

Decision Analysis in R for Technologies in Health

Appendix A

Microsimulation modeling for health decision sciences using R: A tutorial.

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Please cite our papers when using any of the material

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Decision Analysis in R for Technologies in Health

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Appendix A - Microsimulation modeling for health decision sciences using R: A tutorial.

Krijkamp EM, Alarid-Escudero F, Enns EA, Jalal HJ, Hunink MGM, Pechlivanoglou P.

2018

This code forms the basis for the microsimulation model of the article: Krijkamp EM, Alarid-Escudero F, Enns EA, Jalal HJ, Hunink MGM, Pechlivanoglou P. Microsimulation modeling for health decision sciences using R: A tutorial. Med Decis Making. 2018;38(3):400-22.

Please cite the article when using this code.

See GitHub for more information or code updates <https://github.com/DARTH-git/Microsimulation-tutorial>

To program this tutorial we made use of R: 3.3.0 GUI 1.68 Mavericks build (7202) RStudio: Version 1.0.136 2009-2016 RStudio, Inc.

Code of Appendix A

```
#rm(List = ls()) # remove any variables in R's memory
```

Model input

```
# Model input
n.i   <- 100000          # number of simulated individuals
n.t   <- 30              # time horizon, 30 cycles
v.n   <- c("H","S1","S2","D") # the model states: Healthy (H), Sick (S1), Sicker (S2), Dead (D)
n.s   <- length(v.n)     # the number of health states
v.M_1 <- rep("H", n.i)   # everyone begins in the healthy state
d.c   <- d.e <- 0.03     # equal discounting of costs and QALYs by 3%
v.Trt <- c("No Treatment", "Treatment") # store the strategy names
```

Transition probabilities (per cycle)

```
p.HD   <- 0.005          # probability to die when healthy
p.HS1  <- 0.15           # probability to become sick when healthy
p.S1H  <- 0.5            # probability to become healthy when sick
p.S1S2 <- 0.105         # probability to become sicker when sick
rr.S1  <- 3              # rate ratio of death when sick vs healthy
rr.S2  <- 10             # rate ratio of death when sicker vs healthy
r.HD   <- -log(1 - p.HD)  # rate of death when healthy
r.S1D  <- rr.S1 * r.HD    # rate of death when sick
r.S2D  <- rr.S2 * r.HD    # rate of death when sicker
```

```
p.S1D <- 1 - exp(- r.S1D) # probability to die when sick
p.S2D <- 1 - exp(- r.S2D) # probability to die when sicker
```

Cost and utility inputs

```
c.H <- 2000 # cost of remaining one cycle healthy
c.S1 <- 4000 # cost of remaining one cycle sick
c.S2 <- 15000 # cost of remaining one cycle sicker
c.Trt <- 12000 # cost of treatment (per cycle)

u.H <- 1 # utility when healthy
u.S1 <- 0.75 # utility when sick
u.S2 <- 0.5 # utility when sicker
u.Trt <- 0.95 # utility when sick(er) and being treated
```

Functions

The MicroSim function for the simple microsimulation of the 'Sick-Sicker' model keeps track of what happens to each individual during each cycle.

```
MicroSim <- function(v.M_1, n.i, n.t, v.n, d.c, d.e, TR.out = TRUE, TS.out =
TRUE, Trt = FALSE, seed = 1) {
# Arguments:
# v.M_1: vector of initial states for individuals
# n.i: number of individuals
# n.t: total number of cycles to run the model
# v.n: vector of health state names
# d.c: discount rate for costs
# d.e: discount rate for health outcome (QALYs)
# TR.out: should the output include a microsimulation trace? (default is T
RUE)
# TS.out: should the output include a matrix of transitions between states
? (default is TRUE)
# Trt: are the n.i individuals receiving treatment? (scalar with a Bool
ean value, default is FALSE)
# seed: starting seed number for random number generator (default is 1)
# Makes use of:
# Probs: function for the estimation of transition probabilities
# Costs: function for the estimation of cost state values
# Effs: function for the estimation of state specific health outcomes (Q
ALYs)

v.dwc <- 1 / (1 + d.c) ^ (0:n.t) # calculate the cost discount weight bas
ed on the discount rate d.c
v.dwe <- 1 / (1 + d.e) ^ (0:n.t) # calculate the QALY discount weight bas
ed on the discount rate d.e

# create the matrix capturing the state name/costs/health outcomes for all i
ndividuals at each time point
```

```

m.M <- m.C <- m.E <- matrix(nrow = n.i, ncol = n.t + 1,
                             dimnames = list(paste("ind", 1:n.i, sep = " "),
                                                paste("cycle", 0:n.t, sep = " ")),
                             byrow = TRUE)

m.M[, 1] <- v.M_1 # indicate the initial health state

for (i in 1:n.i) {
  set.seed(seed + i) # set the seed for every individual for
  # or the random number generator
  m.C[i, 1] <- Costs(m.M[i, 1], Trt) # estimate costs per individual for t
  # he initial health state conditional on treatment
  m.E[i, 1] <- Effs(m.M[i, 1], Trt) # estimate QALYs per individual for t
  # he initial health state conditional on treatment

  for (t in 1:n.t) {
    v.p <- Probs(m.M[i, t]) # calculate the transition probabilities
    # at cycle t

    m.M[i, t + 1] <- sample(v.n, prob = v.p, size = 1) # sample the next health
    # state and store that state in matrix m.M
    m.C[i, t + 1] <- Costs(m.M[i, t + 1], Trt) # estimate costs per individual
    # during cycle t + 1 conditional on treatment
    m.E[i, t + 1] <- Effs(m.M[i, t + 1], Trt) # estimate QALYs per individual
    # during cycle t + 1 conditional on treatment

  } # close the loop for the time points
  if(i/100 == round(i/100,0)) { # display the progress of the simulation
    cat('\r', paste(i/n.i * 100, "% done", sep = " "))
  }
} # close the loop for the individuals

tc <- m.C %>% v.dwc # total (discounted) cost per individual
te <- m.E %>% v.dwe # total (discounted) QALYs per individual

tc_hat <- mean(tc) # average (discounted) cost
te_hat <- mean(te) # average (discounted) QALYs

if (TS.out == TRUE) { # create a matrix of transitions across states
  TS <- paste(m.M, cbind(m.M[, -1], NA), sep = "->") # transitions from one
  # state to the other
  TS <- matrix(TS, nrow = n.i)
  rownames(TS) <- paste("Ind", 1:n.i, sep = " ") # name the rows
  colnames(TS) <- paste("Cycle", 0:n.t, sep = " ") # name the columns
} else {
  TS <- NULL
}

```

```

    if (TR.out == TRUE) { # create a trace from the individual trajectories
      TR <- t(apply(m.M, 2, function(x) table(factor(x, levels = v.n, ordered =
TRUE))))
      TR <- TR / n.i # create a distribution trace
      rownames(TR) <- paste("Cycle", 0:n.t, sep = " ") # name the rows
      colnames(TR) <- v.n # name the columns
    } else {
      TR <- NULL
    }

    results <- list(m.M = m.M, m.C = m.C, m.E = m.E, tc = tc, te = te, tc_hat =
tc_hat, te_hat = te_hat, TS = TS, TR = TR) # store the results from the simulation in a list
    return(results) # return the results
  } # end of the MicroSim function

```

Probability function

The Probs function that updates the transition probabilities of every cycle is shown below.

```

Probs <- function(M_it) {
  # M_it: health state occupied by individual i at cycle t (character variable)

  v.p.it <- rep(NA, n.s) # create vector of state transition probabilities
  names(v.p.it) <- v.n # name the vector

  # update v.p.it with the appropriate probabilities
  v.p.it[M_it == "H"] <- c(1 - p.HS1 - p.HD, p.HS1, 0, p.HD)
  # transition probabilities when healthy
  v.p.it[M_it == "S1"] <- c(p.S1H, 1 - p.S1H - p.S1S2 - p.S1D, p.S1S2, p.S1D)
  # transition probabilities when sick
  v.p.it[M_it == "S2"] <- c(0, 0, 1 - p.S2D, p.S2D)
  # transition probabilities when sicker
  v.p.it[M_it == "D"] <- c(0, 0, 0, 1)
  # transition probabilities when dead
  ifelse(sum(v.p.it) == 1, return(v.p.it), print("Probabilities do not sum to 1")) # return the transition probabilities or produce an error
}

```

Costs function

The Costs function estimates the costs at every cycle.

```

Costs <- function (M_it, Trt = FALSE) {
  # M_it: health state occupied by individual i at cycle t (character variable)

```

```

# Trt: is the individual being treated? (default is FALSE)

c.it <- 0 # by default the cost for everyone is zero
c.it[M_it == "H"] <- c.H # update the cost if healthy
c.it[M_it == "S1"] <- c.S1 + c.Trt * Trt # update the cost if sick conditional on treatment
c.it[M_it == "S2"] <- c.S2 + c.Trt * Trt # update the cost if sicker conditional on treatment
return(c.it) # return the costs
}

```

Health outcome function

The Effs function to update the utilities at every cycle.

```

Effs <- function (M_it, Trt = FALSE, cl = 1) {
  # M_it: health state occupied by individual i at cycle t (character variable)
  # Trt: is the individual treated? (default is FALSE)
  # cl: cycle length (default is 1)

  u.it <- 0 # by default the utility for everyone is zero
  u.it[M_it == "H"] <- u.H # update the utility if healthy
  u.it[M_it == "S1"] <- Trt * u.Trt + (1 - Trt) * u.S1 # update the utility if sick conditional on treatment
  u.it[M_it == "S2"] <- u.S2 # update the utility if sicker
  QALYs <- u.it * cl # calculate the QALYs during cycle t
  return(QALYs) # return the QALYs
}

```

Run the simulation

```

sim_no_trt <- MicroSim(v.M_1, n.i, n.t, v.n, d.c, d.e, Trt = FALSE) # run for no treatment
sim_trt <- MicroSim(v.M_1, n.i, n.t, v.n, d.c, d.e, Trt = TRUE) # run for treatment

```

Cost-effectiveness analysis

```

# store the mean costs (and the MCSE) of each strategy in a new variable v.C (vector costs)
v.C <- c(sim_no_trt$tc_hat, sim_trt$tc_hat)
se.C <- c(sd(sim_no_trt$tc), sd(sim_trt$tc)) / sqrt(n.i)
# store the mean QALYs (and the MCSE) of each strategy in a new variable v.E (vector health outcomes)
v.E <- c(sim_no_trt$te_hat, sim_trt$te_hat)
se.E <- c(sd(sim_no_trt$te), sd(sim_trt$te)) / sqrt(n.i)

```

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```

delta.C <- v.C[2] - v.C[1] # calculate incremental costs
delta.E <- v.E[2] - v.E[1] # calculate incremental QALYs
se.delta.E <- sd(sim_trt$te - sim_no_trt$te) / sqrt(n.i) # Monte Carlo square
d error (MCSE) of incremental costs
se.delta.C <- sd(sim_trt$tc - sim_no_trt$tc) / sqrt(n.i) # Monte Carlo square
d error (MCSE) of incremental QALYs
ICER <- delta.C / delta.E # calculate the ICER
results <- c(delta.C, delta.E, ICER) # store the values in a new variable

```

Create full incremental cost-effectiveness analysis table

```

table_micro <- data.frame(
  c(round(v.C, 0), ""), # costs per arm
  c(round(se.C, 0), ""), # MCSE for costs
  c(round(v.E, 3), ""), # health outcomes per arm
  c(round(se.E, 3), ""), # MCSE for health outcomes
  c("", round(delta.C, 0), ""), # incremental costs
  c("", round(se.delta.C, 0), ""), # MCSE for incremental costs
  c("", round(delta.E, 3), ""), # incremental QALYs
  c("", round(se.delta.E, 3), ""), # MCSE for health outcomes (QALYs) gained
  c("", round(ICER, 0), "") # ICER
)
rownames(table_micro) <- c(v.Trt, "* are MCSE values") # name the rows
colnames(table_micro) <- c("Costs", "*", "QALYs", "*", "Incremental Costs",
  "*", "QALYs Gained", "*", "ICER") # name the columns
kable(table_micro) # print the table

```

	Costs	*	QALYs	*	Incremental Costs	*	QALYs Gained	*	ICER
No Treatment	75996	183	15.823	0.016					
Treatment	141644	343	16.384	0.016	65648	164	0.561	0.001	117087
* are MCSE values									