

# Example: Difference of biases

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## Summary

This is an example report showing an MCMC simulation for the difference in biases for Bernoulli trials.

## Code for Bayes Analysis

```
source("DBDA2E-utilities.R")
```

```
##
## *****
## Kruschke, J. K. (2015). Doing Bayesian Data Analysis, Second Edition:
## A Tutorial with R, JAGS, and Stan. Academic Press / Elsevier.
## *****
```

```
## Loading required package: coda
```

```
## Linked to JAGS 4.3.0
```

```
## Loaded modules: basemod,bugs
```

```
source("Jags-Ydich-XnomSsubj-MbernBeta.R")
```

```
##
## *****
## Kruschke, J. K. (2015). Doing Bayesian Data Analysis, Second Edition:
## A Tutorial with R, JAGS, and Stan. Academic Press / Elsevier.
## *****
```

```
### Read in the data
d <- read.csv("z6N8z2N7.csv")
y <- d$y
s <- as.numeric(d$s)
```

```
### Bundle data into a list
```

```
Ntotal <- length(y)
Nsubj  <- length(unique(s))

dList <- list( y = y,
               s = s,
               Ntotal = Ntotal,
               Nsubj  = Nsubj )
```

Specify model (check against TEMPmodel.txt)

```
model { for ( i in 1:Ntotal ) { y[i] ~ dbern( theta[s[i]] ) } for ( s in 1:Nsubj ) { theta[s] ~ dbeta(2,2) } }
```

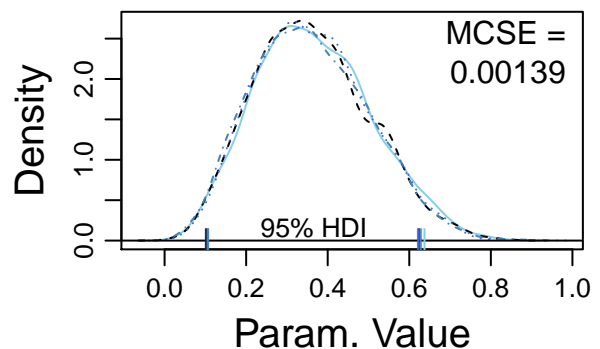
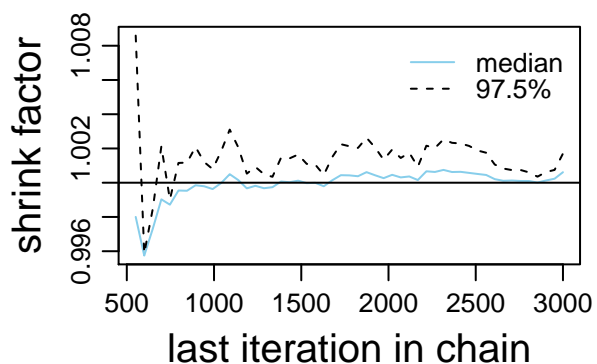
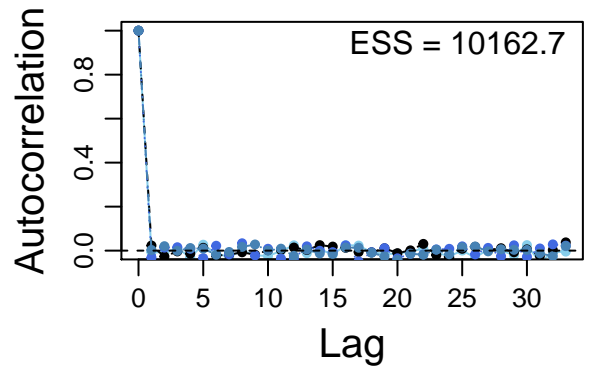
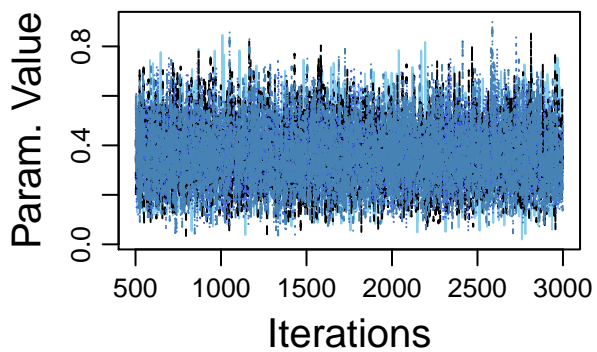
```
### High-level script for JAGS
```

```
mcmcCoda = genMCMC(data=d, numSavedSteps=10000)
```

```
## Compiling model graph
##   Resolving undeclared variables
##   Allocating nodes
## Graph information:
##   Observed stochastic nodes: 15
##   Unobserved stochastic nodes: 2
##   Total graph size: 35
##
## Initializing model
##
## Burning in the MCMC chain...
## Sampling final MCMC chain...
```

```
diagMCMC(mcmcCoda, parName="theta[2]")
smryMCMC(mcmcCoda, compVal=NULL, compValDiff=0.0)
```

## theta[2]



```
##                               Mean      Median      Mode      ESS HDImass
```

```
## theta[1]          0.6674226 0.6773664 0.7044958 9647.1    0.95
## theta[2]          0.3622349 0.3527293 0.3302016 10000.0   0.95
## theta[1]-theta[2] 0.3051877 0.3132239 0.3535127 10000.0   0.95
##                HDIlow HDIhigh CompVal PcntGtCompVal ROPElow
## theta[1]          0.41206513 0.902506      NA          NA      NA
## theta[2]          0.10196671 0.629327      NA          NA      NA
## theta[1]-theta[2] -0.08268338 0.660683      0          93.47     NA
##                ROPEhigh PcntLtROPE PcntInROPE PcntGtROPE
## theta[1]          NA          NA          NA          NA
## theta[2]          NA          NA          NA          NA
## theta[1]-theta[2] NA          NA          NA          NA
```

```
##                Mean      Median      Mode      ESS HDImass
## theta[1]          0.6674226 0.6773664 0.7044958 9647.1    0.95
## theta[2]          0.3622349 0.3527293 0.3302016 10000.0   0.95
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##                ROPEhigh PcntLtROPE PcntInROPE PcntGtROPE
## theta[1]          NA          NA          NA          NA
## theta[2]          NA          NA          NA          NA
## theta[1]-theta[2] NA          NA          NA          NA
```

```
plotMCMC(mcmcCoda, data=d, compVal=NULL, compValDiff=0.0, saveName="TEMP.jpg", saveType="jpg")
```

```
### Test of proportions
table(d)
```

```
##      s
## y    Reginald Tony
## 0      2      5
## 1      6      2
```

```
prop.test(x = c(6,2), n = c(8,7), correct=TRUE)
```

```
## Warning in prop.test(x = c(6, 2), n = c(8, 7), correct = TRUE): Chi-squared
## approximation may be incorrect
```

```
##
## 2-sample test for equality of proportions with continuity
## correction
##
## data:  c(6, 2) out of c(8, 7)
## X-squared = 1.637, df = 1, p-value = 0.2007
## alternative hypothesis: two.sided
## 95 percent confidence interval:
## -0.1191204 1.0000000
## sample estimates:
##      prop 1      prop 2
## 0.7500000 0.2857143
```

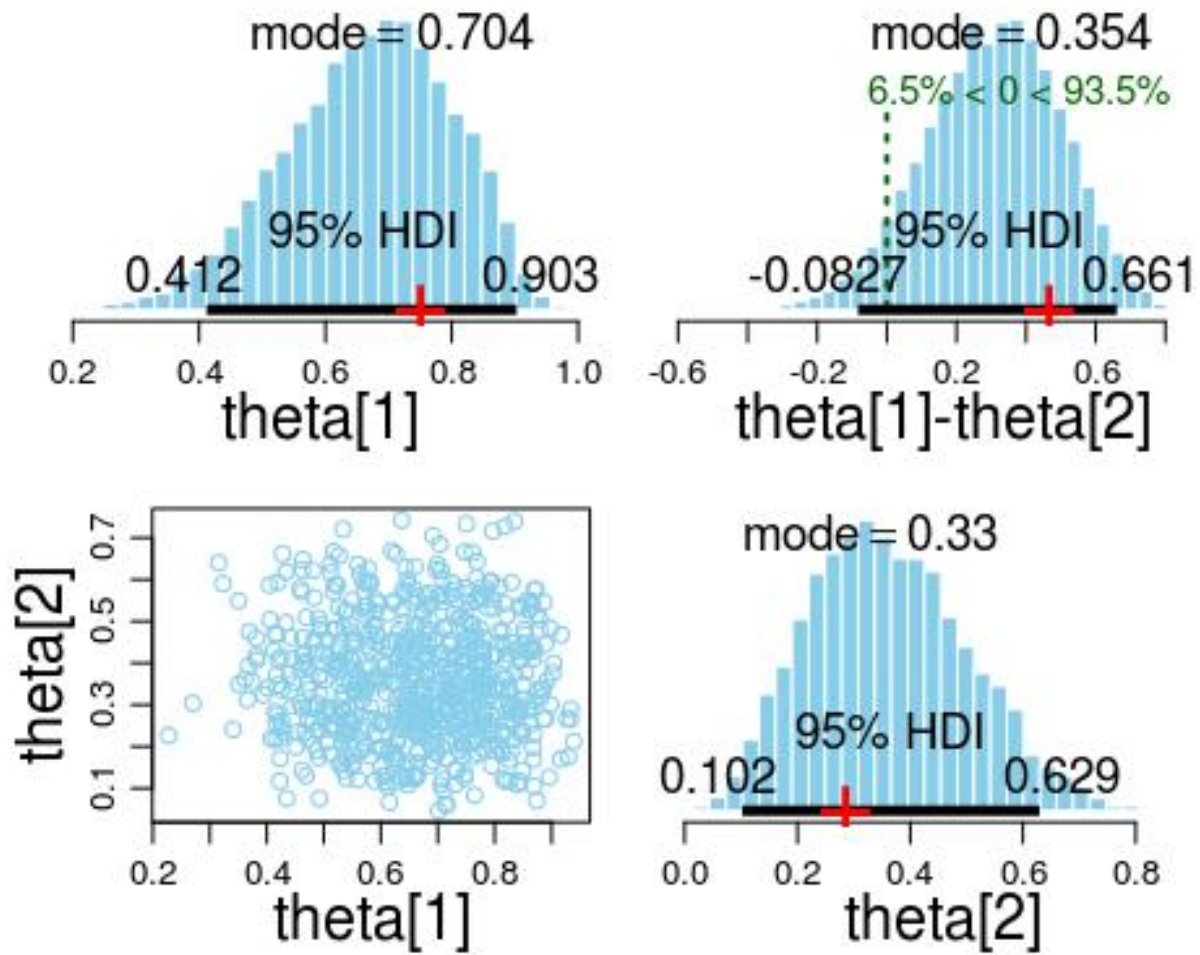


Figure 1: Posterior plot