lab1_code

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#1 a.
s=22
n=70
f=70-22
#probability of success
prob=s/n
alpha0<-beta0<-8
#posterior mean and standard deviation
posterior_stats<-function(samples){</pre>
  posterior_mean<-c()</pre>
  posterior sd<-c()</pre>
  for(i in 1:samples){
    #drawing samples
    poseterior_draws=rbeta(i,alpha0+s,beta0+f)
    #calculating accumulating samples mean and sd
    posterior_mean<-c(posterior_mean,mean(poseterior_draws))</pre>
    posterior_sd<-c(posterior_sd,sd(poseterior_draws))</pre>
  return(list(posterior_mean,posterior_sd))
}
#means and sds of accumulating samples
posterior_means<-posterior_stats(10000)[[1]]</pre>
posterior_standardev<-posterior_stats(10000)[[2]]</pre>
# plot means and comparing with true value
plot(posterior_means, type = 'l', main = "the posterior mean vs the true mean",
     xlab = "the number of samples" )
true mean value=(alpha0+s)/((alpha0+s)+(beta0+f))
abline(h=true_mean_value,col='red')
legend("topright",legend="true mean value",col="red", lty=1, cex=0.8)
# plot sd and comparing with true value
plot(posterior_standardev,type = 'l',main = "the posterior sd vs the true sd" ,
     xlab = "the number of samples")
\label{true_sd_value} $$ true_sd_value=((alpha0+s)*(beta0+f))/((((alpha0+s)+(beta0+f))^2)*((alpha0+s)+(beta0+f)+1)) $$
abline(h=sqrt(true_sd_value),col='red')
legend("topright",legend="true sd value",col="red", lty=1, cex=0.8)
#1. b
poseterior_draws=rbeta(10000,alpha0+s,beta0+f)
pr=length(poseterior_draws[poseterior_draws>0.3])/10000
exact_pr=pbeta(0.3,alpha0+s,beta0+f,lower.tail = FALSE)
cat("The posterior probability is",pr,"\n")
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cat("The exact value from the Beta posterior is",exact_pr)
#both probabilities are almost same
#1 c
poseterior_draws=rbeta(10000,alpha0+s,beta0+f)
phi<-poseterior_draws/(1-poseterior_draws)</pre>
#histogram
hist(phi,breaks=50)
density_phi <- density(phi)</pre>
#with density curve
p<- ggplot()+
  geom_histogram(aes(x = phi,y=..density..),bins=50,color='blue',fill='white')+
  geom_line(aes(x = density_phi$x, y = density_phi$y), color = 'red')+
  ggtitle('The density and distribution of posterior of the odds')
print(p)
#2 a
y < -c(33, 24, 48, 32, 55, 74, 23, 17)
n<-length(y)
#drawing from chi square distribution
draw_chi \leftarrow rchisq(n = 10000, df = n)
mu < -3.6
#calculating tau by formula
tau < -sum((log(y)-mu)^2)/n
sigma 2<-n*tau/draw chi
hist(sigma_2, breaks=180, xlim = c(0,1), main = "the posterior distribution")
#2 b
#qinnicoefficient
ginni_coef<-2*pnorm(sqrt(sigma_2/2),mean = 0,sd=1)-1</pre>
hist(ginni_coef, breaks=50, main="the posterior"
distribution of the Gini coefficient G")
#2 c
#calculating credible interval
intervals <- quantile (ginni_coef, probs=c(0.025, 0.975))
hist(ginni_coef,breaks=50)
abline(v=intervals[1])
abline(v=intervals[2])
legend("topright",legend="95% credible interval",pch="|", cex=0.4)
cat("The 95% equal tail credible interval for G is [",intervals[1],",",intervals[2],"]")
#2 d
#getting densities
posterior<-density(ginni_coef)</pre>
#sorting and ordering posterior x and y values
posterior_sorted <- sort(posterior$y,decreasing = TRUE)</pre>
x_sorted<-posterior$x[order(posterior$y,decreasing = TRUE)]</pre>
#function calculates area under curve and get the hpdi_intervals
hpdi_interval<-function(x,density){</pre>
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for(i in 1:length(x)){
    if((i+1)>length(x)){
      break
    }
    #approx function and integrating to get area under curve by giving intervals
    fn <- approxfun(x, density)</pre>
    area_curve <- integrate(fn, x[i], x[i+1])</pre>
    #breaking out of loop when area is greater than or equal to 0.95
    if(area curve$value>=0.95){
      break
    }
  }
  return(list(x[i],x[i+1]))
hpdi_lower<-hpdi_interval(x_sorted,posterior_sorted)[[1]]
hpdi_upper<-hpdi_interval(x_sorted,posterior_sorted)[[2]]
\textit{\# Plot posterior density with 95\% HPD interval}
plot(posterior, main = "Posterior Density with 95% HPD Interval")
#ablines for hdpi and credible intervals
abline(v = hpdi_lower, lty = 2,col='red')
abline(v = hpdi_upper, lty = 2,col='red')
abline(v=intervals[1])
abline(v=intervals[2])
legend("topright",legend=c("95% credible interval","95% HPDI interval"),col=c('black','red'),lty = c(1,
#hdpi interval is moved little left where density is more, which makes sense
#3 a
y < -c(-2.79, 2.33, 1.83, -2.44, 2.23, 2.33, 2.07, 2.02, 2.14, 2.54)
mu < -2.4
k < -seq(0.01, 10, by=0.01)
*posterior distribution function values
function_posterior<-function(k){</pre>
  y<-c(-2.79, 2.33, 1.83, -2.44, 2.23, 2.33, 2.07, 2.02, 2.14, 2.54)
  mu < -2.4
 p_r \sim \exp(k*sum(\cos(y-mu))-0.5*k)/bessell(x=k,nu=0)^length(y)
 return(p r)
}
#posterior density values
posterior_result<-function_posterior(k)</pre>
#normalizing the posterior distribution by integrating
integration<-integrate(function_posterior,lower=0.01,upper=10)</pre>
normalize_posterior<-posterior_result/integration$value
#area under curve integrates approximate to 1
fn <- approxfun(k, normalize_posterior)</pre>
area_curve <- integrate(fn, min(k), max(k))</pre>
#plotting density curve
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plot(k,normalize_posterior,type='l',main = "The normalized posterior distribution of k")

#3 b
#posterior mode of k means,k value for which
#posterior density value is maximum(i.e in normalize_posterior)
print(k[which.max(normalize_posterior)])
```