# Simple 3-state microsimulation model

Includes sex specific probability of dying when healthy and state occupation: probability of dying when sick depends on the time of being sick

## The DARTH workgroup

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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 Making. 2020 Online first. https://doi.org/10.1177/0272989X19893973

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

```
rm(list = ls())  # clear memory (removes all the variables from the workspace)
```

# 01 Load packages

```
if (!require('pacman')) install.packages('pacman'); library(pacman)

# load (install if required) packages from CRAN
p_load("here", "devtools", "dplyr", "scales", "ellipse", "ggplot2", "lazyeval", "igraph", "truncnorm",
# load (install if required) packages from GitHub
# install_github("DARTH-git/dampack", force = TRUE) # Uncomment if there is a newer version
# install_github("DARTH-git/darthtools", force = TRUE) # Uncomment if there is a newer version
p_load_gh("DARTH-git/dampack", "DARTH-git/darthtools")
```

# 02 Load functions

```
# No functions needed
```

# 03 Input model parameters

```
set.seed(1) # set the seed
# Model structure
v_n <- c("healthy", "sick", "dead") # vector with state names
n_states <- length(v_n)
                                             # number of states
          <- 60
                                             # number of cycles
n t
          <- 10000
                                             # number of individuals
n i
d_e <- d_c <- 0.03
                                             # equal discount of costs and QALYs by 3%
\# calculate discount weights for costs for each cycle based on discount rate d_{\_}c
v_dwc \leftarrow 1 / (1 + d_e) ^ (0:n_t)
\# calculate discount weights for effectiveness for each cycle based on discount rate d_-e
v_dwe \leftarrow 1 / (1 + d_c) ^ (0:n_t)
#### Deterministic analysis ####
# Transition probabilities
# (all non-dead probabilities are conditional on survival)
           <- 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
           <- data.frame(Sex = c("Female", "Male"), p_HD = c(p_HD_female, p_HD_male))</pre>
m_p_HD
# probability to die in sick state by cycle of being sick
p_SD \leftarrow c(0.1, 0.2, 0.3, 0.4, 0.5, rep(0.7, n_t - 5))
```

```
# Costs inputs
c_H <- 1500  # cost of one cycle in healthy state
c_S <- 5000  # cost of one cycle in sick state
c_D <- 0

# utility inputs
u_H <- 1  # utility when healthy
u_S <- 0.85  # utility when sick
u_D <- 0  # utility when dead</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)</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)
v_Ts_init <- rep(0, n_i) # a vector with the time of being sick at the start of the model</pre>
```

#### 04.3 Create a dataframe with the individual characteristics

```
# create a data frame with each individual's
# ID number, treatment effect modifier, age and initial time in sick state
df_X <- data.frame(ID = 1:n_i, Sex = v_sex, Ts_init = v_Ts_init, M_init = v_M_init)
head(df_X) # print the first rows of the dataframe</pre>
```

# 05 Define Simulation Functions

#### 05.1 Probability function

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

```
Probs <- function(M_t, df_X, v_Ts) {
    # Arguments:
    # M_t: health state occupied at cycle t (character variable)
    # df_X: data frame with individual characteristics data
    # v_Ts: vector with the duration of being sick
# Returns:</pre>
```

```
# transition probabilities for that cycle
  # create matrix of state transition probabilities
          <- matrix(0, nrow = n_states, ncol = n_i)</pre>
  # give the state names to the rows
  rownames(m_p_t) <- v_n
  # 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>
  p_HD
  # update m_p_t with the appropriate probabilities
  # (all non-death probabilities are conditional on survival)
  # transition probabilities when healthy
  m_p_t[, M_t == "healthy"] <- rbind((1 - p_HD) * (1 - p_HS) ,
                                     (1 - p_HD) *
                                                   p_HS ,
                                          p_HD)
  # transition probabilities when sick
  m_p_t[, M_t == "sick"] <- rbind(0,</pre>
                                     1 - p_SD[v_Ts],
                                         p_SD[v_Ts])
  # transition probabilities when dead
  m_p_t[, M_t == "dead"] <- rbind(0,
                                     0,
                                     1)
 return(t(m_p_t))
}
```

#### 05.2 Cost function

The Costs function estimates the costs at every cycle.

```
Costs <- function (M_t) {
    # Arguments:
    # M_t: health state occupied at cycle t (character variable)
    # Returns:
    # costs accrued in this cycle

c_t <- c()
    c_t[M_t == "healthy"] <- c_H # costs accrued by being healthy this cycle
    c_t[M_t == "sick"] <- c_S # costs accrued by being sick this cycle
    c_t[M_t == "dead"] <- c_D # costs at dead state

return(c_t)
}</pre>
```

## 05.3 Health outcome function

The Effs function to update the utilities at every cycle.

```
Effs <- function (M_t) {
    # Arguments:
    # M_t: health state occupied at cycle t (character variable)
# Returns:
    # QALYs accrued this cycle

q_t <- c()
q_t[M_t == "healthy"] <- u_H # QALYs accrued by being healthy this cycle
q_t[M_t == "sick"] <- u_S # QALYs accrued by being sick this cycle
q_t[M_t == "dead"] <- u_D # QALYs at dead state

return(q_t)
}</pre>
```

#### 05.4 Microsimulation function

Below we develop the microsimulation function that allows the model to be run.

```
MicroSim <- function(n_i, df_X, seed = 1) {</pre>
  # Arguments:
    # n_i: number of individuals
    # df X: data frame with individual data
    # seed: seed for the random number generator, default is 1
    # results: data frame with total cost and QALYs
  set.seed(seed) # set a seed to be able to reproduce the same results
  \# 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_it
  # (the initial state and all the n_t cycles)
  \# m<sub>M</sub> 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 = " ")))
  m_M[, 1] <- as.character(df_X$M_init) # initial health state</pre>
         <- df X$Ts init # initialize time since illness onset</pre>
  m_C[, 1] <- Costs(m_M[, 1]) # costs accrued during cycle 0</pre>
  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, v_Ts)</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_states = n_i, n_t = n_t, verbose = TRUE)
```

```
\# sample the next health state and store that state in matrix m\_M
   m_M[, t + 1] <- samplev(m_P)</pre>
    # calculate costs per individual during cycle t + 1
   m C[, t + 1] <- Costs(m M[, t + 1])
    \# calculate QALYs per individual during cycle t+1
   m_E[, t + 1] <- Effs (m_M[, t + 1])</pre>
    # update time since illness onset for t + 1
   v_Ts \leftarrow if_else(m_M[, t + 1] == "sick", v_Ts + 1, 0)
    # 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
  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) QAL
  # store the results from the simulation in a list
 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)
 return(results) # return the results
} # end of the `MicroSim` function
```

#### 06 Run Microsimulation

```
# 06 Run Microsimulation

# By specifying all the arguments in the `MicroSim()` the simulation can be started

# Run the simulation model
outcomes <- MicroSim(n_i = n_i, df_X = df_X, seed = 1)

# Show results
results <- data.frame("Total Cost" = outcomes$tc_hat, "Total QALYs" = outcomes$te_hat)
results</pre>
```

## 07 Visualize results

```
options(scipen = 999)  # disable scientific notation in R
plot(density(outcomes$tc), main = paste("Total cost per person"), xlab = "Cost ($)")
plot(density(outcomes$te), main = paste("Total QALYs per person"), xlab = "QALYs")
plot_m_TR(outcomes$m_M)  # health state trace
```