

# Vignette for dampack package

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## Install dampack

The R package **dampack** is not available (yet!) from CRAN. To use this package, you need to install the developer version from my github account. To do so, run the following commands.

```
install.packages("devtools") # Only if you don't have devtools installed already
library(devtools)
install_github("feralae/dampack")
```

You will have to wait for a few seconds to run and you should be good to go!

Given that this package is under current active development, future iterations of it might be significantly different. So use this package and it's functions at your own risk. I will try to keep future versions of this package as consistent as possible, though.

## Cost-effectiveness acceptability curves (CEAC) and Frontier (CEAF)

The cost-effectiveness acceptability curves (CEAC) determine the probability of each strategy being cost effective by willingness-to-pay (WTP) threshold (Van Hout et al. 1994). The cost-effectiveness acceptability frontier (CEAF) determines the optimal strategy –defined as the strategy with the highest net benefit– by WTP (Fenwick, Claxton, and Sculpher 2001). The function **ceaf** computes both the CEAC and CEAF and returns a data frame and a **ggplot2** objects of these outcomes.

To run the **ceaf** function, you need a PSA, specify the strategies, the cost and effectiveness of these, and different values of WTP thresholds. The following code produces the CEAC and CEAF of the breast cancer CEA example included in this package.

```
# Load PSA dataset
data(psa)
# Name of strategies
strategies <- c("Chemo", "Radio", "Surgery")
# Vector of WTP thresholds
v.wtp <- seq(1000, 150000, by = 10000)
# Matrix of costs
m.c <- psa[, c(2, 4, 6)]
# Matrix of effectiveness
m.e <- psa[, c(3, 5, 7)]

out <- ceaf(v.wtp = v.wtp, strategies = strategies, m.e = m.e , m.c = m.c)
gg.ceaf <- out$gg.ceaf
plot(gg.ceaf)
```

## Expected value of perfect information (EVPI)

The expected value of perfect information (EVPI) represents the upper limit that a decision maker should be willing to pay to eliminate uncertainty in a decision model.

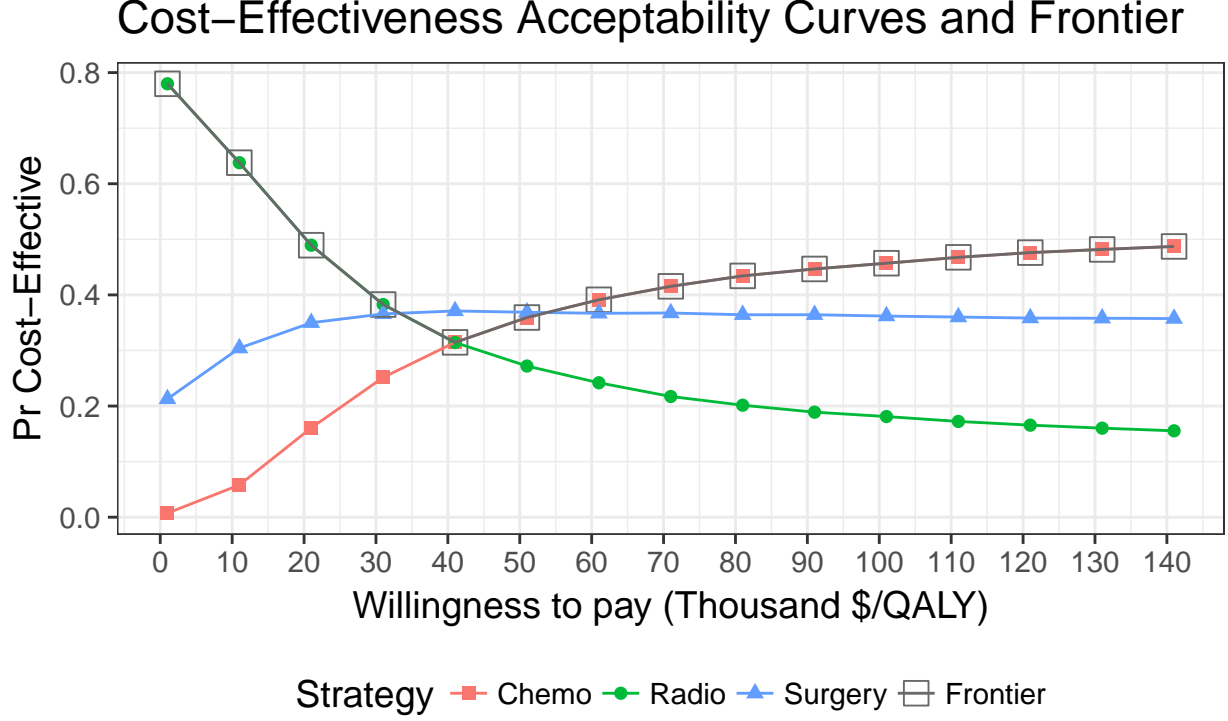


Figure 1: Cost-effectiveness acceptability curves (CEAC) and forntier (CEAF).

The function `evpi` computes the EVPI for different WTP thresholds and returns a data frame with these values. Similar to the `ceaf` function, to run the `evpi` function you need a PSA, the cost and effectiveness of these, and different values of WTP thresholds. The following code produces the EVPI of the breast cancer CEA example included in this package.

```
df.evpi <- evpi(v.wtp = v.wtp, m.e = m.e , m.c = m.c)
```

To plot the EVPI, we use the function `plot.evpi` that takes `df.evpi` as argument.

```
gg.evpi <- plot.evpi(evpi = df.evpi)
gg.evpi
```

## EVPPPI using linear regression metamodeling

The expected value of partial perfect information (EVPPPI) is the expected value of perfect information from a subset of parameters of interest,  $\theta_I$  of a cost-effectiveness analysis (CEA) of  $D$  different strategies with parameters  $\theta = \{\theta_I, \theta_C\}$ , where  $\theta_C$  is the set of complimentary parameters of the CEA. The function `evppi_lrmm` computes the EVPPPI of  $\theta_I$  from a matrix of net monetary benefits  $B$  of the CEA. Each column of  $B$  corresponds to the net benefit  $B_d$  of strategy  $d$ . The function `evppi_lrmm` computes the EVPPPI using a linear regression metamodel (Strong, Oakley, and Brennan 2014; Jalal and Alarid-Escudero 2017) approach following these steps:

1. Determine the optimal strategy  $d^*$  from the expected net benefits  $\bar{B}$

$$d^* = \arg \max_d \{\bar{B}\}$$

2. Compute the opportunity loss for each  $d$  strategy,  $L_d$

$$L_d = B_d - B_{d^*}$$

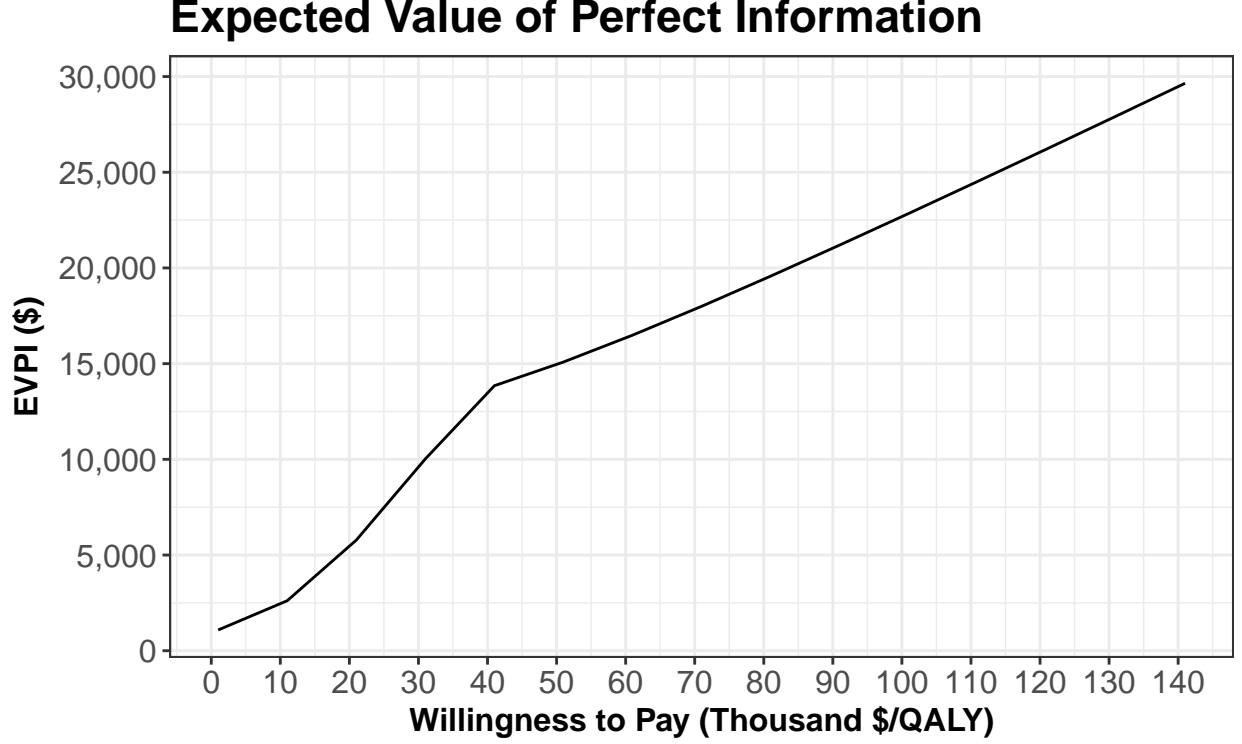


Figure 2: Expected value of perfect information (EVPI).

3. Estimate a linear metamodel for the opportunity loss of each  $d$  strategy,  $L_d$ , by regressing them on the spline basis functions of  $\theta_I$ ,  $f(\theta_I)$

$$L_d = \beta_0 + f(\theta_I) + \epsilon,$$

where  $\epsilon$  is the residual term that captures the complementary parameters  $\theta_C$  and the difference between the original simulation model and the metamodel.

4. Compute the EVPPI of  $\theta_I$  using the estimated losses for each  $d$  strategy,  $\hat{L}_d$  from the linear regression metamodel and applying the following equation:

$$\text{EVPPI}_{\theta_I} = \frac{1}{K} \sum_{i=1}^K \max_d (\hat{L}_d)$$

The spline model in step 3 is fitted using the `mgcv` package.

### Estimation of Dirichlet parameters using MoM

The function `dirichlet_params` computes the  $\alpha$  parameters of a Dirichlet distribution following the method of moments (MoM) proposed by Fielitz and Myers (1975) and Narayanan (1992).

If  $\mu$  is a vector of means and  $\sigma$  is a vector of standard deviations of the random variable, then the second moment  $X_2$  is defined by  $\sigma^2 + \mu^2$ . Using the mean and the second moment, the  $J$  alpha parameters are computed as follows

$$\alpha_i = \frac{(\mu_1 - X_{2_1})\mu_i}{X_{2_1} - \mu_1^2}$$

for  $i = 1, \dots, J - 1$ , and

$$\alpha_J = \frac{(\mu_1 - X_{2_1})(1 - \sum_{i=1}^{J-1} \mu_i)}{X_{2_1} - \mu_1^2}$$

## Estimation of Log-normal parameters uisng MoM

The function `lnorm_params` computes the location,  $\mu$ , and scale,  $\sigma$ , parameters of a log-normal distribution from the mean and variance of a random variable following the method of moments (MoM).

Given the non-logarithmized mean and variance  $m$  and  $v$  of the random variable, respectively, the location,  $\mu$ , and scale,  $\sigma$ , of a log-normal distribution are given by the following equations

$$\mu = \ln \left( \frac{m}{\sqrt{1 + \frac{v}{m^2}}} \right)$$

and

$$\sigma = \sqrt{\ln \left( 1 + \frac{v}{m^2} \right)}$$

## Vignette Info

Note the various macros within the `vignette` section of the metadata block above. These are required in order to instruct R how to build the vignette. Note that you should change the `title` field and the `\VignetteIndexEntry` to match the title of your vignette.

## Styles

The `html_vignette` template includes a basic CSS theme. To override this theme you can specify your own CSS in the document metadata as follows:

```
output:
  markdown::html_vignette:
    css: mystyles.css
```

## More Examples

You can write math expressions, e.g.  $Y = X\beta + \epsilon$ , footnotes<sup>1</sup>, and tables, e.g. using `knitr::kable()`.

	mpg	cyl	disp	hp	drat	wt	qsec	vs	am	gear	carb
Mazda RX4	21.0	6	160.0	110	3.90	2.620	16.46	0	1	4	4
Mazda RX4 Wag	21.0	6	160.0	110	3.90	2.875	17.02	0	1	4	4
Datsun 710	22.8	4	108.0	93	3.85	2.320	18.61	1	1	4	1
Hornet 4 Drive	21.4	6	258.0	110	3.08	3.215	19.44	1	0	3	1
Hornet Sportabout	18.7	8	360.0	175	3.15	3.440	17.02	0	0	3	2
Valiant	18.1	6	225.0	105	2.76	3.460	20.22	1	0	3	1
Duster 360	14.3	8	360.0	245	3.21	3.570	15.84	0	0	3	4
Merc 240D	24.4	4	146.7	62	3.69	3.190	20.00	1	0	4	2

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<sup>1</sup>A footnote here.

	mpg	cyl	disp	hp	drat	wt	qsec	vs	am	gear	carb
Merc 230	22.8	4	140.8	95	3.92	3.150	22.90	1	0	4	2
Merc 280	19.2	6	167.6	123	3.92	3.440	18.30	1	0	4	4

Also a quote using >:

“He who gives up [code] safety for [code] speed deserves neither.” (via)

## References

- Fenwick, Elisabeth, Karl Claxton, and Mark Sculpher. 2001. “Representing Uncertainty: The Role of Cost-Effectiveness Acceptability Curves.” *Health Economics* 10 (May): 779–87. doi:10.1002/hec.635.
- Fielitz, Bruce D., and Buddy L. Myers. 1975. “Estimation of parameters in the beta distribution.” *Decision Sciences* 6 (1): 1–13.
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- Van Hout, Ben A., Maiwenn J. Al, Gilad S. Gordon, and Frans FH Rutten. 1994. “Costs, Effects and C/E-Ratios Alongside A Clinical Trial.” *Health Economics* 3 (5): 309–19.