

# Simple 3-state Digitized Partitioned Survival model in R

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. *BioRxiv* 670612 2019.<https://www.biorxiv.org/content/10.1101/670612v1>

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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", "gems", "flexsurv", "survHE", "ggplot2", "msm", "igraph", "mstate",
# 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 function needed
```

## 03 Input model parameters

```
v_names_states <- c("Healthy", "Sick", "Dead") # state names
n_states <- length(v_names_states) # No of states
c_l <- 1 / 12 # cycle length (a month)
n_t <- 20 # number of years (20 years)
times <- seq(0, n_t, c_l) # the cycles in years
d_r <- 0.03 # discount rate
set.seed(2009) # set the seed

c_H <- 200 # cost of remaining one cycle healthy
c_S <- 500 # cost of remaining one cycle sick
c_D <- 0 # cost of remaining one cycle dead
v_c <- c(c_H, c_S, c_D) # store in a vector

u_H <- 0.75 # utility when healthy
u_S <- 0.30 # utility when sick
u_D <- 0 # utility when dead
v_u <- c(u_H, u_S, u_D) # store in a vector

v_dw <- 1 / (1 + d_r) ^ (times) # discount weight
```

## 04 Digitized Data

Use the function `digitise()` to translate the digitised OS and PFS data into patient level information.

```
# Create IPD and KM data for the OS curves
digitise(surv_inp = "OS_Examp.txt",
        nrisk_inp = "OS_Examp_AtRisk.txt",
        km_output = "KMdata_OS.txt",
```

```

    ipd_output = "IPDdata_OS.txt")

# Create IPD and KM data for the PFS curves (there have been 99 events in the cohort)
digitise(surv_inp   = "PFS_Examp.txt",
        nrisk_inp  = "PFS_Examp_AtRisk.txt",
        km_output  = "KMdata_PFS.txt",
        ipd_output = "IPDdata_PFS.txt")

# Link the IPD files across the two arms of the trial for OS and PFS
IPD_OS  <- make.ipd(ipd_files = c("IPDdata_OS.txt"), ctr = 1,
                  var.labs   = c("time", "event", "arm"))
IPD_PFS <- make.ipd(ipd_files = c("IPDdata_PFS.txt"), ctr = 1,
                  var.labs   = c("time", "event", "arm"))

```

## 05 Analysis

### 05.1 Partitioned Survival model

```

# fit all parametric models to the data and extract the AIC/BIC
# Select the one with the most appropriate fit
# Repeat for PFS and OS
fit_PFS <- fit.fun(time = "time", status = "event", data = IPD_PFS, times = times,
                  extrapolate = T)
fit_OS  <- fit.fun(time = "time", status = "event", data = IPD_OS, times = times,
                  extrapolate = T)

# Check AIC of each model to assess goodness-of-fit
GoF_PFS <- data.frame(AIC = fit_PFS$AIC, BIC = fit_PFS$BIC)
GoF_OS  <- data.frame(AIC = fit_OS$AIC, BIC = fit_OS$BIC)

# "Exponential", "Weibull (AFT)", "Gamma", "log-Normal", "log-Logistic", "Gen. Gamma"
choose_PFS <- rownames(GoF_PFS)[which.min(GoF_PFS$AIC)]
choose_OS  <- rownames(GoF_OS)[which.min(GoF_OS$AIC)]

# construct a partitioned survival model out of the chosen models
m_M_PSM <- partsurv(pfs_survHE = fit_PFS, os_survHE = fit_OS,
                  choose_PFS = choose_PFS, choose_OS = choose_OS,
                  time = times, v_names_states = v_names_states)

# plot the results of PSM and the true data
plot_trace_PSM(time = times, partsurv.model = m_M_PSM, v_names_states = v_names_states)

```

Calculate total cost and QALYs per cycle.

```

v_c_t <- m_M_PSM$trace %*% v_c
v_u_t <- m_M_PSM$trace %*% v_u

tot_c <- t(v_c_t) %*% v_dw
tot_u <- t(v_u_t) %*% v_dw

```

```
# display results
results <- data.frame(Total_Cost = tot_c, Total_QALYs = tot_u)
results
```