Vignette for dampack package

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Estimation of Dirichlet parameters using MoM

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

If μ is a vector of means and σ 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, ..., 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, μ , and scale, σ , 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, μ , and scale, σ , of a log-normal distribution are given by the following equations

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

and

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

EVPPI using linear regression metamodeling

The expected value of partial perfect information (EVPPI) is the expected value of perfect information from a subset of parameters of interest, θ_I of a cost-effectiveness analysis (CEA) of D different strategies with parameters $\theta = \{\theta_I, \theta_C\}$, where θ_C is the set of complimentary parameters of the CEA. The function evppi_lrmm computes the EVPPI of θ_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 EVPPI 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}

$$d^{*}=\arg\max_{d}\left\{ \bar{B}\right\}$$

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

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

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

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

where ϵ is the residual term that captures the complementary parameters θ_C and the difference between the original simulation model and the metamodel.

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

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

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

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. The cost-effectiveness acceptability frontier (CEAF) determines the optimal strategy –defined as the strategy with the highest net benefit– by WTP. 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.ceac
plot(gg.ceaf)</pre>
```

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:

Cost-Effectiveness Acceptability Curves and Frontier

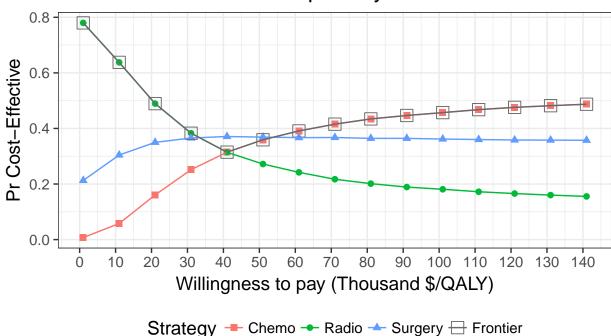


Figure 1: Cost-effectiveness acceptability curves (CEAC) and forntier (CEAF).

output:

rmarkdown::html_vignette:
 css: mystyles.css

More Examples

You can write math expressions, e.g. $Y = X\beta + \epsilon$, footnotes¹, 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
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

Fielitz, Bruce D., and Buddy L. Myers. 1975. "Estimation of parameters in the beta distribution." *Decision Sciences* 6 (1): 1–13.

Jalal, Hawre, and Fernando Alarid-Escudero. 2017. "A Gaussian Approximation Approach for Value of Information Analysis." *Medical Decision Making* Forthcomin: 1–16. http://journals.sagepub.com/doi/full/10. 1177/0272080X17715627