40.017 PROBABILITY & STATISTICS

Homework 3

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Section 2

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Consider two disjoint timelines shown in Figure 1 below.



Figure 1: Disjoint Intervals from 0 to t

Using the definition of conditional probability, we have:

$$\begin{split} \mathbb{P}(1 \text{ arrival in } [0,s]|1 \text{ arrival in } [0,t]) &= \frac{\mathbb{P}(1 \text{ arrival in } [0,s] \cap 1 \text{ arrival in } [0,t])}{\mathbb{P}(1 \text{ arrival in } [0,t])} \\ &= \frac{\mathbb{P}(1 \text{ arrival in } [0,s] \cap 0 \text{ arrival in } [s,t])}{\mathbb{P}(1 \text{ arrival in } [0,t])} \\ &= \frac{\mathbb{P}(N(s)=1) \times \mathbb{P}(N(t-s)=0)}{\mathbb{P}(N(t)=1)} \\ &= \frac{e^{-\lambda s} \frac{(\lambda s)^1}{1!} \times e^{\lambda (t-s)} \frac{[\lambda (t-s)]^0}{0!}}{e^{-\lambda t} \frac{(\lambda t)^1}{1!}} \\ &= \frac{s}{t} e^{-\lambda s} \times e^{-\lambda t} \times e^{\lambda s} \times e^{\lambda t} \\ &= \boxed{\frac{s}{t}} \end{split}$$

(a)

The arrival time of a Poisson process is exponentially distributed, with parameter λ for a time period t. If we have a merged Poisson process, the resulting distribution is $Poi(\lambda_1 + \lambda_2)$. The arrival time X of a Poisson process follows an exponential distribution, so we have

$$X \sim \exp(\lambda_1 + \lambda_2)$$

and they are i.i.d. for both processes. To get the distribution of the 2nd event in the merged process, we add the two arrival times X_1 , X_2 which then follows a Gamma distribution (sum of exponentials):

$$X_1 + X_2 \sim \text{Gamma}(2, \lambda_1 + \lambda_2)$$

(b)

It is not necessarily true: The arrival time of the second event in the merged process will follow a gamma distribution from part (a). Consider also the memoryless-ness property, where the second arrival will occur X_2 time after, which is not necessarily $\max(Y_1, Y_2)$.

Since the enemy tanks are numbered from 0 to N, then we have a total of N+1 tanks. From the hint, we first compute $\mathbb{E}(X_i)$:

$$\mathbb{E}(X_i) = \sum_{i=0}^{N} x_i \mathbb{P}(X = x_i)$$

$$= 0 \times \frac{1}{N+1} + 1 \times \frac{1}{N+1} + \dots + N \times \frac{1}{N+1}$$

$$= \frac{1}{N+1} (0 + 1 + \dots + N)$$

$$= \frac{1}{N+1} \left(\frac{N+1}{2} (N) \right)$$

$$= \frac{N(N+1)}{2(N+1)}$$

$$= \frac{N}{2}$$

For a total of n observations with replacement, we have:

$$\mathbb{E}(\bar{X}_n) = \mathbb{E}(\frac{1}{n} \sum_{i=1}^n X_i)$$

$$= \frac{1}{n} \left[\mathbb{E}(X_1) + \mathbb{E}(X_2) + \dots + \mathbb{E}(X_n) \right]$$

$$= \frac{1}{n} \left[n\mathbb{E}(X_i) \right]$$

$$= \frac{1}{n} \left[n\left(\frac{N}{2}\right) \right]$$

$$= \left[\frac{N}{2} \right]$$

To find an unbiased estimator of the total number of tanks, we need by definition $\mathbb{E}(\hat{N}) = (N+1)$. We note that the total number is N+1, and thus:

$$\mathbb{E}(\bar{X}_n) = \frac{N}{2}$$

$$2\mathbb{E}(\bar{X}_n) = N$$

$$(N+1) = 2\mathbb{E}(\bar{X}_n) + 1$$

$$(N+1) = \boxed{\mathbb{E}(2\bar{X}_n+1)} \quad \text{(linearity of expectation)}$$

Thus, we conclude that an unbiased estimator for the total number of tanks is $\hat{N} = 2\bar{X}_n + 1$.

(a)

We are given that X_1, X_2, \ldots, X_n are i.i.d uniformly distributed with parameter θ . From the hint, we note that if the maximum of X_i 's is < x, then each of the X_i 's is < x, we use the property of independence here:

$$\mathbb{P}(X_{\max} < x) = \mathbb{P}(\max(X_1, X_2, \dots, X_n) < x)$$

$$= \mathbb{P}(X_1 < x) \cap \mathbb{P}(X_2 < x) \cap \dots \cap \mathbb{P}(X_n < x)$$

$$= \prod_{i} \mathbb{P}(X_i < x)$$

$$= \boxed{\left(\frac{x}{\theta}\right)^n} \quad \text{(using cdf of uniform distribution)}$$

where x is between 0 and θ .

(b)

From the hint, we differentiate part (a) to get the pdf of X_{max} :

$$\frac{d}{dx}\left[\left(\frac{x}{\theta}\right)^n\right] = n\frac{x^{n-1}}{\theta^n}$$

Now to find $\mathbb{E}(X_{\text{max}})$:

$$\mathbb{E}(X_{\text{max}}) = \int_{-\infty}^{\infty} x f(x) \, dx$$

$$= \int_{0}^{\infty} x \left(n \cdot \frac{x^{n-1}}{n} \right) \, dx$$

$$= \frac{n}{\theta^{n}} \int_{0}^{\infty} x^{n} \, dx$$

$$= \frac{n}{\theta^{n}} \cdot \frac{x^{n+1}}{n+1} \Big|_{0}^{\theta}$$

$$= \frac{n(\theta^{n+1})}{\theta^{n}(n+1)}$$

$$= \left[\frac{n}{n+1} \cdot \theta \right]$$

(c)

To find an unbiased estimator for θ , we need to show that $\mathbb{E}(\hat{\theta}) - \theta = 0$, i.e.

$$\begin{split} \mathbb{E}(X_{\max}) &= \frac{n}{n+1} \cdot \theta \\ \frac{n+1}{n} \cdot \mathbb{E}(X_{\max}) &= \theta \\ \mathbb{E}\left(\frac{n+1}{n} \cdot X_{\max}\right) &= \theta \quad \text{(by linearity of expectation)} \\ \Rightarrow \hat{\theta} &= \boxed{\frac{n+1}{n} \cdot X_{\max}} \end{split}$$

Thus $\frac{n+1}{n} \cdot X_{\text{max}}$ is an unbiased estimator for θ .

Let $X \sim \text{uniform}(\theta_1, \theta_2)$. Using the method of moments,

$$\mathbb{E}(X) = \bar{x} = \frac{\theta_1 + \theta_2}{2}, \quad \text{Var}(X) = s_x^2 = \frac{(\theta_2 - \theta_1)^2}{12}$$

Combining the two equations, where $\theta_1 = 2\bar{x} - \theta_2$, we start with s_x^2 :

$$s_x^2 = \frac{(\theta_2 - 2\bar{x} + \theta_2)^2}{12}$$

$$= \frac{(2\theta_2 - 2\bar{x})^2}{12}$$

$$= \frac{4(\theta_2 - \bar{x})^2}{12}$$

$$3s_x^2 = (\theta_2 - \bar{x})^2$$

$$\sqrt{3}s_x = \theta_2 - \bar{x}$$

$$\theta_2 = \sqrt{3}s_x + \bar{x}$$

$$\Rightarrow \theta_1 = 2\bar{x} - \sqrt{3}s_x + \bar{x}$$

$$= \bar{x} - \sqrt{3}s_x$$

Thus we have the estimators for θ_1 and θ_2 :

$$\hat{\theta}_1 = \bar{x} - \sqrt{3}s_x, \quad \hat{\theta}_2 = \bar{x} + \sqrt{3}s_x$$

(a)

From Week 6 Class 1 page 11, the Likelihood function of an exponential distribution is given by:

$$L(\theta) = f(x_1|\theta)f(x_2|\theta)\dots f(x_n|\theta)$$

$$= \theta^n \exp\left(-\theta \sum_{i=1}^n x_i\right)$$

$$= \theta^4 \exp\left(-\theta(1.3 + 0.6 + 0.3 + 0.8)\right)$$

$$= \theta^4 \exp\left(-3\theta\right)$$

where the x_i 's are i.i.d, and there are 4 observations of x.

(b)

The prior for θ has a gamma(5,2) distribution, given by:

$$f(\theta) = C_0 \theta^{5-1} e^{-2\theta}$$
$$= C_0 \theta^4 e^{-2\theta}$$

Using the formula in Week 6 Class 2, posterior = constant \times likelihood function \times prior, we have:

posterior = constant × likelihood function × prior
=
$$C_1 \times \theta^4 e^{-3\theta} \times C_0 \theta^4 e^{-2\theta}$$

= $C_2 \theta^{4+4} e^{-3\theta-2\theta}$
= $C_2 \theta^8 e^{-5\theta}$

(c)

The standard gamma function with parameters α, λ is given by:

$$f(\theta) = C\theta^{\alpha - 1}e^{-\lambda\theta}$$

for any real constant C. From (b), we see that the posterior is also clearly gamma distributed, where:

posterior =
$$C_2 \theta^{9-1} e^{-5\theta}$$

and with parameters $\alpha = 9$, $\lambda = 5$. From Week 3 Class 1, the mean of a $\theta \sim \text{Gamma}(\alpha, \lambda)$ distribution is given by:

$$\mathbb{E}(\theta) = \frac{\alpha}{\lambda}$$
$$= \boxed{\frac{9}{5}}$$

We start with the general definition of expectation:

$$\begin{split} \mathbb{E}(X) &= \int_{-\infty}^{\infty} x \cdot f(x) \, \mathrm{d}x \\ &= \int_{0}^{1} x \cdot \frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)} x^{a-1} (1-x)^{b-1} \, \mathrm{d}x \\ &= \int_{0}^{1} \frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)} \cdot x^{(a+1)-1} (1-x)^{b-1} \, \mathrm{d}x \\ &= \int_{0}^{1} \frac{\Gamma(a+b)}{\Gamma(a)} \times \frac{\Gamma(a+1)}{\Gamma((a+1)+b)} \times \frac{\Gamma((a+1)+b)}{\Gamma(a+1)\Gamma(b)} x^{(a+1)-1} (1-x)^{b-1} \, \mathrm{d}x \\ &= \int_{0}^{1} \frac{\Gamma(a+b)}{\Gamma(a)} \times \frac{a\Gamma(a)}{\Gamma((a+1)+b)} \times \frac{\Gamma((a+1)+b)}{\Gamma(a+1)\Gamma(b)} x^{(a+1)-1} (1-x)^{b-1} \, \mathrm{d}x \quad (\because \Gamma(\alpha+1) = \alpha \Gamma(\alpha)) \\ &= \int_{0}^{1} \frac{a\Gamma(a+b)}{(a+b)\Gamma(a+b)} \times \frac{\Gamma((a+1)+b)}{\Gamma(a+1)\Gamma(b)} x^{(a+1)-1} (1-x)^{b-1} \, \mathrm{d}x \\ &= \frac{a}{a+b} \int_{0}^{1} \frac{\Gamma((a+1)+b)}{\Gamma(a+1)\Gamma(b)} x^{(a+1)-1} (1-x)^{b-1} \, \mathrm{d}x \\ &= \boxed{\frac{a}{a+b}} \quad \text{(total area under beta} (a+1,b) \text{ r.v. is 1)} \end{split}$$