

BST 5230 Bayesian Statistics - Spring 2018

Mid-term Exam

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1.(10 Points 5,5) A disease has base rate 3%. The sensitivity of a test is 0.80 and the specificity is 0.995.

- (a) Find the probability that a person selected at random tests positive for the disease.
- (b) Find the positive predictive value.

Answers:

(a)

$$\begin{aligned}P(T = +) &= P(T = +, D = +) + P(T = +, D = -) \\&= P(T = +|D = +) * P(D = +) + P(T = +|D = -) * P(D = -) \\&= P(T = +|D = +) * P(D = +) + [1 - P(T = -|D = -)] * [1 - P(D = -)] \\&= 0.8 * 0.03 + (1 - 0.995)(1 - 0.03) \\&= 0.02885\end{aligned}$$

(b)

$$\begin{aligned}PPV &= P(D = +|T = +) \\&= \frac{P(D = +)P(T = +|D = +)}{P(T = +)} \\&= \frac{0.03 * 0.80}{0.02885} \\&= 0.8318891\end{aligned}$$

2. (20 Points) Suppose that $X_1, X_2, \dots, X_n \sim \text{i.i.d GEOMETRIC}(\theta)$. That is, they are geometric random variables. (That is, each is the number of trials needed to get one success where the probability of success is θ .) The probability mass function for the geometric distribution is

$$f(x|\theta) = \theta(1 - \theta)^{x-1}, \quad x = 1, 2, 3, \dots$$

Suppose that we assume a $\text{BETA}(\alpha, \beta)$ prior distribution for θ .

- (a) Find the posterior distribution for θ .
- (b) Is the $\text{BETA}(\alpha, \beta)$ prior distribution a conjugate prior? Explain.
- (c) Suppose that $n = 5$ and we obtain $x_1 = 1, x_2 = 5, x_3 = 4, x_4 = 2, x_5 = 8$. Assume a $\text{BETA}(2, 3)$ prior distribution. Find the posterior distribution and plot it.
- (d) Find the posterior mean and standard deviation.

Answers:

(a)

$$\begin{aligned} P(\theta|X) &= \frac{c_1 \theta^{\alpha-1} (1-\theta)^{\beta-1} * \theta(1-\theta)^{x-1}}{c_2} \\ &= c_3 \theta^{\alpha} (1-\theta)^{x+\beta-1} \quad c_1, c_2, c_3 \text{ are supposed to be unknown constants} \\ &\sim \text{BETA}(\alpha + 1, \beta + x - 1) \end{aligned}$$

(b) Yes, $\text{BETA}(\alpha, \beta)$ is a conjugate prior for geometric data because if we have prior $\text{BETA}(\alpha, \beta)$ and geometric data, we will get posterior $\text{BETA}(\alpha + 1, \beta + X - 1)$, which is in the same family of prior distribution.

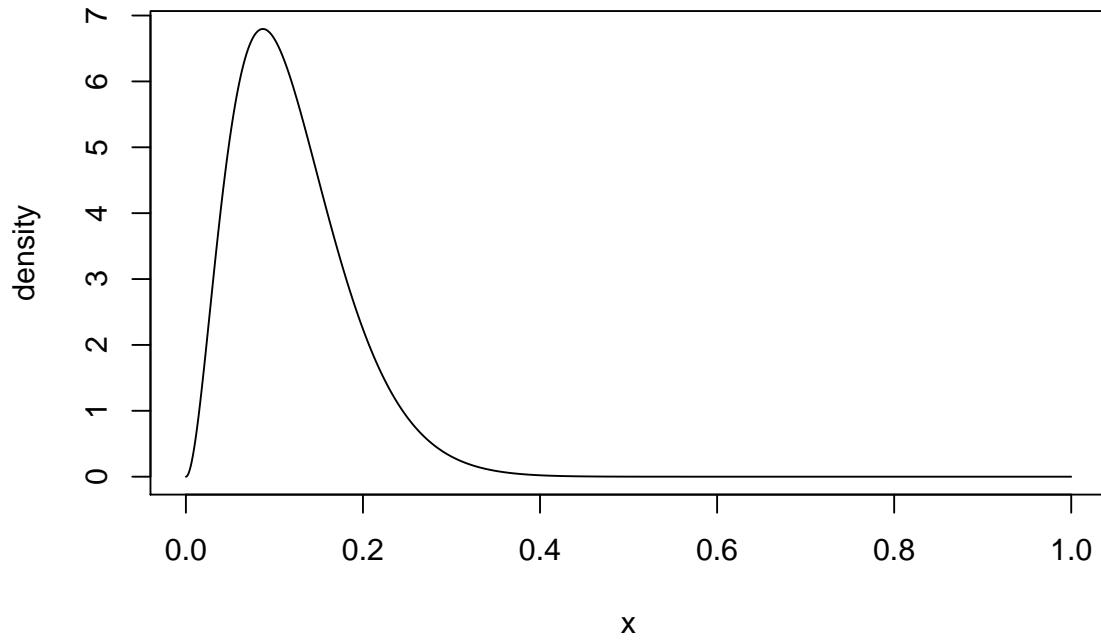
(c) According to the result from part (a), we have:

$$\begin{aligned} P(\theta|X) &\sim \text{BETA}(\alpha + 1, \beta + \sum_{i=1}^5 x_i - 1) \\ &\sim \text{BETA}(3, 22) \end{aligned}$$

The plot of the posterior distribution is as follows:

```
x = seq(0, 1, 0.0001)
d = dbeta(x, 3, 22)
plot(x, d, type = "l", main = "2.(c) Posterior ~ BETA(3, 22)", ylab = "density")
```

2.(c) Posterior ~ BETA(3, 22)



(d) According to the properties of BETA distribution:

$$Mean = \frac{\alpha}{\alpha + \beta} = \frac{3}{3 + 22} = \frac{3}{25}$$

$$s.d. = \sqrt{Var} = \sqrt{\frac{\alpha\beta}{(\alpha + \beta)^2(\alpha + \beta + 1)}} = \sqrt{\frac{3 * 22}{(3 + 22)^2(3 + 22 + 1)}} = 0.0637302$$

3. (6 Points) Choose and plot a prior that describes your belief in the following circumstances. Explain your reasoning in both cases.

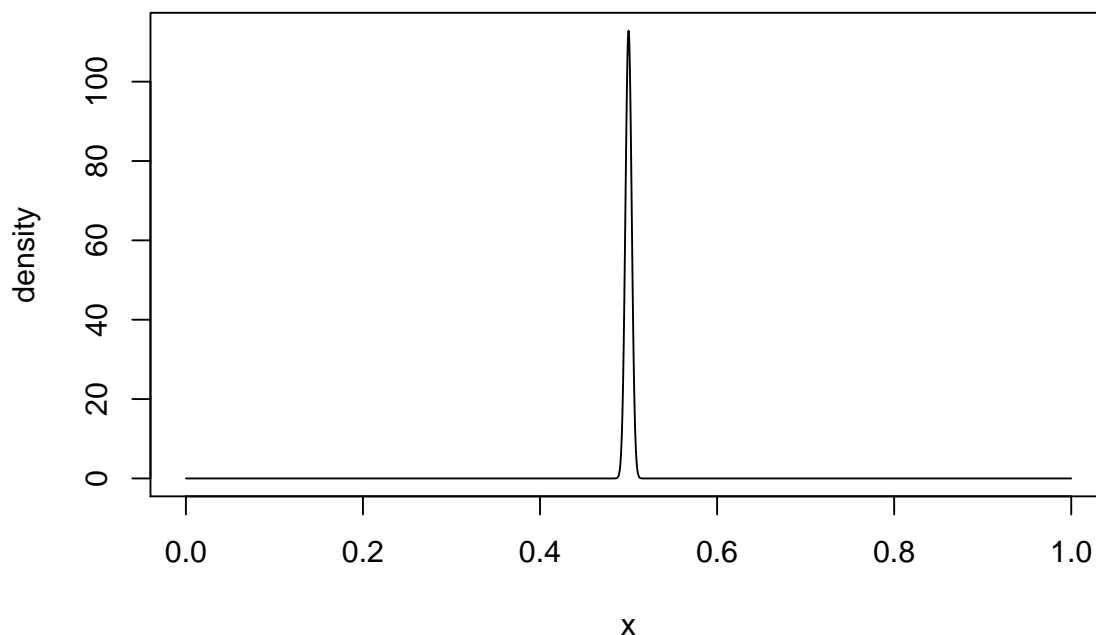
- (a) Suppose you have a coin that you know is minted by the US government and has not been tampered with. Therefore you have a strong prior belief that the coin is fair.
- (b) Now you have a different coin, this one made of some strange material and marked “Patent Pending: International Magic Company.”

Answers:

- (a) Since we have a strong belief that the coin should be fair, the mean should be 0.5, the variance should be very small and the support should be within (0, 1). So I choose the prior $\text{BETA}(10000, 10000)$. Here is the plot of the prior distribution I choose:

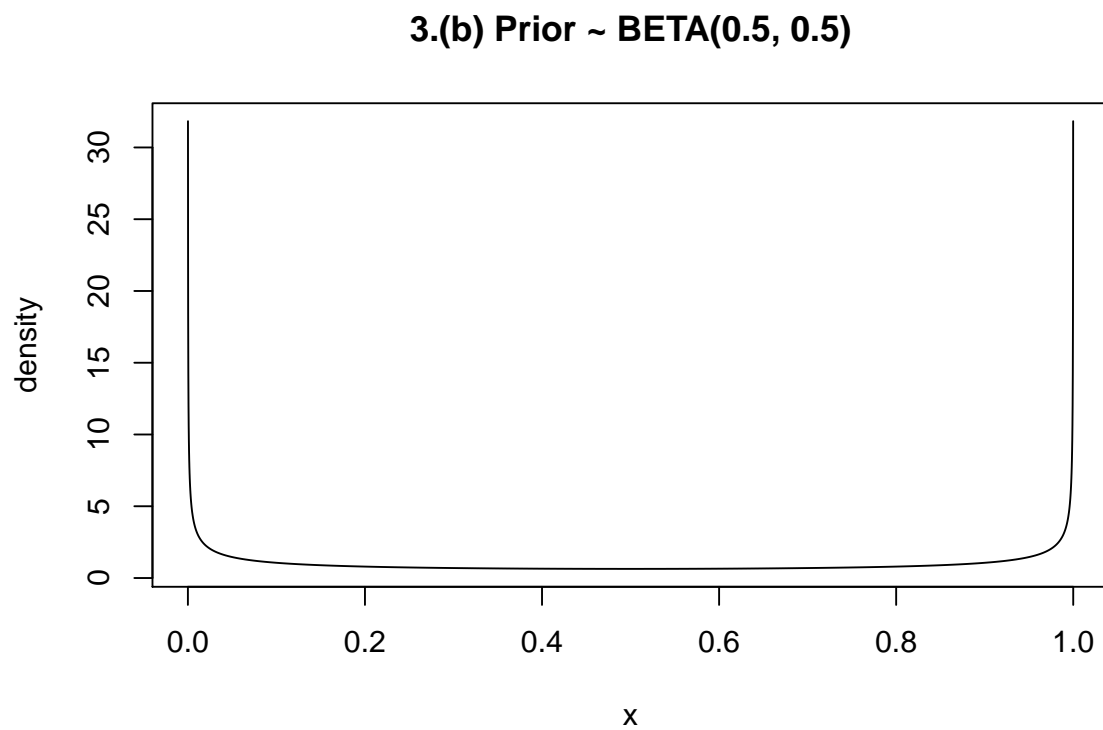
```
x = seq(0, 1, 0.0001)
d = dbeta(x, 10000, 10000)
plot(x, d, type = "l", main = "3.(a) Prior ~ BETA(10000, 10000)", ylab = "density")
```

3.(a) Prior ~ BETA(10000, 10000)



- (b) Since we have a strong belief that the coin should be either 0 or 1, the density should be high at either 0 or 1, and the density should be very small in between 0 and 1 and the support should be within (0, 1). So I choose the prior $\text{BETA}(0.5, 0.5)$. Here is the plot of the prior distribution I choose:

```
x = seq(0, 1, 0.0001)
d = dbeta(x, 0.5, 0.5)
plot(x, d, type = "l", main = "3.(b) Prior ~ BETA(0.5, 0.5)", ylab = "density")
```



4. (16 Points) Suppose that $X|\theta \sim \text{BIN}(4, \theta)$, that is binomial with 4 trials and probability of success. Assume a $\text{BETA}(2,3)$ prior distribution for θ . We observed $x = 1$ success out of the 4 trials.

- Suppose we apply the Metropolis-Hastings algorithm beginning with $\theta^{(1)} = 0.6$. Suppose we use as the proposal density normal distribution with mean $\theta^{(t)}$ and variance $\sigma^2 = 0.04$. Find the acceptance probability if the first proposal is $\theta^{(\text{prop})} = 0.56$.
- Suppose we apply the Metropolis-Hastings algorithm beginning with $\theta^{(1)} = 0.6$. Suppose we use as the proposal density normal distribution with mean $\theta^{(t)}$ and variance $\sigma^2 = 0.04$. Find the acceptance probability if the first proposal is $\theta^{(\text{prop})} = 0.62$.
- Suppose we apply the Metropolis-Hastings algorithm beginning with $\theta^{(1)} = 0.6$. Suppose we use as the proposal density a $\text{BETA}(2,3)$ prior distribution. Find the acceptance probability if the first proposal is $\theta^{(\text{prop})} = 0.56$.
- Among (a) & (c) which is (are) the simpler Metropolis and which is (are) the Metropolis-Hastings algorithm.

Answers:

- The R codes for calculating the acceptance probability and the result are as follows:

```
# Essential part of BETA distribution
pBeta = function(p,a,b)
{
  z = 0
  if (p >= 0 && p <= 1) { z = p^(a-1) * (1-p)^(b-1) }
  return(z)
}

# Binomial likelihood function
pLik = function(x,n,p) { p^x * (1-p)^(n-x) }

# Proposal distribution
pProp = function(p)
{
  rnorm(n=1,mean=p,sd=sqrt(0.04))
}

n = 4 # the number of trials
x = 1 # the number of success
a = 2 # alpha for beta distribution
b = 3 # beta for beta distribution
```

```

pp = 0.56 # current value
pc = 0.6  # proposal value

num = pBeta(pp,a,b)*pLik(x,n,pp) # numerator
den = pBeta(pc,a,b)*pLik(x,n,pc) # denominator

(alpha = num/den) # acceptance probability - alpha

## [1] 1.402933

```

(b) The R codes for calculating the acceptance probability and the result are as follows:

```

pp = 0.62 # current value
pc = 0.6  # proposal value

num = pBeta(pp,a,b)*pLik(x,n,pp) # numerator
den = pBeta(pc,a,b)*pLik(x,n,pc) # denominator

(alpha = num/den) # acceptance probability - alpha

## [1] 0.8262261

```

(c) The R codes for calculating the acceptance probability and the result are as follows:

```

# Proposal distribution
pProp = function(p)
{
  rbeta(n = 1, shape1 = 2, shape2 = 3)
}

pp = 0.56 # current value
pc = 0.6  # proposal value

num = pBeta(pp,a,b)*pLik(x,n,pp)*pc*(1-pc)^2 # numerator
den = pBeta(pc,a,b)*pLik(x,n,pc)*pp*(1-pc)^2 # denominator

(alpha = num/den) # acceptance probability - alpha

## [1] 1.242267

```

- (d) In this question, (a) and (b) are the simpler Metropolis algorithm because the proposal density are normal distributions which are symmetric, while (c) is the Metropolis-Hastings algorithm because the proposal density is beta which is not symmetric.