.001 -12.6761057931288 7"
## [[3]]
## [1] "0.1 -12.6381565046395 9.4"
## [[4]]
## [1] "10 -12.5399738370197 13"
##
## [[5]]
## [1] "100 -13.4446760922504 15"
stopCluster(cl)
```

Render this .rmd into a pdf

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