# Simple 3-state microsimulation model

Includes age and sex specific probability of dying when Healthy

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- Jalal H, Pechlivanoglou P, Krijkamp E, Alarid-Escudero F, Enns E, Hunink MG. An Overview of R in Health Decision Sciences. Med Decis Making. 2017; 37(3): 735-746. https://journals.sagepub.com/doi/abs/10.1177/0272989X16686559
- 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. https://journals.sagepub.com/doi/abs/10.1177/0272989X18754513
- Krijkamp EM, Alarid-Escudero F, Enns E, Pechlivanoglou P, Hunink MM, Jalal H. A Multidimensional Array Representation of State-Transition Model Dynamics. Med Decis Mak. 2020;40(2):242-248. https://doi.org/10.1177/0272989X19893973

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Change eval to TRUE if you want to knit this document.

# 01 Load packages

```
if (!require('pacman')) install.packages('pacman'); library(pacman) # use this package to conveniently
# load (install if required) packages from CRAN
p_load("here", "dplyr", "devtools", "scales", "ellipse", "ggplot2", "lazyeval", "igraph", "re
# load (install if required) packages from GitHub
# install_github("DARTH-git/darthtools", force = TRUE) # Uncomment if there is a newer version
p_load_gh("DARTH-git/darthtools")
```

# 02 Load functions

```
# No functions needed
```

# 03 Input model parameters

```
set.seed(1) # set the seed
# Model structure
v_names_states <- c("Healthy", "Sick", "Dead") # vector with state names
n_states <- length(v_names_states) # number of states</pre>
n_t
          <- 60
                                    # number of cycles
          <- 10000
                                    # number of individuals
n_i
          - d_c < 0.03
d_e
                                    # equal discount of costs and QALYs
#### Deterministic analysis ####
# Transition probabilities
       <- 0.05 # probability Healthy -> Sick
p_HD_female <- 0.0382 # probability health -> Dead when female
p HD male <- 0.0463 # probability health -> Dead when male
           <- 0.1 # probability Sick -> Dead
p_SD
# Costs inputs
c_H <- 1500 # cost of one cycle in Healthy state
          <- 5000 # cost of one cycle in Sick state
c_S
           <- 0
                    # cost of one cycle in Dead state
c D
# utility inputs
          <- 1
                # utility when Healthy
u_H
u_S
           <- 0.85
                      # utility when Sick
\mathtt{u}_{\mathtt{D}}
          <- 0
                      # utility when Dead
\# calculate discount weights for costs for each cycle based on discount rate d_c
```

```
v_dwc      <- 1 / (1 + d_e) ^ (0:n_t)
# calculate discount weights for effectiveness for each cycle based on discount rate d_e
v_dwe      <- 1 / (1 + d_c) ^ (0:n_t)

m_p_HD       <- data.frame(Sex = c("Female", "Male"), p_HD = c(p_HD_female, p_HD_male))</pre>
```

# 04 Sample individual level characteristics

#### 04.1 Static characteristics

```
# randomly sample the sex of an individual (50% female)
v_sex <- sample(x = c("Female", "Male"), prob = c(0.5, 0.5), size = n_i, replace = TRUE)
df_X <- data.frame(ID = 1:n_i, Sex = v_sex)</pre>
```

## 04.2 Dynamic characteristics

```
# Specify the initial health state of the individuals
# everyone begins in the Healthy state (in this example)
v_M_init <- rep("Healthy", times = n_i)</pre>
```

## 05 Define Simulation Functions

#### 05.1 Probability function

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

```
Probs <- function(M_t, df_X) {</pre>
  # Arguments:
    # M_t: health state occupied at cycle t (character variable)
    # df X: data frame with individual characteristics data
  # Returns:
    # transition probabilities for that cycle
  # create matrix of state transition probabilities
                  <- matrix(0, nrow = n_states, ncol = n_i)</pre>
  rownames(m_p_t) <- v_names_states # give the state names to the rows
  # lookup baseline probability and rate of dying based on individual characteristics
  p_HD_all <- inner_join(df_X, m_p_HD, by = c("Sex"))</pre>
          <- p_HD_all[M_t == "Healthy", "p_HD"]</pre>
  \# update m_p_t with the appropriate probabilities
  # transition probabilities when Healthy
  m_p_t[, M_t == "Healthy"] \leftarrow rbind((1 - p_HD) * (1 - p_HS),
                                       (1 - p_HD) *
                                                         p_HS ,
                                            p_HD
```

#### 05.2 Cost function

The Costs function estimates the costs at every cycle.

#### 05.3 Health outcome function

The Effs function to update the utilities at every cycle.

## 06 Run Microsimulation

```
MicroSim <- function(n_i, df_X, seed = 1) {</pre>
# Arguments:
 # n_i:
           number of individuals
  # df X:
           data frame with individual data
  # seed:
           defauls is 1
# Returns
  # a list with information about the individuals transitions, associated costs and
  # effects and total costs and rewards
  set.seed(seed) # set the seed
  \# create three matrices called m\_M, m\_C and m\_E
  # number of rows is equal to the n_i, the number of columns is equal to n_t
  # (the initial state and all the n_t cycles)
  # m M is used to store the health state information over time for every individual
  # m_C is used to store the costs information over time for every individual
  # m_E is used to store the effects information over time for every individual
  m_M \leftarrow m_C \leftarrow m_E \leftarrow matrix(nrow = n_i, ncol = n_t + 1,
                                        dimnames = list(paste("ind" , 1:n_i, sep = " "),
                                                        paste("cycle", 0:n_t, sep = " ")))
                              # initial health state
  m_M[, 1] <- v_M_init
  m_C[, 1] \leftarrow Costs(m_M[, 1]) # costs accrued during cycle 0
  m_E[, 1] <- Effs(m_M[, 1]) # QALYs accrued during cycle 0</pre>
  # open a loop for time running cycles 1 to n t
  for (t in 1:n t) {
    # calculate the transition probabilities for the cycle based on health state t
   m_P <- Probs(m_M[, t], df_X)</pre>
   # check if transition probabilities are between 0 and 1
    check transition probability(m P, verbose = TRUE)
    # check if each of the rows of the transition probabilities matrix sum to one
   check_sum_of_transition_array(m_P, n_rows = n_i, n_cycles = n_t, verbose = TRUE)
    \# sample the current health state and store that state in matrix m\_M
   m_M[, t + 1] \leftarrow samplev(m_P, 1)
   m_C[, t + 1] \leftarrow Costs(m_M[, t + 1]) # calculate costs
   m_E[, t + 1] \leftarrow Effs (m_M[, t + 1]) + calculate QALYs
    # Display simulation progress
   if(t/(n_t/10) = round(t/(n_t/10), 0))  { # display progress every 10%
      cat('\r', paste(t/n_t * 100, "% done", sep = " "))
   }
  } # close the loop for the time points
  # calculate
         <- m_C %*% v_dwc # total (discounted) cost per individual</pre>
         <- m E %*% v dwe # total (discounted) QALYs per individual
  tc_hat <- mean(tc) # average (discounted) cost
 te_hat <- mean(te) # average (discounted) QALYs
```

### 07 Visualize results

```
plot(density(outcomes$tc), main = paste("Total cost per person"), xlab = "Cost ($)")
plot(density(outcomes$te), main = paste("Total QALYs per person"), xlab = "QALYs")
plot_trace_microsim(outcomes$m_M)  # health state trace
```

Small note: The difference between paste() and paste0() is that the argument sep by default is " " (paste) and " " (paste0). In conclusion, paste0() is faster than paste() if our objective is concatenate strings without spaces because we don't have to specify the argument sep.