

BCEA: a R package to run Bayesian cost-effectiveness analysis

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What is BCEA **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
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So what *is* it, then?

- BCEA is a package that 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/>) to achieve this
 - R is particularly good at interacting with standard MCMC software
 - WinBUGS (<http://www.mrc-bsu.cam.ac.uk/bugs/winbugs/contents.shtml>)
 - JAGS (<http://mcmc-jags.sourceforge.net/>)
 - R is **free** and there is a very large community of contributors
 - R is specifically designed for statistical analysis and has very good graphical capabilities

- 1 Pre-process the data
 - Create, aggregate, modify original variables

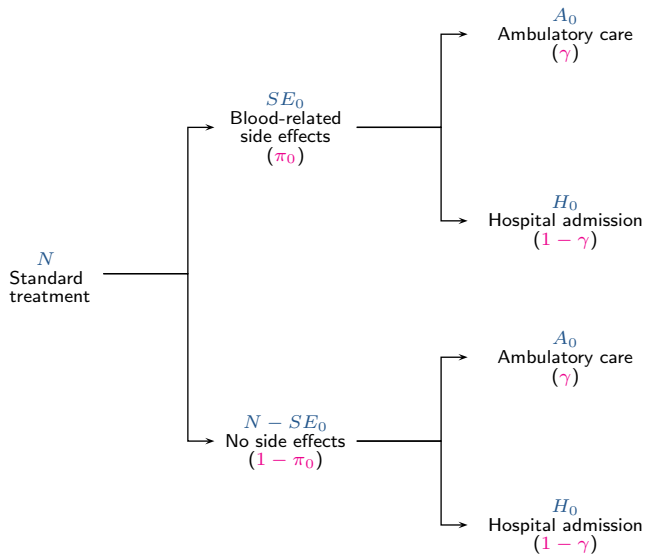
- ① Pre-process the data
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- ② Code the Bayesian model to describe the distributional assumptions
 - In BUGS/JAGS this is usually done in a `txt` file

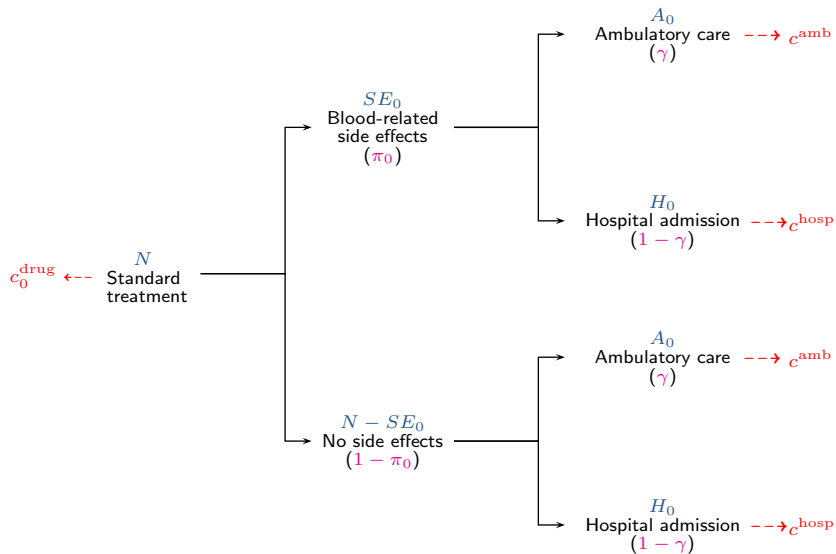
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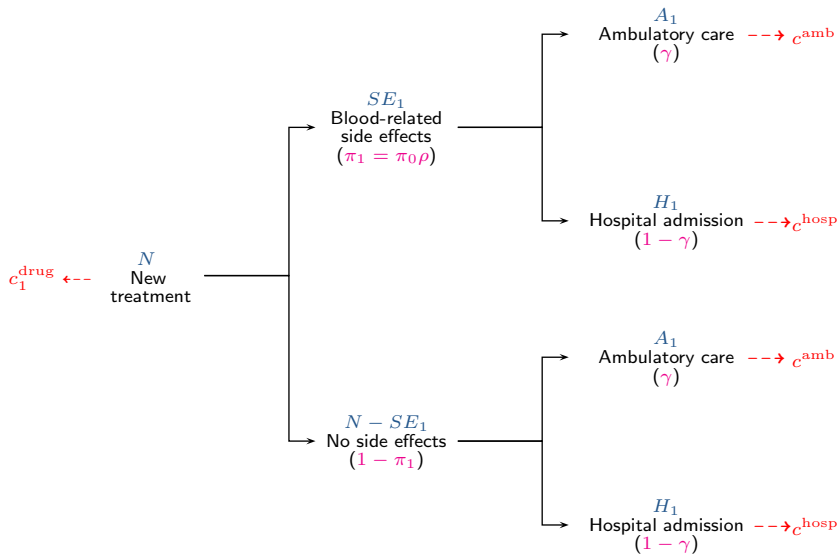
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 - Compute expected utilities and identify the most cost-effective intervention
 - Compute EIB, ICER, CEAC, EVPI, . . .
 - Cost-effectiveness plot and other graphical summaries

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- The model can be coded in BUGS/JAGS

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model {
  pi[1] ~ dbeta(a.pi,b.pi)           # Baseline probability of side effects (t=0)
  pi[2] <- pi[1]*rho                 # Decreased probability of side effects (t=1)
  rho ~ dnorm(m.rho,tau.rho)        # Decrement rate in side effects for t=1
  gamma ~ dbeta(a.gamma,b.gamma)    # Probability of ambulatory care
  c.amb ~ dlnorm(m.amb,tau.amb)      # Unit cost of ambulatory care
  c.hosp ~ dlnorm(m.hosp,tau.hosp)   # Unit cost of hospitalisation
  for (t in 1:2) {
    SE[t] ~ dbin(pi[t],N)            # Predicted no. patients with side effects
    A[t] ~ dbin(gamma,SE[t])         # Predicted no. patients needing ambulatory care
    H[t] <- SE[t] - A[t]            # Predicted no. patients needing hospitalisation
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- This will generate samples from the posterior distributions of the relevant quantities, eg: $\theta_t = (\pi_t, \gamma, \rho, SE_t, A_t, H_t)$

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- Using one of the libraries above, the simulations will be available in the R workspace and can be used to post-process the results of the Bayesian model to run the health economic evaluation

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`install.packages("BCEA")`

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`treats <- c("Old Chemotherapy", "New Chemotherapy")`

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```
treats <- c("Old Chemotherapy", "New Chemotherapy")
```

- Then, need to define the variables of cost and effectiveness in suitable matrices [*if not already available from the model*]

```
e <- c <- matrix(NA,1000,2)
e <- N - SE
for (t in 1:2) {
  c[,t] <- c.drug[t]*(N-SE[,t]) +
           (c.amb+c.drug[t])*A[,t] +
           (c.hosp+c.drug[t])*H[,t]
}
```

- Finally, we are ready to call the function `bcea` that runs the economic analysis
`m <- bcea(e=e,c=c,ref=2,interventions=treats,Kmax=50000)`

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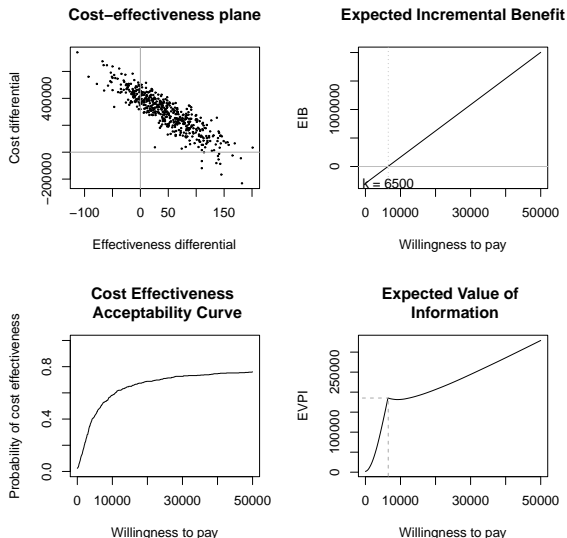
```
m <- bcea(e=e,c=c,ref=2,interventions=treats,Kmax=50000)
```
- The inputs of the function are
 - **e**: the matrix containing the simulations from the posterior distributions of the variable of clinical effectiveness ($n_{\text{sim}} \times n_{\text{int}}$ values)
 - **c**: the matrix containing the simulations from the posterior distributions of the variable of cost ($n_{\text{sim}} \times n_{\text{int}}$ values)
 - **ref**: an indication of which intervention is to be taken as reference (default: the intervention in the first column of `e` or `c`)
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- The output is an object `m`, containing several elements

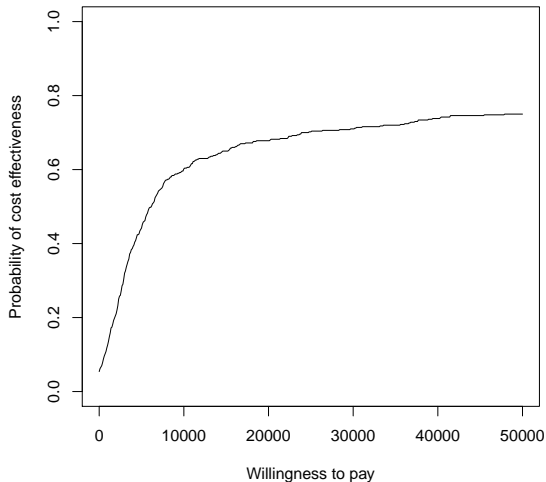
<code>m\$n.sim</code>	<code>m\$delta.c</code>	<code>m\$ceac</code>	<code>m\$best</code>	<code>m\$ol</code>	<code>m\$comp</code>
<code>m\$n.comparators</code>	<code>m\$ICER</code>	<code>m\$ib</code>	<code>m\$U</code>	<code>m\$evi</code>	<code>m\$step</code>
<code>m\$n.comparisons</code>	<code>m\$Kmax</code>	<code>m\$eib</code>	<code>m\$vi</code>	<code>m\$interventions</code>	
<code>m\$delta.e</code>	<code>m\$k</code>	<code>m\$kstar</code>	<code>m\$Ustar</code>	<code>m\$ref</code>	

- These elements are processed to produce relevant summaries and graphs
- For example, the code `plot(m)` generates

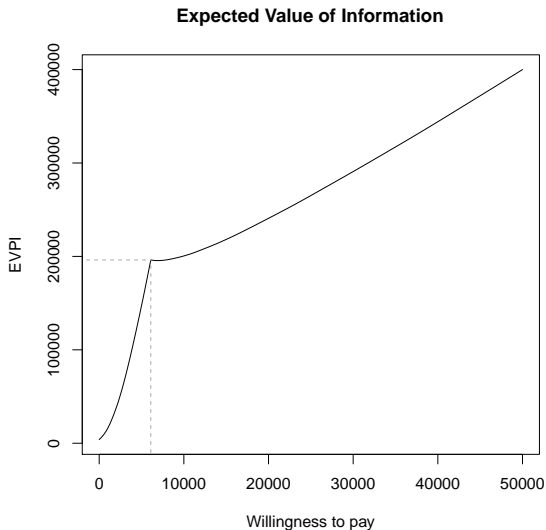


- The plots can be obtained separately
- For example: `ceac.plot(m)`

Cost Effectiveness Acceptability Curve



- The plots can be obtained separately
- For example: `evi.plot(m)`



- Similarly, the code `summary(m)` produces a synthesis of the main results

```
Cost-effectiveness analysis summary
```

```
Reference intervention: New Chemotherapy
```

```
Comparator intervention: Old Chemotherapy
```

```
Optimal decision: choose Old Chemotherapy for k<6500 and  
                  New Chemotherapy for k>=6500
```

```
Analysis for willingness to pay parameter k = 25000
```

```
                Expected utility  
Old Chemotherapy    18608376  
New Chemotherapy    19410840
```

```
                EIB  CEAC  ICER  
New Chemotherapy vs Old Chemotherapy 802465 0.728 6497.1
```

```
Optimal intervention (max expected utility) for k=25000: New Chemotherapy
```

```
EVPI 237560
```

- It is possible to specify additional options, eg `summary(m, wtp=1000)`

Cost-effectiveness analysis summary

Reference intervention: New Chemotherapy

Comparator intervention: Old Chemotherapy

Optimal decision: choose Old Chemotherapy for $k < 6500$ and
New Chemotherapy for $k \geq 6500$

Analysis for willingness to pay parameter $k = 1000$

	Expected utility
Old Chemotherapy	-44691
New Chemotherapy	-288267

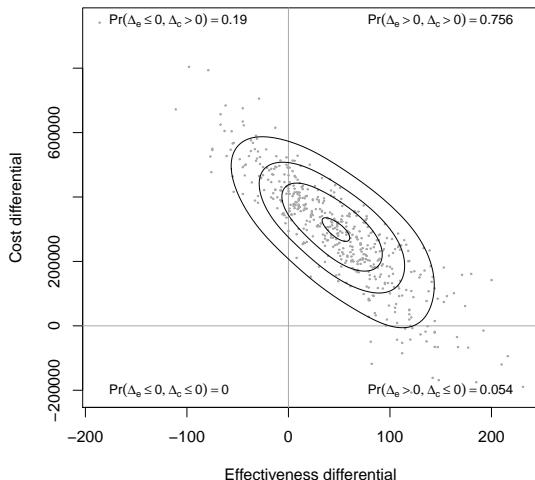
	EIB	CEAC	ICER
New Chemotherapy vs Old Chemotherapy	-243576	0.126	6497.1

Optimal intervention (max expected utility) for $k=1000$: Old Chemotherapy

EVPI 15877

- The code `contour(m)` generates a contourplot of the cost-effectiveness plane and also estimates the proportion of points lying in each quadrant

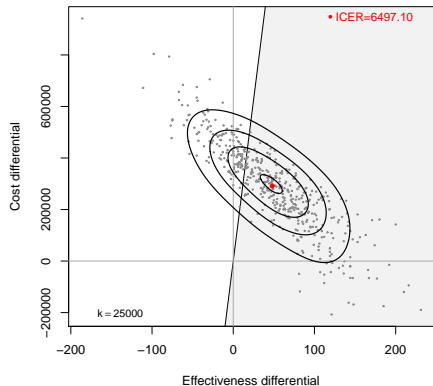
Cost effectiveness plane contour plot
New Chemotherapy vs Old Chemotherapy



The specialised function `contour2` also shows the **sustainability area**

- `contour2(m)`

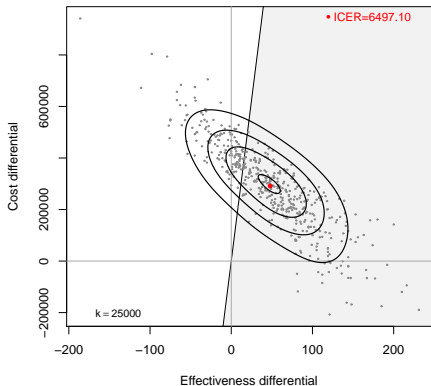
Cost effectiveness plane
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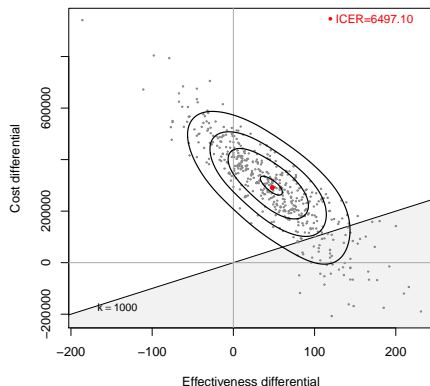
- `contour2(m)`

Cost effectiveness plane
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- `contour2(m, wtp=1000)`

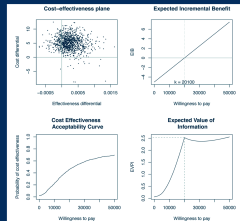
Cost effectiveness plane
New Chemotherapy vs Old Chemotherapy



- BCEA features heavily in the brilliant, forthcoming book on Bayesian methods in health economics (written by me 😊)
- In the book, I describe the entire process of *making* Bayesian analysis in health economics, including pre-processing of the data and running the model

Chapman & Hall/CRC Biostatistics Series

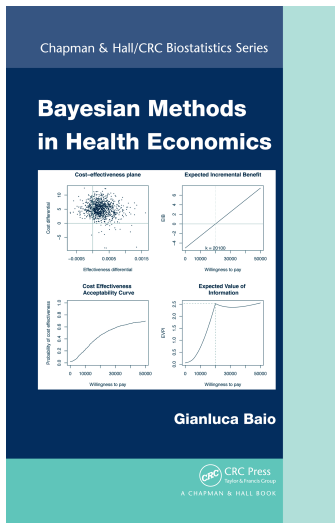
Bayesian Methods in Health Economics



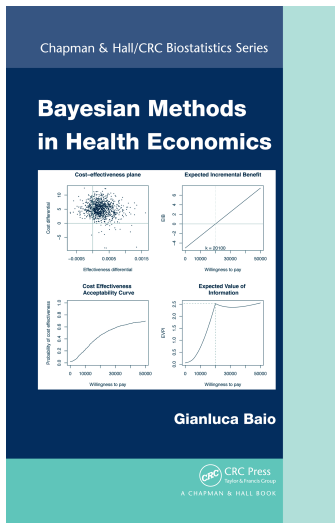
Gianluca Baio

 CRC Press
Taylor & Francis Group
A CHAPMAN & HALL BOOK

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 - Inclusion of risk aversion parameters
 - Mixed strategies
 - Possible extensions to compute the expected value of partial information



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- More info is available at the webpages www.statistica.it/gianluca/BMHE and www.statistica.it/gianluca/BCEA
- Also, some discussion (and more to come) in a few posts on gianlubai0.blogspot.co.uk
- And you can even ask me for the promo code!



Thank you!