

Additional file 1: Models, Cluster-Level Mortality Data, R Source Code and Output

This additional file contains the following:

Pages 2-3 and 4-5: the two statistical models (MKmodel_final and MKmodelFC_final) for modelling the effects of the community and facility interventions and community and facility interventions combined – see the lines they are referred to in the R Source Code.

Page 6 and Page 7: the two cluster-level mortality datasets, the first, MK.csv, for the community and facility interventions, the two-by-two arm trial [hf denotes cluster (health facility) number; fi = facility intervention; ci = community intervention; urb = urban (1) or rural (0); llw and sal are variables for the two districts the sample was also stratified by; mvital = maternal deaths; bvital = baby deaths (stillbirth or neonatal death); obs = the total number of observations in each cluster]; the second, MKFC.csv, for the combined intervention, the one-arm trial [variables as before; fici = combined facility and community intervention] – see the lines they are referred to in the R Source Code.

Pages 8-10 and 11-21: R Source Code for the main analysis, and the future scenario analysis, respectively.

Pages 22-27, 28-33 and 34-48: R Output for the main analysis, the local life expectancy sensitivity analysis [note that for this analysis the only difference in the R Source file from the main analysis – see pages 8-10 - is that 86.0 and 53.27 are replaced by 45 and 28.1, respectively], and the future scale-up scenarios analysis, respectively.

```

# MKmodel_final

model {
  for (i in 1:length(hf)) {
    # combined baby deaths
    bdeath[i] ~ dbin(b[i],obs[i])
    logit(b[i]) <- alpha0_b + fac[1]*fi[i] + com[1]*ci[i] + strata2[1]*llw[i] +
strata3[1]*sal[i] + urban[1]*urb[i]
    # Maternal deaths
    mvital[i] ~ dbin(m[i],obs[i])
    logit(m[i]) <- alpha0_m + fac[2]*fi[i] + com[2]*ci[i] + strata2[2]*llw[i] +
strata3[2]*sal[i] + urban[2]*urb[i]
  }
  #Blocking of coefficients (Priors)
  fac[1:2] ~ dnorm(mu[1:2],P[1:2,1:2])
  com[1:2] ~ dnorm(mu[1:2],P[1:2,1:2])
  strata2[1:2] ~ dnorm(mu[1:2],P[1:2,1:2])
  strata3[1:2] ~ dnorm(mu[1:2],P[1:2,1:2])
  urban[1:2] ~ dnorm(mu[1:2],P[1:2,1:2])

  # Priors
  alpha0_b ~ dnorm(0.0, 1.0E-6)
  alpha0_m ~ dnorm(0.0, 1.0E-6)

#cost-effectiveness

pop <- 108000 # population of babies and mothers in 2 arms (CI, No CI, FI, no FI) in 2.25
years (1.2m * CBR of 0.04 * 27/12)

bmr_nc <- 0.059443785 # baby mortality rate in No CI arm: 219 SB + 232 NND out of 7587 births
bmr_nf <- 0.061760305 # baby mortality rate in No FI arm: 237 SB + 217 NND out of 7351 births
mmr_nc <- 0.0038223277 # mother mortality rate in No CI arm: 29 MD out of 7587 births
mmr_nf <- 0.0039450415 # mother mortality rate in No FI arm: 29 MD out of 7351 births

bm_nc ~ dbin(bmr_nc,pop) # baby mortality in No CI arm
bm_nf ~ dbin(bmr_nf,pop) # baby mortality in No FI arm
mm_nc ~ dbin(mmr_nc,pop) # mother mortality in No CI arm
mm_nf ~ dbin(mmr_nf,pop) # mother mortality in No FI arm

bs_nc <- pop - bm_nc # baby survivors in No CI arm
bs_nf <- pop - bm_nf # baby survivors in No FI arm
ms_nc <- pop - mm_nc # mother survivors in No CI arm
ms_nf <- pop - mm_nf # mother survivors in No CI arm

ob_nc <- bm_nc / bs_nc # odds of baby death in No CI areas
ob_nf <- bm_nf / bs_nf # odds of baby death in No FI areas
om_nc <- mm_nc / ms_nc # odds of mother death in No CI areas
om_nf <- mm_nf / ms_nf # odds of mother death in No FI areas

orb_c <- exp(com[1]) # Odds Ratio of CI on BMR
orb_f <- exp(fac[1]) # Odds Ratio of FI on BMR
orm_c <- exp(com[2]) # Odds Ratio of CI on MMR
orm_f <- exp(fac[2]) # Odds Ratio of FI on MMR

ob_c <- ob_nc * orb_c # odds of baby death in CI arm (odds in No CI arms * Odds Ratio of CI on
BMR)
ob_f <- ob_nf * orb_f # odds of baby death in FI arm (odds in No FI arms * Odds Ratio of FI on
BMR)
om_c <- om_nc * orm_c # odds of mother death in CI arm (odds in No CI * Odds Ratio of CI on
MMR)
om_f <- om_nf * orm_f # odds of mother death in FI arm (odds in No FI * Odds Ratio of FI on
MMR)

ob_c1 <- ob_c + 1 # odds of baby death in CI areas + 1 (for calculation of survivors in
intervention arm)
ob_f1 <- ob_f + 1 # odds of baby death in FI areas + 1 (for calculation of survivors in
intervention arm)
om_c1 <- om_c + 1 # odds of mother death in CI areas + 1 (for calculation of survivors in
intervention arm)
om_f1 <- om_f + 1 # odds of mother death in FI areas + 1 (for calculation of survivors in
intervention arm)

bs_c <- pop / ob_c1 # baby survivors in CI arm
bs_f <- pop / ob_f1 # baby survivors in FI arm
ms_c <- pop / om_c1 # mother survivors in CI arm

```

```

ms_f <- pop / om_f1 # mother survivors in FI arm

bm_c <- pop - bs_c # baby deaths in CI arm
bm_f <- pop - bs_f # baby deaths in FI arm
mm_c <- pop - ms_c # mother deaths in CI arm
mm_f <- pop - ms_f # mother deaths in FI arm

daly_b <- 86.0 # Disability Adjusted Life Years (DALYs) per baby death
daly_m <- 53.27 # Disability Adjusted Life Years (DALYs) per mother death

lb_c <- bm_nc - bm_c # lives of babies saved (deaths averted) by CI (difference in mortality
between No CI and CI arms)
db_c <- lb_c * daly_b # DALYs of babies averted by CI

lm_c <- mm_nc - mm_c # lives of mothers saved (deaths averted) by CI (difference in
mortality between No CI and CI arms)
dm_c <- lm_c * daly_m # DALYs of mothers averted by CI
d_c <- db_c + dm_c # total DALYs (mothers and babies) averted by CI

lb_f <- bm_nf - bm_f # lives of babies saved (deaths averted) by FI (difference in mortality
between No FI and FI arms)
db_f <- lb_f * daly_b # DALYs of babies averted by FI

lm_f <- mm_nf - mm_f # lives of mothers saved (deaths averted) by FI (difference in
mortality between No FI and FI arms)
dm_f <- lm_f * daly_m # DALYs of mothers averted by FI
d_f <- db_f + dm_f # total DALYs (mothers and babies) averted by FI

}

```

```

# MKmodelFC_final

model {
  for (i in 1:length(hf)) {
    # combined baby deaths
    bdeath[i] ~ dbin(b[i],obs[i])
    logit(b[i]) <- alpha0_b + fc[1]*fici[i] + strata2[1]*llw[i] +
strata3[1]*sal[i] + urban[1]*urb[i]
    # Maternal deaths
    mvital[i] ~ dbin(m[i],obs[i])
    logit(m[i]) <- alpha0_m + fc[2]*fici[i] + strata2[2]*llw[i] +
strata3[2]*sal[i] + urban[2]*urb[i]
  }
  #Blocking of coefficients
  fc[1:2] ~ dnorm(mu[1:2],P[1:2,1:2])
  strata2[1:2] ~ dnorm(mu[1:2],P[1:2,1:2])
  strata3[1:2] ~ dnorm(mu[1:2],P[1:2,1:2])
  urban[1:2] ~ dnorm(mu[1:2],P[1:2,1:2])

  # Priors
  alpha0_b ~ dnorm(0.0, 1.0E-6)
  alpha0_m ~ dnorm(0.0, 1.0E-6)

#cost-effectiveness

pop <- 54000 # population of babies and mothers in 1 control arm in 2.25 years (600,000 * CBR
of 0.04 * 27/12)

bmr_con <- 0.05809575 # baby mortality rate in control arm: 105 SB + 111 NND out of 3718
births
mmr_con <- 0.0040344271 # mother mortality rate in control arm: 15 MD out of 3718 births

bm_con ~ dbin(bmr_con,pop) # baby mortality in control arm
mm_con ~ dbin(mmr_con,pop) # mother mortality in control arm

bs_con <- pop - bm_con # baby survivors in control arm
ms_con <- pop - mm_con # mother survivors in control arm

ob_con <- bm_con / bs_con # odds of baby death in control arm
om_con <- mm_con / ms_con # odds of mother death in control arm

orb_fc <- exp(fc[1]) # Odds Ratio of FICI on BMR
orm_fc <- exp(fc[2]) # Odds Ratio of FICI on MMR

ob_fc <- ob_con * orb_fc # odds of baby death in FICI arm (odds in control arm * Odds Ratio of
FICI on BMR)
om_fc <- om_con * orm_fc # odds of mother death in FICI arm (odds in control arm * Odds Ratio
of FICI on MMR)

ob_fc1 <- ob_fc + 1 # odds of baby death in FICI arm + 1 (for calculation of survivors in
intervention arm)
om_fc1 <- om_fc + 1 # odds of mother death in FICI arm + 1 (for calculation of survivors in
intervention arm)

bs_fc <- pop / ob_fc1 # baby survivors in FICI arm
ms_fc <- pop / om_fc1 # mother survivors in FICI arm

bm_fc <- pop - bs_fc # baby deaths in FICI arm
mm_fc <- pop - ms_fc # mother deaths in FICI arm

daly_b <- 86.0 # Disability Adjusted Life Years (DALYs) per baby death
daly_m <- 53.27 # Disability Adjusted Life Years (DALYs) per mother death

lb_fc <- bm_con - bm_fc # lives of babies saved (deaths averted) by FICI (difference in
mortality between control and FICI arms)
db_fc <- lb_fc * daly_b # DALYs of babies averted by FICI

lm_fc <- mm_con - mm_fc # lives of mothers saved (deaths averted) by FICI (difference in
mortality between control and FICI arms)
dm_fc <- lm_fc * daly_m # DALYs of mothers averted by FICI
d_fc <- db_fc + dm_fc # total DALYs (mothers and babies) averted by FICI

#For Table 1 credible intervals for two-arm FICI:
lb_fc_ <- lb_fc * 2
db_fc_ <- db_fc * 2

```

```
lm_fc_ <- lm_fc * 2  
dm_fc_ <- dm_fc * 2  
d_fc_ <- d_fc * 2  
}
```

MK.csv

hf	fi	ci	urb	llw	sal	mvital	bdeath	obs
1	1	1	0	0	1	2	24	335
2	0	0	1	1	0	1	8	131
3	1	0	1	1	0	0	8	224
4	1	0	0	0	0	2	20	302
5	1	0	0	1	0	0	7	159
6	0	0	0	1	0	0	18	258
7	1	1	0	1	0	0	13	315
8	0	0	0	1	0	2	4	77
9	1	0	0	0	1	1	29	463
10	1	1	0	0	1	1	33	524
11	1	0	0	0	1	1	24	373
12	1	0	0	1	0	1	34	359
13	0	1	0	1	0	1	28	387
14	1	1	0	1	0	2	24	425
15	0	0	0	0	0	0	32	397
16	0	0	0	1	0	1	18	215
17	0	1	0	1	0	1	22	407
18	0	0	0	1	0	0	11	123
19	1	0	0	0	0	3	24	526
20	0	0	0	0	0	0	16	226
21	1	0	0	0	0	0	36	456
22	0	1	0	0	1	1	43	691
23	0	0	0	1	0	0	12	309
24	0	0	1	1	0	0	5	132
25	1	1	0	0	0	0	29	352
26	0	1	0	0	0	0	14	200
27	1	1	0	0	1	1	8	268
28	1	1	0	1	0	1	25	577
29	1	0	0	0	0	3	17	503
30	1	0	0	1	0	0	34	474
31	0	0	0	0	1	3	31	582
32	0	0	0	0	1	1	34	526
33	1	1	0	1	0	1	14	215
34	1	0	0	1	0	0	26	421
35	1	0	0	1	0	1	8	262
36	1	1	0	1	0	0	6	188
37	0	1	0	1	0	0	8	249
38	0	1	0	1	0	1	23	407
39	1	1	0	1	0	0	14	204
40	0	1	0	1	0	0	10	187
41	0	1	0	0	1	1	32	515
42	0	0	0	1	0	0	20	268
43	0	0	0	0	0	0	19	221
44	0	1	0	0	0	2	11	284
45	0	0	0	1	0	0	16	371
46	1	1	0	0	0	3	19	389
47	1	1	0	0	0	0	20	358
48	0	0	0	1	0	0	20	323
49	0	1	0	1	0	0	7	213
50	1	1	0	1	0	0	19	423
51	1	0	0	0	1	2	25	490
52	1	1	0	1	0	0	20	332
53	0	1	0	0	0	3	24	529
54	0	1	0	1	0	0	10	148
55	0	1	0	1	0	0	7	235
56	1	1	0	0	0	0	12	345
57	0	0	0	0	1	1	7	146
58	0	1	0	0	0	1	24	298
59	0	1	0	0	1	0	17	335
60	1	0	0	1	0	1	24	327
61	0	0	0	0	0	1	37	610

MKFC.csv

strata	arm	hf	fici	urb	llw	sal	mvital	bdeath	obs
3	3	1	1	0	0	1	2	24	335
2	0	2	0	1	1	0	1	8	131
2	0	3	0	0	1	0	0	18	258
2	3	4	1	0	1	0	0	13	315
2	0	5	0	0	1	0	2	4	77
3	3	6	1	0	0	1	1	33	524
2	3	7	1	0	1	0	2	24	425
1	0	8	0	0	0	0	0	32	397
2	0	9	0	0	1	0	1	18	215
2	0	10	0	0	1	0	0	11	123
1	0	11	0	0	0	0	0	16	226
2	0	12	0	0	1	0	0	12	309
2	0	13	0	1	1	0	0	5	132
1	3	14	1	0	0	0	0	29	352
3	3	15	1	0	0	1	1	8	268
2	3	16	1	0	1	0	1	25	577
3	0	17	0	0	0	1	3	31	582
3	0	18	0	0	0	1	1	34	526
2	3	19	1	0	1	0	1	14	215
2	3	20	1	0	1	0	0	6	188
2	3	21	1	0	1	0	0	14	204
2	0	22	0	0	1	0	0	20	268
1	0	23	0	0	0	0	0	19	221
2	0	24	0	0	1	0	0	16	371
1	3	25	1	0	0	0	3	19	389
1	3	26	1	0	0	0	0	20	358
2	0	27	0	0	1	0	0	20	323
2	3	28	1	0	1	0	0	19	423
2	3	29	1	0	1	0	0	20	332
1	3	30	1	0	0	0	0	12	345
3	0	31	0	0	0	1	1	7	146
1	0	32	0	0	0	0	1	37	610

```

setwd("/Users/timothycolbourn/Documents/Papers/MaiKhanda CEA paper/JAGS revised") #set working directory for
saving outputs
sink(file="MaiKhanda_final_log_.txt",split=TRUE)
# MaiKhanda Economic Evaluation v4 in R and JAGS Tim Colbourn 2nd Dec 2014
install.packages("R2jags")
library(R2jags)
install.packages("BCEA")
library(BCEA)

# Community Intervention and Facility Intervention model

mk <- read.csv('MK.csv') # data exported from Stata using 'MK_WinBUGS_final_cluster.do' and also stored in
'MKcluster.txt'

h <- 100
mu <- c(0,0)
P <- (1/h^2)*diag(2)

data <- list(mu=mu,P=P,hf=mk$hf,fi=mk$fi,ci=mk$ci,urb=mk$urb,llw=mk$llw,sal=mk$sal,mvital=mk$mvital,bdeath=mk
$bdeath,obs=mk$obs)

filein <- "/Users/timothycolbourn/Documents/Papers/MaiKhanda CEA paper/JAGS revised/MKmodel_final.txt" #filepath
to text file containing model

inits <- function(){
  list(alpha0_b=rnorm(1), alpha0_m=rnorm(1), strata2=rnorm(2,0,1), strata3=rnorm(2,0,1), urban=rnorm(2,0,1),
  fac=rnorm(2,0,1), com=rnorm(2,0,1))
  } # initial values of paramters

params <-
c("bs_nc","bs_c","ms_nc","ms_c","bs_nf","bs_f","ms_nf","ms_f","orb_c","orm_c","orb_f","orm_f","lb_c","db_c","lm_c",
,"dm_c","lb_f","db_f","lm_f","dm_f","d_c","d_f") #paramters to monitor

model <- jags(data,inits,params,model.file=filein,n.chains=2,n.iter=510000,n.burnin=10000,n.thin=10,DIC=TRUE)

print(model,digits=3,intervals=c(0.025,0.500,0.975))

m <- as.mcmc(model) #converts the object model to an MCMC structure, for plotting and diagnostics
raftery.diag(m) #Raftery-Lewis diagnostic estimates of (for each chain[[1]],[[2]]..) number of iterations
required to estimate 2.5th quantile (=q) of posterior dist, with accuracy (tolerance) r of +/-0.005 and
probability s of 0.95 (defaults)
summary(m) #MCMC error and other indications of convergence. MCMC error is how close posterior SD is to Naive SE.
Time-series SE adjusts Naive SE for autocorrelation; lower values = better convergence

#Health Economic analysis, including EVPPi (p.145-151) - via BCEA package loaded earlier
attach.jags(model)
# Community Intervention (CI) vs Do Nothing
ec <- cc <- matrix(NA,model$BUGSoutput$n.sims,2)
ec <- matrix(c((bs_nc*86.0 + ms_nc*53.27),(bs_c*86.0 + ms_c*53.27)),model$BUGSoutput$n.sims,2) # effects = baby
(times 86.0 DALYs each) survivors + mother survivors (times 53.27 DALYs each) in No CI, baby (times 86.0 DALYs
each) survivors + mother survivors (times 53.27 DALYs each) in CI
cc <- matrix(c(rep(0,model$BUGSoutput$n.sims),rep(5348791,model$BUGSoutput$n.sims)),model$BUGSoutput$n.sims,2) #
costs = cost of do nothing, cost of community intervention
treatsc <- c("do nothing","Community Intervention")
ci <- bcea(e=ec,c=cc,ref=2,interventions=treatsc,Kmax=2500) # Max cost-effectivness threshold is over 3 times
GDPpc per DALY averted
layout(matrix(1:4,2,2)) #so the four health economic analysis plots from the following commands are on one page
ceplane.plot(ci,comparison=1,wtp=779.8) # Cost-effectiveness plane - Willingness to Pay (WTP) threshold is
$779.8 GDPpc (per DALY averted)
eib.plot(ci) # Expected Incremental Benefit
ceac.plot(ci) # Cost-effectiveness Acceptability Curve (prob cost-effective by WTP)
evi.plot(ci) # Expected value of information by willingness to pay
quartz.save("Figure1.pdf", type="pdf")
summary(ci,wtp=780) # Increments of 5 are only allowed in this command therefore 779.8 is not allowed
# Facility Intervention (FI) vs Do Nothing
ef <- cf <- matrix(NA,model$BUGSoutput$n.sims,2)
ef <- matrix(c((bs_nf*86.0 + ms_nf*53.27),(bs_f*86.0 + ms_f*53.27)),model$BUGSoutput$n.sims,2) # effects = baby
(times 86.0 DALYs each) survivors + mother survivors (times 53.27 DALYs each) in No FI, baby (times 86.0 DALYs
each) survivors + mother survivors (times 53.27 DALYs each) in FI
cf <- matrix(c(rep(0,model$BUGSoutput$n.sims),rep(5592212,model$BUGSoutput$n.sims)),model$BUGSoutput$n.sims,2) #
costs = cost of do nothing, cost of community intervention
treatsf <- c("do nothing","Facility Intervention")
fi <- bcea(e=ef,c=cf,ref=2,interventions=treatsf,Kmax=2500)
layout(matrix(1:4,2,2)) #so the four health economic analysis plots from the following commands are on one page
ceplane.plot(fi,comparison=1,wtp=779.8) # Cost-effectiveness plane
eib.plot(fi) # Expected Incremental Benefit
ceac.plot(fi) # Cost-effectiveness Acceptability Curve (prob cost-effective by WTP)
evi.plot(fi) # Expected value of information by willingness to pay

```



```

quartz.save("Figure2.pdf", type="pdf")
summary(fi,wtp=780) # Increments of 5 are only allowed in this command therefore 779.8 is not allowed
# Community Intervention (CI) vs Facility Intervention (FI)
e2 <- c2 <- matrix(NA,model$BUGSoutput$n.sims,2)
e2 <- matrix(c((bs_c*86.0 + ms_c*53.27),(bs_f*86.0 + ms_f*53.27)),model$BUGSoutput$n.sims,2) # effects = baby
(times 86.0 DALYs each) survivors + mother survivors (times 53.27 DALYs each) in CI, baby (times 86.0 DALYs each)
survivors + mother survivors (times 53.27 DALYs each) in FI
c2 <- matrix(c(rep(5348791,model$BUGSoutput$n.sims),rep(5592212,model$BUGSoutput$n.sims)),model$BUGSoutput$n.sims,
2) # costs = cost of community intervention, cost of facility intervention
treatscvf <- c("Community Intervention","Facility Intervention")
cvf <- bcea(e=e2,c=c2,ref=1,interventions=treatscvf,Kmax=2500)
layout(matrix(1:4,2,2)) #so the four health economic analysis plots from the following commands are on one page
ceplane.plot(cvf,comparison=1,wtp=779.8) # Cost-effectiveness plane
eib.plot(cvf) # Expected Incremental Benefit
ceac.plot(cvf) # Cost-effectiveness Acceptability Curve (prob cost-effective by WTP)
evi.plot(cvf) # Expected value of information by willingness to pay
quartz.save("plots_CvF.pdf", type="pdf")
summary(cvf,wtp=780) # Increments of 5 are only allowed in this command therefore 779.8 is not allowed

# combined Facility Intervention & Community Intervention (FICI) model
mkfc <- read.csv('MKFC.csv') # data exported from Stata using 'MK_WinBUGS_final_cluster_FC.do' and also stored in
'MKclusterFC.txt'

data2 <- list(mu=mu,P=P,hf=mkfc$hf,fici=mkfc$fici,urb=mkfc$urb,llw=mkfc$llw,sal=mkfc$sal,mvital=mkfc
$mvital,bdeath=mkfc$bdeath,obs=mkfc$obs)

filein2 <- "/Users/timothycolbourn/Documents/Papers/MaiKhanda CEA paper/JAGS revised/MKmodelFC_final.txt"
#filepath to text file containing model

inits2 <- function(){
list(alpha0_b=rnorm(1), alpha0_m=rnorm(1), strata2=rnorm(2,0,1), strata3=rnorm(2,0,1), urban=rnorm(2,0,1),
fc=rnorm(2,0,1))
} # initial values of paramters

params2 <-
c("bs_con","bs_fc","ms_con","ms_fc","orb_fc","orm_fc","lb_fc","db_fc","lm_fc","dm_fc","d_fc","lb_fc_","db_fc_","lm
_fc_","dm_fc_","d_fc_") #paramters to monitor

model2 <- jags(data2,inits2,params2,model.file=filein2,n.chains=2,n.iter=510000,n.burnin=10000,n.thin=10,DIC=TRUE)

print(model2,digits=3,intervals=c(0.025,0.500,0.975)) #produces customised output

m2 <- as.mcmc(model2) #converts the object model to an MCMC structure, for plotting and diagnostics
raftery.diag(m2) #Raftery-Lewis diagnostic estimates of (for each chain[[1]],[[2]]..) number of iterations
required to estimate 2.5th quantile (=q) of posterior dist, with accuracy (tolerance) r of +/-0.005 and
probability s of 0.95 (defaults)
summary(m2) #MCMC error and other indications of convergence. MCMC error is how close posterior SD is to Naive
SE. Time-series SE adjusts Naive SE for autocorrelation; lower values = better convergence

#Health Economic analysis, including EVPPI (p.145-151) - via BCEA package loaded earlier

# combined Facility Intervention & Community Intervention (FICI) vs Do Nothing
s_c <- bs_c*86.0 + ms_c*53.27 # need these estimates from the above model (for the CI v FI v FICI comparison
below) before we attach the estimates from the second model (see line 121 below). Note the number of baby and
mother survivors are each already multiplied by their respective DALYs averted here
s_f <- bs_f*86.0 + ms_f*53.27

attach.jags(model2)
efc <- cfc <- matrix(NA,model2$BUGSoutput$n.sims,2)
efc <- matrix(c((bs_con*86.0 + ms_con*53.27),(bs_fc*86.0 + ms_fc*53.27)),model2$BUGSoutput$n.sims,2) # effects =
baby (times 86.0 DALYs each) survivors + mother survivors (times 53.27 DALYs each) in control, baby (times 86.0
DALYs each) survivors + mother survivors (times 53.27 DALYs each) in FICI
cfc <- matrix(c(rep(0,model2$BUGSoutput$n.sims),rep(5470501,model2$BUGSoutput$n.sims)),model2$BUGSoutput$n.sims,2)
# costs = cost of do nothing, cost of combined facility and community intervention
treatscfc <- c("do nothing","Facility & Community Interventions")
fc <- bcea(e=efc,c=cfc,ref=2,interventions=treatscfc,Kmax=2500)
layout(matrix(1:4,2,2)) #so the four health economic analysis plots from the following commands are on one page
ceplane.plot(fc,comparison=1,wtp=779.8) # Cost-effectiveness plane
eib.plot(fc) # Expected Incremental Benefit
ceac.plot(fc) # Cost-effectiveness Acceptability Curve (prob cost-effective by WTP)
evi.plot(fc) # Expected value of information by willingness to pay
quartz.save("Figure3.pdf", type="pdf")
summary(fc,wtp=780) # Increments of 5 are only allowed in this command therefore 779.8 is not allowed
# Community Intervention (CI) vs Facility Intervention (FI) vs. combined FICI
e3 <- c3 <- matrix(NA,model$BUGSoutput$n.sims,3)
e3 <- matrix(c(s_c,s_f,(bs_fc*86.0*2 + ms_fc*53.27*2)),model$BUGSoutput$n.sims,3) # effects = baby+mother
survivors (*DALYs, see line 103) in CI, baby+mother survivors (*DALYs, see line 104) in FI, baby+mother survivors
in FICI (*DALYs) multiplied by two so the population size is the same as CI, and FI

```

```

c3 <- matrix(c(rep(5348791,model$BUGSoutput$n.sims),rep(5592212,model$BUGSoutput$n.sims),rep(5470501*2,model
$BUGSoutput$n.sims)),model$BUGSoutput$n.sims,3) # costs = cost of community intervention, cost of facility
intervention, cost of FICI multiplied by two so the population size is the same as CI, and FI
treats3 <- c("Community Intervention","Facility Intervention","FICI")
cvfvfc <- bcea(e=e3,c=c3,ref=1,interventions=treats3,Kmax=2500)
layout(matrix(1:4,2,2)) #so the four health economic analysis plots from the following commands are on one page
ceplane.plot(cvfvfc,comparison=2,wtp=779.8) # Cost-effectiveness plane
eib.plot(cvfvfc) # Expected Incremental Benefit
ceac.plot(cvfvfc) # Cost-effectiveness Acceptability Curve (prob cost-effective by WTP)
evi.plot(cvfvfc) # Expected value of information by willingness to pay
quartz.save("Figure4.pdf", type="pdf")
summary(cvfvfc,wtp=780) # Increments of 5 are only allowed in this command therefore 779.8 is not allowed
# Facility Intervention (FI) vs Community Intervention (CI) vs combined FICI
fvcvfc <- bcea(e=e3,c=c3,ref=2,interventions=treats3,Kmax=2500)
layout(matrix(1:4,2,2)) #so the four health economic analysis plots from the following commands are on one page
ceplane.plot(fvcvfc,comparison=2,wtp=779.8) # Cost-effectiveness plane
eib.plot(fvcvfc) # Expected Incremental Benefit
ceac.plot(fvcvfc) # Cost-effectiveness Acceptability Curve (prob cost-effective by WTP)
evi.plot(fvcvfc) # Expected value of information by willingness to pay
quartz.save("plots_FvCvFC.pdf", type="pdf")
summary(fvcvfc,wtp=780) # Increments of 5 are only allowed in this command therefore 779.8 is not allowed
# combined FICI vs Facility Intervention (FI) vs Community Intervention (CI)
fcvfc <- bcea(e=e3,c=c3,ref=3,interventions=treats3,Kmax=2500)
layout(matrix(1:4,2,2)) #so the four health economic analysis plots from the following commands are on one page
ceplane.plot(fcvfc,comparison=1,wtp=779.8) # Cost-effectiveness plane
eib.plot(fcvfc) # Expected Incremental Benefit
ceac.plot(fcvfc) # Cost-effectiveness Acceptability Curve (prob cost-effective by WTP)
evi.plot(fcvfc) # Expected value of information by willingness to pay
quartz.save("plots_FvCvFC.pdf", type="pdf")
summary(fcvfc,wtp=780) # Increments of 5 are only allowed in this command therefore 779.8 is not allowed

sink()

```

```

setwd("/Users/timothycolbourn/Documents/MaiKhanda CEA paper/JAGS") #set working directory for saving outputs
sink(file="MaiKhanda_future_final_log.txt",split=TRUE)
# MaiKhanda Economic Evaluation v4 in R and JAGS Tim Colbourn 2nd Dec 2014
#install.packages("R2jags")
library(R2jags)
#install.packages("BCEA")
library(BCEA)

# Community Intervention and Facility Intervention model

mk <- read.csv('MK.csv') # data exported from Stata using 'MK_WinBUGS_final_cluster.do' and also stored in
'MKcluster.txt'

h <- 100
mu <- c(0,0)
P <- (1/h^2)*diag(2)

data <- list(mu=mu,P=P,hf=mk$hf,fi=mk$fi,ci=mk$ci,urb=mk$urb,llw=mk$llw,sal=mk$sal,mvital=mk$mvital,bdeath=mk
$bdeath,obs=mk$obs)

filein <- "/Users/timothycolbourn/Documents/MaiKhanda CEA paper/JAGS/MKmodel.txt" #filepath to text file
containing model

inits <- function(){
  list(alpha0_b=rnorm(1), alpha0_m=rnorm(1), strata2=rnorm(2,0,1), strata3=rnorm(2,0,1), urban=rnorm(2,0,1),
  fac=rnorm(2,0,1), com=rnorm(2,0,1))
  } # initial values of paramters

params <-
c("bs_nc", "bs_c", "ms_nc", "ms_c", "bs_nf", "bs_f", "ms_nf", "ms_f", "orb_c", "orm_c", "orb_f", "orm_f", "lb_c", "db_c", "lm_c",
, "dm_c", "lb_f", "db_f", "lm_f", "dm_f", "d_c", "d_f") #paramters to monitor

model <- jags(data,inits,params,model.file=filein,n.chains=2,n.iter=21000,n.burnin=1000,n.thin=2,DIC=TRUE)

print(model,digits=3,intervals=c(0.025,0.500,0.975))

m <- as.mcmc(model) #converts the object model to an MCMC structure, for plotting and diagnostics
raftery.diag(m) #Raftery-Lewis diagnostic estimates of (for each chain[[1]],[[2]]..) number of iterations
required to estimate 2.5th quantile (=q) of posterior dist, with accuracy (tolerance) r of +/-0.005 and
probability s of 0.95 (defaults)
summary(m) #MCMC error and other indications of convergence. MCMC error is how close posterior SD is to Naive SE.
Time-series SE adjusts Naive SE for autocorrelation; lower values = better convergence

#Health Economic analysis, including EVPPI (p.145-151) - via BCEA package loaded earlier
attach.jags(model)
# Community Intervention (CI) vs Do Nothing
ec <- cc <- matrix(NA,model$BUGSoutput$n.sims,2)
ec <- matrix(c((bs_nc*86.0 + ms_nc*53.27),(bs_c*86.0 + ms_c*53.27)),model$BUGSoutput$n.sims,2) # effects = baby
(times 86.0 DALYs each) survivors + mother survivors (times 53.27 DALYs each) in No CI, baby (times 86.0 DALYs
each) survivors + mother survivors (times 53.27 DALYs each) in CI
cc <- matrix(c(rep(0,model$BUGSoutput$n.sims),rep(5348791,model$BUGSoutput$n.sims)),model$BUGSoutput$n.sims,2) #
costs = cost of do nothing, cost of community intervention
treatsc <- c("do nothing","Community Intervention")
ci <- bcea(e=ec,c=cc,ref=2,interventions=treatsc,Kmax=2500) # Max cost-effectivness threshold is over 3 times
GDPpc per DALY averted
layout(matrix(1:4,2,2)) #so the four health economic analysis plots from the following commands are on one page
ceplane.plot(ci,comparison=1,wtp=779.8) # Cost-effectiveness plane - Willingness to Pay (WTP) threshold is
$779.8 GDPpc (per DALY averted)
eib.plot(ci) # Expected Incremental Benefit
ceac.plot(ci) # Cost-effectiveness Acceptability Curve (prob cost-effective by WTP)
evi.plot(ci) # Expected value of information by willingness to pay
quartz.save("plots_C_.pdf", type="pdf")
summary(ci,wtp=780) # Increments of 5 are only allowed in this command therefore 779.8 is not allowed

# Facility Intervention (FI) vs Do Nothing
ef <- cf <- matrix(NA,model$BUGSoutput$n.sims,2)
ef <- matrix(c((bs_nf*86.0 + ms_nf*53.27),(bs_f*86.0 + ms_f*53.27)),model$BUGSoutput$n.sims,2) # effects = baby
(times 86.0 DALYs each) survivors + mother survivors (times 53.27 DALYs each) in No FI, baby (times 86.0 DALYs
each) survivors + mother survivors (times 53.27 DALYs each) in FI
cf <- matrix(c(rep(0,model$BUGSoutput$n.sims),rep(5592212,model$BUGSoutput$n.sims)),model$BUGSoutput$n.sims,2) #
costs = cost of do nothing, cost of community intervention
treatsf <- c("do nothing","Facility Intervention")
fi <- bcea(e=ef,c=cf,ref=2,interventions=treatsf,Kmax=2500)
layout(matrix(1:4,2,2)) #so the four health economic analysis plots from the following commands are on one page
ceplane.plot(fi,comparison=1,wtp=779.8) # Cost-effectiveness plane
eib.plot(fi) # Expected Incremental Benefit
ceac.plot(fi) # Cost-effectiveness Acceptability Curve (prob cost-effective by WTP)
evi.plot(fi) # Expected value of information by willingness to pay
quartz.save("plots_F_.pdf", type="pdf")

```

```

summary(fi,wtp=780) # Increments of 5 are only allowed in this command therefore 779.8 is not allowed
# Community Intervention (CI) vs Facility Intervention (FI)
e2 <- c2 <- matrix(NA,model$BUGSoutput$n.sims,2)
e2 <- matrix(c((bs_c*86.0 + ms_c*53.27),(bs_f*86.0 + ms_f*53.27)),model$BUGSoutput$n.sims,2) # effects = baby
(times 86.0 DALYs each) survivors + mother survivors (times 53.27 DALYs each) in CI, baby (times 86.0 DALYs each)
survivors + mother survivors (times 53.27 DALYs each) in FI
c2 <- matrix(c(rep(5348791,model$BUGSoutput$n.sims),rep(5592212,model$BUGSoutput$n.sims)),model$BUGSoutput$n.sims,
2) # costs = cost of community intervention, cost of facility intervention
treatscvf <- c("Community Intervention","Facility Intervention")
cvf <- bcea(e=e2,c=c2,ref=1,interventions=treatscvf,Kmax=2500)
layout(matrix(1:4,2,2)) #so the four health economic analysis plots from the following commands are on one page
ceplane.plot(cvf,comparison=1,wtp=779.8) # Cost-effectiveness plane
eib.plot(cvf) # Expected Incremental Benefit
ceac.plot(cvf) # Cost-effectiveness Acceptability Curve (prob cost-effective by WTP)
evi.plot(cvf) # Expected value of information by willingness to pay
quartz.save("plots_CvF_.pdf", type="pdf")
summary(cvf,wtp=780) # Increments of 5 are only allowed in this command therefore 779.8 is not allowed

# combined Facility Intervention & Community Intervention (FICI) model
mkfc <- read.csv('MKFC.csv') # data exported from Stata using 'MK_WinBUGS_final_cluster_FC.do' and also stored in
'MKclusterFC.txt'

data2 <- list(mu=mu,P=P,hf=mkfc$hf,fici=mkfc$fici,urb=mkfc$urb,llw=mkfc$llw,sal=mkfc$sal,mvital=mkfc
$mvital,bdeath=mkfc$bdeath,obs=mkfc$obs)

filein2 <- "/Users/timothycolbourn/Documents/MaiKhanda CEA paper/JAGS/MKmodelFC.txt" #filepath to text file
containing model

inits2 <- function(){
list(alpha0_b=rnorm(1), alpha0_m=rnorm(1), strata2=rnorm(2,0,1), strata3=rnorm(2,0,1), urban=rnorm(2,0,1),
fc=rnorm(2,0,1))
} # initial values of paramters

params2 <- c("bs_con","bs_fc","ms_con","ms_fc","orb_fc","orm_fc","lb_fc","db_fc","lm_fc","dm_fc","d_fc")
#paramters to monitor

model2 <- jags(data2,inits2,params2,model.file=filein2,n.chains=2,n.iter=21000,n.burnin=1000,n.thin=2,DIC=TRUE)

print(model2,digits=3,intervals=c(0.025,0.500,0.975)) #produces customised output

m2 <- as.mcmc(model2) #converts the object model to an MCMC structure, for plotting and diagnostics
raftery.diag(m2) #Raftery-Lewis diagnostic estimates of (for each chain[[1]],[[2]]..) number of iterations
required to estimate 2.5th quantile (=q) of posterior dist, with accuracy (tolerance) r of +/-0.005 and
probability s of 0.95 (defaults)
summary(m2) #MCMC error and other indications of convergence. MCMC error is how close posterior SD is to Naive
SE. Time-series SE adjusts Naive SE for autocorrelation; lower values = better convergence

#Health Economic analysis, including EVPPi (p.145-151) - via BCEA package loaded earlier

# combined Facility Intervention & Community Intervention (FICI) vs Do Nothing
s_c <- bs_c*86.0 + ms_c*53.27 # need these estimates from the above model (for the CI v FI v FICI comparison
below) before we attach the estimates from the second model (see line 121 below). Note the number of baby and
mother survivors are each already multiplied by their respective DALYs averted here
s_f <- bs_f*86.0 + ms_f*53.27

attach.jags(model2)
efc <- cfc <- matrix(NA,model2$BUGSoutput$n.sims,2)
efc <- matrix(c((bs_con*86.0 + ms_con*53.27),(bs_fc*86.0 + ms_fc*53.27)),model2$BUGSoutput$n.sims,2) # effects =
baby (times 86.0 DALYs each) survivors + mother survivors (times 53.27 DALYs each) in control, baby (times 86.0
DALYs each) survivors + mother survivors (times 53.27 DALYs each) in FICI
cfc <- matrix(c(rep(0,model2$BUGSoutput$n.sims),rep(5470501,model2$BUGSoutput$n.sims)),model$BUGSoutput$n.sims,2)
# costs = cost of do nothing, cost of combined facility and community intervention
treatscfc <- c("do nothing","Facility & Community Interventions")
fc <- bcea(e=efc,c=cfc,ref=2,interventions=treatscfc,Kmax=2500)
layout(matrix(1:4,2,2)) #so the four health economic analysis plots from the following commands are on one page
ceplane.plot(fc,comparison=1,wtp=779.8) # Cost-effectiveness plane
eib.plot(fc) # Expected Incremental Benefit
ceac.plot(fc) # Cost-effectiveness Acceptability Curve (prob cost-effective by WTP)
evi.plot(fc) # Expected value of information by willingness to pay
quartz.save("plots_FC_.pdf", type="pdf")
summary(fc,wtp=780) # Increments of 5 are only allowed in this command therefore 779.8 is not allowed
# Community Intervention (CI) vs Facility Intervention (FI) vs. combined FICI
e3 <- c3 <- matrix(NA,model$BUGSoutput$n.sims,3)
e3 <- matrix(c((s_c,s_f),(bs_fc*86.0*2 + ms_fc*53.27*2)),model$BUGSoutput$n.sims,3) # effects = baby+mother
survivors (*DALYs, see line 103) in CI, baby+mother survivors (*DALYs, see line 104) in FI, baby+mother survivors
in FICI (*DALYs) multiplied by two so the population size is the same as CI, and FI
c3 <- matrix(c(rep(5348791,model$BUGSoutput$n.sims),rep(5592212,model$BUGSoutput$n.sims),rep(5470501*2,model
$BUGSoutput$n.sims)),model$BUGSoutput$n.sims,3) # costs = cost of community intervention, cost of facility

```

```

intervention, cost of FICI multiplied by two so the population size is the same as CI, and FI
treats3 <- c("Community Intervention", "Facility Intervention", "FICI")
cvfvfc <- bcea(e=e3, c=c3, ref=1, interventions=treats3, Kmax=2500)
layout(matrix(1:4, 2, 2)) #so the four health economic analysis plots from the following commands are on one page
ceplane.plot(cvfvfc, comparison=2, wtp=779.8) # Cost-effectiveness plane
eib.plot(cvfvfc) # Expected Incremental Benefit
ceac.plot(cvfvfc) # Cost-effectiveness Acceptability Curve (prob cost-effective by WTP)
evi.plot(cvfvfc) # Expected value of information by willingness to pay
quartz.save("plots_CvFvFC_.pdf", type="pdf")
summary(cvfvfc, wtp=780) # Increments of 5 are only allowed in this command therefore 779.8 is not allowed

#Future scale-up 1
years <- 5 # Time Horizon: Years of scale-up projection
cdr <- 0.03 # Cost Discount Rate (per year)
edr <- 0.00 # Effect Discount Rate (per year)
cycle <- c(1:years)
scale <- (15013694*0.04*2)/(1200000*0.04*2) # scale-up conversion factor - see Table 2 of paper - ratio of trial
population to total population of Malawi (assume budget for intervention scales proportionally with increased
population so that the total beneficiaries per year and total cost per year (not discounted) remains the same)

s_cm <- (s_c/2.25)*scale # DALYs averted per year by the Community Intervention if scaled-up to the whole of
Malawi
fut_ec_ = NA * c(1:(length(s_cm)*length(cycle)))
fut_ec_5 <- fut_ec_4 <- fut_ec_3 <- fut_ec_2 <- fut_ec_1 <- NULL
for (a in 1:length(s_cm)) {
  for (i in 1:length(cycle)) fut_ec__[i] <- c(s_cm[a]/((1+edr)^i))
  fut_ec_5[a] <- fut_ec__[i]
}
for (a in 1:length(s_cm)) {
  for (i in 1:length(cycle)) fut_ec__[i] <- c(s_cm[a]/((1+edr)^i))
  fut_ec_4[a] <- fut_ec__[i-1]
}
for (a in 1:length(s_cm)) {
  for (i in 1:length(cycle)) fut_ec__[i] <- c(s_cm[a]/((1+edr)^i))
  fut_ec_3[a] <- fut_ec__[i-2]
}
for (a in 1:length(s_cm)) {
  for (i in 1:length(cycle)) fut_ec__[i] <- c(s_cm[a]/((1+edr)^i))
  fut_ec_2[a] <- fut_ec__[i-3]
}
for (a in 1:length(s_cm)) {
  for (i in 1:length(cycle)) fut_ec__[i] <- c(s_cm[a]/((1+edr)^i))
  fut_ec_1[a] <- fut_ec__[i-4]
}
}
fut_ec_ <- c(fut_ec_1, fut_ec_2, fut_ec_3, fut_ec_4, fut_ec_5)
fut_ec <- tapply( fut_ec_, (seq_along(fut_ec_)-1) %/% years, sum) # Future effects of CI for all years in time
horizon added together (for each of the simulations)
start_c <- 362083*scale # Start-up costs of Community Intervention for the whole of Malawi
annual_c <- (27250+2068997)*scale # Annual running costs of Community Intervention for the whole of Malawi
annual_ext_c <- 120067*scale # Annual external costs of Community Intervention for the whole of Malawi
fut_cc_ <- NULL
for (i in 1:length(cycle)) fut_cc_[i] <- c((annual_c + annual_ext_c)/((1+cdr)^i)) # Future costs of CI for each
year in time horizon
fut_cc <- start_c + tapply( fut_cc_, (seq_along(fut_cc_)-1) %/% years, sum) # Future costs of CI for all years
in time horizon added together (this is constant so needs to be replicated by the number of simulations in the
BCEA)

s_fm <- (s_f/2.25)*scale # DALYs averted per year by the Facility Intervention if scaled-up to the whole of
Malawi
fut_ef_ = NA * c(1:(length(s_fm)*length(cycle)))
fut_ef_5 <- fut_ef_4 <- fut_ef_3 <- fut_ef_2 <- fut_ef_1 <- NULL
for (a in 1:length(s_fm)) {
  for (i in 1:length(cycle)) fut_ef__[i] <- c(s_fm[a]/((1+edr)^i))
  fut_ef_5[a] <- fut_ef__[i]
}
for (a in 1:length(s_fm)) {
  for (i in 1:length(cycle)) fut_ef__[i] <- c(s_fm[a]/((1+edr)^i))
  fut_ef_4[a] <- fut_ef__[i-1]
}
for (a in 1:length(s_fm)) {
  for (i in 1:length(cycle)) fut_ef__[i] <- c(s_fm[a]/((1+edr)^i))
  fut_ef_3[a] <- fut_ef__[i-2]
}
for (a in 1:length(s_fm)) {
  for (i in 1:length(cycle)) fut_ef__[i] <- c(s_fm[a]/((1+edr)^i))
  fut_ef_2[a] <- fut_ef__[i-3]
}
for (a in 1:length(s_fm)) {

```

```

    for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
    fut_ef_1[a] <- fut_ef__[i-4]
  }
fut_ef_ <- c(fut_ef_1,fut_ef_2,fut_ef_3,fut_ef_4,fut_ef_5)
fut_ef <- tapply( fut_ef_, (seq_along(fut_ef_)-1) %/% years, sum) # Future effects of FI for all years in time
horizon added together, for each of the simulations
start_f <- 362860*scale # Start-up costs of Facility Intervention for the whole of Malawi
annual_f <- (25328+2026811)*scale # Annual running costs of Facility Intervention for the whole of Malawi
annual_ext_f <- 272017*scale # Annual external costs of Facility Intervention for the whole of Malawi
fut_cf_ <- NULL
for (i in 1:length(cycle)) fut_cf_[i] <-c((annual_f + annual_ext_f)/((1+edr)^i)) # Future costs of FI for each
year in time horizon
fut_cf <- start_f + tapply( fut_cf_, (seq_along(fut_cf_)-1) %/% years, sum) # Future costs of FI for all years
in time horizon added together, for each of the simulations

s_fc <- (bs_fc*86.0*2 + ms_fc*53.27*2)/2.25 #DALYs averted per year by FICI, doubled so the same as CI or FI
two-arm trial
s_fcm <- s_fc*scale # DALYs saved by the combined Facility and Community Intervention if scaled-up to the whole
of Malawi
fut_efc_<-NA*c(1:(length(s_fcm)*length(cycle)))
fut_efc__ <- fut_efc_5 <- fut_efc_4 <- fut_efc_3 <- fut_efc_2 <- fut_efc_1 <- NULL
for (a in 1:length(s_fcm)) {
  for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
  fut_efc_5[a] <- fut_efc__[i]
}
for (a in 1:length(s_fcm)) {
  for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
  fut_efc_4[a] <- fut_efc__[i-1]
}
for (a in 1:length(s_fcm)) {
  for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
  fut_efc_3[a] <- fut_efc__[i-2]
}
for (a in 1:length(s_fcm)) {
  for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
  fut_efc_2[a] <- fut_efc__[i-3]
}
for (a in 1:length(s_fcm)) {
  for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
  fut_efc_1[a] <- fut_efc__[i-4]
}
}
fut_efc_ <- c(fut_efc_1,fut_efc_2,fut_efc_3,fut_efc_4,fut_efc_5)
fut_efc <- tapply( fut_efc_, (seq_along(fut_efc_)-1) %/% years, sum) # Future effects of FICI for all years in
time horizon added together, for each of the simulations
start_fc <- 362472*2*scale # Start-up costs of the combined Facility and Community Intervention for the whole of
Malawi
annual_fc <- (26289+2047904)*2*scale # Annual running costs of the combined Facility and Community Intervention
for the whole of Malawi
annual_ext_fc <- 196042*2*scale # Annual external costs of FICI for the whole of Malawi
fut_cfc_ <- NULL
for (i in 1:length(cycle)) fut_cfc_[i] <-c((annual_fc + annual_ext_fc)/((1+edr)^i)) # Future costs of FI for
each year in time horizon
fut_cfc <- start_fc + tapply( fut_cfc_, (seq_along(fut_cfc_)-1) %/% years, sum) # Future costs of FICI for all
years in time horizon added together, for each of the simulations

# Future scale-up: Community Intervention (CI) vs Facility Intervention (FI) vs. combined FICI
e4 <- c4 <- matrix(NA,model$BUGSoutput$n.sims,3)
e4 <- matrix(c(fut_ec,fut_ef,fut_efc),model$BUGSoutput$n.sims,3) # effects - see lines 142, 151 and 161 above
c4 <- matrix(c(rep(fut_cc,model$BUGSoutput$n.sims),rep(fut_cf,model$BUGSoutput$n.sims),rep(fut_cfc,model
$BUGSoutput$n.sims)),model$BUGSoutput$n.sims,3) # costs - see lines 147, 156 and 166 above
future <- bcea(e=e4,c=c4,ref=1,interventions=treats3,Kmax=2500)
layout(matrix(1:4,2,2)) #so the four health economic analysis plots from the following commands are on one page
ceplane.plot(future,comparison=2,wtp=779.8) # Cost-effectiveness plane
eib.plot(future) # Expected Incremental Benefit
ceac.plot(future) # Cost-effectiveness Acceptability Curve (prob cost-effective by WTP)
evi.plot(future) # Expected value of information by willingness to pay
quartz.save("plots_future_5_003_000.pdf", type="pdf")
print("time horizon in years")
years
print("cost discount rate")
cdr
print("effects discount rate")
edr
summary(future,wtp=780)

# ... Future scale-up scenarios 2-9 omitted to save space, the code is identical to that for Future scale-up
scenario 1 above, except for changing the years cdr and edr inputs in the first 3 lines and the
quartz.save("plots_future_X_XXX_XXX.pdf" command eight lines from the end ...

```

```

#Future scale-up 10
years <- 10 # Time Horizon: Years of scale-up projection
cdr <- 0.03 # Cost Discount Rate (per year)
edr <- 0.00 # Effect Discount Rate (per year)
cycle <- c(1:years)
scale <- (15013694*0.04*2)/(1200000*0.04*2) # scale-up conversion factor - see Table 2 of paper - ratio of trial
population to total population of Malawi (assume budget for intervention scales proportionally with increased
population so that the total beneficiaries per year and total cost per year (not discounted) remains the same)

s_cm <- (s_c/2.25)*scale # DALYs averted per year by the Community Intervention if scaled-up to the whole of
Malawi
fut_ec_ = NA*c(1:(length(s_cm)*length(cycle)))
fut_ec__ <- fut_ec_10 <- fut_ec_9 <- fut_ec_8 <- fut_ec_7 <- fut_ec_6 <- fut_ec_5 <- fut_ec_4 <- fut_ec_3 <-
fut_ec_2 <- fut_ec_1 <- NULL
for (a in 1:length(s_cm)) {
  for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
  fut_ec_10[a] <- fut_ec__[i]
}
for (a in 1:length(s_cm)) {
  for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
  fut_ec_9[a] <- fut_ec__[i-1]
}
for (a in 1:length(s_cm)) {
  for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
  fut_ec_8[a] <- fut_ec__[i-2]
}
for (a in 1:length(s_cm)) {
  for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
  fut_ec_7[a] <- fut_ec__[i-3]
}
for (a in 1:length(s_cm)) {
  for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
  fut_ec_6[a] <- fut_ec__[i-4]
}
for (a in 1:length(s_cm)) {
  for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
  fut_ec_5[a] <- fut_ec__[i-5]
}
for (a in 1:length(s_cm)) {
  for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
  fut_ec_4[a] <- fut_ec__[i-6]
}
for (a in 1:length(s_cm)) {
  for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
  fut_ec_3[a] <- fut_ec__[i-7]
}
for (a in 1:length(s_cm)) {
  for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
  fut_ec_2[a] <- fut_ec__[i-8]
}
for (a in 1:length(s_cm)) {
  for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
  fut_ec_1[a] <- fut_ec__[i-9]
}
fut_ec_ <- c(fut_ec_1,fut_ec_2,fut_ec_3,fut_ec_4,fut_ec_5,fut_ec_6,fut_ec_7,fut_ec_8,fut_ec_9,fut_ec_10)
fut_ec <- tapply( fut_ec_, (seq_along(fut_ec_)-1) %/% years, sum) # Future effects of CI for all years in time
horizon added together (for each of the simulations)
start_c <- 362083*scale # Start-up costs of Community Intervention for the whole of Malawi
annual_c <- (27250+2068997)*scale # Annual running costs of Community Intervention for the whole of Malawi
annualext_c <- 120067*scale # Annual external costs of Community Intervention for the whole of Malawi
fut_cc_ <- NULL
for (i in 1:length(cycle)) fut_cc_[i] <-c((annual_c + annualext_c)/((1+cdr)^i)) # Future costs of CI for each
year in time horizon
fut_cc <- start_c + tapply( fut_cc_, (seq_along(fut_cc_)-1) %/% years, sum) # Future costs of CI for all years
in time horizon added together (this is constant so needs to be replicated by the number of simulations in the
BCEA)

s_fm <- (s_f/2.25)*scale # DALYs averted per year by the Facility Intervention if scaled-up to the whole of
Malawi
fut_ef_ = NA*c(1:(length(s_fm)*length(cycle)))
fut_ef__ <- fut_ef_10 <- fut_ef_9 <- fut_ef_8 <- fut_ef_7 <- fut_ef_6 <- fut_ef_5 <- fut_ef_4 <- fut_ef_3 <-
fut_ef_2 <- fut_ef_1 <- NULL
for (a in 1:length(s_fm)) {
  for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
  fut_ef_10[a] <- fut_ef__[i]
}
for (a in 1:length(s_fm)) {

```

```

    for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
    fut_ef_9[a] <- fut_ef__[i-1]
  }
  for (a in 1:length(s_fm)) {
    for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
    fut_ef_8[a] <- fut_ef__[i-2]
  }
  for (a in 1:length(s_fm)) {
    for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
    fut_ef_7[a] <- fut_ef__[i-3]
  }
  for (a in 1:length(s_fm)) {
    for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
    fut_ef_6[a] <- fut_ef__[i-4]
  }
  for (a in 1:length(s_fm)) {
    for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
    fut_ef_5[a] <- fut_ef__[i-5]
  }
  for (a in 1:length(s_fm)) {
    for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
    fut_ef_4[a] <- fut_ef__[i-6]
  }
  for (a in 1:length(s_fm)) {
    for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
    fut_ef_3[a] <- fut_ef__[i-7]
  }
  for (a in 1:length(s_fm)) {
    for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
    fut_ef_2[a] <- fut_ef__[i-8]
  }
  for (a in 1:length(s_fm)) {
    for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
    fut_ef_1[a] <- fut_ef__[i-9]
  }
  fut_ef_ <- c(fut_ef_1,fut_ef_2,fut_ef_3,fut_ef_4,fut_ef_5,fut_ef_6,fut_ef_7,fut_ef_8,fut_ef_9,fut_ef_10)
  fut_ef <- tapply( fut_ef_, (seq_along(fut_ef_)-1) %/% years, sum) # Future effects of CI for all years in time
  horizon added together (for each of the simulations)
  start_f <- 362860*scale # Start-up costs of Facility Intervention for the whole of Malawi
  annual_f <- (25328+2026811)*scale # Annual running costs of Facility Intervention for the whole of Malawi
  annualext_f <- 272017*scale # Annual external costs of Facility Intervention for the whole of Malawi
  fut_cf_ <- NULL
  for (i in 1:length(cycle)) fut_cf_[i] <-c((annual_f + annualext_f)/((1+edr)^i)) # Future costs of FI for each
  year in time horizon
  fut_cf <- start_f + tapply( fut_cf_, (seq_along(fut_cf_)-1) %/% years, sum) # Future costs of FI for all years
  in time horizon added together, for each of the simulations

  s_fc <- (bs_fc*86.0*2 + ms_fc*53.27*2)/2.25 #DALYs averted per year by FICI, doubled so the same as CI or FI
  two-arm trial
  s_fcm <- s_fc*scale # DALYs saved by the combined Facility and Community Intervention if scaled-up to the whole
  of Malawi
  fut_efc_<NA*c(1:(length(s_fcm)*length(cycle)))
  fut_efc__ <- fut_efc_10 <- fut_efc_9 <- fut_efc_8 <- fut_efc_7 <- fut_efc_6 <- fut_efc_5 <- fut_efc_4 <-
  fut_efc_3 <- fut_efc_2 <- fut_efc_1 <- NULL
  for (a in 1:length(s_fcm)) {
    for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
    fut_efc_10[a] <- fut_efc__[i]
  }
  for (a in 1:length(s_fcm)) {
    for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
    fut_efc_9[a] <- fut_efc__[i-1]
  }
  for (a in 1:length(s_fcm)) {
    for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
    fut_efc_8[a] <- fut_efc__[i-2]
  }
  for (a in 1:length(s_fcm)) {
    for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
    fut_efc_7[a] <- fut_efc__[i-3]
  }
  for (a in 1:length(s_fcm)) {
    for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
    fut_efc_6[a] <- fut_efc__[i-4]
  }
  for (a in 1:length(s_fcm)) {
    for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
    fut_efc_5[a] <- fut_efc__[i-5]
  }
  for (a in 1:length(s_fcm)) {

```



```

    for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
    fut_efc_4[a] <- fut_efc__[i-6]
  }
  for (a in 1:length(s_fcm)) {
    for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
    fut_efc_3[a] <- fut_efc__[i-7]
  }
  for (a in 1:length(s_fcm)) {
    for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
    fut_efc_2[a] <- fut_efc__[i-8]
  }
  for (a in 1:length(s_fcm)) {
    for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
    fut_efc_1[a] <- fut_efc__[i-9]
  }
}
fut_efc_ <-
c(fut_efc_1,fut_efc_2,fut_efc_3,fut_efc_4,fut_efc_5,fut_efc_6,fut_efc_7,fut_efc_8,fut_efc_9,fut_efc_10)
fut_efc <- tapply( fut_efc_, (seq_along(fut_efc_)-1) %/% years, sum) # Future effects of FICI for all years in
time horizon added together, for each of the simulations
start_fc <- 362472*2*scale # Start-up costs of the combined Facility and Community Intervention for the whole of
Malawi
annual_fc <- (26289+2047904)*2*scale # Annual running costs of the combined Facility and Community Intervention
for the whole of Malawi
annual_ext_fc <- 196042*2*scale # Annual external costs of FICI for the whole of Malawi
fut_cfc_ <- NULL
for (i in 1:length(cycle)) fut_cfc_[i] <-c((annual_fc + annual_ext_fc)/((1+edr)^i)) # Future costs of FI for
each year in time horizon
fut_cfc <- start_fc + tapply( fut_cfc_, (seq_along(fut_cfc_)-1) %/% years, sum) # Future costs of FICI for all
years in time horizon added together, for each of the simulations

# Future scale-up: Community Intervention (CI) vs Facility Intervention (FI) vs. combined FICI
e4 <- c4 <- matrix(NA,model$BUGSoutput$n.sims,3)
e4 <- matrix(c(fut_ec,fut_ef,fut_efc),model$BUGSoutput$n.sims,3) # effects - see lines 142, 151 and 161 above
c4 <- matrix(c(rep(fut_cc,model$BUGSoutput$n.sims),rep(fut_cf,model$BUGSoutput$n.sims),rep(fut_cfc,model
$BUGSoutput$n.sims)),model$BUGSoutput$n.sims,3) # costs - see lines 147, 156 and 166 above
future <- bcea(e=e4,c=c4,ref=1,interventions=treats3,Kmax=2500)
layout(matrix(1:4,2,2)) #so the four health economic analysis plots from the following commands are on one page
ceplane.plot(future,comparison=2,wt=779.8) # Cost-effectiveness plane
eib.plot(future) # Expected Incremental Benefit
ceac.plot(future) # Cost-effectiveness Acceptability Curve (prob cost-effective by WTP)
evi.plot(future) # Expected value of information by willingness to pay
quartz.save("plots_future_10_003_000.pdf", type="pdf")
print("time horizon in years")
years
print("cost discount rate")
cdr
print("effects discount rate")
edr
summary(future,wt=780)

# ... Future scale-up scenarios 11-18 omitted to save space, the code is identical to that for Future scale-up
scenario 10 above, except for changing the years cdr and edr inputs in the first 3 lines and the
quartz.save("plots_future_X_XXX_XXX.pdf" command eight lines from the end ...

#Future scale-up 19
years <- 20 # Time Horizon: Years of scale-up projection
cdr <- 0.03 # Cost Discount Rate (per year)
edr <- 0.00 # Effect Discount Rate (per year)
cycle <- c(1:years)
scale <- (15013694*0.04^2)/(1200000*0.04^2) # scale-up conversion factor - see Table 2 of paper - ratio of trial
population to total population of Malawi (assume budget for intervention scales proportionally with increased
population so that the total beneficiaries per year and total cost per year (not discounted) remains the same)

s_cm <- (s_c/2.25)*scale # DALYs averted per year by the Community Intervention if scaled-up to the whole of
Malawi
fut_ec_ =NA*c(1:(length(s_cm)*length(cycle)))
fut_ec__ <- fut_ec_20 <- fut_ec_19 <- fut_ec_18 <- fut_ec_17 <- fut_ec_16 <- fut_ec_15 <- fut_ec_14 <- fut_ec_13
<- fut_ec_12 <- fut_ec_11 <- fut_ec_10 <- fut_ec_9 <- fut_ec_8 <- fut_ec_7 <- fut_ec_6 <- fut_ec_5 <- fut_ec_4 <-
fut_ec_3 <- fut_ec_2 <- fut_ec_1 <- NULL
for (a in 1:length(s_cm)) {
  for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
  fut_ec_20[a] <- fut_ec__[i]
}
for (a in 1:length(s_cm)) {
  for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
  fut_ec_19[a] <- fut_ec__[i-1]
}
for (a in 1:length(s_cm)) {

```

```

    for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
    fut_ec_18[a] <- fut_ec__[i-2]
  }
  for (a in 1:length(s_cm)) {
    for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
    fut_ec_17[a] <- fut_ec__[i-3]
  }
  for (a in 1:length(s_cm)) {
    for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
    fut_ec_16[a] <- fut_ec__[i-4]
  }
  for (a in 1:length(s_cm)) {
    for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
    fut_ec_15[a] <- fut_ec__[i-5]
  }
  for (a in 1:length(s_cm)) {
    for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
    fut_ec_14[a] <- fut_ec__[i-6]
  }
  for (a in 1:length(s_cm)) {
    for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
    fut_ec_13[a] <- fut_ec__[i-7]
  }
  for (a in 1:length(s_cm)) {
    for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
    fut_ec_12[a] <- fut_ec__[i-8]
  }
  for (a in 1:length(s_cm)) {
    for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
    fut_ec_11[a] <- fut_ec__[i-9]
  }
  for (a in 1:length(s_cm)) {
    for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
    fut_ec_10[a] <- fut_ec__[i-10]
  }
  for (a in 1:length(s_cm)) {
    for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
    fut_ec_9[a] <- fut_ec__[i-11]
  }
  for (a in 1:length(s_cm)) {
    for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
    fut_ec_8[a] <- fut_ec__[i-12]
  }
  for (a in 1:length(s_cm)) {
    for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
    fut_ec_7[a] <- fut_ec__[i-13]
  }
  for (a in 1:length(s_cm)) {
    for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
    fut_ec_6[a] <- fut_ec__[i-14]
  }
  for (a in 1:length(s_cm)) {
    for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
    fut_ec_5[a] <- fut_ec__[i-15]
  }
  for (a in 1:length(s_cm)) {
    for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
    fut_ec_4[a] <- fut_ec__[i-16]
  }
  for (a in 1:length(s_cm)) {
    for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
    fut_ec_3[a] <- fut_ec__[i-17]
  }
  for (a in 1:length(s_cm)) {
    for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
    fut_ec_2[a] <- fut_ec__[i-18]
  }
  for (a in 1:length(s_cm)) {
    for (i in 1:length(cycle)) fut_ec__[i] <-c(s_cm[a]/((1+edr)^i))
    fut_ec_1[a] <- fut_ec__[i-19]
  }
}
fut_ec_ <-
c(fut_ec_1,fut_ec_2,fut_ec_3,fut_ec_4,fut_ec_5,fut_ec_6,fut_ec_7,fut_ec_8,fut_ec_9,fut_ec_10,fut_ec_11,fut_ec_12,f
ut_ec_13,fut_ec_14,fut_ec_15,fut_ec_16,fut_ec_17,fut_ec_18,fut_ec_19,fut_ec_20)
fut_ec <- tapply( fut_ec_, (seq_along(fut_ec_)-1) %/% years, sum) # Future effects of CI for all years in time
horizon added together (for each of the simulations)
start_c <- 362083*scale # Start-up costs of Community Intervention for the whole of Malawi
annual_c <- (27250+2068997)*scale # Annual running costs of Community Intervention for the whole of Malawi
annualext_c <- 120067*scale # Annual external costs of Community Intervention for the whole of Malawi

```

```

fut_cc_ <- NULL
for (i in 1:length(cycle)) fut_cc_[i] <-c((annual_c + annualect_c)/((1+cdr)^i)) # Future costs of CI for each
year in time horizon
fut_cc <- start_c + tapply( fut_cc_, (seq_along(fut_cc_)-1) %/% years, sum) # Future costs of CI for all years
in time horizon added together (this is constant so needs to be replicated by the number of simulations in the
BCEA)

s_fm <- (s_f/2.25)*scale # DALYs averted per year by the Facility Intervention if scaled-up to the whole of
Malawi
fut_ef_ = NA*c(1:(length(s_fm)*length(cycle)))
fut_ef__ <- fut_ef_20 <- fut_ef_19 <- fut_ef_18 <- fut_ef_17 <- fut_ef_16 <- fut_ef_15 <- fut_ef_14 <- fut_ef_13
<- fut_ef_12 <- fut_ef_11 <- fut_ef_10 <- fut_ef_9 <- fut_ef_8 <- fut_ef_7 <- fut_ef_6 <- fut_ef_5 <- fut_ef_4 <-
fut_ef_3 <- fut_ef_2 <- fut_ef_1 <- NULL
for (a in 1:length(s_fm)) {
  for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
  fut_ef_20[a] <- fut_ef__[i]
}
for (a in 1:length(s_fm)) {
  for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
  fut_ef_19[a] <- fut_ef__[i-1]
}
for (a in 1:length(s_fm)) {
  for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
  fut_ef_18[a] <- fut_ef__[i-2]
}
for (a in 1:length(s_fm)) {
  for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
  fut_ef_17[a] <- fut_ef__[i-3]
}
for (a in 1:length(s_fm)) {
  for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
  fut_ef_16[a] <- fut_ef__[i-4]
}
for (a in 1:length(s_fm)) {
  for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
  fut_ef_15[a] <- fut_ef__[i-5]
}
for (a in 1:length(s_fm)) {
  for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
  fut_ef_14[a] <- fut_ef__[i-6]
}
for (a in 1:length(s_fm)) {
  for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
  fut_ef_13[a] <- fut_ef__[i-7]
}
for (a in 1:length(s_fm)) {
  for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
  fut_ef_12[a] <- fut_ef__[i-8]
}
for (a in 1:length(s_fm)) {
  for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
  fut_ef_11[a] <- fut_ef__[i-9]
}
for (a in 1:length(s_fm)) {
  for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
  fut_ef_10[a] <- fut_ef__[i-10]
}
for (a in 1:length(s_fm)) {
  for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
  fut_ef_9[a] <- fut_ef__[i-11]
}
for (a in 1:length(s_fm)) {
  for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
  fut_ef_8[a] <- fut_ef__[i-12]
}
for (a in 1:length(s_fm)) {
  for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
  fut_ef_7[a] <- fut_ef__[i-13]
}
for (a in 1:length(s_fm)) {
  for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
  fut_ef_6[a] <- fut_ef__[i-14]
}
for (a in 1:length(s_fm)) {
  for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
  fut_ef_5[a] <- fut_ef__[i-15]
}
for (a in 1:length(s_fm)) {
  for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
}

```

```

    fut_ef_4[a] <- fut_ef__[i-16]
  }
  for (a in 1:length(s_fm)) {
    for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
    fut_ef_3[a] <- fut_ef__[i-17]
  }
  for (a in 1:length(s_fm)) {
    for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
    fut_ef_2[a] <- fut_ef__[i-18]
  }
  for (a in 1:length(s_fm)) {
    for (i in 1:length(cycle)) fut_ef__[i] <-c(s_fm[a]/((1+edr)^i))
    fut_ef_1[a] <- fut_ef__[i-19]
  }
  fut_ef_ <-
  c(fut_ef_1,fut_ef_2,fut_ef_3,fut_ef_4,fut_ef_5,fut_ef_6,fut_ef_7,fut_ef_8,fut_ef_9,fut_ef_10,fut_ef_11,fut_ef_12,fut_ef_13,fut_ef_14,fut_ef_15,fut_ef_16,fut_ef_17,fut_ef_18,fut_ef_19,fut_ef_20)
  fut_ef <- tapply( fut_ef_, (seq_along(fut_ef_)-1) %/% years, sum) # Future effects of CI for all years in time horizon added together (for each of the simulations)
  start_f <- 362860*scale # Start-up costs of Facility Intervention for the whole of Malawi
  annual_f <- (25328+2026811)*scale # Annual running costs of Facility Intervention for the whole of Malawi
  annualext_f <- 272017*scale # Annual external costs of Facility Intervention for the whole of Malawi
  fut_cf_ <- NULL
  for (i in 1:length(cycle)) fut_cf_[i] <-c((annual_f + annualext_f)/((1+edr)^i)) # Future costs of FI for each year in time horizon
  fut_cf <- start_f + tapply( fut_cf_, (seq_along(fut_cf_)-1) %/% years, sum) # Future costs of FI for all years in time horizon added together, for each of the simulations

  s_fc <- (bs_fc*86.0*2 + ms_fc*53.27*2)/2.25 #DALYs averted per year by FICI, doubled so the same as CI or FI two-arm trial
  s_fcm <- s_fc*scale # DALYs saved by the combined Facility and Community Intervention if scaled-up to the whole of Malawi
  fut_efc_<-NA*c(1:(length(s_fcm)*length(cycle)))
  fut_efc__ <- fut_efc_20 <- fut_efc_19 <- fut_efc_18 <- fut_efc_17 <- fut_efc_16 <- fut_efc_15 <- fut_efc_14 <- fut_efc_13 <- fut_efc_12 <- fut_efc_11 <- fut_efc_10 <- fut_efc_9 <- fut_efc_8 <- fut_efc_7 <- fut_efc_6 <- fut_efc_5 <- fut_efc_4 <- fut_efc_3 <- fut_efc_2 <- fut_efc_1 <- NULL
  for (a in 1:length(s_fcm)) {
    for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
    fut_efc_20[a] <- fut_efc__[i]
  }
  for (a in 1:length(s_fcm)) {
    for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
    fut_efc_19[a] <- fut_efc__[i-1]
  }
  for (a in 1:length(s_fcm)) {
    for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
    fut_efc_18[a] <- fut_efc__[i-2]
  }
  for (a in 1:length(s_fcm)) {
    for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
    fut_efc_17[a] <- fut_efc__[i-3]
  }
  for (a in 1:length(s_fcm)) {
    for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
    fut_efc_16[a] <- fut_efc__[i-4]
  }
  for (a in 1:length(s_fcm)) {
    for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
    fut_efc_15[a] <- fut_efc__[i-5]
  }
  for (a in 1:length(s_fcm)) {
    for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
    fut_efc_14[a] <- fut_efc__[i-6]
  }
  for (a in 1:length(s_fcm)) {
    for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
    fut_efc_13[a] <- fut_efc__[i-7]
  }
  for (a in 1:length(s_fcm)) {
    for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
    fut_efc_12[a] <- fut_efc__[i-8]
  }
  for (a in 1:length(s_fcm)) {
    for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
    fut_efc_11[a] <- fut_efc__[i-9]
  }
  for (a in 1:length(s_fcm)) {
    for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
    fut_efc_10[a] <- fut_efc__[i-10]
  }

```

```

}
for (a in 1:length(s_fcm)) {
  for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
  fut_efc_9[a] <- fut_efc__[i-11]
}
for (a in 1:length(s_fcm)) {
  for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
  fut_efc_8[a] <- fut_efc__[i-12]
}
for (a in 1:length(s_fcm)) {
  for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
  fut_efc_7[a] <- fut_efc__[i-13]
}
for (a in 1:length(s_fcm)) {
  for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
  fut_efc_6[a] <- fut_efc__[i-14]
}
for (a in 1:length(s_fcm)) {
  for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
  fut_efc_5[a] <- fut_efc__[i-15]
}
for (a in 1:length(s_fcm)) {
  for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
  fut_efc_4[a] <- fut_efc__[i-16]
}
for (a in 1:length(s_fcm)) {
  for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
  fut_efc_3[a] <- fut_efc__[i-17]
}
for (a in 1:length(s_fcm)) {
  for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
  fut_efc_2[a] <- fut_efc__[i-18]
}
for (a in 1:length(s_fcm)) {
  for (i in 1:length(cycle)) fut_efc__[i] <-c(s_fcm[a]/((1+edr)^i))
  fut_efc_1[a] <- fut_efc__[i-19]
}
fut_efc_ <-
c(fut_efc_1,fut_efc_2,fut_efc_3,fut_efc_4,fut_efc_5,fut_efc_6,fut_efc_7,fut_efc_8,fut_efc_9,fut_efc_10,fut_efc_11,
fut_efc_12,fut_efc_13,fut_efc_14,fut_efc_15,fut_efc_16,fut_efc_17,fut_efc_18,fut_efc_19,fut_efc_20)
fut_efc <- tapply( fut_efc_, (seq_along(fut_efc_)-1) %/% years, sum) # Future effects of FICI for all years in
time horizon added together, for each of the simulations
start_fc <- 362472*2*scale # Start-up costs of the combined Facility and Community Intervention for the whole of
Malawi
annual_fc <- (26289+2047904)*2*scale # Annual running costs of the combined Facility and Community Intervention
for the whole of Malawi
annualext_fc <- 196042*2*scale # Annual external costs of FICI for the whole of Malawi
fut_cfc_ <- NULL
for (i in 1:length(cycle)) fut_cfc_[i] <-c((annual_fc + annualext_fc)/((1+edr)^i)) # Future costs of FI for
each year in time horizon
fut_cfc <- start_fc + tapply( fut_cfc_, (seq_along(fut_cfc_)-1) %/% years, sum) # Future costs of FICI for all
years in time horizon added together, for each of the simulations

# Future scale-up: Community Intervention (CI) vs Facility Intervention (FI) vs. combined FICI
e4 <- c4 <- matrix(NA,model$BUGSoutput$n.sims,3)
e4 <- matrix(c(fut_ec,fut_ef,fut_efc),model$BUGSoutput$n.sims,3) # effects - see lines 142, 151 and 161 above
c4 <- matrix(c(rep(fut_cc,model$BUGSoutput$n.sims),rep(fut_cf,model$BUGSoutput$n.sims),rep(fut_cfc,model
$BUGSoutput$n.sims)),model$BUGSoutput$n.sims,3) # costs - see lines 147, 156 and 166 above
future <- bcea(e=e4,c=c4,ref=1,interventions=treats3,Kmax=2500)
layout(matrix(1:4,2,2)) #so the four health economic analysis plots from the following commands are on one page
ceplane.plot(future,comparison=2,wdp=779.8) # Cost-effectiveness plane
eib.plot(future) # Expected Incremental Benefit
ceac.plot(future) # Cost-effectiveness Acceptability Curve (prob cost-effective by WTP)
evi.plot(future) # Expected value of information by willingness to pay
quartz.save("plots_future_20_003_000.pdf", type="pdf")
print("time horizon in years")
years
print("cost discount rate")
cdr
print("effects discount rate")
edr
summary(future,wdp=780)

# ... Future scale-up scenarios 20-27 omitted to save space, the code is identical to that for Future scale-up
scenario 19 above, except for changing the years cdr and edr inputs in the first 3 lines and the
quartz.save("plots_future_X_XXX_XXX.pdf" command eight lines from the end

sink()

```

MaiKhanda_final_log_.txt output from MaiKhanda_final_submit.R 3rd Dec 10:12

The downloaded binary packages are in
/var/folders/ps/63xh8zd90wbdcl2v8l10_npc0000gq/T/RtmpavJqyt/downloaded_packages

The downloaded binary packages are in
/var/folders/ps/63xh8zd90wbdcl2v8l10_npc0000gq/T/RtmpavJqyt/downloaded_packages

Compiling model graph
Resolving undeclared variables
Allocating nodes
Graph Size: 702

Initializing model

Inference for Bugs model at "/Users/timothycolbourn/Documents/Papers/MaiKhanda CEA paper/JAGS revised/MKmodel_final.txt", fit using jags,
2 chains, each with 510000 iterations (first 10000 discarded), n.thin = 10
n.sims = 100000 iterations saved

	mu.vect	sd.vect	2.5%	50%	97.5%	Rhat	n.eff
bs_c	102352.276	336.338	101673.738	102361.079	102978.330	1.001	100000
bs_f	101621.174	368.950	100873.777	101630.223	102316.649	1.001	33000
bs_nc	101580.075	77.652	101427.000	101580.000	101731.000	1.001	76000
bs_nf	101329.914	79.146	101175.000	101330.000	101485.000	1.001	100000
d_c	67360.641	29006.136	8807.651	68124.694	121507.556	1.001	100000
d_f	19900.924	32103.889	-44768.651	20638.361	80586.481	1.001	21000
db_c	66409.268	28324.616	9415.513	67146.479	119034.719	1.001	100000
db_f	25048.424	31049.592	-37739.441	25770.955	83762.741	1.001	34000
dm_c	951.373	6523.040	-14369.912	1835.218	11169.105	1.001	57000
dm_f	-5147.499	8559.492	-25540.628	-3935.924	7883.364	1.004	480
lb_c	772.201	329.356	109.483	780.773	1384.125	1.001	100000
lb_f	291.261	361.042	-438.831	299.662	973.985	1.001	34000
lm_c	17.859	122.452	-269.756	34.451	209.670	1.001	57000
lm_f	-96.630	160.681	-479.456	-73.886	147.989	1.004	480
ms_c	107605.037	123.931	107313.273	107622.046	107797.303	1.001	64000
ms_f	107477.274	162.687	107089.672	107500.481	107724.933	1.004	500
ms_nc	107587.178	20.335	107547.000	107587.000	107626.000	1.001	100000
ms_nf	107573.905	20.467	107533.000	107574.000	107614.000	1.001	37000
orb_c	0.873	0.054	0.774	0.872	0.982	1.001	100000
orb_f	0.954	0.057	0.846	0.952	1.070	1.001	33000
orm_c	0.957	0.298	0.494	0.916	1.657	1.001	30000
orm_f	1.228	0.379	0.652	1.175	2.135	1.004	430
deviance	509.790	274.274	500.653	507.739	519.946	1.019	6700

For each parameter, n.eff is a crude measure of effective sample size,
and Rhat is the potential scale reduction factor (at convergence, Rhat=1).

DIC info (using the rule, $pD = \text{var}(\text{deviance})/2$)
 $pD = 37613.2$ and $DIC = 38123.0$
DIC is an estimate of expected predictive error (lower deviance is better).
[[1]]

Quantile (q) = 0.025
Accuracy (r) = +/- 0.005
Probability (s) = 0.95

	Burn-in (M)	Total (N)	Lower bound (Nmin)	Dependence factor (I)
bs_c	60	90720	3746	24.2
bs_f	60	86860	3746	23.2
bs_nc	20	38160	3746	10.2
bs_nf	20	37480	3746	10.0
d_c	60	90920	3746	24.3
d_f	80	94800	3746	25.3
db_c	60	87780	3746	23.4
db_f	80	91840	3746	24.5
deviance	180	240360	3746	64.2
dm_c	910	1058070	3746	282.0
dm_f	1080	1107540	3746	296.0
lb_c	60	87780	3746	23.4
lb_f	80	91840	3746	24.5
lm_c	910	1058070	3746	282.0
lm_f	1080	1107540	3746	296.0
ms_c	840	1026480	3746	274.0
ms_f	990	972090	3746	260.0
ms_nc	20	41880	3746	11.2
ms_nf	10	37530	3746	10.0
orb_c	60	90160	3746	24.1
orb_f	80	91960	3746	24.5
orm_c	1430	1476530	3746	394.0
orm_f	900	944800	3746	252.0

[[2]]

Quantile (q) = 0.025
 Accuracy (r) = +/- 0.005
 Probability (s) = 0.95

	Burn-in (M)	Total (N)	Lower bound (Nmin)	Dependence factor (I)
bs_c	60	86700	3746	23.1
bs_f	60	89360	3746	23.9
bs_nc	10	38720	3746	10.3
bs_nf	20	38510	3746	10.3
d_c	80	98980	3746	26.4
d_f	60	91900	3746	24.5
db_c	60	90760	3746	24.2
db_f	60	96760	3746	25.8
deviance	210	289240	3746	77.2
dm_c	1120	1281840	3746	342.0
dm_f	840	943460	3746	252.0
lb_c	60	90760	3746	24.2
lb_f	60	96760	3746	25.8
lm_c	1120	1281840	3746	342.0
lm_f	840	943460	3746	252.0
ms_c	800	878400	3746	234.0
ms_f	910	987350	3746	264.0
ms_nc	20	39660	3746	10.6
ms_nf	20	38060	3746	10.2
orb_c	60	91020	3746	24.3
orb_f	90	129420	3746	34.5
orm_c	1320	1526400	3746	407.0
orm_f	1080	1112940	3746	297.0

Iterations = 10001:509991
 Thinning interval = 10
 Number of chains = 2
 Sample size per chain = 50000

1. Empirical mean and standard deviation for each variable,
 plus standard error of the mean:

	Mean	SD	Naive SE	Time-series SE
bs_c	1.024e+05	3.363e+02	1.064e+00	2.098e+00
bs_f	1.016e+05	3.690e+02	1.167e+00	2.319e+00
bs_nc	1.016e+05	7.765e+01	2.456e-01	2.456e-01
bs_nf	1.013e+05	7.915e+01	2.503e-01	2.513e-01
d_c	6.736e+04	2.901e+04	9.173e+01	2.288e+02
d_f	1.990e+04	3.210e+04	1.015e+02	3.015e+02
db_c	6.641e+04	2.832e+04	8.957e+01	1.807e+02
db_f	2.505e+04	3.105e+04	9.819e+01	1.970e+02
deviance	5.098e+02	2.743e+02	8.673e-01	1.030e+00
dm_c	9.514e+02	6.523e+03	2.063e+01	2.049e+02
dm_f	-5.147e+03	8.559e+03	2.707e+01	2.643e+02
lb_c	7.722e+02	3.294e+02	1.042e+00	2.101e+00
lb_f	2.913e+02	3.610e+02	1.142e+00	2.291e+00
lm_c	1.786e+01	1.225e+02	3.872e-01	3.847e+00
lm_f	-9.663e+01	1.607e+02	5.081e-01	4.962e+00
ms_c	1.076e+05	1.239e+02	3.919e-01	3.884e+00
ms_f	1.075e+05	1.627e+02	5.145e-01	4.995e+00
ms_nc	1.076e+05	2.033e+01	6.430e-02	6.430e-02
ms_nf	1.076e+05	2.047e+01	6.472e-02	6.472e-02
orb_c	8.733e-01	5.390e-02	1.704e-04	3.430e-04
orb_f	9.538e-01	5.740e-02	1.815e-04	3.640e-04
orm_c	9.570e-01	2.976e-01	9.410e-04	9.251e-03
orm_f	1.228e+00	3.790e-01	1.199e-03	1.176e-02

2. Quantiles for each variable:

	2.5%	25%	50%	75%	97.5%
bs_c	1.017e+05	1.021e+05	1.024e+05	1.026e+05	1.030e+05
bs_f	1.009e+05	1.014e+05	1.016e+05	1.019e+05	1.023e+05
bs_nc	1.014e+05	1.015e+05	1.016e+05	1.016e+05	1.017e+05
bs_nf	1.012e+05	1.013e+05	1.013e+05	1.014e+05	1.015e+05
d_c	8.808e+03	4.842e+04	6.812e+04	8.727e+04	1.215e+05
d_f	-4.477e+04	-1.213e+03	2.064e+04	4.175e+04	8.059e+04
db_c	9.416e+03	4.786e+04	6.715e+04	8.586e+04	1.190e+05
db_f	-3.774e+04	4.530e+03	2.577e+04	4.624e+04	8.376e+04
deviance	5.007e+02	5.048e+02	5.077e+02	5.113e+02	5.199e+02
dm_c	-1.437e+04	-2.689e+03	1.835e+03	5.568e+03	1.117e+04
dm_f	-2.554e+04	-9.837e+03	-3.936e+03	9.889e+02	7.883e+03
lb_c	1.095e+02	5.565e+02	7.808e+02	9.984e+02	1.384e+03
lb_f	-4.388e+02	5.268e+01	2.997e+02	5.376e+02	9.740e+02
lm_c	-2.698e+02	-5.049e+01	3.445e+01	1.045e+02	2.097e+02
lm_f	-4.795e+02	-1.847e+02	-7.389e+01	1.856e+01	1.480e+02
ms_c	1.073e+05	1.075e+05	1.076e+05	1.077e+05	1.078e+05

```

ms_f      1.071e+05  1.074e+05  1.075e+05  1.076e+05  1.077e+05
ms_nc     1.075e+05  1.076e+05  1.076e+05  1.076e+05  1.076e+05
ms_nf     1.075e+05  1.076e+05  1.076e+05  1.076e+05  1.076e+05
orb_c     7.740e-01  8.363e-01  8.717e-01  9.083e-01  9.819e-01
orb_f     8.459e-01  9.145e-01  9.523e-01  9.916e-01  1.070e+00
orm_c     4.943e-01  7.454e-01  9.160e-01  1.123e+00  1.657e+00
orm_f     6.517e-01  9.563e-01  1.175e+00  1.436e+00  2.135e+00

```

```

quartz
  2

```

Cost-effectiveness analysis summary

Reference intervention: Community Intervention
 Comparator intervention: do nothing

Optimal decision: choose do nothing for $k < 80$ and Community Intervention for $k \geq 80$

Analysis for willingness to pay parameter $k = 780$

```

                Expected utility
do nothing      11284303221
Community Intervention  11331495730

```

```

                EIB    CEAC    ICER
Community Intervention vs do nothing  47192509  0.97832  79.405

```

Optimal intervention (max expected utility) for $k=780$: Community Intervention

```

EVPI 210423
quartz
  2

```

Cost-effectiveness analysis summary

Reference intervention: Facility Intervention
 Comparator intervention: do nothing

Optimal decision: choose do nothing for $k < 285$ and Facility Intervention for $k \geq 285$

Analysis for willingness to pay parameter $k = 780$

```

                Expected utility
do nothing      11266970900
Facility Intervention  11276901409

```

```

                EIB    CEAC    ICER
Facility Intervention vs do nothing  9930509  0.66163  281

```

Optimal intervention (max expected utility) for $k=780$: Facility Intervention

```

EVPI 5851010
quartz
  2

```

Cost-effectiveness analysis summary

Reference intervention: Community Intervention
 Comparator intervention: Facility Intervention

Community Intervention dominates for all k in $[0 - 2500]$

Analysis for willingness to pay parameter $k = 780$

```

                Expected utility
Community Intervention  11331495730
Facility Intervention  11276901409

```

```

                EIB    CEAC    ICER
Community Intervention vs Facility Intervention  54594321  0.94599  -3.4934

```

Optimal intervention (max expected utility) for $k=780$: Community Intervention

```

EVPI 793159
Compiling model graph
  Resolving undeclared variables
  Allocating nodes
  Graph Size: 357

```

Initializing model

Inference for Bugs model at "/Users/timothycolbourn/Documents/Papers/MaiKhanda CEA paper/JAGS revised/MKmodelFC_final.txt", fit using jags,
 2 chains, each with 510000 iterations (first 10000 discarded), n.thin = 10
 n.sims = 100000 iterations saved

	mu.vect	sd.vect	2.5%	50%	97.5%	Rhat	n.eff
bs_con	50862.940	54.468	50756.000	50863.000	50969.000	1.001	46000
bs_fc	51338.360	223.893	50873.579	51346.994	51750.893	1.001	21000
d_fc	37590.203	20198.564	-4641.764	38498.852	74617.980	1.001	3100
d_fc_	75180.406	40397.128	-9283.528	76997.703	149235.960	1.001	3100
db_fc	40886.076	18860.691	1982.103	41544.463	75674.721	1.001	29000
db_fc_	81772.153	37721.381	3964.205	83088.927	151349.441	1.001	29000
dm_fc	-3295.873	7298.995	-21134.687	-1728.280	5950.436	1.003	830
dm_fc_	-6591.747	14597.989	-42269.374	-3456.560	11900.872	1.003	830
lb_fc	475.419	219.310	23.048	483.075	879.939	1.001	29000
lb_fc_	950.839	438.621	46.095	966.150	1759.877	1.001	29000
lm_fc	-61.871	137.019	-396.747	-32.444	111.703	1.003	830
lm_fc_	-123.742	274.038	-793.493	-64.888	223.407	1.003	830
ms_con	53782.118	14.783	53753.000	53782.000	53811.000	1.001	100000
ms_fc	53720.247	138.299	53383.775	53749.854	53894.724	1.003	840
orb_fc	0.841	0.073	0.707	0.838	0.992	1.001	28000
orm_fc	1.287	0.635	0.488	1.150	2.839	1.003	600
deviance	258.437	64.142	251.107	257.322	269.176	1.001	8100

For each parameter, n.eff is a crude measure of effective sample size,
 and Rhat is the potential scale reduction factor (at convergence, Rhat=1).

DIC info (using the rule, $pD = \text{var}(\text{deviance})/2$)
 $pD = 2057.1$ and $DIC = 2315.5$
 DIC is an estimate of expected predictive error (lower deviance is better).
 [[1]]

Quantile (q) = 0.025
 Accuracy (r) = +/- 0.005
 Probability (s) = 0.95

	Burn-in (M)	Total (N)	Lower bound (Nmin)	Dependence factor (I)
bs_con	20	39230	3746	10.5
bs_fc	60	89260	3746	23.8
d_fc	180	267960	3746	71.5
d_fc_	180	267960	3746	71.5
db_fc	60	89260	3746	23.8
db_fc_	60	89260	3746	23.8
deviance	200	255800	3746	68.3
dm_fc	990	1039610	3746	278.0
dm_fc_	990	1039610	3746	278.0
lb_fc	60	89260	3746	23.8
lb_fc_	60	89260	3746	23.8
lm_fc	990	1039610	3746	278.0
lm_fc_	990	1039610	3746	278.0
ms_con	10	41710	3746	11.1
ms_fc	990	968880	3746	259.0
orb_fc	60	90560	3746	24.2
orm_fc	960	1050720	3746	280.0

[[2]]

Quantile (q) = 0.025
 Accuracy (r) = +/- 0.005
 Probability (s) = 0.95

	Burn-in (M)	Total (N)	Lower bound (Nmin)	Dependence factor (I)
bs_con	20	38760	3746	10.3
bs_fc	80	97620	3746	26.1
d_fc	120	138270	3746	36.9
d_fc_	120	138270	3746	36.9
db_fc	80	97600	3746	26.1
db_fc_	80	97600	3746	26.1
deviance	180	261540	3746	69.8
dm_fc	1200	1454600	3746	388.0
dm_fc_	1200	1454600	3746	388.0
lb_fc	80	97600	3746	26.1
lb_fc_	80	97600	3746	26.1
lm_fc	1200	1454600	3746	388.0
lm_fc_	1200	1454600	3746	388.0
ms_con	10	40440	3746	10.8
ms_fc	1200	1289400	3746	344.0
orb_fc	60	91300	3746	24.4
orm_fc	900	932100	3746	249.0

Iterations = 10001:509991
 Thinning interval = 10
 Number of chains = 2
 Sample size per chain = 50000

1. Empirical mean and standard deviation for each variable,
 plus standard error of the mean:

	Mean	SD	Naive SE	Time-series SE
bs_con	50862.9402	54.46801	0.1722430	0.1744591
bs_fc	51338.3597	223.89290	0.7080115	1.4048950
d_fc	37590.2028	20198.56425	63.8734685	236.9286552
d_fc_	75180.4056	40397.12850	127.7469370	473.8573103
db_fc	40886.0763	18860.69058	59.6427405	119.9372858
db_fc_	81772.1526	37721.38116	119.2854809	239.8745716
deviance	258.4367	64.14191	0.2028345	0.2908747
dm_fc	-3295.8735	7298.99468	23.0814478	241.7167825
dm_fc_	-6591.7470	14597.98936	46.1628956	483.4335650
lb_fc	475.4195	219.31036	0.6935202	1.3946196
lb_fc_	950.8390	438.62071	1.3870405	2.7892392
lm_fc	-61.8711	137.01886	0.4332917	4.5375780
lm_fc_	-123.7422	274.03772	0.8665834	9.0751561
ms_con	53782.1179	14.78296	0.0467478	0.0465820
ms_fc	53720.2468	138.29911	0.4373402	4.5709052
orb_fc	0.8409	0.07287	0.0002304	0.0004641
orm_fc	1.2869	0.63451	0.0020065	0.0215007

2. Quantiles for each variable:

	2.5%	25%	50%	75%	97.5%
bs_con	50756.0000	50826.0000	50863.000	50900.000	50969.0000
bs_fc	50873.5789	51193.7442	51346.994	51494.207	51750.8933
d_fc	-4641.7638	24695.1064	38498.852	51591.503	74617.9801
d_fc_	-9283.5275	49390.2129	76997.703	103183.005	149235.9602
db_fc	1982.1025	28624.4984	41544.463	53944.712	75674.7206
db_fc_	3964.2050	57248.9968	83088.927	107889.423	151349.4412
deviance	251.1071	254.6414	257.322	260.585	269.1764
dm_fc	-21134.6871	-6462.9401	-1728.280	1628.723	5950.4359
dm_fc_	-42269.3742	-12925.8801	-3456.560	3257.446	11900.8718
lb_fc	23.0477	332.8430	483.075	627.264	879.9386
lb_fc_	46.0954	665.6860	966.150	1254.528	1759.8772
lm_fc	-396.7465	-121.3242	-32.444	30.575	111.7033
lm_fc_	-793.4930	-242.6484	-64.888	61.150	223.4066
ms_con	53753.0000	53772.0000	53782.000	53792.000	53811.0000
ms_fc	53383.7752	53660.0125	53749.854	53814.128	53894.7242
orb_fc	0.7073	0.7902	0.838	0.888	0.9922
orm_fc	0.4877	0.8587	1.150	1.564	2.8393

quartz
2

Cost-effectiveness analysis summary

Reference intervention: Facility & Community Interventions
 Comparator intervention: do nothing

Optimal decision: choose do nothing for k<150 and Facility & Community Interventions for k>=150

Analysis for willingness to pay parameter k = 780

	Expected utility
do nothing	5646565298
Facility & Community Interventions	5670415155

	EIB	CEAC	ICER
Facility & Community Interventions vs do nothing	23849857	0.92922	145.53

Optimal intervention (max expected utility) for k=780: Facility & Community Interventions

EVPI 598177
quartz
2

Cost-effectiveness analysis summary

Reference intervention: Community Intervention
 Comparator intervention(s): Facility Intervention
 : FICI

Optimal decision: choose Community Intervention for k<295 and FICI for k>=295

Analysis for willingness to pay parameter k = 780

	Expected utility
Community Intervention	11331495730
Facility Intervention	11276901409
FICI	11340830310

	EIB	CEAC	ICER
Community Intervention vs Facility Intervention	54594321	0.94599	-3.4934
Community Intervention vs FICI	-9334580	0.39826	292.2212

Optimal intervention (max expected utility) for k=780: FICI

EVPI 11708726
quartz
2

Cost-effectiveness analysis summary

Reference intervention: Facility Intervention
Comparator intervention(s): Community Intervention
: FICI

Optimal decision: choose Community Intervention for k<295 and FICI for k>=295

Analysis for willingness to pay parameter k = 780

	Expected utility
Community Intervention	11331495730
Facility Intervention	11276901409
FICI	11340830310

	EIB	CEAC	ICER
Facility Intervention vs Community Intervention	-54594321	0.05401	-3.4934
Facility Intervention vs FICI	-63928901	0.06219	60.2222

Optimal intervention (max expected utility) for k=780: FICI

EVPI 11708726
quartz
2

Cost-effectiveness analysis summary

Reference intervention: FICI
Comparator intervention(s): Community Intervention
: Facility Intervention

Optimal decision: choose Community Intervention for k<295 and FICI for k>=295

Analysis for willingness to pay parameter k = 780

	Expected utility
Community Intervention	11331495730
Facility Intervention	11276901409
FICI	11340830310

	EIB	CEAC	ICER
FICI vs Community Intervention	9334580	0.60174	292.221
FICI vs Facility Intervention	63928901	0.93781	60.222

Optimal intervention (max expected utility) for k=780: FICI

EVPI 11708726

MaiKhanda_LLE_log_.txt output from MaiKhanda_final local life expectancy.R 3rd Dec 2014 12:22

The downloaded binary packages are in
/var/folders/ps/63xh8zd90wbdc12v8l10_npc0000gq/T/RtmpBaqeiP/downloaded_packages

The downloaded binary packages are in
/var/folders/ps/63xh8zd90wbdc12v8l10_npc0000gq/T/RtmpBaqeiP/downloaded_packages

Compiling model graph
Resolving undeclared variables
Allocating nodes
Graph Size: 702

Initializing model

Inference for Bugs model at "/Users/timothycolbourn/Documents/Papers/MaiKhanda CEA paper/JAGS revised/MKmodel_final_LLE.txt", fit using jags,
2 chains, each with 510000 iterations (first 10000 discarded), n.thin = 10
n.sims = 100000 iterations saved

	mu.vect	sd.vect	2.5%	50%	97.5%	Rhat	n.eff
bs_c	102352.341	331.911	101674.022	102362.440	102972.802	1.001	25000
bs_f	101621.975	366.363	100878.279	101631.773	102306.900	1.001	88000
bs_nc	101580.265	77.680	101427.000	101580.000	101733.000	1.001	49000
bs_nf	101330.257	79.035	101175.000	101331.000	101485.000	1.001	100000
d_c	35285.654	15001.749	4752.910	35746.440	63440.101	1.001	4700
d_f	10212.566	16795.611	-24133.017	10662.580	41678.647	1.001	48000
db_c	34743.418	14632.151	4874.324	35195.247	62263.601	1.001	18000
db_f	13127.297	16129.191	-19746.894	13522.160	43374.305	1.001	89000
dm_c	542.237	3363.687	-7339.775	1007.084	5761.899	1.003	940
dm_f	-2914.732	4580.281	-13964.244	-2189.957	3956.138	1.003	1200
lb_c	772.076	325.159	108.318	782.117	1383.636	1.001	18000
lb_f	291.718	358.426	-438.820	300.492	963.873	1.001	89000
lm_c	19.297	119.704	-261.202	35.839	205.050	1.003	940
lm_f	-103.727	162.999	-496.948	-77.934	140.788	1.003	1200
ms_c	107606.522	121.252	107322.105	107623.268	107794.310	1.003	980
ms_f	107470.139	164.833	107072.305	107496.211	107716.739	1.003	1200
ms_nc	107587.225	20.248	107547.000	107587.000	107627.000	1.001	100000
ms_nf	107573.866	20.582	107533.000	107574.000	107614.000	1.001	51000
orb_c	0.873	0.053	0.774	0.871	0.982	1.001	18000
orb_f	0.954	0.057	0.847	0.952	1.071	1.001	94000
orm_c	0.953	0.291	0.504	0.913	1.636	1.003	950
orm_f	1.245	0.385	0.669	1.184	2.178	1.002	1500
deviance	509.134	103.082	500.590	507.722	520.363	1.001	65000

For each parameter, n.eff is a crude measure of effective sample size,
and Rhat is the potential scale reduction factor (at convergence, Rhat=1).

DIC info (using the rule, $pD = \text{var}(\text{deviance})/2$)
 $pD = 5313.0$ and $DIC = 5822.1$
DIC is an estimate of expected predictive error (lower deviance is better).
[[1]]

Quantile (q) = 0.025
Accuracy (r) = +/- 0.005
Probability (s) = 0.95

	Burn-in (M)	Total (N)	Lower bound (Nmin)	Dependence factor (I)
bs_c	60	90360	3746	24.1
bs_f	80	95060	3746	25.4
bs_nc	20	37790	3746	10.1
bs_nf	20	38550	3746	10.3
d_c	90	130140	3746	34.7
d_f	80	98720	3746	26.4
db_c	60	91540	3746	24.4
db_f	80	97460	3746	26.0
deviance	180	282120	3746	75.3
dm_c	1100	1105390	3746	295.0
dm_f	1200	1298040	3746	347.0
lb_c	60	91540	3746	24.4
lb_f	80	97460	3746	26.0
lm_c	1100	1105390	3746	295.0
lm_f	1200	1298040	3746	347.0
ms_c	1200	1396560	3746	373.0
ms_f	1080	1221840	3746	326.0
ms_nc	20	41140	3746	11.0
ms_nf	20	37820	3746	10.1
orb_c	60	91300	3746	24.4
orb_f	60	87880	3746	23.5
orm_c	1210	1363340	3746	364.0
orm_f	880	1016730	3746	271.0

[[2]]

Quantile (q) = 0.025
 Accuracy (r) = +/- 0.005
 Probability (s) = 0.95

	Burn-in (M)	Total (N)	Lower bound (Nmin)	Dependence factor (I)
bs_c	60	86760	3746	23.2
bs_f	80	93280	3746	24.9
bs_nc	10	37930	3746	10.1
bs_nf	10	38120	3746	10.2
d_c	80	90520	3746	24.2
d_f	80	98180	3746	26.2
db_c	60	88860	3746	23.7
db_f	80	94400	3746	25.2
deviance	240	296100	3746	79.0
dm_c	1120	1194200	3746	319.0
dm_f	990	1028520	3746	275.0
lb_c	60	88860	3746	23.7
lb_f	80	94400	3746	25.2
lm_c	1120	1194200	3746	319.0
lm_f	990	1028520	3746	275.0
ms_c	910	959530	3746	256.0
ms_f	990	1084680	3746	290.0
ms_nc	20	40290	3746	10.8
ms_nf	20	40510	3746	10.8
orb_c	60	84420	3746	22.5
orb_f	60	89700	3746	23.9
orm_c	1080	1168320	3746	312.0
orm_f	1040	1110850	3746	297.0

Iterations = 10001:509991
 Thinning interval = 10
 Number of chains = 2
 Sample size per chain = 50000

1. Empirical mean and standard deviation for each variable,
 plus standard error of the mean:

	Mean	SD	Naive SE	Time-series SE
bs_c	1.024e+05	3.319e+02	1.050e+00	2.090e+00
bs_f	1.016e+05	3.664e+02	1.159e+00	2.295e+00
bs_nc	1.016e+05	7.768e+01	2.456e-01	2.465e-01
bs_nf	1.013e+05	7.904e+01	2.499e-01	2.489e-01
d_c	3.529e+04	1.500e+04	4.744e+01	1.209e+02
d_f	1.021e+04	1.680e+04	5.311e+01	1.562e+02
db_c	3.474e+04	1.463e+04	4.627e+01	9.357e+01
db_f	1.313e+04	1.613e+04	5.100e+01	1.029e+02
deviance	5.091e+02	1.031e+02	3.260e-01	4.700e-01
dm_c	5.422e+02	3.364e+03	1.064e+01	1.053e+02
dm_f	-2.915e+03	4.580e+03	1.448e+01	1.446e+02
lb_c	7.721e+02	3.252e+02	1.028e+00	2.079e+00
lb_f	2.917e+02	3.584e+02	1.133e+00	2.287e+00
lm_c	1.930e+01	1.197e+02	3.785e-01	3.748e+00
lm_f	-1.037e+02	1.630e+02	5.154e-01	5.144e+00
ms_c	1.076e+05	1.213e+02	3.834e-01	3.685e+00
ms_f	1.075e+05	1.648e+02	5.212e-01	5.148e+00
ms_nc	1.076e+05	2.025e+01	6.403e-02	6.380e-02
ms_nf	1.076e+05	2.058e+01	6.509e-02	6.509e-02
orb_c	8.733e-01	5.304e-02	1.677e-04	3.394e-04
orb_f	9.537e-01	5.698e-02	1.802e-04	3.637e-04
orm_c	9.534e-01	2.911e-01	9.204e-04	8.926e-03
orm_f	1.245e+00	3.848e-01	1.217e-03	1.210e-02

2. Quantiles for each variable:

	2.5%	25%	50%	75%	97.5%
bs_c	1.017e+05	1.021e+05	1.024e+05	1.026e+05	1.030e+05
bs_f	1.009e+05	1.014e+05	1.016e+05	1.019e+05	1.023e+05
bs_nc	1.014e+05	1.015e+05	1.016e+05	1.016e+05	1.017e+05
bs_nf	1.012e+05	1.013e+05	1.013e+05	1.014e+05	1.015e+05
d_c	4.753e+03	2.538e+04	3.575e+04	4.559e+04	6.344e+04
d_f	-2.413e+04	-8.719e+02	1.066e+04	2.181e+04	4.168e+04
db_c	4.874e+03	2.510e+04	3.520e+04	4.477e+04	6.226e+04
db_f	-1.975e+04	2.563e+03	1.352e+04	2.422e+04	4.337e+04
deviance	5.006e+02	5.048e+02	5.077e+02	5.114e+02	5.204e+02
dm_c	-7.340e+03	-1.341e+03	1.007e+03	2.936e+03	5.762e+03
dm_f	-1.396e+04	-5.364e+03	-2.190e+03	3.369e+02	3.956e+03
lb_c	1.083e+02	5.577e+02	7.821e+02	9.949e+02	1.384e+03
lb_f	-4.388e+02	5.696e+01	3.005e+02	5.382e+02	9.639e+02
lm_c	-2.612e+02	-4.772e+01	3.584e+01	1.045e+02	2.050e+02
lm_f	-4.969e+02	-1.909e+02	-7.793e+01	1.199e+01	1.408e+02

```

ms_c      1.073e+05  1.075e+05  1.076e+05  1.077e+05  1.078e+05
ms_f      1.071e+05  1.074e+05  1.075e+05  1.076e+05  1.077e+05
ms_nc     1.075e+05  1.076e+05  1.076e+05  1.076e+05  1.076e+05
ms_nf     1.075e+05  1.076e+05  1.076e+05  1.076e+05  1.076e+05
orb_c     7.742e-01  8.368e-01  8.714e-01  9.081e-01  9.821e-01
orb_f     8.474e-01  9.144e-01  9.521e-01  9.909e-01  1.071e+00
orm_c     5.040e-01  7.454e-01  9.128e-01  1.116e+00  1.636e+00
orm_f     6.688e-01  9.717e-01  1.184e+00  1.452e+00  2.178e+00

```

```
quartz
  2
```

Cost-effectiveness analysis summary

Reference intervention: Community Intervention
 Comparator intervention: do nothing

Optimal decision: choose do nothing for k<155 and Community Intervention for k>=155

Analysis for willingness to pay parameter k = 780

```

                Expected utility
do nothing      5923564111
Community Intervention  5945738130

```

```

                EIB    CEAC    ICER
Community Intervention vs do nothing 22174019 0.96623 151.59

```

Optimal intervention (max expected utility) for k=780: Community Intervention

```
EVPI 170677
quartz
  2
```

Cost-effectiveness analysis summary

Reference intervention: Facility Intervention
 Comparator intervention: do nothing

Optimal decision: choose do nothing for k<550 and Facility Intervention for k>=550

Analysis for willingness to pay parameter k = 780

```

                Expected utility
do nothing      5914496043
Facility Intervention  5916869632

```

```

                EIB    CEAC    ICER
Facility Intervention vs do nothing 2373589 0.58166 547.58

```

Optimal intervention (max expected utility) for k=780: Facility Intervention

```
EVPI 4144723
quartz
  2
```

Cost-effectiveness analysis summary

Reference intervention: Community Intervention
 Comparator intervention: Facility Intervention

Community Intervention dominates for all k in [0 - 2500]

Analysis for willingness to pay parameter k = 780

```

                Expected utility
Community Intervention  5945738130
Facility Intervention  5916869632

```

```

                EIB    CEAC    ICER
Community Intervention vs Facility Intervention 28868498 0.94837 -6.6329

```

Optimal intervention (max expected utility) for k=780: Community Intervention

```
EVPI 383986
```

```

Compiling model graph
  Resolving undeclared variables
  Allocating nodes
  Graph Size: 357

```

```
Initializing model
```

Inference for Bugs model at "/Users/timothycolbourn/Documents/Papers/MaiKhanda CEA paper/JAGS revised/MKmodelFC_final_LLE.txt", fit using jags,
 2 chains, each with 510000 iterations (first 10000 discarded), n.thin = 10
 n.sims = 100000 iterations saved

	mu.vect	sd.vect	2.5%	50%	97.5%	Rhat	n.eff
bs_con	50862.710	54.501	50756.000	50863.000	50969.000	1.001	100000
bs_fc	51338.166	224.031	50873.106	51346.741	51748.058	1.001	100000
d_fc	19738.609	10538.349	-2049.754	20219.078	38975.523	1.001	41000
d_fc_	39477.217	21076.698	-4099.509	40438.156	77951.046	1.001	41000
db_fc	21395.535	9873.907	982.407	21794.699	39603.014	1.001	100000
db_fc_	42791.069	19747.813	1964.815	43589.398	79206.028	1.001	100000
dm_fc	-1656.926	3648.879	-10858.291	-914.776	3185.935	1.004	2500
dm_fc_	-3313.852	7297.758	-21716.583	-1829.552	6371.871	1.004	2500
lb_fc	475.456	219.420	21.831	484.327	880.067	1.001	100000
lb_fc_	950.913	438.840	43.663	968.653	1760.134	1.001	100000
lm_fc	-58.965	129.853	-386.416	-32.554	113.378	1.004	2500
lm_fc_	-117.931	259.707	-772.832	-65.109	226.757	1.004	2500
ms_con	53782.154	14.741	53753.000	53782.000	53810.000	1.001	63000
ms_fc	53723.189	131.158	53391.857	53749.976	53896.268	1.003	2600
orb_fc	0.841	0.073	0.708	0.838	0.993	1.001	100000
orm_fc	1.273	0.600	0.482	1.150	2.794	1.001	7000
deviance	258.836	118.929	251.181	257.482	269.122	1.001	5700

For each parameter, n.eff is a crude measure of effective sample size, and Rhat is the potential scale reduction factor (at convergence, Rhat=1).

DIC info (using the rule, $pD = \text{var}(\text{deviance})/2$)
 $pD = 7072.1$ and $DIC = 7330.9$
 DIC is an estimate of expected predictive error (lower deviance is better).
 [[1]]

Quantile (q) = 0.025
 Accuracy (r) = +/- 0.005
 Probability (s) = 0.95

	Burn-in (M)	Total (N)	Lower bound (Nmin)	Dependence factor (I)
bs_con	20	38400	3746	10.3
bs_fc	60	89460	3746	23.9
d_fc	120	187840	3746	50.1
d_fc_	120	187840	3746	50.1
db_fc	80	90000	3746	24.0
db_fc_	80	90000	3746	24.0
deviance	180	242460	3746	64.7
dm_fc	990	1044340	3746	279.0
dm_fc_	990	1044340	3746	279.0
lb_fc	80	90000	3746	24.0
lb_fc_	80	90000	3746	24.0
lm_fc	990	1044340	3746	279.0
lm_fc_	990	1044340	3746	279.0
ms_con	20	41690	3746	11.1
ms_fc	990	1024760	3746	274.0
orb_fc	60	87540	3746	23.4
orm_fc	1080	1238220	3746	331.0

[[2]]

Quantile (q) = 0.025
 Accuracy (r) = +/- 0.005
 Probability (s) = 0.95

	Burn-in (M)	Total (N)	Lower bound (Nmin)	Dependence factor (I)
bs_con	20	37610	3746	10.0
bs_fc	60	86400	3746	23.1
d_fc	120	139470	3746	37.2
d_fc_	120	139470	3746	37.2
db_fc	60	87580	3746	23.4
db_fc_	60	87580	3746	23.4
deviance	160	195520	3746	52.2
dm_fc	1200	1289520	3746	344.0
dm_fc_	1200	1289520	3746	344.0
lb_fc	60	87580	3746	23.4
lb_fc_	60	87580	3746	23.4
lm_fc	1200	1289520	3746	344.0
lm_fc_	1200	1289520	3746	344.0
ms_con	20	42020	3746	11.2
ms_fc	1300	1450930	3746	387.0
orb_fc	80	93480	3746	25.0
orm_fc	990	1044340	3746	279.0

Iterations = 10001:509991
 Thinning interval = 10
 Number of chains = 2
 Sample size per chain = 50000

1. Empirical mean and standard deviation for each variable,
 plus standard error of the mean:

	Mean	SD	Naive SE	Time-series SE
bs_con	50862.7095	54.5009	0.1723471	0.1707505
bs_fc	51338.1658	224.0307	0.7084472	1.4103686
d_fc	19738.6087	10538.3488	33.3251851	122.0902297
d_fc_	39477.2174	21076.6977	66.6503703	244.1804594
db_fc	21395.5346	9873.9067	31.2240347	63.2247828
db_fc_	42791.0692	19747.8135	62.4480693	126.4495656
deviance	258.8364	118.9290	0.3760866	0.4633666
dm_fc	-1656.9259	3648.8792	11.5387691	120.3292760
dm_fc_	-3313.8517	7297.7584	23.0775382	240.6585520
lb_fc	475.4563	219.4201	0.6938674	1.4049952
lb_fc_	950.9126	438.8403	1.3877349	2.8099903
lm_fc	-58.9653	129.8534	0.4106324	4.2821806
lm_fc_	-117.9307	259.7067	0.8212647	8.5643613
ms_con	53782.1539	14.7411	0.0466154	0.0467904
ms_fc	53723.1886	131.1580	0.4147580	4.2651317
orb_fc	0.8409	0.0729	0.0002305	0.0004673
orm_fc	1.2734	0.6000	0.0018974	0.0196008

2. Quantiles for each variable:

	2.5%	25%	50%	75%	97.5%
bs_con	50756.0000	50826.0000	50863.0000	50899.000	50969.0000
bs_fc	50873.1057	51192.8322	51346.7408	51494.565	51748.0582
d_fc	-2049.7544	12937.2223	20219.0779	27065.169	38975.5228
d_fc_	-4099.5089	25874.4447	40438.1558	54130.338	77951.0456
db_fc	982.4074	14980.6830	21794.6989	28283.757	39603.0142
db_fc_	1964.8149	29961.3660	43589.3978	56567.513	79206.0285
deviance	251.1813	254.7882	257.4819	260.805	269.1220
dm_fc	-10858.2915	-3355.3934	-914.7762	876.682	3185.9354
dm_fc_	-21716.5830	-6710.7868	-1829.5525	1753.365	6371.8707
lb_fc	21.8313	332.9041	484.3266	628.528	880.0670
lb_fc_	43.6626	665.8081	968.6533	1257.056	1760.1340
lm_fc	-386.4161	-119.4090	-32.5543	31.199	113.3785
lm_fc_	-772.8321	-238.8180	-65.1086	62.397	226.7570
ms_con	53753.0000	53772.0000	53782.0000	53792.000	53810.0000
ms_fc	53391.8568	53661.7696	53749.9756	53814.902	53896.2684
orb_fc	0.7078	0.7900	0.8375	0.888	0.9925
orm_fc	0.4817	0.8554	1.1499	1.553	2.7940

quartz
2

Cost-effectiveness analysis summary

Reference intervention: Facility & Community Interventions
 Comparator intervention: do nothing

Optimal decision: choose do nothing for k<280 and Facility & Community Interventions for k>=280

Analysis for willingness to pay parameter k = 780

	Expected utility
do nothing	2964078353
Facility & Community Interventions	2974003967

	EIB	CEAC	ICER
Facility & Community Interventions vs do nothing	9925614	0.88326	277.15

Optimal intervention (max expected utility) for k=780: Facility & Community Interventions

EVPI 534291
quartz
2

Cost-effectiveness analysis summary

Reference intervention: Community Intervention
 Comparator intervention(s): Facility Intervention
 : FICI

Optimal decision: choose Community Intervention for k<555 and FICI for k>=555

Analysis for willingness to pay parameter k = 780

	Expected utility
Community Intervention	5945738130
Facility Intervention	5916869632
FICI	5948007934

	EIB	CEAC	ICER
Community Intervention vs Facility Intervention	28868498	0.94837	-6.6329
Community Intervention vs FICI	-2269803	0.44973	554.8100

Optimal intervention (max expected utility) for k=780: FICI

EVPI 7212170
quartz
2

Cost-effectiveness analysis summary

Reference intervention: Facility Intervention
Comparator intervention(s): Community Intervention
: FICI

Optimal decision: choose Community Intervention for k<555 and FICI for k>=555

Analysis for willingness to pay parameter k = 780

	Expected utility
Community Intervention	5945738130
Facility Intervention	5916869632
FICI	5948007934

	EIB	CEAC	ICER
Facility Intervention vs Community Intervention	-28868498	0.05163	-6.6329
Facility Intervention vs FICI	-31138302	0.07416	114.3433

Optimal intervention (max expected utility) for k=780: FICI

EVPI 7212170
quartz
2

Cost-effectiveness analysis summary

Reference intervention: FICI
Comparator intervention(s): Community Intervention
: Facility Intervention

Optimal decision: choose Community Intervention for k<555 and FICI for k>=555

Analysis for willingness to pay parameter k = 780

	Expected utility
Community Intervention	5945738130
Facility Intervention	5916869632
FICI	5948007934

	EIB	CEAC	ICER
FICI vs Community Intervention	2269803	0.55027	554.81
FICI vs Facility Intervention	31138302	0.92584	114.34

Optimal intervention (max expected utility) for k=780: FICI

EVPI 7212170

Initializing model

Inference for Bugs model at "/Users/timothycolbourn/Documents/Papers/MaiKhanda CEA paper/JAGS revised/MKmodel_final.txt", fit using jags,

2 chains, each with 21000 iterations (first 1000 discarded), n.thin = 2
n.sims = 20000 iterations saved

	mu.vect	sd.vect	2.5%	50%	97.5%	Rhat	n.eff
bs_c	102352.659	343.406	101661.403	102364.806	102974.464	1.002	1600
bs_f	101642.576	364.048	100913.382	101652.590	102322.593	1.001	20000
bs_nc	101580.385	77.117	101429.000	101581.000	101730.000	1.001	20000
bs_nf	101329.767	78.958	101174.000	101330.000	101485.000	1.001	20000
d_c	67298.347	29427.529	8538.188	68147.311	121191.979	1.014	120
d_f	20716.243	31641.196	-42734.637	21267.961	79527.764	1.001	20000
db_c	66415.528	28949.779	7855.385	67343.667	119294.148	1.002	1500
db_f	26901.607	30614.663	-34747.922	27731.431	84027.008	1.001	20000
dm_c	882.819	6334.383	-12815.524	2235.752	11123.963	1.206	12
dm_f	-6185.364	9067.853	-26875.173	-5220.987	9104.277	1.001	3400
lb_c	772.274	336.625	91.342	783.066	1387.141	1.002	1500
lb_f	312.809	355.984	-404.046	322.459	977.058	1.001	20000
lm_c	16.573	118.911	-240.577	41.970	208.822	1.206	12
lm_f	-116.113	170.224	-504.509	-98.010	170.908	1.001	3400
ms_c	107603.759	120.510	107340.560	107628.444	107796.684	1.200	12
ms_f	107457.681	172.146	107063.392	107477.079	107744.663	1.001	3800
ms_nc	107587.186	20.354	107547.000	107587.000	107626.000	1.001	20000
ms_nf	107573.795	20.677	107532.000	107574.000	107614.000	1.001	20000
orb_c	0.873	0.055	0.773	0.871	0.985	1.002	1500
orb_f	0.950	0.057	0.845	0.949	1.065	1.001	20000
orm_c	0.960	0.289	0.491	0.897	1.588	1.191	12
orm_f	1.274	0.401	0.596	1.231	2.195	1.006	1500
deviance	520.565	358.490	500.567	508.428	527.019	1.001	3800

For each parameter, n.eff is a crude measure of effective sample size, and Rhat is the potential scale reduction factor (at convergence, Rhat=1).

DIC info (using the rule, $pD = \text{var}(\text{deviance})/2$)

$pD = 64259.6$ and $DIC = 64780.1$

DIC is an estimate of expected predictive error (lower deviance is better).

[[1]]

Quantile (q) = 0.025
Accuracy (r) = +/- 0.005
Probability (s) = 0.95

	Burn-in (M)	Total (N)	Lower bound (Nmin)	Dependence factor (I)
bs_c	48	45760	3746	12.20
bs_f	40	43320	3746	11.60
bs_nc	4	7508	3746	2.00
bs_nf	4	7622	3746	2.03
d_c	48	47814	3746	12.80
d_f	40	47800	3746	12.80
db_c	44	43108	3746	11.50
db_f	36	38520	3746	10.30
deviance	40	50768	3746	13.60
dm_c	216	248166	3746	66.20
dm_f	224	249584	3746	66.60
lb_c	44	43108	3746	11.50
lb_f	36	38520	3746	10.30
lm_c	216	248166	3746	66.20
lm_f	224	249584	3746	66.60
ms_c	216	241380	3746	64.40
ms_f	176	180528	3746	48.20
ms_nc	4	8052	3746	2.15
ms_nf	4	7890	3746	2.11
orb_c	50	53240	3746	14.20
orb_f	54	65454	3746	17.50
orm_c	272	271540	3746	72.50
orm_f	328	374936	3746	100.00

[[2]]

Quantile (q) = 0.025
Accuracy (r) = +/- 0.005
Probability (s) = 0.95

	Burn-in (M)	Total (N)	Lower bound (Nmin)	Dependence factor (I)
bs_c	42	52080	3746	13.90
bs_f	42	43542	3746	11.60
bs_nc	4	7886	3746	2.11
bs_nf	4	7950	3746	2.12

d_c	42	43602	3746	11.60
d_f	36	37626	3746	10.00
db_c	48	49218	3746	13.10
db_f	36	41876	3746	11.20
deviance	48	61096	3746	16.30
dm_c	300	369552	3746	98.70
dm_f	228	225672	3746	60.20
lb_c	48	49218	3746	13.10
lb_f	36	41876	3746	11.20
lm_c	300	369552	3746	98.70
lm_f	228	225672	3746	60.20
ms_c	140	145390	3746	38.80
ms_f	408	380136	3746	101.00
ms_nc	4	8248	3746	2.20
ms_nf	4	8124	3746	2.17
orb_c	42	49164	3746	13.10
orb_f	42	40614	3746	10.80
orm_c	402	430980	3746	115.00
orm_f	160	167400	3746	44.70

Iterations = 1001:20999
Thinning interval = 2
Number of chains = 2
Sample size per chain = 10000

1. Empirical mean and standard deviation for each variable,
plus standard error of the mean:

	Mean	SD	Naive SE	Time-series SE
bs_c	1.024e+05	3.434e+02	2.428e+00	1.037e+01
bs_f	1.016e+05	3.640e+02	2.574e+00	1.114e+01
bs_nc	1.016e+05	7.712e+01	5.453e-01	5.327e-01
bs_nf	1.013e+05	7.896e+01	5.583e-01	5.583e-01
d_c	6.730e+04	2.943e+04	2.081e+02	9.264e+02
d_f	2.072e+04	3.164e+04	2.237e+02	1.022e+03
db_c	6.642e+04	2.895e+04	2.047e+02	8.986e+02
db_f	2.690e+04	3.061e+04	2.165e+02	9.489e+02
deviance	5.206e+02	3.585e+02	2.535e+00	6.547e+00
dm_c	8.828e+02	6.334e+03	4.479e+01	8.641e+02
dm_f	-6.185e+03	9.068e+03	6.412e+01	1.400e+03
lb_c	7.723e+02	3.366e+02	2.380e+00	1.045e+01
lb_f	3.128e+02	3.560e+02	2.517e+00	1.103e+01
lm_c	1.657e+01	1.189e+02	8.408e-01	1.622e+01
lm_f	-1.161e+02	1.702e+02	1.204e+00	2.628e+01
ms_c	1.076e+05	1.205e+02	8.521e-01	1.630e+01
ms_f	1.075e+05	1.721e+02	1.217e+00	2.543e+01
ms_nc	1.076e+05	2.035e+01	1.439e-01	1.439e-01
ms_nf	1.076e+05	2.068e+01	1.462e-01	1.505e-01
orb_c	8.732e-01	5.513e-02	3.898e-04	1.693e-03
orb_f	9.504e-01	5.653e-02	3.997e-04	1.756e-03
orm_c	9.600e-01	2.888e-01	2.042e-03	3.876e-02
orm_f	1.274e+00	4.014e-01	2.838e-03	6.203e-02

2. Quantiles for each variable:

	2.5%	25%	50%	75%	97.5%
bs_c	1.017e+05	1.021e+05	1.024e+05	1.026e+05	1.030e+05
bs_f	1.009e+05	1.014e+05	1.017e+05	1.019e+05	1.023e+05
bs_nc	1.014e+05	1.015e+05	1.016e+05	1.016e+05	1.017e+05
bs_nf	1.012e+05	1.013e+05	1.013e+05	1.014e+05	1.015e+05
d_c	8.538e+03	4.854e+04	6.815e+04	8.709e+04	1.212e+05
d_f	-4.273e+04	1.053e+02	2.127e+04	4.210e+04	7.953e+04
db_c	7.855e+03	4.836e+04	6.734e+04	8.554e+04	1.193e+05
db_f	-3.475e+04	6.991e+03	2.773e+04	4.768e+04	8.403e+04
deviance	5.006e+02	5.049e+02	5.084e+02	5.127e+02	5.270e+02
dm_c	-1.282e+04	-3.169e+03	2.236e+03	5.369e+03	1.112e+04
dm_f	-2.688e+04	-1.192e+04	-5.221e+03	6.142e+02	9.104e+03
lb_c	9.134e+01	5.623e+02	7.831e+02	9.946e+02	1.387e+03
lb_f	-4.040e+02	8.129e+01	3.225e+02	5.544e+02	9.771e+02
lm_c	-2.406e+02	-5.948e+01	4.197e+01	1.008e+02	2.088e+02
lm_f	-5.045e+02	-2.238e+02	-9.801e+01	1.153e+01	1.709e+02
ms_c	1.073e+05	1.075e+05	1.076e+05	1.077e+05	1.078e+05
ms_f	1.071e+05	1.073e+05	1.075e+05	1.076e+05	1.077e+05
ms_nc	1.075e+05	1.076e+05	1.076e+05	1.076e+05	1.076e+05
ms_nf	1.075e+05	1.076e+05	1.076e+05	1.076e+05	1.076e+05
orb_c	7.734e-01	8.367e-01	8.712e-01	9.074e-01	9.849e-01
orb_f	8.455e-01	9.119e-01	9.486e-01	9.870e-01	1.065e+00
orm_c	4.908e-01	7.555e-01	8.975e-01	1.144e+00	1.588e+00
orm_f	5.958e-01	9.728e-01	1.231e+00	1.531e+00	2.195e+00

quartz
2

Cost-effectiveness analysis summary

Reference intervention: Community Intervention
Comparator intervention: do nothing

Optimal decision: choose do nothing for $k < 80$ and Community Intervention for $k \geq 80$

Analysis for willingness to pay parameter $k = 780$

	Expected utility
do nothing	11284324399
Community Intervention	11331468318

	EIB	CEAC	ICER
Community Intervention vs do nothing	47143920	0.97835	79.479

Optimal intervention (max expected utility) for $k=780$: Community Intervention

EVPI 285672
quartz
2

Cost-effectiveness analysis summary

Reference intervention: Facility Intervention
Comparator intervention: do nothing

Optimal decision: choose do nothing for $k < 270$ and Facility Intervention for $k \geq 270$

Analysis for willingness to pay parameter $k = 780$

	Expected utility
do nothing	11266956465
Facility Intervention	11277522922

	EIB	CEAC	ICER
Facility Intervention vs do nothing	10566457	0.67495	269.94

Optimal intervention (max expected utility) for $k=780$: Facility Intervention

EVPI 5440834
quartz
2

Cost-effectiveness analysis summary

Reference intervention: Community Intervention
Comparator intervention: Facility Intervention

Community Intervention dominates for all k in $[0 - 2500]$

Analysis for willingness to pay parameter $k = 780$

	Expected utility
Community Intervention	11331468318
Facility Intervention	11277522922

	EIB	CEAC	ICER
Community Intervention vs Facility Intervention	53945396	0.94415	-3.5356

Optimal intervention (max expected utility) for $k=780$: Community Intervention

EVPI 903141

Initializing model

Inference for Bugs model at "/Users/timothycolbourn/Documents/Papers/MaiKhanda CEA paper/JAGS revised/MKmodelFC_final.txt", fit using jags,
2 chains, each with 21000 iterations (first 1000 discarded), n.thin = 2
n.sims = 20000 iterations saved

	mu.vect	sd.vect	2.5%	50%	97.5%	Rhat	n.eff
bs_con	50862.986	54.260	50756.000	50863.000	50970.000	1.001	20000
bs_fc	51342.036	224.311	50894.180	51349.031	51754.583	1.001	20000
d_fc	38986.900	20133.229	-1761.312	39646.226	75428.855	1.008	5700
db_fc	41198.284	18868.364	3537.467	41636.420	75926.881	1.001	15000
dm_fc	-2211.385	6632.355	-19802.372	-792.271	6448.254	1.083	250
lb_fc	479.050	219.400	41.133	484.144	882.871	1.001	15000
lm_fc	-41.513	124.505	-371.736	-14.873	121.049	1.083	250
ms_con	53782.142	14.881	53753.000	53782.000	53811.000	1.001	20000
ms_fc	53740.629	125.852	53409.374	53767.722	53904.154	1.081	260

orb_fc	0.840	0.073	0.706	0.838	0.986	1.001	12000
orm_fc	1.192	0.575	0.447	1.069	2.739	1.032	20000
deviance	274.238	684.960	251.663	258.148	271.404	1.005	19000

For each parameter, n.eff is a crude measure of effective sample size, and Rhat is the potential scale reduction factor (at convergence, Rhat=1).

DIC info (using the rule, $pD = \text{var}(\text{deviance})/2$)
 $pD = 234593.1$ and $DIC = 234867.4$
 DIC is an estimate of expected predictive error (lower deviance is better).
 [[1]]

Quantile (q) = 0.025
 Accuracy (r) = +/- 0.005
 Probability (s) = 0.95

	Burn-in (M)	Total (N)	Lower bound (Nmin)	Dependence factor (I)
bs_con	4	7604	3746	2.03
bs_fc	40	49416	3746	13.20
d_fc	48	50352	3746	13.40
db_fc	36	41876	3746	11.20
deviance	72	73504	3746	19.60
dm_fc	260	237692	3746	63.50
lb_fc	36	41876	3746	11.20
lm_fc	260	237692	3746	63.50
ms_con	4	8630	3746	2.30
ms_fc	280	316940	3746	84.60
orb_fc	56	59904	3746	16.00
orm_fc	264	293670	3746	78.40

[[2]]

Quantile (q) = 0.025
 Accuracy (r) = +/- 0.005
 Probability (s) = 0.95

	Burn-in (M)	Total (N)	Lower bound (Nmin)	Dependence factor (I)
bs_con	2	7800	3746	2.08
bs_fc	28	29976	3746	8.00
d_fc	36	35130	3746	9.38
db_fc	36	36876	3746	9.84
deviance	48	65472	3746	17.50
dm_fc	352	433598	3746	116.00
lb_fc	36	36876	3746	9.84
lm_fc	352	433598	3746	116.00
ms_con	4	8078	3746	2.16
ms_fc	308	380842	3746	102.00
orb_fc	48	53016	3746	14.20
orm_fc	224	242196	3746	64.70

Iterations = 1001:20999
 Thinning interval = 2
 Number of chains = 2
 Sample size per chain = 10000

1. Empirical mean and standard deviation for each variable, plus standard error of the mean:

	Mean	SD	Naive SE	Time-series SE
bs_con	50862.9863	54.25979	0.3836746	0.383677
bs_fc	51342.0361	224.31077	1.5861167	6.687838
d_fc	38986.8997	20133.22916	142.3634287	733.900825
db_fc	41198.2844	18868.36368	133.4194791	575.676426
deviance	274.2380	684.95976	4.8433969	9.588212
dm_fc	-2211.3847	6632.35549	46.8978354	1037.517544
lb_fc	479.0498	219.39958	1.5513893	6.693912
lm_fc	-41.5128	124.50451	0.8803799	19.476582
ms_con	53782.1420	14.88095	0.1052242	0.105225
ms_fc	53740.6292	125.85151	0.8899046	19.973687
orb_fc	0.8397	0.07318	0.0005175	0.002215
orm_fc	1.1925	0.57523	0.0040675	0.092813

2. Quantiles for each variable:

	2.5%	25%	50%	75%	97.5%
bs_con	50756.0000	50827.0000	50863.0000	50899.0000	50970.0000
bs_fc	50894.1796	51200.1761	51349.0309	51491.9897	51754.5831
d_fc	-1761.3122	26414.6354	39646.2265	52536.0328	75428.8549
db_fc	3537.4673	29204.5867	41636.4200	53689.1623	75926.8810

deviance	251.6627	255.3545	258.1476	261.4310	271.4043
dm_fc	-19802.3720	-5638.2032	-792.2708	2558.6247	6448.2542
lb_fc	41.1333	339.5882	484.1444	624.2926	882.8707
lm_fc	-371.7359	-105.8420	-14.8727	48.0313	121.0485
ms_con	53753.0000	53772.0000	53782.0000	53792.0000	53811.0000
ms_fc	53409.3744	53675.9711	53767.7223	53831.4781	53904.1537
orb_fc	0.7063	0.7910	0.8377	0.8857	0.9862
orm_fc	0.4472	0.7775	1.0689	1.4901	2.7391

quartz
2

Cost-effectiveness analysis summary

Reference intervention: Facility & Community Interventions
Comparator intervention: do nothing

Optimal decision: choose do nothing for k<145 and Facility & Community Interventions for k>=145

Analysis for willingness to pay parameter k = 780

	Expected utility
do nothing	5646569387
Facility & Community Interventions	5671508668

	EIB	CEAC	ICER
Facility & Community Interventions vs do nothing	24939281	0.94225	140.32

Optimal intervention (max expected utility) for k=780: Facility & Community Interventions

EVPI 488886
quartz
2

Cost-effectiveness analysis summary

Reference intervention: Community Intervention
Comparator intervention(s): Facility Intervention
: FICI

Optimal decision: choose Community Intervention for k<255 and FICI for k>=255

Analysis for willingness to pay parameter k = 780

	Expected utility
Community Intervention	11331468318
Facility Intervention	11277522922
FICI	11343017336

	EIB	CEAC	ICER
Community Intervention vs Facility Intervention	53945396	0.94415	-3.5356
Community Intervention vs FICI	-11549017	0.37850	254.4698

Optimal intervention (max expected utility) for k=780: FICI

EVPI 10716765
quartz
2

[1] "time horizon in years"
[1] 5
[1] "cost discount rate"
[1] 0.03
[1] "effects discount rate"
[1] 0

Cost-effectiveness analysis summary

Reference intervention: Community Intervention
Comparator intervention(s): Facility Intervention
: FICI

Optimal decision: choose Community Intervention for k<230 and FICI for k>=230

Analysis for willingness to pay parameter k = 780

	Expected utility
Community Intervention	31506755790
Facility Intervention	313568285502
FICI	315406426918

	EIB	CEAC	ICER
--	-----	------	------

Community Intervention vs Facility Intervention 1499272288 0.99875 -3.2331
Community Intervention vs FICI -338869127 0.24650 225.3860

Optimal intervention (max expected utility) for k=780: FICI

EVPI 72244693

quartz

2

[1] "time horizon in years"

[1] 5

[1] "cost discount rate"

[1] 0

[1] "effects discount rate"

[1] 0

Cost-effectiveness analysis summary

Reference intervention: Community Intervention

Comparator intervention(s): Facility Intervention

: FICI

Optimal decision: choose Community Intervention for k<250 and FICI for k>=250

Analysis for willingness to pay parameter k = 780

	Expected utility
Community Intervention	315055903400
Facility Intervention	313556064029
FICI	315382551054

	EIB	CEAC	ICER
Community Intervention vs Facility Intervention	1499839371	0.99875	-3.5294
Community Intervention vs FICI	-326647654	0.25400	245.3884

Optimal intervention (max expected utility) for k=780: FICI

EVPI 75301091

quartz

2

[1] "time horizon in years"

[1] 5

[1] "cost discount rate"

[1] 0.1

[1] "effects discount rate"

[1] 0

Cost-effectiveness analysis summary

Reference intervention: Community Intervention

Comparator intervention(s): Facility Intervention

: FICI

Optimal decision: choose Community Intervention for k<190 and FICI for k>=190

Analysis for willingness to pay parameter k = 780

	Expected utility
Community Intervention	315089433936
Facility Intervention	313591226103
FICI	315451243664

	EIB	CEAC	ICER
Community Intervention vs Facility Intervention	1498207833	0.99875	-2.6771
Community Intervention vs FICI	-361809728	0.23225	187.8400

Optimal intervention (max expected utility) for k=780: FICI

EVPI 66762106

quartz

2

[1] "time horizon in years"

[1] 5

[1] "cost discount rate"

[1] 0.03

[1] "effects discount rate"

[1] 0.02

Cost-effectiveness analysis summary

Reference intervention: Community Intervention

Comparator intervention(s): Facility Intervention

: FICI

Optimal decision: choose Community Intervention for $k < 240$ and FICI for $k \geq 240$

Analysis for willingness to pay parameter $k = 780$

	Expected utility			
Community Intervention	297004097950			
Facility Intervention	295590391430			
FICI	297315655188			
		EIB	CEAC	ICER
Community Intervention vs Facility Intervention	1413706520	0.99875	-3.4297	
Community Intervention vs FICI	-311557239	0.25155	239.0876	

Optimal intervention (max expected utility) for $k=780$: FICI

EVPI 70073716
quartz
2
[1] "time horizon in years"
[1] 5
[1] "cost discount rate"
[1] 0
[1] "effects discount rate"
[1] 0.02

Cost-effectiveness analysis summary

Reference intervention: Community Intervention
Comparator intervention(s): Facility Intervention
: FICI

Optimal decision: choose Community Intervention for $k < 265$ and FICI for $k \geq 265$

Analysis for willingness to pay parameter $k = 780$

	Expected utility			
Community Intervention	296992443559			
Facility Intervention	295578169957			
FICI	297291779325			
		EIB	CEAC	ICER
Community Intervention vs Facility Intervention	1414273602	0.99875	-3.744	
Community Intervention vs FICI	-299335766	0.25865	260.306	

Optimal intervention (max expected utility) for $k=780$: FICI

EVPI 73193910
quartz
2
[1] "time horizon in years"
[1] 5
[1] "cost discount rate"
[1] 0.1
[1] "effects discount rate"
[1] 0.02

Cost-effectiveness analysis summary

Reference intervention: Community Intervention
Comparator intervention(s): Facility Intervention
: FICI

Optimal decision: choose Community Intervention for $k < 200$ and FICI for $k \geq 200$

Analysis for willingness to pay parameter $k = 780$

	Expected utility			
Community Intervention	297025974095			
Facility Intervention	295613332031			
FICI	297360471934			
		EIB	CEAC	ICER
Community Intervention vs Facility Intervention	1412642064	0.99875	-2.8398	
Community Intervention vs FICI	-334497839	0.23610	199.2591	

Optimal intervention (max expected utility) for $k=780$: FICI

EVPI 64479001
quartz
2


```
[1] "time horizon in years"
[1] 5
[1] "cost discount rate"
[1] 0.03
[1] "effects discount rate"
[1] 0.03
```

Cost-effectiveness analysis summary

```
Reference intervention: Community Intervention
Comparator intervention(s): Facility Intervention
                        : FICI
```

Optimal decision: choose Community Intervention for $k < 250$ and FICI for $k \geq 250$

Analysis for willingness to pay parameter $k = 780$

```
                Expected utility
Community Intervention  288572376234
Facility Intervention  287198610388
FICI                   288871184738
```

```
                EIB    CEAC    ICER
Community Intervention vs Facility Intervention 1373765846 0.99875 -3.5299
Community Intervention vs FICI                 -298808504 0.25405 246.0703
```

Optimal intervention (max expected utility) for $k=780$: FICI

EVPI 69079152

quartz

2

```
[1] "time horizon in years"
[1] 5
[1] "cost discount rate"
[1] 0
[1] "effects discount rate"
[1] 0.03
```

Cost-effectiveness analysis summary

```
Reference intervention: Community Intervention
Comparator intervention(s): Facility Intervention
                        : FICI
```

Optimal decision: choose Community Intervention for $k < 270$ and FICI for $k \geq 270$

Analysis for willingness to pay parameter $k = 780$

```
                Expected utility
Community Intervention  288560721843
Facility Intervention  287186388915
FICI                   288847308875
```

```
                EIB    CEAC    ICER
Community Intervention vs Facility Intervention 1374332929 0.99875 -3.8533
Community Intervention vs FICI                 -286587031 0.26200 267.9084
```

Optimal intervention (max expected utility) for $k=780$: FICI

EVPI 72231344

quartz

2

```
[1] "time horizon in years"
[1] 5
[1] "cost discount rate"
[1] 0.1
[1] "effects discount rate"
[1] 0.03
```

Cost-effectiveness analysis summary

```
Reference intervention: Community Intervention
Comparator intervention(s): Facility Intervention
                        : FICI
```

Optimal decision: choose Community Intervention for $k < 210$ and FICI for $k \geq 210$

Analysis for willingness to pay parameter $k = 780$

```
                Expected utility
Community Intervention  288594252379
```

Facility Intervention	287221550988			
FICI	288916001484			
		EIB	CEAC	ICER
Community Intervention vs Facility Intervention	1372701391	0.99875	-2.9227	
Community Intervention vs FICI	-321749105	0.23855	205.0786	

Optimal intervention (max expected utility) for k=780: FICI

EVPI 63426113
quartz
2
[1] "time horizon in years"
[1] 10
[1] "cost discount rate"
[1] 0.03
[1] "effects discount rate"
[1] 0

Cost-effectiveness analysis summary

Reference intervention: Community Intervention
Comparator intervention(s): Facility Intervention
: FICI

Optimal decision: choose Community Intervention for k<210 and FICI for k>=210

Analysis for willingness to pay parameter k = 780

	Expected utility			
Community Intervention	630157093293			
Facility Intervention	627159407405			
FICI	630857667962			
		EIB	CEAC	ICER
Community Intervention vs Facility Intervention	2997685888	0.9990	-3.0089	
Community Intervention vs FICI	-700574669	0.1505	206.6982	

Optimal intervention (max expected utility) for k=780: FICI

EVPI 59381827
quartz
2
[1] "time horizon in years"
[1] 10
[1] "cost discount rate"
[1] 0
[1] "effects discount rate"
[1] 0

Cost-effectiveness analysis summary

Reference intervention: Community Intervention
Comparator intervention(s): Facility Intervention
: FICI

Optimal decision: choose Community Intervention for k<245 and FICI for k>=245

Analysis for willingness to pay parameter k = 780

	Expected utility			
Community Intervention	630116336969			
Facility Intervention	627116667949			
FICI	630774172181			
		EIB	CEAC	ICER
Community Intervention vs Facility Intervention	2999669020	0.999	-3.5269	
Community Intervention vs FICI	-657835212	0.166	241.6732	

Optimal intervention (max expected utility) for k=780: FICI

EVPI 66144334
quartz
2
[1] "time horizon in years"
[1] 10
[1] "cost discount rate"
[1] 0.1
[1] "effects discount rate"
[1] 0

Cost-effectiveness analysis summary

Reference intervention: Community Intervention
Comparator intervention(s): Facility Intervention
: FICI

Optimal decision: choose Community Intervention for k<150 and FICI for k>=150

Analysis for willingness to pay parameter k = 780

	Expected utility
Community Intervention	630223245104
Facility Intervention	627228778048
FICI	630993190415

	EIB	CEAC	ICER
Community Intervention vs Facility Intervention	2994467055	0.999	-2.1681
Community Intervention vs FICI	-769945312	0.129	149.9301

Optimal intervention (max expected utility) for k=780: FICI

EVPI 49814291

quartz

2

[1] "time horizon in years"

[1] 10

[1] "cost discount rate"

[1] 0.03

[1] "effects discount rate"

[1] 0.02

Cost-effectiveness analysis summary

Reference intervention: Community Intervention
Comparator intervention(s): Facility Intervention
: FICI

Optimal decision: choose Community Intervention for k<235 and FICI for k>=235

Analysis for willingness to pay parameter k = 780

	Expected utility
Community Intervention	566019439361
Facility Intervention	563325570555
FICI	566623038119

	EIB	CEAC	ICER
Community Intervention vs Facility Intervention	2693868806	0.9990	-3.3497
Community Intervention vs FICI	-603598757	0.1613	230.1099

Optimal intervention (max expected utility) for k=780: FICI

EVPI 57375378

quartz

2

[1] "time horizon in years"

[1] 10

[1] "cost discount rate"

[1] 0

[1] "effects discount rate"

[1] 0.02

Cost-effectiveness analysis summary

Reference intervention: Community Intervention
Comparator intervention(s): Facility Intervention
: FICI

Optimal decision: choose Community Intervention for k<270 and FICI for k>=270

Analysis for willingness to pay parameter k = 780

	Expected utility
Community Intervention	565978683037
Facility Intervention	563282831099
FICI	566539542338

	EIB	CEAC	ICER
Community Intervention vs Facility Intervention	2695851938	0.9990	-3.9263
Community Intervention vs FICI	-560859301	0.1781	269.0464

Optimal intervention (max expected utility) for k=780: FICI

EVPI 64640722

quartz

2

[1] "time horizon in years"

[1] 10

[1] "cost discount rate"

[1] 0.1

[1] "effects discount rate"

[1] 0.02

Cost-effectiveness analysis summary

Reference intervention: Community Intervention

Comparator intervention(s): Facility Intervention
: FICI

Optimal decision: choose Community Intervention for $k < 170$ and FICI for $k \geq 170$

Analysis for willingness to pay parameter $k = 780$

	Expected utility
Community Intervention	566085591172
Facility Intervention	563394941198
FICI	566758560572

	EIB	CEAC	ICER
Community Intervention vs Facility Intervention	2690649974	0.99900	-2.4136
Community Intervention vs FICI	-672969400	0.13335	166.9120

Optimal intervention (max expected utility) for $k=780$: FICI

EVPI 47200900

quartz

2

[1] "time horizon in years"

[1] 10

[1] "cost discount rate"

[1] 0.03

[1] "effects discount rate"

[1] 0.03

Cost-effectiveness analysis summary

Reference intervention: Community Intervention

Comparator intervention(s): Facility Intervention
: FICI

Optimal decision: choose Community Intervention for $k < 245$ and FICI for $k \geq 245$

Analysis for willingness to pay parameter $k = 780$

	Expected utility
Community Intervention	537501350668
Facility Intervention	534942570721
FICI	538061830174

	EIB	CEAC	ICER
Community Intervention vs Facility Intervention	2558779947	0.99900	-3.5273
Community Intervention vs FICI	-560479506	0.16695	242.3134

Optimal intervention (max expected utility) for $k=780$: FICI

EVPI 56627976

quartz

2

[1] "time horizon in years"

[1] 10

[1] "cost discount rate"

[1] 0

[1] "effects discount rate"

[1] 0.03

Cost-effectiveness analysis summary

Reference intervention: Community Intervention

Comparator intervention(s): Facility Intervention
: FICI

Optimal decision: choose Community Intervention for $k < 285$ and FICI for $k \geq 285$

Analysis for willingness to pay parameter k = 780

	Expected utility
Community Intervention	537460594344
Facility Intervention	534899831264
FICI	537978334393

	EIB	CEAC	ICER
Community Intervention vs Facility Intervention	2560763079	0.9990	-4.1345
Community Intervention vs FICI	-517740049	0.1855	283.3147

Optimal intervention (max expected utility) for k=780: FICI

EVPI 64158023

quartz

2

[1] "time horizon in years"

[1] 10

[1] "cost discount rate"

[1] 0.1

[1] "effects discount rate"

[1] 0.03

Cost-effectiveness analysis summary

Reference intervention: Community Intervention

Comparator intervention(s): Facility Intervention
: FICI

Optimal decision: choose Community Intervention for k<180 and FICI for k>=180

Analysis for willingness to pay parameter k = 780

	Expected utility
Community Intervention	537567502478
Facility Intervention	535011941364
FICI	538197352627

	EIB	CEAC	ICER
Community Intervention vs Facility Intervention	2555561115	0.99900	-2.5416
Community Intervention vs FICI	-629850149	0.13725	175.7638

Optimal intervention (max expected utility) for k=780: FICI

EVPI 46091319

quartz

2

[1] "time horizon in years"

[1] 20

[1] "cost discount rate"

[1] 0.03

[1] "effects discount rate"

[1] 0

Cost-effectiveness analysis summary

Reference intervention: Community Intervention

Comparator intervention(s): Facility Intervention
: FICI

Optimal decision: choose Community Intervention for k<180 and FICI for k>=180

Analysis for willingness to pay parameter k = 780

	Expected utility
Community Intervention	1260379247717
Facility Intervention	1254386830995
FICI	1261848413252

	EIB	CEAC	ICER
Community Intervention vs Facility Intervention	5992416723	1.00	-2.6229
Community Intervention vs FICI	-1469165534	0.07	178.8683

Optimal intervention (max expected utility) for k=780: FICI

EVPI 39332414

quartz

2

[1] "time horizon in years"

[1] 20

[1] "cost discount rate"

[1] 0

[1] "effects discount rate"
[1] 0

Cost-effectiveness analysis summary

Reference intervention: Community Intervention
Comparator intervention(s): Facility Intervention
: FICI

Optimal decision: choose Community Intervention for $k < 240$ and FICI for $k \geq 240$

Analysis for willingness to pay parameter $k = 780$

	Expected utility
Community Intervention	1260237204107
Facility Intervention	1254237875788
FICI	1261557414435

	EIB	CEAC	ICER
Community Intervention vs Facility Intervention	5999328319	1.000	-3.5256
Community Intervention vs FICI	-1320210328	0.093	239.8157

Optimal intervention (max expected utility) for $k=780$: FICI

EVPI 51683524

quartz

2

[1] "time horizon in years"
[1] 20
[1] "cost discount rate"
[1] 0.1
[1] "effects discount rate"
[1] 0

Cost-effectiveness analysis summary

Reference intervention: Community Intervention
Comparator intervention(s): Facility Intervention
: FICI

Optimal decision: choose Community Intervention for $k < 105$ and FICI for $k \geq 105$

Analysis for willingness to pay parameter $k = 780$

	Expected utility
Community Intervention	1260555713989
Facility Intervention	1254571883810
FICI	1262209932339

	EIB	CEAC	ICER
Community Intervention vs Facility Intervention	5983830180	1.000	-1.5015
Community Intervention vs FICI	-1654218349	0.051	103.1511

Optimal intervention (max expected utility) for $k=780$: FICI

EVPI 28034329

quartz

2

[1] "time horizon in years"
[1] 20
[1] "cost discount rate"
[1] 0.03
[1] "effects discount rate"
[1] 0.02

Cost-effectiveness analysis summary

Reference intervention: Community Intervention
Comparator intervention(s): Facility Intervention
: FICI

Optimal decision: choose Community Intervention for $k < 220$ and FICI for $k \geq 220$

Analysis for willingness to pay parameter $k = 780$

	Expected utility
Community Intervention	1030374277350
Facility Intervention	1025471383468
FICI	1031495676163

EIB	CEAC	ICER
-----	------	------

Community Intervention vs Facility Intervention 4902893883 1.00000 -3.2082
Community Intervention vs FICI -1121398813 0.08695 218.7800

Optimal intervention (max expected utility) for k=780: FICI

EVPI 38677596

quartz

2

[1] "time horizon in years"

[1] 20

[1] "cost discount rate"

[1] 0

[1] "effects discount rate"

[1] 0.02

Cost-effectiveness analysis summary

Reference intervention: Community Intervention

Comparator intervention(s): Facility Intervention
: FICI

Optimal decision: choose Community Intervention for k<295 and FICI for k>=295

Analysis for willingness to pay parameter k = 780

	Expected utility
Community Intervention	1030232233740
Facility Intervention	1025322428261
FICI	1031204677347

	EIB	CEAC	ICER
Community Intervention vs Facility Intervention	4909805479	1.000	-4.3123
Community Intervention vs FICI	-972443607	0.124	293.3268

Optimal intervention (max expected utility) for k=780: FICI

EVPI 54215061

quartz

2

[1] "time horizon in years"

[1] 20

[1] "cost discount rate"

[1] 0.1

[1] "effects discount rate"

[1] 0.02

Cost-effectiveness analysis summary

Reference intervention: Community Intervention

Comparator intervention(s): Facility Intervention
: FICI

Optimal decision: choose Community Intervention for k<130 and FICI for k>=130

Analysis for willingness to pay parameter k = 780

	Expected utility
Community Intervention	1030550743622
Facility Intervention	1025656436283
FICI	1031857195250

	EIB	CEAC	ICER
Community Intervention vs Facility Intervention	4894307339	1.00000	-1.8365
Community Intervention vs FICI	-1306451628	0.05745	126.1677

Optimal intervention (max expected utility) for k=780: FICI

EVPI 25450120

quartz

2

[1] "time horizon in years"

[1] 20

[1] "cost discount rate"

[1] 0.03

[1] "effects discount rate"

[1] 0.03

Cost-effectiveness analysis summary

Reference intervention: Community Intervention

Comparator intervention(s): Facility Intervention
: FICI

Optimal decision: choose Community Intervention for $k < 245$ and FICI for $k \geq 245$

Analysis for willingness to pay parameter $k = 780$

	Expected utility			
Community Intervention	937456205820			
Facility Intervention	932993460518			
FICI	938437112830			
		EIB	CEAC	ICER
Community Intervention vs Facility Intervention	4462745302	1.00000	-3.526	
Community Intervention vs FICI	-980907010	0.09985	240.455	

Optimal intervention (max expected utility) for $k=780$: FICI

EVPI 39170730

quartz

2

[1] "time horizon in years"

[1] 20

[1] "cost discount rate"

[1] 0

[1] "effects discount rate"

[1] 0.03

Cost-effectiveness analysis summary

Reference intervention: Community Intervention

Comparator intervention(s): Facility Intervention

: FICI

Optimal decision: choose Community Intervention for $k < 325$ and FICI for $k \geq 325$

Analysis for willingness to pay parameter $k = 780$

	Expected utility			
Community Intervention	937314162210			
Facility Intervention	932844505312			
FICI	938146114013			
		EIB	CEAC	ICER
Community Intervention vs Facility Intervention	4469656898	1.00000	-4.7395	
Community Intervention vs FICI	-831951803	0.14235	322.3876	

Optimal intervention (max expected utility) for $k=780$: FICI

EVPI 57149945

quartz

2

[1] "time horizon in years"

[1] 20

[1] "cost discount rate"

[1] 0.1

[1] "effects discount rate"

[1] 0.03

Cost-effectiveness analysis summary

Reference intervention: Community Intervention

Comparator intervention(s): Facility Intervention

: FICI

Optimal decision: choose Community Intervention for $k < 140$ and FICI for $k \geq 140$

Analysis for willingness to pay parameter $k = 780$

	Expected utility			
Community Intervention	937632672092			
Facility Intervention	933178513333			
FICI	938798631917			
		EIB	CEAC	ICER
Community Intervention vs Facility Intervention	4454158759	1.00000	-2.0185	
Community Intervention vs FICI	-1165959825	0.0615	138.6675	

Optimal intervention (max expected utility) for $k=780$: FICI

EVPI 24582113