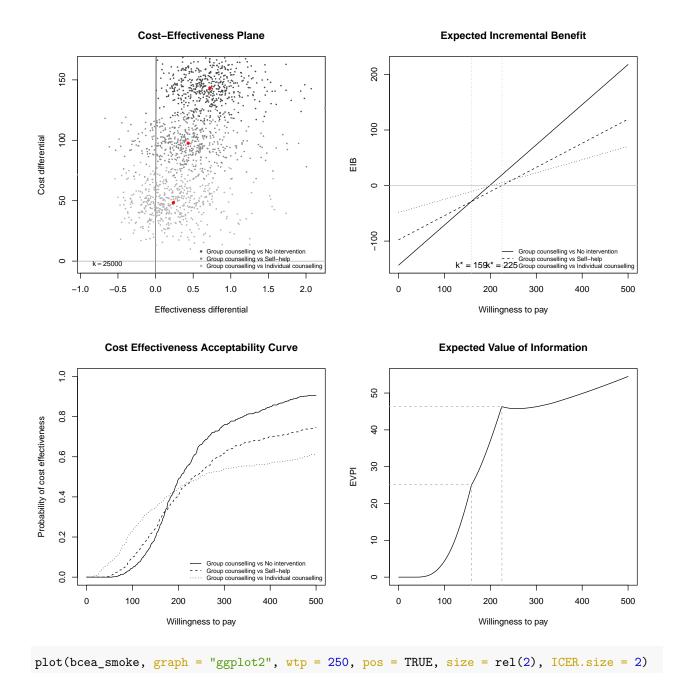
R journal examples

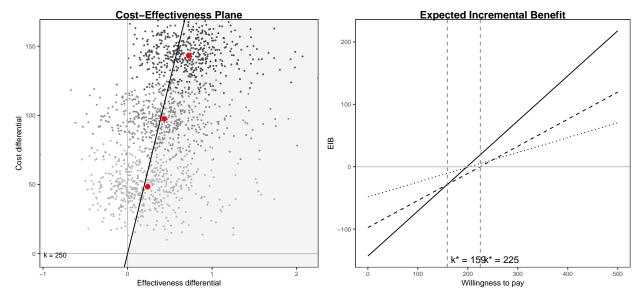
Smoking cessation example

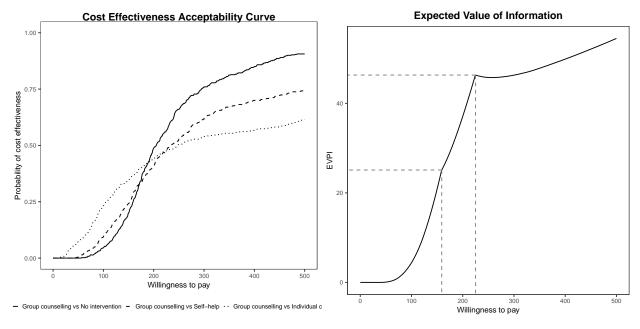
```
data(Smoking, package = "BCEA")

treats <- c("No intervention", "Self-help", "Individual counselling", "Group counselling")
bcea_smoke <- bcea(e, c, ref = 4, interventions = treats, Kmax = 500)

grid of plots
library(ggplot2)
plot(bcea_smoke)</pre>
```



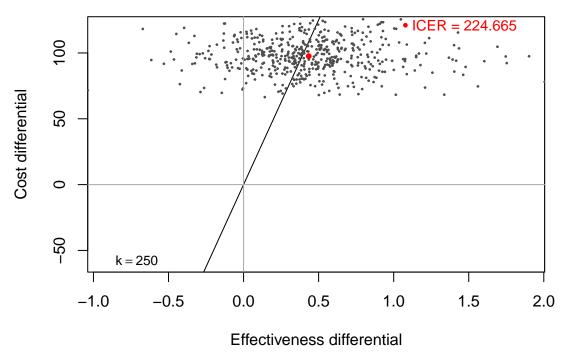




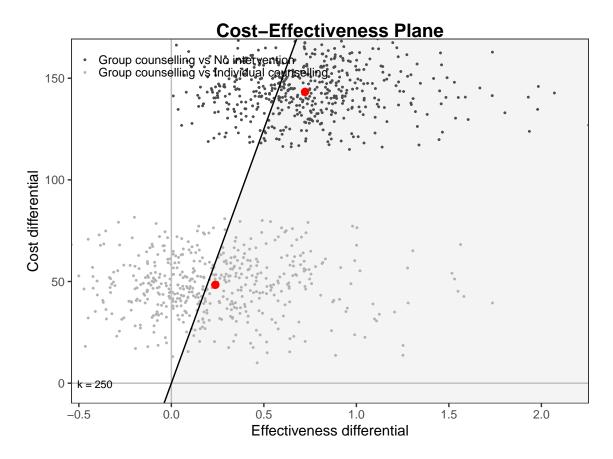
 $individual\ plots\ cost{-effectiveness}\ plane$

ceplane.plot(bcea_smoke, comparison = 2, wtp = 250)

Cost-Effectiveness Plane Group counselling vs Self-help



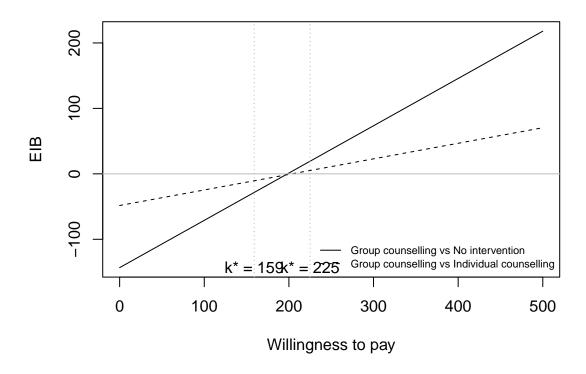
```
setComparisons(bcea_smoke) <- c(1,3)
ceplane.plot(bcea_smoke, wtp = 250, graph = "ggplot2")</pre>
```



other plots

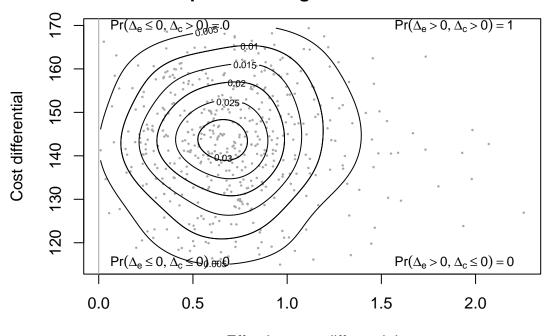
eib.plot(bcea_smoke)

Expected Incremental Benefit



contour(bcea_smoke)

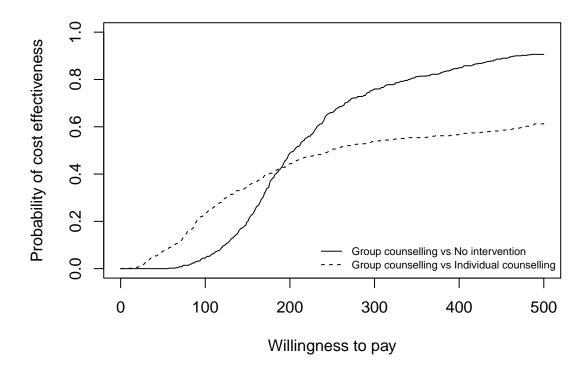
Cost effectiveness plane contour plot Group counselling vs No intervention



Effectiveness differential

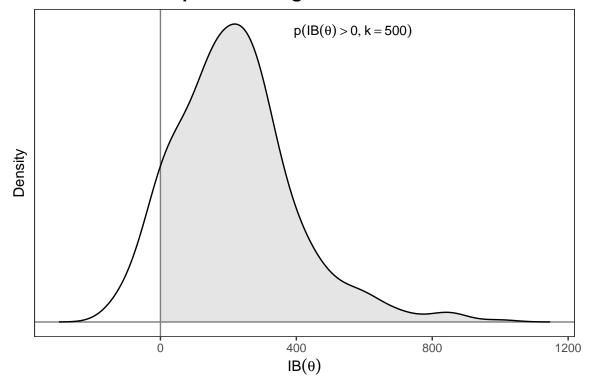
ceac.plot(bcea_smoke)

Cost Effectiveness Acceptability Curve



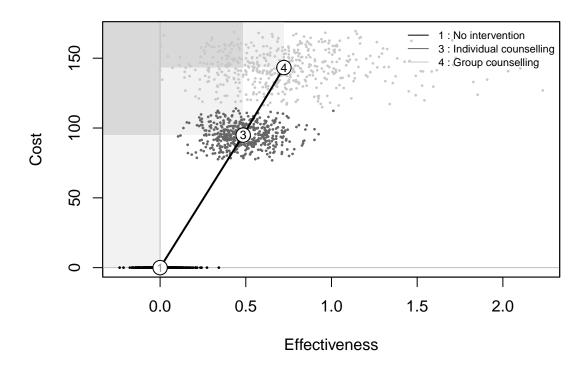
```
ib.plot(bcea_smoke)
#> NB: k (wtp) is defined in the interval [0 - 500]
```

Incremental Benefit Distribution Group counselling vs No intervention



```
ceef.plot(bcea_smoke)
#>
#> Cost-effectiveness efficiency frontier summary
#> Interventions on the efficiency frontier:
                    Effectiveness Costs Increase slope Increase angle
#>
                         0.48486 94.919 195.77
                                                   1.5657
#> Self-help
#> Individual counselling
                         0.72252 143.301
                                            203.57
                                                         1.5659
#> Interventions not on the efficiency frontier:
               Effectiveness Costs Dominance type
```

Cost-effectiveness efficiency frontier

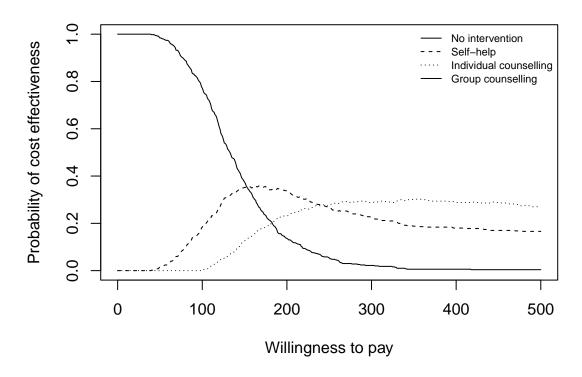


 ${\it multiple \; simultaneous \; comparisons}$

```
bcea_smoke <- multi.ce(bcea_smoke)

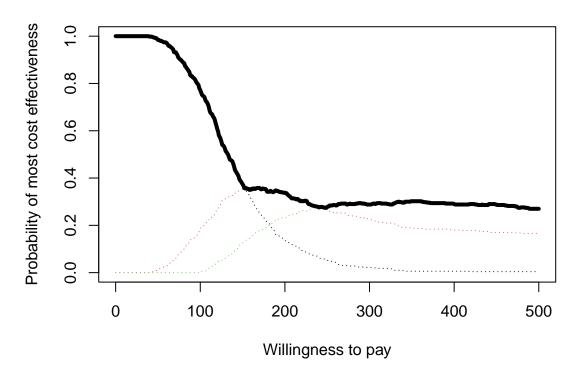
ceac.plot(bcea_smoke, pos = "topright")</pre>
```

Cost Effectiveness Acceptability Curve



ceaf.plot(bcea_smoke)

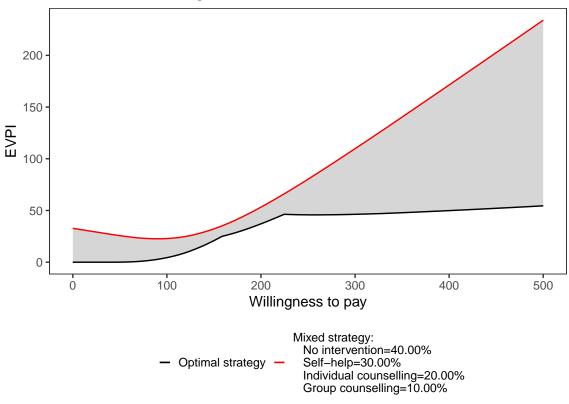
Cost-effectiveness acceptability frontier



mixed strategy

```
mixedAn(bcea_smoke) <- c(0.4, 0.3, 0.2, 0.1)
summary(bcea_smoke, wtp = 250)
#>
#> Analysis of mixed strategy for willingness to pay parameter k = 250
#>
#> Reference intervention: Group counselling (10.00% market share)
#> Comparator intervention(s): No intervention (40.00% market share)
#> : Individual counselling (20.00% market share)
#>
#> Loss in the expected value of information = 34.43
evi.plot(bcea_smoke, graph = "ggplot", pos = "b")
```

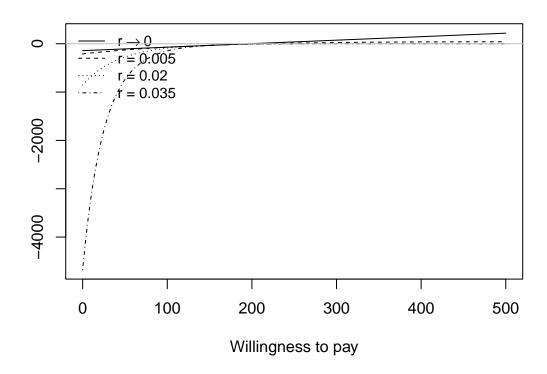
Expected Value of Information



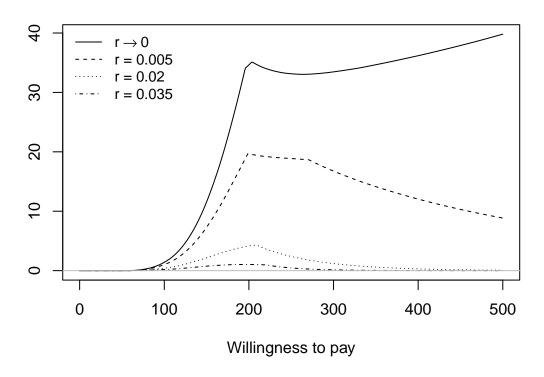
risk aversion

```
r <- c(0, 0.005, 0.020, 0.035)
CEriskav(bcea_smoke) <- r
plot(bcea_smoke)
```

EIB as a function of the risk aversion parameter



EVI as a function of the risk aversion parameter

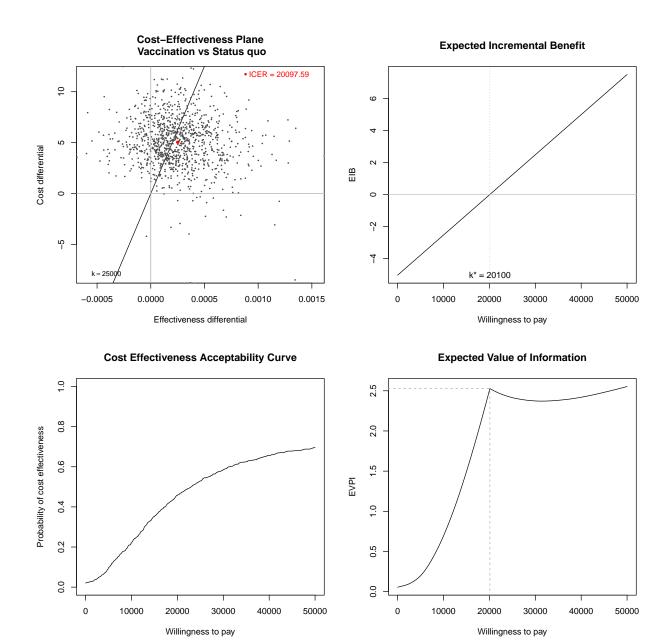


Influenza vaccine data

```
data(Vaccine)

treats <- c("Status quo", "Vaccination")
bcea_vacc <- bcea(e, c, ref = 2, interventions = treats)

grid of plots
plot(bcea_vacc)</pre>
```



summary output

```
summary(bcea_vacc, wtp = 10000)
#>

#> Cost-effectiveness analysis summary
#>

#> Reference intervention: Vaccination
#> Comparator intervention: Status quo
#>

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

#>

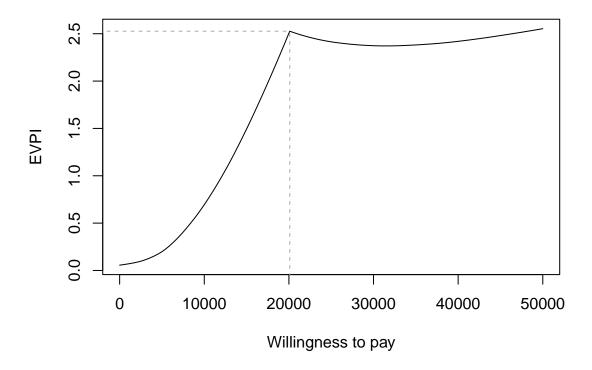
#> Analysis for willingness to pay parameter k = 10000
#>
```

```
Expected utility
                        -36.054
#> Status quo
#> Vaccination
                        -34.826
#>
#>
                               EIB CEAC ICER
#> Vaccination vs Status quo 1.2284 0.529 20098
#>
#> Optimal intervention (max expected utility) for k = 10000: Status quo
#>
#> EVPI 3.0287
head(sim_table(bcea_vacc, wtp = 25000)$Table)
            U1
                      U2
                               U*
                                        IB2_1
                                                               VI
#> 1 -36.57582 -38.71760 -36.57582 -2.1417866 2.141787 -1.135907
#> 2 -27.92514 -27.67448 -27.67448 0.2506573 0.000000
                                                         7.765431
#> 3 -28.03024 -33.37394 -28.03024 -5.3436963 5.343696
                                                         7.409665
#> 4 -53.28408 -47.13734 -47.13734 6.1467384 0.000000 -11.697432
#> 5 -43.58389 -40.40469 -40.40469 3.1791976 0.000000 -4.964782
#> 6 -42.37456 -33.08547 -33.08547 9.2890987 0.000000
```

value of information

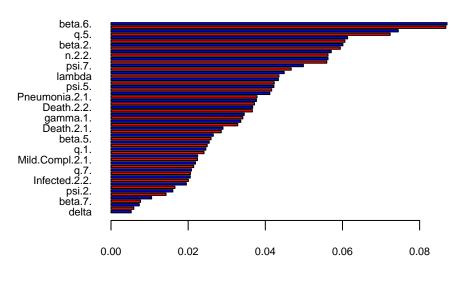
evi.plot(bcea_vacc)

Expected Value of Information



```
inp <- createInputs(vaccine, print_is_linear_comb = FALSE)
info.rank(bcea_vacc, inp)</pre>
```

Info-rank plot for willingness to pay = 20100



Proportion of total EVPI

```
EVPPI <- evppi(bcea_vacc, c("beta.1.", "beta.2."), inp$mat)
#> [1] "method: GAM" "method: GAM"
#>
#> Calculating fitted values for the GAM regression
#>
#> Calculating fitted values for the GAM regression
#> Calculating fitted values for the GAM regression
#> Calculating EVPPI
plot(EVPPI)
```

Expected Value of Perfect Partial Information

