

BCEA: An R Package for Cost-Effectiveness Analysis

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Summary

Health economic cost-effectiveness analyses (CEA) consist of analytic approaches for combining costs and health consequences of intervention(s). These help to understand how much an intervention may cost (per unit of health gained) compared to an alternative intervention, such as a control or status quo. For resource allocation, a decision maker may wish to know if an intervention is cost saving, and if not then how much more would it cost to implement it compared to a less effective intervention.

Current guidance for cost-effectiveness analyses advocates the quantification of uncertainties which can be represented by random samples obtained from a probability sensitivity analysis or, more efficiently, a Bayesian model. The R (R Core Team, 2016) package BCEA can be used to post-process the sampled costs and health impacts to perform advanced analyses producing standardised and highly customisable outputs. BCEA is valuable for statisticians and practitioners working in the field of health economic modeling wanting to simplify and standardise their workflow, for example in the preparation of dossiers in support of marketing authorisation, or academic and scientific publications.

Statement of need

BCEA is a tool for interpreting and presenting the random sample of results from a CEA in a simple, powerful and standardised way, with useful, technically advanced measures and graphical summaries. BCEA was primarily written to use posterior distribution samples from a Bayesian model (e.g., run in WinBUGS or Stan) but can take any probability sensitivity analysis (PSA) random samples as inputs. BCEA also aims to be used in a health economic modeling workflow, meaning that it can be plugged-in as one of the steps in a CEA analysis. BCEA does not provide modeling functionality, like some other CEA packages such as hesim (Incerti & Jansen, 2021) or heemod (Filipović-Pierucci et al., 2017), but the package philosophy (borrowed from UNIX) is to do one thing well by focusing on the analysis following a model run. The package dampack (Alarid-Escudero et al., 2021) is also a decision analytic modeling package that provides a suite of functions for analyzing and visualizing the health economic outputs but takes a different design approach to BCEA. Although it has a focused scope, within this BCEA is designed to be extensible and flexible. Currently, BCEA has base R, ggplot2, and plotly versions of the plotting functions. The code is written so that computation of new statistics and new plotting functionality can be easily added. In BCEA the workflow centers around the bcea() function rather than separate functions for each type of statistic, with the aim to reduce the learning curve and easily expose the package functionality. Finally, BCEA has an expansive suite of functions from basic cost-effectiveness analyses, e.g. increment benefit (IB) and ICER calculation and plotting, to more sophisticated methods, e.g., Expected Value of Perfect Partial Information (EVPPI).

The breadth of models used for CEA is wide and growing in complexity and applications



(Krijkamp et al., 2018, 2019), but their implementation and, in particular, post-processing of their output can (and should be) standardised (Alarid-Escudero et al., 2019). This has the benefit of greater reliability, facilitating assessment and reuse. Decoupling the modeling from the post-processing allows for flexibility in the CEA model but, as long as its output is in a standard format, BCEA can be used. Thus, methodologies in CEA modeling can advance independently of the post-processing and presentation.

For further, in-depth details about BCEA we encourage the package user to consult Baio (2014) and Baio et al. (2017).

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References

- Alarid-Escudero, F., Knowlton, G., Easterly, C., & Enns, E. (2021). *Decision analytic modeling package (dampack)*. https://github.com/DARTH-git/dampack
- Alarid-Escudero, F., Krijkamp, E. M., Pechlivanoglou, P., Jalal, H., Kao, S., Yu, Z., Yang, A., & Enns, E. A. (2019). A need for change! A coding framework for improving transparency in decision modeling. *Pharmacoeconomics*, 37(11), 1329–1339. https://doi.org/10.1007/s40273-019-00837-x
- Baio, G. (2014). Bayesian models for cost-effectiveness analysis in the presence of structural zero costs. *Statistics in Medicine*, *33(11)*, 1900–1913. https://doi.org/10.1002/sim.6074
- Baio, G., Berardi, A., & Heath, A. (2017). *Bayesian cost-effectiveness analysis with the R package BCEA*. Springer International Publishing. https://doi.org/10.1007/978-3-319-55718-2
- Filipović-Pierucci, A., Zarca, K., & Durand-Zaleski, I. (2017). *Markov models for health economic evaluations: The R package heemod.* 1(1). http://arxiv.org/abs/1702.03252
- Incerti, D., & Jansen, J. P. (2021). hesim: Health economic simulation modeling and decision analysis. http://arxiv.org/abs/2102.09437
- Krijkamp, E. M., Alarid-Escudero, F., Enns, E. A., Jalal, H. J., Hunink, M. G., & Pechlivanoglou, P. (2018). Microsimulation modeling for health decision sciences using R: A tutorial. *Med. Decis. Mak.*, 38(3), 400–422. https://doi.org/10.1177/0272989X18754513
- Krijkamp, E. M., Alarid-Escudero, F., Enns, E. A., Pechlivanoglou, P., Hunink, M. G., & Jalal, H. J. (2019). A multidimensional array representation of state-transition model dynamics. bioRxiv, June, 670612. https://doi.org/10.1101/670612
- R Core Team. (2016). R: A language and environment for statistical computing. R Foundation for Statistical Computing. https://www.R-project.org/