Simple 3-state microsimulation model

Includes age and sex specific probability of dying when healthy

The DARTH workgroup

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Please cite our publications when using this code:

- 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

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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) # 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/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")</pre>
                                     # vector with state names
n_t <- 60
                                      # number of cycles
n_i <- 10000
                                      # number of individuals
d_e <- d_c <- 0.03
                                      # equal discount of costs and QALYs by 3%
#### 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_D <- 0
# utility inputs
```

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.

05.2 Cost function

The Costs function estimates the costs at every cycle.

```
Costs <- function (M_t) {
    # M_t: current health state
    c_t <- c()
    c_t[M_t == "dead"] <- c_D  # costs at dead state
    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
    return(c_t)  # return costs accrued this cycle
}</pre>
```

05.3 Health outcome function

The Effs function to update the utilities at every cycle.

```
Effs <- function (M_t) {
    # M_t: current health state
    q_t <- c()
    q_t[M_t == "dead"] <- u_D  # QALYs at dead state
    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

return(q_t)  # return the QALYs accrued this cycle
}</pre>
```

06 Run Microsimulation

```
# 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<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] <- v_M_init
                               # initial health state
  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
  # 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>
    \# sample the current health state and store that state in matrix m\_M
    m_M[, t + 1] <- samplev(m_P)
    m_C[, t + 1] <- Costs(m_M[, t + 1]) # calculate costs</pre>
    m_E[, t + 1] <- Effs (m_M[, t + 1]) # calculate QALYs</pre>
    # 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)</pre>
                        # average (discounted) QALYs
  # 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
# 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)
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
```