

# Working with the R package BCEA

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```

ISPOR Short courses, Internetville

17 June 2021



### BCEA



### An R package for (Bayesian) cost-effectiveness analysis

#### What is BCFA not?

- BCEA is **not** a package to automatically run a Bayesian analysis
  - It cannot build the health economic model for you
  - It does not prepare the data to be used in the model
  - It does not automatically run the MCMC simulations
  - It does not choose the prior distributions for you

#### So what is it then?

- BCEA provides a set of specific functions to systematically post-process the output of a Bayesian health economic model
- Uses R http://cran.r-project.org/
  - Very good at interacting with standard MCMC software
  - BUGS: www.mrc-bsu.cam.ac.uk/bugs/winbugs/contents.shtml
  - JAGS: www.mcmc-jags.sourceforge.net/
  - rstan: https://mc-stan.org/users/interfaces/rstan
  - Free and there is a very large community of contributors
  - Specifically designed for statistical analysis and has very good graphical capabilities





## BCEA



### A R package for (Bayesian) cost-effectiveness analysis

#### BCEA and its use directly in R are designed with these objectives in mind

- Checking the model assumptions
  - Do we mean what we mean (eg in terms of PSA simulations)?...
  - Simulation error (especially, **but not only**, for a Bayesian approach)
- Produce the base-case economic evaluation
  - What's the most cost-effective intervention, given current evidence?
  - Cost-effectiveness plane, Expected Incremental Benefit (as a function of k),...
- Perform uncertainty analysis
  - Standard PSA (mandatory): Cost-effectiveness Plane, CEAC, ...
  - Fairly easy (but not always used): CEAF
  - More advanced/"too difficult" (rarely used): EVP(P)I/EVSI
- Standardised reporting
  - Graphical tools (use excellent R facilities)

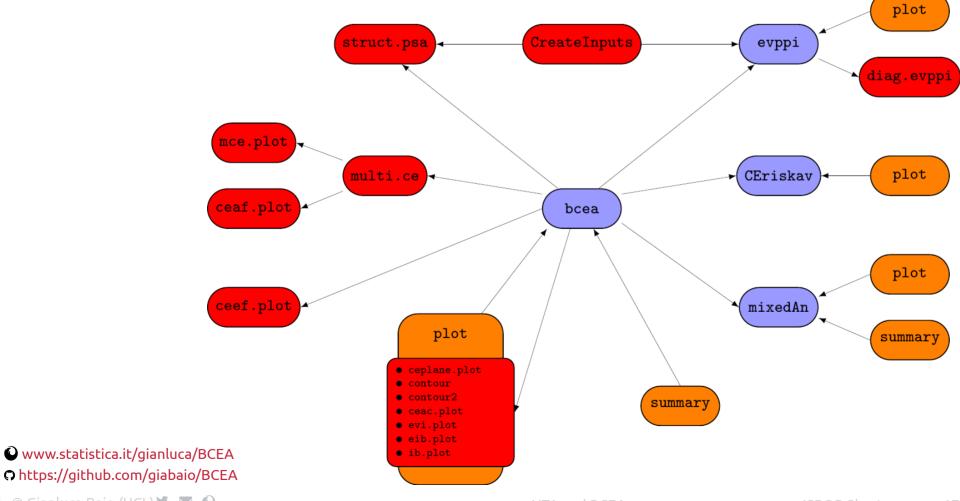




## **BCEA**



### An R package for (Bayesian) cost-effectiveness analysis







# Part I

# Using BCEA to summarise outputs of an economic model







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Installation Fconomic model Cost & effects Using BCEA Show, Me. The. Data

- BCFA is available from CRAN
  - Current *stable* version: 2.3-1.1 (26 August 2019)
- But it is also under constant development in the GitHub repository
  - Current *stable* version: 2.3-2 (30 January 2020)
  - Current *development* version: 2.4-1 (5 February 2021) **major** update (soon to be released!)

```
# Install BCEA (only required once and needs an internet connection!).
# You can either get the "official" version from CRAN
install.packages("BCEA")
# Or the development version from the GitHub repository
devtools::install_github("giabaio/BCEA")
                                                      # stable version (2.3-2)
devtools::install_github("giabaio/BCEA", ref="devel") # development version (2.4-1)
```

NB: The beauty of the GitHub version is that it can be updated on the fly and be immediately available for users!



Using BCEA



library(dplyr) # (Not necessary - helpful for data manipulation!) library(BCEA) # Then loads the package (so you can access its functions) data(Vaccine) # Loads an example dataset

Fconomic model

Cost & effects

- The "Vaccine" example is a fictional cost-effectiveness model for and influenza vaccine, based on evidence synthesis (and a real case)
- 2 treatment options ("Standard of care" vs "Vaccination") and overall 63 parameters

Show, Me. The. Data

- Discussed in details in Baio et al, 2017 and Baio and Dawid, 2011
- In this case, PSA simulations obtained from a full Bayesian model, but could be done in a spreadsheet and imported into R



Installation



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Show, Me. The. Data Economic model Cost & effects Installation Using BCEA # Creates a matrix with the underlying model simulations inp = CreateInputs(vaccine, print.lincom=FALSE) # NB: this option will be slightly different in 2.4-1! # Visualise the output inp\$mat %>% as tibble() # "piping" ('%>%') and formatting nicely # A tibble: 1,000 x 56 Adverse.events Death.1.1. Death.2.1. Death.2.2. GP.1.1. GP.2.1. GP.2.2. Hospital.1.1. Hospital.2.1. Hospital.2.2. Infected.1.1 <dbl><dbl><dbl><dbl><dbl> <dbl> <dbl><dbl><dbl><dbl><db] ... with 990 more rows, and 33 more variables: Trt.1.2.2. <dbl>, Trt.2.2.2. <dbl>, beta.1. <dbl>, beta.2. <dbl>, beta.3. <dbl>, k



pi.1.2. <dbl>, psi.1. <dbl>, psi.2. <dbl>, psi.3. <dbl>, psi.4. <dbl>, psi.5. <dbl>, psi.6. <dbl>, psi.7. <dbl>, psi.8. <dbl>



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Economic model Cost & effects Installation Using BCEA Show, Me. The. Data

```
# Defines the number of simulations considered
                                                      # applies the function 'nrow' (number of rows) to the object 'inp$mat'
n.sims=inp$mat %>% nrow()
                                                       # NB: in R 4.1.0, can also use 'native' pipe ('|>')
                                                       # (probably a bit quicker, but in most cases, may be immaterial...)
# Aggregates the model inputs to compute (e,c)
QALYs.inf = QALYs.pne <- QALYs.hosp <- QALYs.adv <- QALYs.death <- matrix(0,n.sims,2)
for (t in 1:2) {
 QALYs.inf[,t] = ((Infected[,t,1] + Infected[,t,2])*omega[,1]/365)/N
 QALYs.pne[,t] = ((Pneumonia[,t,1] + Pneumonia[,t,2])*omega[,4]/365)/N
 QALYs.hosp[,t] = ((Hospital[,t,1] + Hospital[,t,2])*omega[,5]/365)/N
  QALYs.death[,t] = ((Death[,t,1] + Death[,t,2])*omega[,6])/N
OALYs.adv[.2] = (Adverse.events*omega[.7]/365)/N
e = -(OALYs.inf + OALYs.pne + OALYs.adv + OALYs.hosp + OALYs.death) + ...
```

- NB: The data stored in the Vaccine object (built-in in BCEA) already contains the objects e, c that can be used to run the decision analysis...
- So, this step is actually not needed (but documented in Baio et al, 2017)





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Cost & effects Show. Me. The. Data Economic model Installation Using BCEA

```
cbind(e,c) %>% as_tibble(.name_repair="universal")
                                                           # ensures that the columns are named
# A tibble: 1,000 x 4
   Status. Ouo...1 Vaccination...2 Status. Quo...3 Vaccination...4
                             <dbl>
            <dbl>
                                             <dbl>
                                                              <dbl>
        -0.00105
                         -0.000899
                                             10.4
                                                              16.3
        -0.000884
                         -0.000732
                                              5.83
                                                               9.37
        -0.000890
                         -0.000698
                                              5.78
                                                              15.9
        -0.00164
                         -0.00114
                                             12.2
                                                              18.7
        -0.00135
                         -0.000957
                                              9.79
                                                              16.5
                                                               9.69
        -0.00143
                                              6.56
                         -0.000936
        -0.000960
                         -0.00105
                                              8.45
                                                              11.3
        -0.00181
                         -0.00139
                                              6.76
                                                               9.99
        -0.000842
                                              3.60
                                                              10.1
                         -0.000556
        -0.00168
                         -0.00105
                                              4.09
                                                              11.0
# ... with 990 more rows
```

• These calculations can be done also in a spreadsheet (nothing more than algebra, once you have the simulations)





• At this point, we are ready to call the function been that runs the economic analysis, for example something like

```
treats = c("Status quo", "Vaccination")
m = bcea(e=e,c=c,ref=2,interventions=treats.Kmax=50000)
```

- The inputs to the function are
  - e: a **matrix** containing the simulations for the clinical benefits (that is  $n_{
    m sim} imes n_{
    m int}$  values)
  - c: a **matrix** containing the simulations for the costs (that is  $n_{
    m sim} imes n_{
    m int}$  values)
  - ref: an indication of which intervention is to be taken as reference (default: the intervention in the first column of e or c)
  - interventions: a vector of labels for the interventions being compared
  - Kmax: the maximum value of k, the parameter of willingness to pay
- The output is an object m containing several elements

```
names(m)
                                                                                                                            "k"
[1] "n.sim"
                      "n.comparators" "n.comparisons" "delta.e"
                                                                         "delta.c"
                                                                                           "ICER"
                                                                                                            "Kmax"
[19] "interventions" "ref"
                                       "comp"
                                                        "step"
```







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Can visualise the output in various formats (tables/graphs)

```
# The 'summary' "method" produces a tabular output
summary(m)
```

Cost-effectiveness analysis summary

Reference intervention: Vaccination Comparator intervention: Status quo

Optimal decision: choose Status quo for k<20100 and Vaccination for k>=20100

Analysis for willingness to pay parameter k = 25000

Expected utility

Status quo -36.054 Vaccination -34.826

CEAC ICER

Vaccination vs Status quo 1.2284 0.529 20098

Optimal intervention (max expected utility) for k=25000: Vaccination

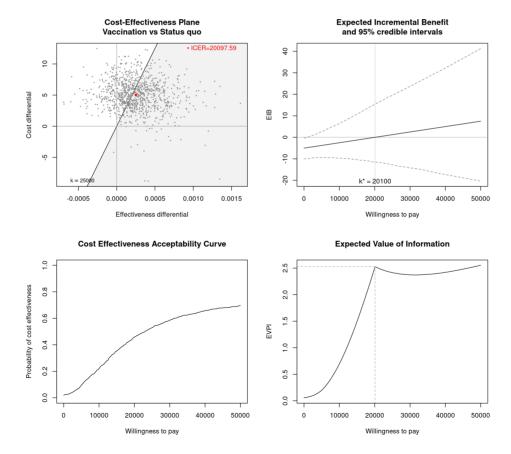
EVPI 2.4145





Can visualise the output in various formats (tables/graphs)

# The 'plot' "method" produces a \*specific\* version of graphical output plot(m)

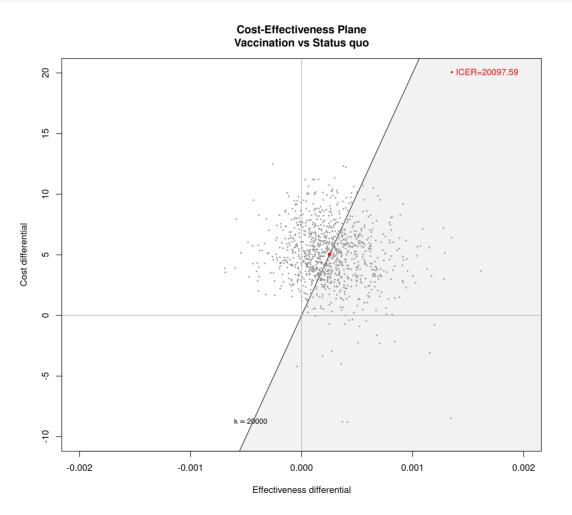








ceplane.plot(m,wtp=20000,xlim=c(-.002,.002),ylim=c(-10,20))







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# Using 'ggplot', you can go crazy with customisation...
ceplane.plot(m,wtp=10000,graph="gg",point\_colors="blue",point\_sizes=.8,area\_color="springgreen3")



• https://ggplot2.tidyverse.org/

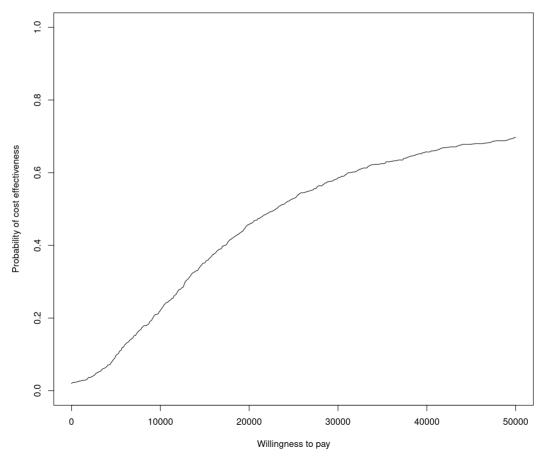






# Plots the Cost-Effectiveness Acceptability Curve ceac.plot(m)

#### **Cost Effectiveness Acceptability Curve**

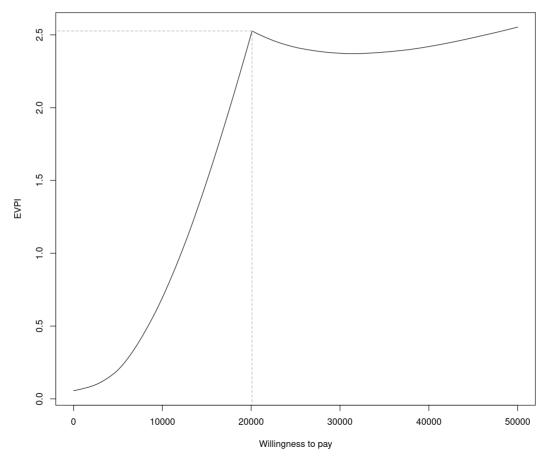






# Plots the Expected Value of Partial Information (EVPI) evi.plot(m)

#### **Expected Value of Information**





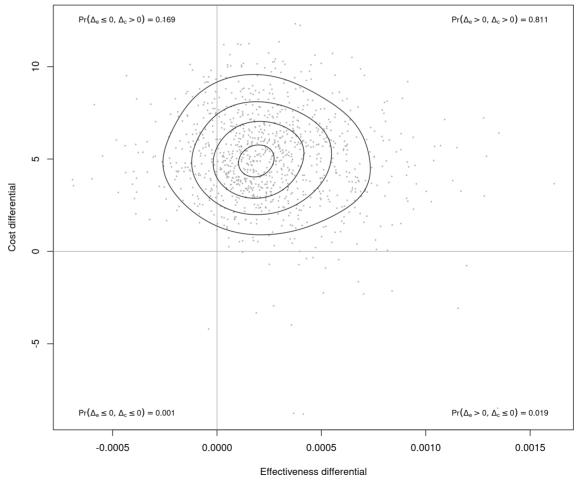
# Specialised plots



• Can generate a *contour* plot of the cost-effectiveness plane and estimate the proportion of points in each quadrant

# "Basic" contourplot contour(m)

#### Cost effectiveness plane contour plot Vaccination vs Status quo



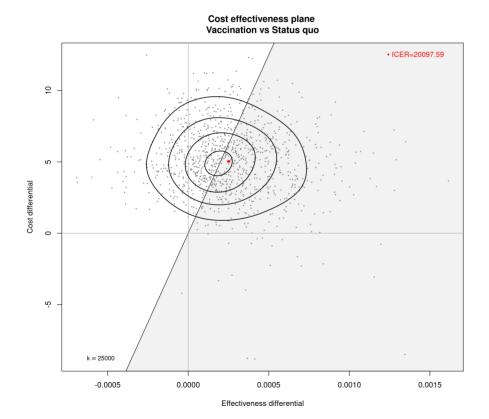


# Specialised plots

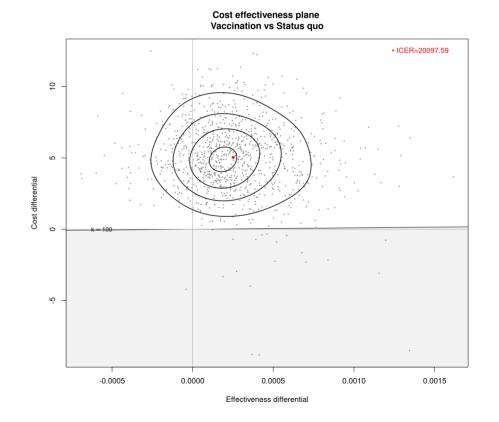


• The specialised function contour 2 also shows the sustainability area

contour2(m)



contour2(m,wtp=100)



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# Specialised plots



#### Cost-effectiveness efficiency frontier

ceef.plot(m,print.plot=FALSE)

Cost-effectiveness efficiency frontier summary

Interventions on the efficiency frontier:

Costs Increase slope Increase ang Effectiveness Status quo -0.00105595 9.6555 NA 1.5

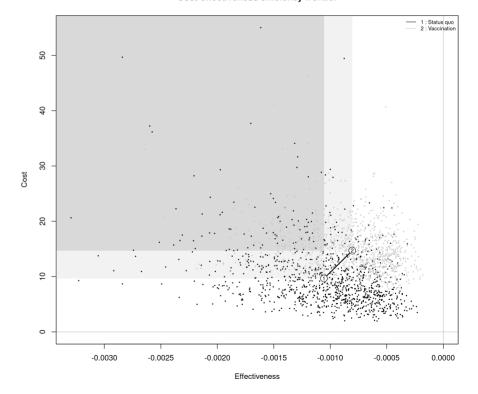
Vaccination

-0.00080537 14.6914

20098

ceef.plot(m,print.summary=FALSE)

#### Cost-effectiveness efficiency frontier





# Exporting graphical output



• R has excellent graphical facilities and the graphs produced by BCEA can be easily exported to many different formats

```
# "Opens" the graphical device
pdf("NAME_OF_THE_FILE", width=8, height=8)
                                               # for 'pdf', units are in inches
# Makes the plot
ceplane.plot(BCEA OBJECT)
                                              # of course, specify whatever name you've chosen when creating the object...
# "Closes" the graphical device
dev.off()
# "Open" the graphical device"
jpeg("NAME_OF_FILE.jpg",width=480,height=480) # for 'jpeg' units are in px
# Makes the plot
ceplane.plot(BCEA_OBJECT)
# "Closes" the graphical device
dev.off()
```

NB: Rstudio and rmarkdown can do even more – that's for another time...



# Part II

# Advanced use of BCEA







### Probabilistic "depression model"

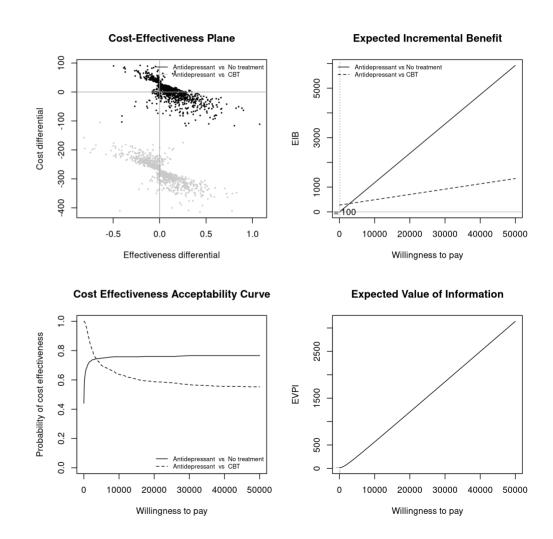
- Fictional model comparing antidepressants to cognitive behaviour therapy (CBT) and no treatment in people with depression
- Statistical modelling based on evidence synthesis
  - Benefits: based on QALYs
  - Costs: associated with treatments and various resources use
- Economic modelling: two matrices with relevant population summaries
  - effects
  - costs
- NB: The details of the actual modelling are *not* important for the purposes of demonstrating the example...





### Probabilistic "depression model"

```
# Intervention labels
t.names<-c("No treatment", "CBT", "Antidepressant")</pre>
# "Standard" analysis: pairwise comparisons
depression.bcea = bcea(effects,costs,
                       interventions=t.names, ref=3)
  the third intervention is the reference
# Plots the results
plot(depression.bcea)
```



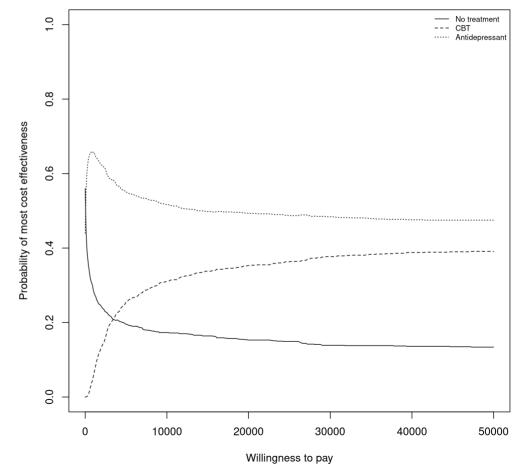




### Probabilistic "depression model"

```
# For multiple treatment comparison
depression.multi.ce = multi.ce(depression.bcea)
# Specialised plot method
```

#### Cost-effectiveness acceptability curve for multiple comparisons





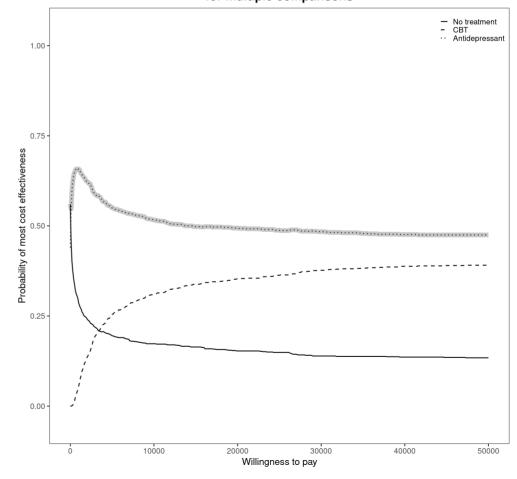


### Probabilistic "depression model"

Can use ggplot to customise the graph

```
mce.plot(depression.multi.ce,pos=c(1,1),graph="ggplot2") +
  ggplot2::stat_summary(fun=max, geom="line",
                        colour="grey25", alpha=.3, lwd=2.5)
```

#### Cost-effectiveness acceptability curve for multiple comparisons





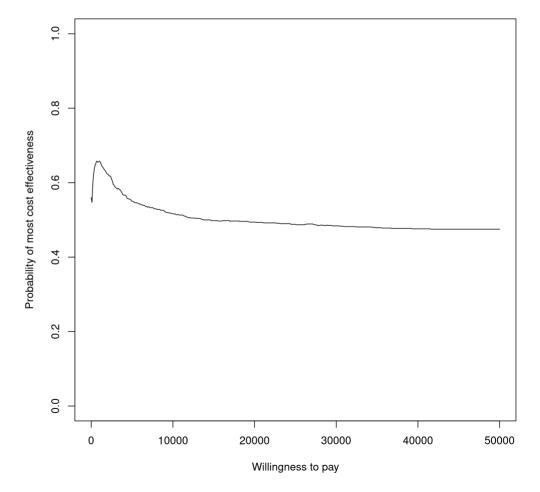


### Probabilistic "depression model"

• Can use a specialised plotting method for the costeffectiveness acceptability **frontier** 

```
# Specialised plot
ceaf.plot(depression.multi.ce)
```

#### Cost-effectiveness acceptability frontier







## **BCEAweb**



- Inspired by similar projects eg SAVI
- Create a web interface to use BCEA without even opening R (or even having it installed on your computer!)
- Typical work flow
  - Design the economic model (eg Markov model, decision tree, ...)
  - Run the statistical analysis to estimate the quantities of interest (eg survival analysis, evidence synthesis, ...)
  - Run the economic model and obtain "PSA samples"
  - Upload "PSA samples", including values for (e,c) to <code>BCEAweb</code>
  - Use BCEA in the background to do all the economic analysis
  - Create reports that can be used as the basis for papers, reimbursement files, ...



## **BCEAweb**



```
# Creates a matrix with the underlying model simulations
inp = CreateInputs(vaccine, print.lincom=FALSE)
# Runs BCEAweb
BCEAweb(e=e,
                            # matrix of simulations for the effectiveness
                            # matrix of simulations for the costs
       c=c.
                            # matrix of simulations for all the model parameters
       parameters=inp$mat
```

- BCEAweb exists as a standalone webapp
  - Can access it Ohere
- Or, you can launch your own "local" version from the BCEA package (as in the code above)!
  - This will launch a web page from which you can manipulate your output ( Live Demo)

