

Assignment 1

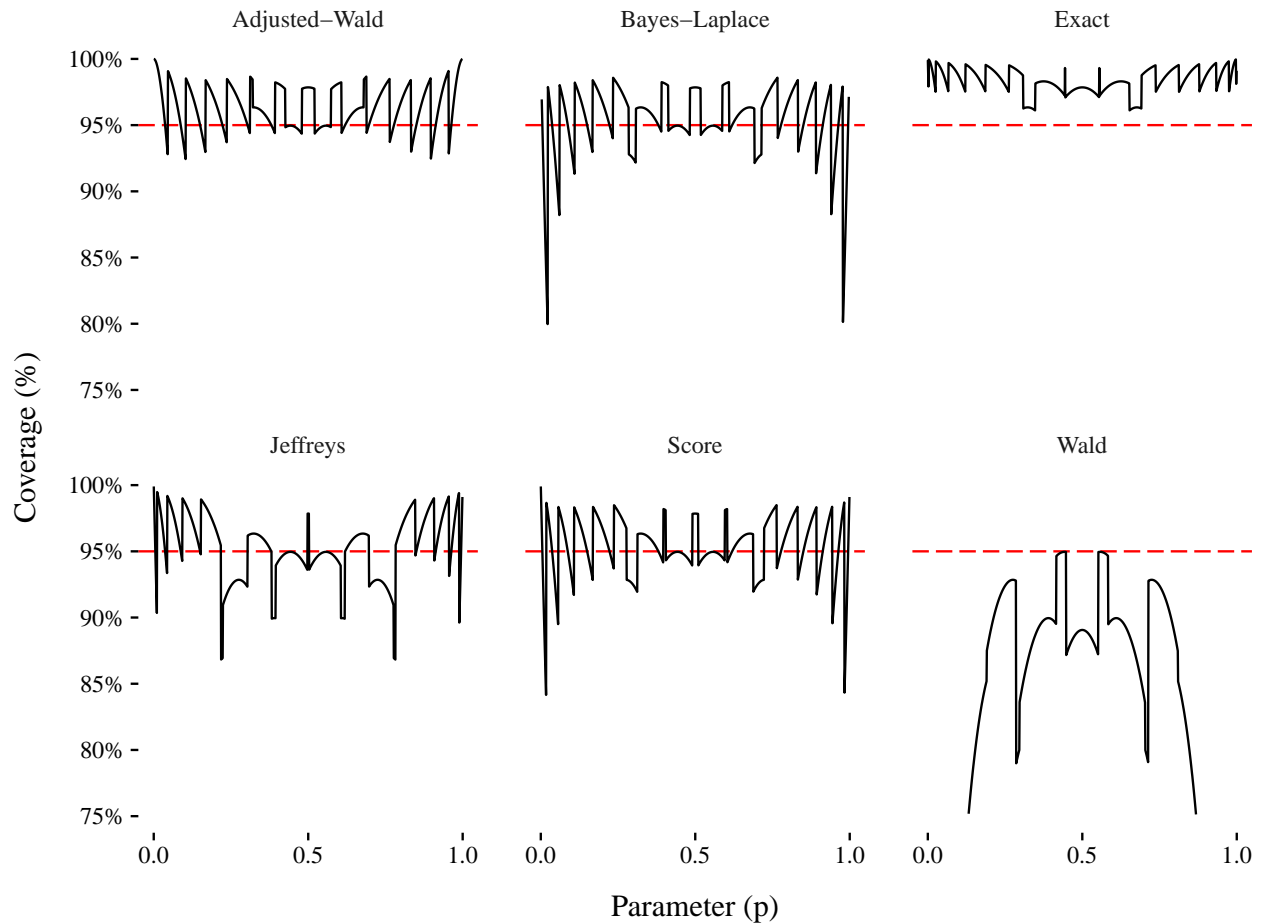
2017-03-29

Question 1: Inference for the binomial parameter:

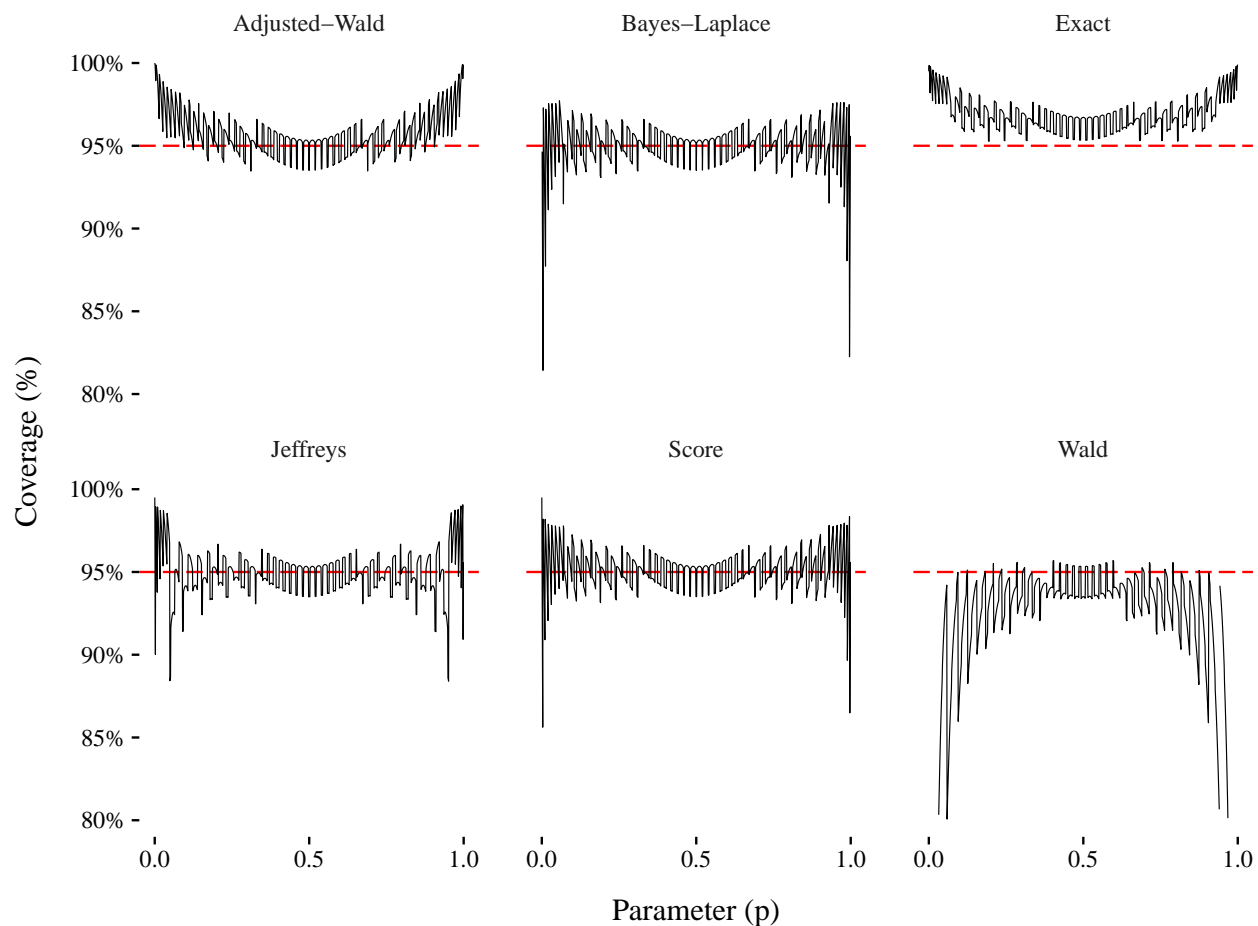
- (a) Develop an R function to calculate HPD intervals for data (x, n) , given a $\text{beta}(a, b)$ prior.

Running the above code gives the following output

- (b) Reproduce Agresti & Coull's (1998) Figure 4 ($n = 10$), and replicate for the Score and Bayes-Laplace & Jeffreys HPD intervals



- (c) Repeat (b) for $n = 50$.



- (d) Compare the minimum coverage of the six graphs at (c)
- (e) The adjusted Wald interval appears to perform well with respect to frequentist coverage, if close to nominal combined with reasonable minimum coverage is aimed for. From a Bayesian point of view, performance of individual intervals is just as, if not more, important. Given $x = 0$, compare the adjusted Wald interval with the exact & Score intervals (all two-sided), and with the Bayes-Laplace & Jeffreys HPD intervals, for a range of values of n and α , comment on its limitations, and give an appropriate graphical illustration.

Question 2: Inference for the Cauchy parameter:

- (a) Develop an R function to find percentiles of a (general) Cauchy posterior as discussed by Jaynes (1976, Example 6) and Box & Tiao (1973, p.64), to be used for the examples below.

```
source("Question2/2a.R")
CauchyPercentage <- function(x, # a vector of samples
                             p, # a vector of possible parameters
                             y = NULL # a value to test  $Pr[p < y]$ ) {
  cauchydens <- function(x,p){

    cauchydist <- function(x,p) {
      H = vector(mode = "numeric",length = length(x))
      for(i in 1:length(x)){
        H[i] = (1+(x[i] - p)^2)^(-1)
      }
      return(prod(H))
    }
    dens = vapply(p,cauchydist,min(p), x = x)
    c = (integrate(dens, -Inf, Inf)$value)^(-1)
    data.frame(vparams = p, postdens = c*dens(p))
  }

  df = cauchydens(x,p)

  yind = ifelse(length(which(df$vparams == y))==1,
                which(df$vparams == y),
                max(which(df$vparams < y)))

  plot(df$vparams, df$postdens, type = 'l')
  abline(v = df$vparams[yind])
  abline(h = df[yind,"postdens"])

  cumdist = c*integrate(dens,-Inf,y)$value

  return((c(yind,df[yind,"postdens"],cumdist)))
}

CauchyHPD <- function(x, # vector of samples
                      p, # vector of possible parameter values
                      alpha = 0.95, # HPD interval value
                      tol = 0.0001) { # level of tolerance for exact HPD interval

  cauchydens <- function(x,p){

    cauchydist <- function(x,p) {
      H = vector(mode = "numeric",length = length(x))
      for(i in 1:length(x)){
        H[i] = (1+(x[i] - p)^2)^(-1)
      }
      return(prod(H))
    }
    dens = function(p) vapply(p,cauchydist,min(p), x = x)
```

```

c = (integrate(dens, -Inf, Inf)$value)^(-1)
data.frame(vparams = p, postdens = c*dens(p))
}

df = cauchydens(x,p)

cumdist = cumsum(df$postdens)*diff(df$vparams)[1]
post_median = which.min(abs(cumdist-0.5))

HPDlimits <- function(post_dens) { ## find lower and upper values for which
  ## prob dens is closest to target value
  lower = which.min(abs(df$postdens[1:post_median]-post_dens))
  upper = which.min(abs(df$postdens[(post_median+1):length(df$postdens)]-post_dens))+post_median
  limits = c(lower,upper)
}

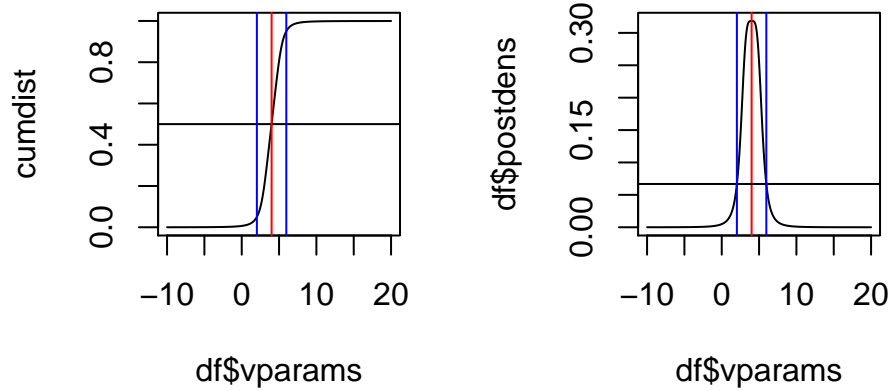
HPDlimitarea <- function(post_dens) {
  limitints = HPDlimits(post_dens)
  limitarea = sum(df$postdens[limitints[1]:limitints[2]])*diff(df$vparams)[1]
}

## find credible interval
v2 = seq(0,max(df$postdens),by=tol)
vals = sapply(v2,HPDlimitarea)
w = which.min(abs(vals-alpha))
r = c(df$vparams[HPDlimits(v2[w])])
names(r) = c("lower","upper")
par(mfrow = c(1,2))
plot(df$vparams, cumdist, type = 'l')
abline(h = 0.5)
abline(v = df$vparams[post_median], col = 'red')
abline(v = r["upper"], col = 'blue')
abline(v = r["lower"], col = 'blue')
plot(df$vparams, df$postdens, type = 'l')
abline(v = df$vparams[post_median], col = 'red')
abline(h = df[HPDlimits(v2[w])][1], "postdens")
abline(v = r["upper"], col = 'blue')
abline(v = r["lower"], col = 'blue')
return(r)
}

```

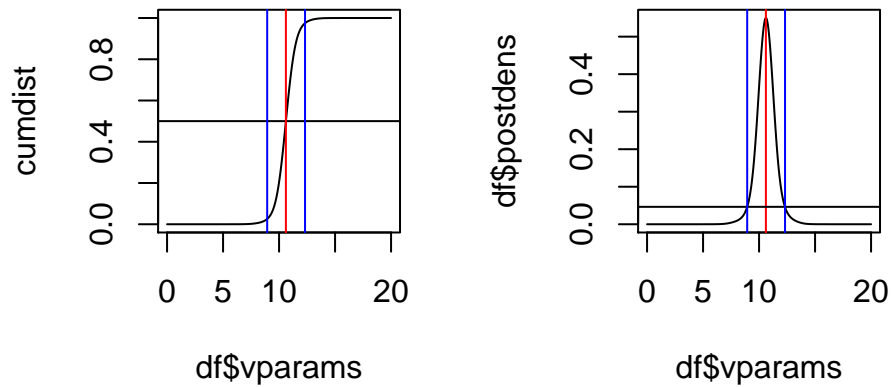
- (b) Consider Jaynes' example of $n = 2$ observations (3, 5): plot the posterior and calculate the 90% central credible interval. Explain why it is quite different from the confidence interval derived by Jaynes (p.202).

```
source("Question2/2b.R")
```



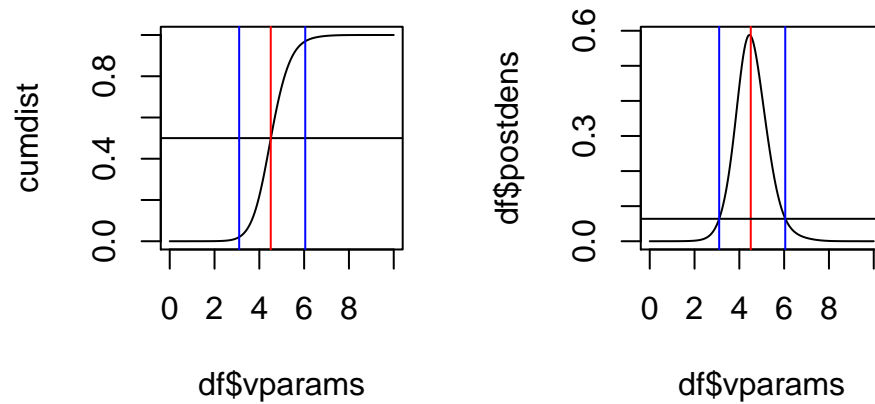
(c) Consider Box & Tiao's example of $n = 5$ observations (11.4, 7.3, 9.8, 13.7, 10.6): plot the posterior and calculate 95% central and HPD credible intervals and check $Pr[\theta < 11.5]$ given by Box & Tiao.

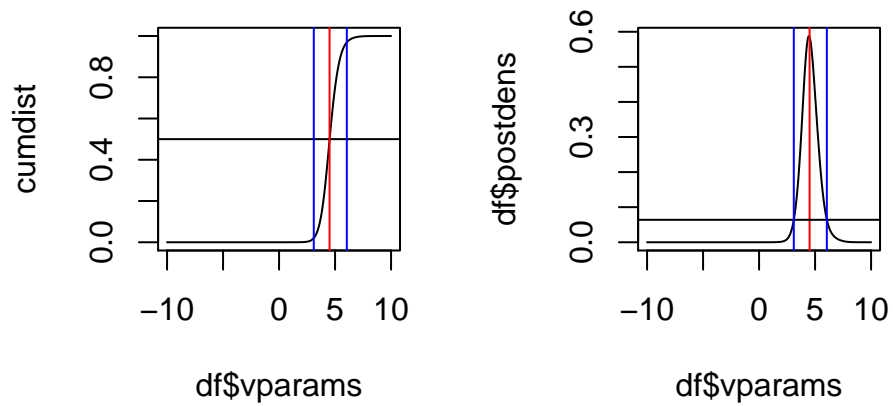
```
source("Question2/2c.R")
```



(d) Consider Berger's (1985, p.141) example of $n = 5$ observations (4.0, 5.5, 7.5, 4.5, 3.0): calculate 95% central and HPD credible intervals, with and without Berger's restriction ($\theta > 0$).

```
source("Question2/2d.R")
```





- (e) Clearly, Berger's restriction ($\theta > 0$) will sometimes lead to a posterior quite different from the unrestricted posterior. Plot this restricted posterior for the hypothetical negative version of Berger's example: i.e. $(-4.0, -5.5, -7.5, -4.5, -3.0)$, and calculate the 95% HPD interval.

```
source("Question2/2e.R")
```

