

```
#####
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```
#Accompanying material to the paper:
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```
#Garcia-Donato, Castellanos and Quirón (2021), Bayesian Variable Selection with  
Applications in Health Sciences
```

```
#published in Mathematics
```

```
#####
```

```
#This code contains algorithm that illustrates the use of the Gibbs sampling  
#algorithm to explore the model space. For illustrative purposes, the Bayes factor  
#is taken as defined by the unit information prior but you are supposed to  
#make use of your own Bayes factor  
#(The standard way of using BayesVarSel is much more  
#efficient than this code)
```

```
library(MASS)
```

```
library(BayesVarSel)
```

```
data(UCrime)
```

```
#47 16
```

```
y<- UCrime$y
```

```
X<- UCrime[,1:15]
```

```
p<- dim(X)[2]
```

```
#Factor Bayes x Prior
```

```
B.PM<- function(model){
```

```
  data<- data.frame(y=y, X=X[,model==1])
```

```
  tt<- Btest(models=list(M1=y~., M0=y~1), prior.betas="g", priorprob="c",
```

```
data=data)
```

```
  return(tt$BFi0[1]/choose(length(model), sum(model)))
```

```
}
```

```
###
```

```
#B.PM(sample(c(1,0), rep=T, size=16))
```

```
#null model
```

```
modelsB.PM<- rep(0, p+1) #the last column contains BF*Pr(M)
```

```
modelsB.PM[p+1]<- 1 #B.PM(rep(0,p)) Null vs Null
```

```
modelsB.PM[p+2]<- 0 #Label for each model
```

```
d.ini=8 #This is the dimension of the initial model, we have choose an  
intermediate dimension, it can be modified or random generated.
```

```
current.model<- sample(c(rep(1, d.ini), rep(0, p-d.ini)))
```

```
B.PMcurrent<- B.PM(current.model)
```

```
proposal.model<- current.model
```

```
pow2=2^((p-1):0)
```

```
#matrix of unique models and BFxP(M)
```

```
modelsB.PM<- rbind(modelsB.PM, c(current.model,
```

```
B.PMcurrent, ind=c(pow2%*%current.model)))
```

```
visitedmodels.PM<- modelsB.PM[,-(p+2)]
```

```
set.seed(12)
```

```
N<- 300
```

```

for (i in 1:N){
  #cat("It:",i,"\n")
  for (j in 1:p){
    proposal.model<- current.model; proposal.model[j]<- 1-current.model[j]
    #check if it is in your list
    ind.prop=c(pow2%*%proposal.model)
    coincident<- which(modelsB.PM[,p+2]==ind.prop)
    #cat("coincident", coincident, "\n")
    if (length(coincident)==0) {B.PMproposal<- B.PM(proposal.model);
modelsB.PM<- rbind(modelsB.PM, c(proposal.model, B.PMproposal,ind.prop))}
    else B.PMproposal<- modelsB.PM[coincident, p+1]

    #B.PMproposal<- B.PM(proposal.model)
    ratio<- B.PMproposal/(B.PMproposal+B.PMcurrent)
    #cat("ratio", ratio, "\n")

    if (runif(1)<ratio) {current.model[j]<- proposal.model[j]; B.PMcurrent<-
B.PMproposal}

  }

  visitedmodels.PM<- rbind(visitedmodels.PM, c(current.model, B.PMcurrent))
}

round(colMeans(visitedmodels.PM[,-(p+1)]), 2)

#0.59 0.13 0.82 0.87 0.24 0.12 0.29 0.15 0.09 0.15 0.33 0.18 0.98 0.55 0.11

#the real
crimeBvs<- Bvs(formula=y~., n.keep=10, prior.models="S", prior.betas="g",
data=UScrime)
round(crimeBvs$inclprob, 2)
  M   So   Ed  Po1  Po2   LF  M.F  Pop   NW   U1   U2  GDP  Ineq  Prob  Time
0.59 0.13 0.80 0.84 0.27 0.13 0.29 0.16 0.12 0.17 0.34 0.22 0.97 0.54 0.14

```