University of Duisburg-Essen Faculty of Business Administration and Economics Chair of Econometrics



A Functional Approach to (Parallelised) Monte Carlo Simulation

Advanced R for Econometricians

Final Project

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from:

Alexander Langnau, Öcal Kaptan, Sunyoung Ji

Matriculation Number: 232907, 230914, 229979

Study Path: M.Sc. Econometircs

Reviewer: Prof. Dr. Christoph Hanck

Secondary Reviewer: M.Sc. Martin C. Arnold, M.Sc. Jens Klenke

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1 Introduction

Monte Carlo, named after a casino in Monaco, simulates complex probabilistic events using simple random events, such as the tossing of a pair of dice to simulate the casino's overall business model. In Monte Carlo computing, a pseudo-random number generator is repeatedly called which returns a real number in [0, 1], and the results are used to generate a distribution of samples that is a fair representation of the target probability distribution under study. (**Barbu**) Monte Carlo Method is combined with programming in modern research and contributes to various studies. The task in this paper is to create the wrapper functions providing a convenient interface for Monte Carlo Simulations.

2 Preprocess

2.1 Function for creating grid

create_grid is the function hat creates a parameter grid with all permutations of the given parameters. This is necessary to try all possible combinations to find the optimal parameters. This function tunes parameters to improve performance of Monte Carlo Simulation function.

```
create_grid <- function(parameters, nrep){</pre>
  input <- parameters
  storage <- list()</pre>
  name_vec <- c()</pre>
  for(i in 1:length(input)){ #1:3
    a <- as.numeric(input[[i]][[2]])</pre>
    b <- as.numeric(input[[i]][[3]])</pre>
    c <- as.numeric(input[[i]][[4]])</pre>
    output <- seq(from=a, to=b, by=c)
    storage[[i]] <- output
    name_vec[i] <- input[[i]][[1]]</pre>
  }
  grid <- expand_grid(unlist(storage[1])</pre>
                         , unlist(storage[2])
                         , unlist(storage[3])
                         , unlist(storage[4])
                         , unlist(storage[5])
                         , c(1:nrep))
  names(grid) <- c(name_vec, "rep")</pre>
```

```
return(grid)
}
create_grid() Example:
#One parameter (works)
param_list1 <- list(c("n", 10, 20, 10))</pre>
tail(create_grid(param_list1, nrep=10), 2)
## # A tibble: 2 x 2
##
         n
             rep
##
     <dbl> <int>
## 1
        20
## 2
        20
              10
tail(create_grid(param_list1, nrep=1), 2)
## # A tibble: 2 x 2
##
         n
             rep
##
     <dbl> <int>
## 1
        10
               1
## 2
        20
               1
#two parameter (works)
param_list2 <- list(c("n", 10, 20, 10)</pre>
                     ,c("mu", 0, 1, 0.25))
tail(create_grid(param_list1, nrep=10), 2)
## # A tibble: 2 x 2
##
         n
             rep
##
     <dbl> <int>
        20
               9
## 1
## 2
        20
              10
#three parameters (works)
param_list3 <- list(c("n", 10, 20, 10)
                     ,c("mu", 0, 1, 0.25)
                     ,c("sd", 0, 0.3, 0.1))
tail(create_grid(param_list3, nrep=10), 2)
```

A tibble: 2 x 4

```
##
                      sd
         n
               mu
                           rep
##
     <dbl> <dbl> <int>
## 1
         20
                1
                     0.3
                             9
## 2
         20
                     0.3
                            10
                1
#four parameters (works)
param_list4 <- list(c("n", 10, 20, 10)</pre>
                      ,c("mu", 0, 1, 0.25)
                      ,c("sd", 0, 0.3, 0.1)
                      ,c("gender", 0, 1, 1))
tail(create_grid(param_list4, nrep=5),2)
## # A tibble: 2 x 5
##
                      sd gender
         n
               mu
                                   rep
##
     <dbl> <dbl> <dbl>
                          <dbl> <int>
## 1
         20
                     0.3
## 2
         20
                     0.3
                              1
                                     5
                1
grid_4 <- create_grid(param_list4, nrep=50)</pre>
tail(grid_4,2)
## # A tibble: 2 x 5
##
                      sd gender
               mu
                                   rep
##
     <dbl> <dbl> <dbl>
                          <dbl> <int>
## 1
         20
                     0.3
                                    49
## 2
         20
                1
                     0.3
                               1
                                    50
```

2.2 Data generation function

data_generation allows users to flexibly change data while keeping the summary statistics and to choose the number of inputs by using different purrr mapping functions: map, map2, and pmap for a input, two inputs, and p inputs respectively.

In the function below, simulation means a distribution of data, and grid is a list of parameters.

```
data_generation <- function(simulation, grid){
    #this is for use inside the function

if(ncol(grid)==2){
    var1 <- c(unlist(grid[,1]))
    data <- map(var1, simulation)
    #different purrr-functions depending on how many input variables we use</pre>
```

```
}
  if(ncol(grid)==3){
    var1 <- c(unlist(grid[,1]))</pre>
    var2 <- c(unlist(grid[,2]))</pre>
    data <- map2(var1, var2, simulation)</pre>
  }
  if(ncol(grid)==4){
    var1 <- c(unlist(grid[,1]))</pre>
    var2 <- c(unlist(grid[,2]))</pre>
    var3 <- c(unlist(grid[,3]))</pre>
    list1 <- list(var1,var2,var3)</pre>
    data <- pmap(list1, .f=simulation)</pre>
  }
  return(data)
}
data_generation() Example:
grid1 <- create_grid(param_list1, nrep=3)</pre>
tail(data_generation(simulation=rnorm, grid=grid1),1)
## $n6
   [1] -0.491031166 -2.309168876 1.005738524 -0.709200763 -0.688008616
   [6] 1.025571370 -0.284773007 -1.220717712 0.181303480 -0.138891362
## [11] 0.005764186 0.385280401 -0.370660032 0.644376549 -0.220486562
## [16] 0.331781964 1.096839013 0.435181491 -0.325931586 1.148807618
grid2 <- create_grid(param_list2, nrep=3)</pre>
tail(data_generation(simulation=rnorm, grid=grid2),1)
## $n30
   [1] 1.9672673 0.8917199 0.3015793 0.7240548 2.1146485 1.5500440
##
## [7] 2.2366758 1.1390979 1.4102751 0.4415431 1.6053707
                                                                  0.4936665
## [13] -0.4205655 1.1279930 2.9458512 1.8009143 2.1652534 1.3588557
## [19] 0.3914428 0.7977591
```

Users can apply many distributions such as normal, uniform, poisson distributions by putting existing functions in r as simulation.

```
# Application to Uniform distribution
param_list_runif <- list(c("n", 10, 30, 10)</pre>
                         ,c("min", 0, 0, 0)
                         ,c("max", 1, 1, 0))
grid_unif <- create_grid(param_list_runif, nrep=3)</pre>
tail(data_generation(simulation=runif, grid=grid_unif),1)
## $n9
  [1] 0.004638151 0.277560080 0.325203143 0.588706277 0.249684701 0.043117281
## [7] 0.110678788 0.703753812 0.939021239 0.311169018 0.078492930 0.321744091
## [13] 0.624905537 0.440241850 0.801345301 0.279283805 0.570713193 0.042128012
## [19] 0.190717455 0.727086471 0.826690050 0.510721075 0.567726166 0.001155820
## [25] 0.143778103 0.865967083 0.082561061 0.244570682 0.981543157 0.577581279
# Application to Poisson distribution
param_list_rpois <- list(c("n", 10, 30, 10)</pre>
                         , c("lambda", 0, 10, 1))
grid_pois <- create_grid(param_list_rpois, nrep=3)</pre>
tail(grid_pois,2) # nrow(grid_pois) = 99
## # A tibble: 2 x 3
##
         n lambda
                    rep
##
     <dbl> <dbl> <int>
## 1
        30
               10
## 2
        30
               10
                      3
tail(data_generation(simulation=rpois, grid=grid_pois),1)
## $n99
## [1] 8 8 8 12 7 6 10 9 5 8 19 12 7 12 13 7 3 9 7 15 6 13 11 15 8
## [26] 13 9 7 9 5
```

2.3 Summary function

summary_function offers summary statistics that users can choose.

```
#summary function for one input
summary_function <- function(sum_fun, data_input){

count <- length(data_input)
summary_matrix <- matrix(nrow=count, ncol=1)

for(i in 1:count){
  input <- list(data_input[[i]])
  output <- sapply(sum_fun, do.call, input)
  summary_matrix[i] <- output
}

#output <- as.data.frame(summary_matrix)
#names(output) <- sum_fun
colnames(summary_matrix) <- sum_fun
return(summary_matrix)
}</pre>
```

summary_function Example:

```
grid_test <- create_grid(param_list3, nrep=3)
test_data <- data_generation(simulation=rnorm, grid=grid_test)
tail(summary_function(sum_fun=list("mean"), data_input=test_data),2)</pre>
```

```
## mean
## [119,] 1.03361
## [120,] 1.01786
```

2.4 Summary array funcation

The outcome of create_array_function illustrates the combination of user defined grid and the summary statistics. This function product dataframes with all permutations and results that allow, thus users can look any possible parameter regarding specific grid.

```
create_array_function <- function(comb, parameters, nrep){
  storage <- list()
  name_vec <- c()

for(i in 1:length(parameters)){
    #this creates the sequences of parameters
    a <- as.numeric(parameters[[i]][[2]])
    b <- as.numeric(parameters[[i]][[3]])
    c <- as.numeric(parameters[[i]][[4]])</pre>
```

```
output <- seq(from=a, to=b, by=c)</pre>
  storage[[i]] <- output</pre>
  name_vec[i] <- parameters[[i]][[1]]</pre>
  #this just stores the names of the variables
}
matrix.numeration <- paste("rep","=", 1:nrep, sep = "")</pre>
if(length(parameters)==1){
  comb_ordered <- comb %>% arrange(comb[,2])
  seq1 <- c(unlist(storage[1]))</pre>
  row.names <- paste(name_vec[1],"=",seq1, sep = "")</pre>
  dimension_array <- c(length(seq1), nrep)</pre>
  dim_names_list <- list(row.names, matrix.numeration)</pre>
}
if(length(parameters)==2){
  comb_ordered <- comb %>% arrange(comb[,2]) %>% arrange(comb[,3])
  seq1 <- c(unlist(storage[1]))</pre>
  seq2 <- c(unlist(storage[2]))</pre>
  row.names <- paste(name_vec[1],"=",seq1, sep = "")</pre>
  column.names <- paste(name_vec[2],"=",seq2, sep = "")</pre>
  dimension_array <- c(length(seq1), length(seq2), nrep)</pre>
  dim_names_list <- list(row.names, column.names, matrix.numeration)</pre>
}
if(length(parameters)==3){
  comb ordered <- comb %>% arrange(comb[,2]) %>%
    arrange(comb[,3]) %>% arrange(comb[,4])
  seq1 <- c(unlist(storage[1]))</pre>
  seq2 <- c(unlist(storage[2]))</pre>
  seq3 <- c(unlist(storage[3]))</pre>
  row.names <- paste(name_vec[1],"=",seq1, sep = "")</pre>
  column.names <- paste(name_vec[2],"=",seq2, sep = "")</pre>
  matrix.names1 <- paste(name_vec[3],"=",seq3, sep = "")</pre>
```

create_array_function Example:

```
# PREP TEST `create_array_function`
main_function_array_test <- function(parameters #list of parameters</pre>
                                         , nrep #number of repetitions
                                         , simulation #data genereation
                                         , sum_fun){ #summary statistics
  grid <- create_grid(parameters, nrep) #Step 1: create grid</pre>
  raw_data <- data_generation(simulation, grid) #Step 2: simlate data
  summary <- summary_function(sum_fun, data_input=raw_data) #Step 3: Summary statistics</pre>
  comb <- cbind(grid, summary) #Step 4: Combine resuluts with parameters</pre>
  array_1 <- create_array_function(comb, parameters, nrep) #Step 5: Create array</pre>
  return(comb)
}
param_list3x <- list(c("n", 10, 20, 10)</pre>
                      ,c("mu", 0, 5, 1)
                      ,c("sd", 0, 1, 1))
comb1 <- main_function_array_test(parameters=param_list3x</pre>
                                     , nrep = 1
                                     , simulation = rnorm
                                     , sum_fun="mean")
head(comb1,2)
```

n mu sd rep mean

```
## 1 10 0 0 1 0.0000000
## 2 10 0 1 1 -0.6031898
```

create_array_function(comb=comb1, parameters=param_list3x, nrep=1)

```
## , , sd=0, rep=1
##
        mu=0 mu=1 mu=2 mu=3 mu=4 mu=5
## n=10
           0
                1
                     2
                          3
## n=20
                1
           0
                     2
                          3
                               4
                                    5
##
## , , sd=1, rep=1
##
##
              mu=0
                        mu=1
                                 mu=2
                                          mu=3
                                                   mu=4
                                                             mu=5
## n=10 -0.6031898 1.1547493 1.768505 2.799209 4.297611 5.240045
## n=20 -0.1950611 0.7933902 1.609615 2.815089 4.066077 4.798390
```

3 Monte Carlo Simulation Funcion

4 Examples

5 Conclusion

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