

# Sick-Sicker case study

*DARTH workgroup*

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In this document we showcase the framework via a fully functional decision model. In this case-study we perform a cost-effectiveness analysis (CEA) using a previously published 4-state model called the Sick-Sicker model.(Enns et al. 2015) The hypothetical disease affects individuals with an average age of 25 years and results in increased mortality, increased treatment costs and reduced quality of life. The model is used to quantify the expected costs and quality-adjusted life years (QALYs) for individuals with a hypothetical disease with two different stages, Sick and Sicker. We calibrate the model using data for three parameters, our targets. We then evaluate the cost-effectiveness of a hypothetical treatment that increases quality of life (QoL) in one of the disease states.(Krijkamp et al. 2018) We identify the uncertainty around our decision based on the CEA using sensitivity analysis. Finally, we perform a value of information (VoI) analysis to see if it is worth investing in extra research projects with the aim to reduce the uncertainty around our decision.

## The Sick-Sicker model

In the Sick-Sicker model, we simulate a hypothetical cohort of 25-year-old individuals over a lifetime (or reaching age 90 years old) using 65 annual cycles, represented with `n.t.` The cohort start in the “Healthy” health state (denoted “H”). Healthy individuals are at risk of developing the illness, at which point they would transition to the first stage of the disease (the “Sick” health state, denoted “S1”). Sick individuals are at risk of further progressing to a more severe stage (the “Sicker” health state, denoted “S2”), which is constant in this case example. There is a chance that individuals in the Sick state eventually recover and return back to the Healthy healthy state. However, once an individual reaches the Sicker health state, they cannot recover; that is, the probability of transitioning to the Sick or Healthy health states from the Sicker health state is zero. Individuals in the Healthy state face background mortality that is age-specific (i.e., time-dependent). Sick and Sicker individuals face an increased mortality in the form of a hazard rate ratio (HR) of 3 and 10 times, respectively, on the background mortality rate. Sick and Sicker individuals also experience increased health care costs and reduced QoL compared to healthy individuals. Once simulated individuals die, they transition to the “Dead” health state (denoted “D”), where they remain. Figure 1 shows the state-transition diagram of the Sick-Sicker model. The evolution of the cohort is simulated in one-year discrete-time cycles. Both costs and QALYs are discounted at an annual rate of 0.03 %.

Two alternative strategies exist for this hypothetical disease: a no-treatment and a treatment strategy. Under the treatment strategy, Sick and Sicker individuals receive treatment and continue doing so until they recover or die. The cost of the treatment is additional to the cost of being Sick or Sicker for one year. The treatment improves QoL for those individuals who are Sick but has no effect on the QoL of those who are sicker. We evaluate the cost-effectiveness of the treatment assuming a willingness to pay of \$80000.

## 01 Define model inputs

The input for the Sick-Sicker model is informed by external data. All model parameter values and R variable names, for both the general set up and the external data, are presented in Table 1. A base-case parameters set is generated by the `01_Model-inputs_function.R` file that is using the external data stored in the `01_basecase-params.csv` dataset in the data folder. In this same data folder, we can find the file `01_all-cause-mortality-USA-2015.csv`. This file is derived from the Human Mortality data base and include the age specific mortality rated for the cohort based on the 2015 data about all-cause mortality rates for the USA population. The `01_model-inputs.R` file stored in the R folder is source to load all information, about the general setup of the model, the age-specific mortality rates and the base-case parameter set. This base-case parameters set includes placeholder values for the parameters we need to calibrated. This means

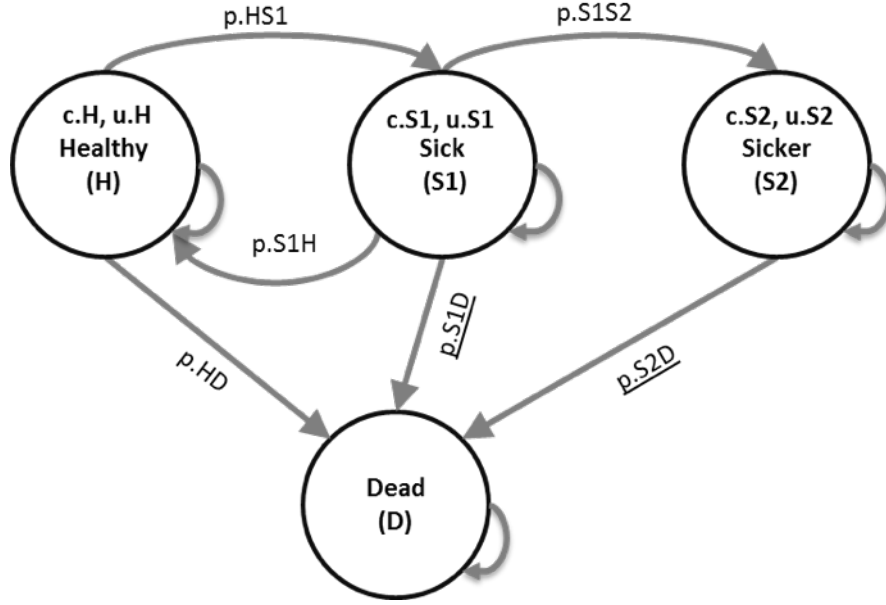


Figure 1: Sick-Sicker

that we will use these values for now, but these values will be replaced by the best-fitted calibrated values after we performed the calibration.

```
source("R/01_model-inputs.R") # source the inputs file
```

Table 1: Description of parameters with their R name and value.

Parameter	R name	Value
Time horizon ( $n_t$ )	n.t	65 years
Names of health states ( $n$ )	v.n	H, S1, S2, D
Annual discount rate (costs/QALYs)	d.c/d.e	3%
Annual transition probabilities		
- Disease onset (H to S1)	p.HS1	0.15
- Recovery (S1 to H)	p.S1H	0.5
- Disease progression (S1 to S2) in the time-homogenous model	p.S1S2	0.105
Annual mortality		
- All-cause mortality (H to D)	p.HD	age-specific
- Hazard ratio of death in S1 vs H	hr.S1	3
- Hazard ratio of death in S2 vs H	hr.S2	3
Annual costs		
- Healthy individuals	c.H	\$2,000
- Sick individuals in S1	c.S1	\$4,000
- Sick individuals in S2	c.S2	\$15,000
- Dead individuals	c.D	\$0
- Additional costs of sick individuals treated in S1 or S2	c.Trt	\$12,000
Utility weights		
- Healthy individuals	u.H	1.00
- Sick individuals in S1	u.S1	0.75
- Sick individuals in S2	u.S2	0.50
- Dead individuals	u.D	0.00
Intervention effect		

Parameter	R name	Value
- Utility for treated individuals in S1	<code>u.Trt</code>	0.95

Enns, EA, LE Cipriano, CT Simons, and CY Kong. 2015. “Identifying Best-Fitting Inputs in Health-Economic Model Calibration: A Pareto Frontier Approach.” *Medical Decision Making* 35 (2): 170–82. doi:10.1177/0272989X14528382.

Krijkamp, EM, F Alarid-Escudero, EA Enns, H Jalal, MGM Hunink, and P Pechlivanoglou. 2018. “Microsimulation Modeling for Health Decision Sciences Using R: A Tutorial.” *Medical Decision Making : An International Journal of the Society for Medical Decision Making* 38 (3): 400–422. doi:10.1177/0272989X18754513.