

Midterm [20 pts]

March 18th, 2024 in class

PUBH 8442: Bayes Decision Theory and Data Analysis

Give your final answers in simplified, closed form wherever possible. However, partial credit will be awarded for incomplete solutions. Good luck!

1. *In-vitro fertilization* [11 pts]

Let θ be the unknown success rate of a novel *in vitro* fertilization method. Let X be the number of failed attempts before the first success. Then X has a geometric distribution with probability θ :

$$P(X = k | \theta) = (1 - \theta)^k \theta$$

for $k = 0, 1, 2, 3, \dots$

(a) (3 points) Show that the Jeffreys prior for θ is

$$p(\theta) \propto \theta^{-1}(1 - \theta)^{-\frac{1}{2}}$$

for $\theta \in (0, 1)$.

The log-likelihood is

$$L = \log P(X | \theta) = X \log(1 - \theta) + \log \theta.$$

Differentiating twice gives

$$\frac{d^2 L}{d\theta^2} = -\frac{1}{\theta^2} - \frac{X}{(1 - \theta)^2}.$$

The Fisher information $I(\theta)$ is the negative expectation of the log-likelihood. Using $E(X | \theta) = (1 - \theta)\theta^{-1}$,

$$I(\theta) = \frac{1}{\theta^2} + \frac{1 - \theta}{\theta(1 - \theta)^2} = \theta^{-2}(1 - \theta)^{-1}.$$

The Jeffreys prior is the square root of $I(\theta)$:

$$p(\theta) \propto \sqrt{I(\theta)} = \theta^{-1}(1 - \theta)^{-\frac{1}{2}}.$$

For parts (b-e), use the Jeffreys prior given in part (a).

(b) (1 points) Is the prior proper?

No, the prior is improper because its integral is not finite on $[0, 1]$.

(c) (2 points) For which values of X , if any, will the posterior be proper?

The posterior is

$$p(\theta | X) \propto (1 - \theta)^{X-1/2}.$$

Thus the posterior is a proper $\text{Beta}\{1, X + 1/2\}$ distribution for all possible X .

(d) (2 points) What is the Bayes estimator for θ , given X , under squared error loss?

The Bayes estimator under squared error loss is the conditional expectation. Using $\theta | X \sim \text{Beta}\{1, X + 1/2\}$, the Bayes estimator is

$$E(\theta | X) = \frac{1}{1 + X + 1/2} = \frac{2}{2X + 3}.$$

(e) (3 points) Let X_2 and X_3 be distributed identically to X , and assume X, X_2 , and X_3 are independent given θ . If $X = 1$, what is $P(X_2 + X_3 = 0 | X = 1)$, the probability that there are no failed attempts in the next two trials?

The posterior after observing $X = 1$ is $p(\theta | X = 1) = \text{Beta}(1, 3/2)$, and after observing $X = 1, X_2 = 0$ is $\text{Beta}(2, 3/2)$ distributed. Thus,

$$\begin{aligned} P(X_2 + X_3 = 0 | X = 1) &= P(X_2 = 0 | X = 1)P(X_3 = 0 | X_2 = 0, X = 1) \\ &= \frac{1}{B(1, 3/2)} \int_0^1 \theta \cdot (1 - \theta)^{1/2} d\theta \times \frac{1}{B(2, 3/2)} \int_0^1 \theta \cdot \theta(1 - \theta)^{1/2} d\theta \\ &= \frac{B(3, 3/2)}{B(1, 3/2)} \\ &= 8/35, \end{aligned}$$

where $B(\cdot, \cdot)$ denotes the Beta function.

2. Bayesian P-Value [3 pts]

Consider $y_1, \dots, y_n \stackrel{iid}{\sim} N(\theta, \sigma^2)$, with σ^2 known and a flat prior on θ : $p(\theta) = 1 \forall \theta$. Let $\mathbf{y} = (y_1, \dots, y_n)$, and let $T(\cdot)$ give the sample mean: $T(\mathbf{y}) = \bar{y}$. What is the posterior predictive p-value,

$$P(T(\mathbf{Y}) \geq T(\mathbf{y}) | \mathbf{y})?$$

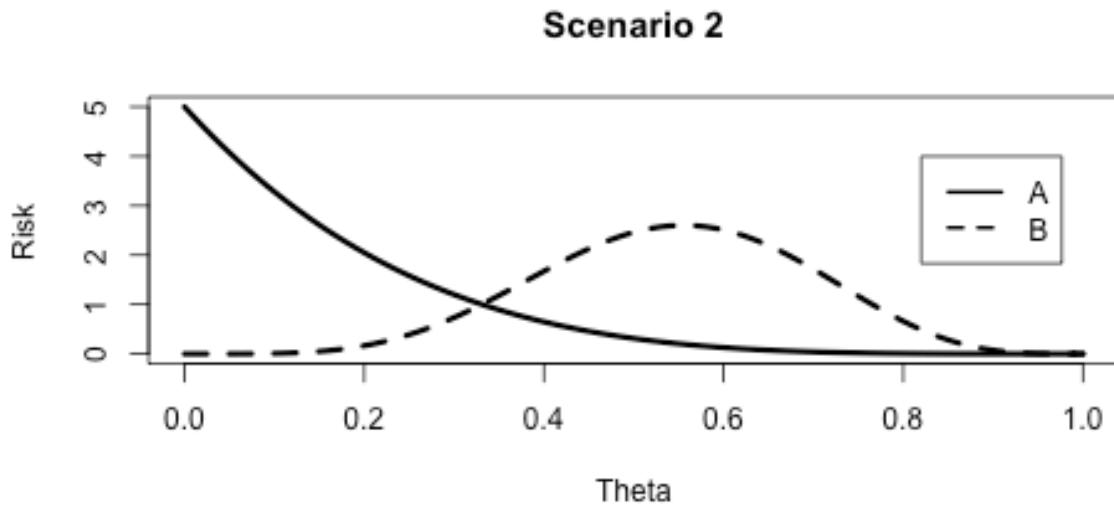
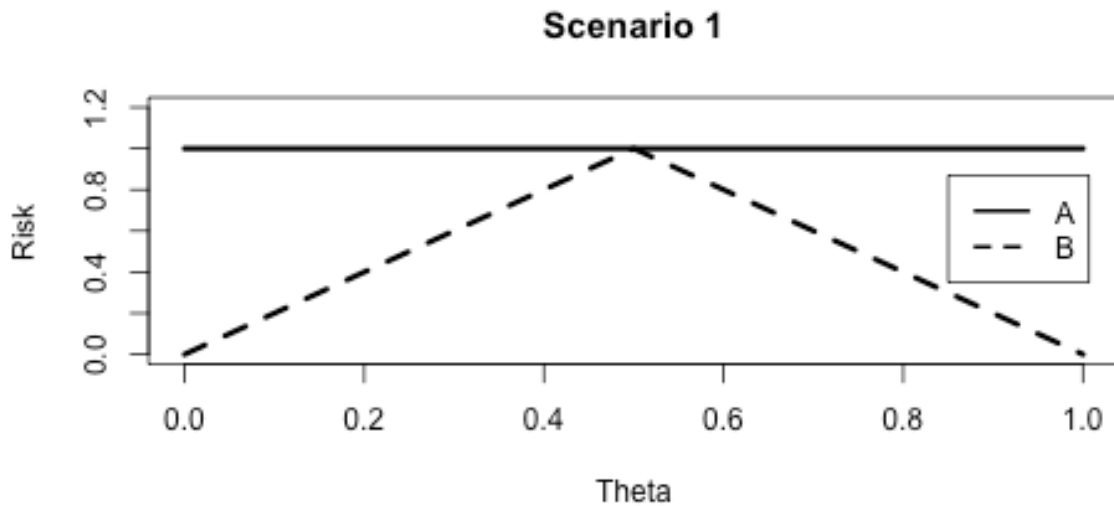
The posterior distribution of θ given \mathbf{y} is

$$\theta | \mathbf{y} \sim N(\bar{y}, \sigma^2/n)$$

(see slide 9 of the “More on Interval Estimation” notes on 2/18). So, because $\bar{Y} \sim N(\theta, \sigma^2/n)$, it follows from the marginal distribution of the normal-normal model that

$$\bar{Y} | \mathbf{y} \sim N(\bar{y}, 2\sigma^2/n).$$

Thus, $P(\bar{Y} \geq \bar{y} | \mathbf{y}) = \frac{1}{2}$.



3. *Decision rules* [6 pts]

The above figure shows the frequentist risk of two decision rules (A and B) as a function of an unknown parameter θ , under two different scenarios. For each scenario, A and B are the only two potential decisions (action space = $\{A,B\}$), θ can take any value between 0 and 1, and θ has a prior that is positive over the unit interval $p(\theta) > 0 \forall \theta \in [0, 1]$.

For each of the following, answer “yes”, “no”, or “not enough informations”:

- (a) Is A admissible in Scenario 1? *No*
- (b) Is A minimax in Scenario 1? *Yes*
- (c) Is A a Bayes rule in Scenario 1? *No*
- (d) Is B admissible in Scenario 1? *Yes*
- (e) Is B minimax in Scenario 1? *Yes*
- (f) Is B a Bayes rule in Scenario 1? *Yes*

- (g) Is A admissible in Scenario 2? *Yes*
- (h) Is A minimax in Scenario 2? *No*
- (i) Is A a Bayes rule in Scenario 2? *Not enough information*
- (j) Is B admissible in Scenario 2? *Yes*
- (k) Is B minimax in Scenario 2? *Yes*
- (l) Is B a Bayes rule in Scenario 2? *Not enough information*