

STA602 Exam I Spring 2022

Instructions

- This exam is **open note and open book** (it is closed to all other resources).
- Write your name, NetID, and signature below if you use a tablet to complete the exam; otherwise put this information with question 1.
- Only what is on the exam will be graded (or written work submitted as one pdf file).
- Show all work and back up all your results for full credit.
- You must label/assign pages when submitting to Gradescope to avoid losing points.
- The exam should be submitted via Gradescope; make sure to “assign pages” as you do on your homework so we can find your solutions!

Community Standard

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- I will not lie, cheat, or steal in my academic endeavors;
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- I will act if the Standard is compromised.

I have adhered to the Duke Community Standard in completing this exam.

Name: _____

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Signature: _____

Score

(For TA use only — leave this section blank.)

1. _____/8

2. _____/2

3. _____/7

4. _____/11

Overall: _____/28

List of common distributions

$$\text{Geometric}(x|\theta) = \theta(1 - \theta)^x \mathbb{1}(x \in \{0, 1, 2, \dots\}) \text{ for } 0 < \theta < 1$$

$$\text{Bernoulli}(x|\theta) = \theta^x(1 - \theta)^{1-x} \mathbb{1}(x \in \{0, 1\}) \text{ for } 0 < \theta < 1$$

$$\text{Binomial}(x|n, \theta) = \binom{n}{x} \theta^x(1 - \theta)^{n-x} \mathbb{1}(x \in \{0, 1, \dots, n\}) \text{ for } 0 < \theta < 1$$

$$\text{Poisson}(x|\theta) = \frac{e^{-\theta}\theta^x}{x!} \mathbb{1}(x \in \{0, 1, 2, \dots\}) \text{ for } \theta > 0$$

$$\text{Exp}(x|\theta) = \theta e^{-\theta x} \mathbb{1}(x > 0) \text{ for } \theta > 0$$

$$\text{Uniform}(x|a, b) = \frac{1}{b - a} \mathbb{1}(a < x < b) \text{ for } a < b$$

$$\text{Gamma}(x|a, b = \text{rate}) = \frac{b^a}{\Gamma(a)} x^{a-1} e^{-bx} \mathbb{1}(x > 0) \text{ for } a, b > 0,$$

$$\text{Gamma}(x|a, b = \text{scale}) = \frac{1}{b^a \Gamma(a)} x^{a-1} e^{-x/b} \mathbb{1}(x > 0) \text{ for } a, b > 0,$$

$$\text{Pareto}(x|\alpha, c) = \frac{\alpha c^\alpha}{x^{\alpha+1}} \mathbb{1}(x > c) \text{ for } \alpha, c > 0$$

$$\text{Beta}(x|a, b) = \frac{1}{B(a, b)} x^{a-1} (1 - x)^{b-1} \mathbb{1}(0 < x < 1) \text{ for } a, b > 0$$

$$\mathcal{N}(x|\mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{1}{2\sigma^2}(x - \mu)^2\right) \text{ for } \mu \in \mathbb{R}, \sigma^2 > 0$$

$$\mathcal{N}(x|\mu, \lambda^{-1}) = \sqrt{\frac{\lambda}{2\pi}} \exp\left(-\frac{1}{2}\lambda(x - \mu)^2\right) \text{ for } \mu \in \mathbb{R}, \lambda > 0$$

1. **[8 points]** Suppose an insurer is interested in modeling insurance claims above a deductible limit of \$1000. Adopt the sampling model for claims data

$$Y_i \mid \alpha \stackrel{i.i.d.}{\sim} \text{Pareto}(\alpha, c = 1000), \quad i = 1, \dots, n.$$

Assume the shape parameter α has prior $\alpha \sim \text{Gamma}(a, b = \text{rate})$.

- (a) [3 points]** What is the posterior distribution of α ? Show all your work.

Hint: For $c > 0$, $c^\alpha = e^{\alpha \ln c}$.

- (b) [3 points]** One can show that the maximum likelihood estimator (MLE) of α is:

$$\hat{\alpha}_n = \frac{n}{\sum_{i=1}^n \ln\left(\frac{y_i}{1000}\right)}.$$

The MLE $\hat{\alpha}_n$ is a good estimate of α based on *data alone*. Show that the posterior mean of α can be expressed as a weighted average of its prior mean and its MLE:

$$\mathbb{E}[\alpha \mid Y_1 = y_1, \dots, Y_n = y_n] = w\mathbb{E}[\alpha] + (1 - w)\hat{\alpha}_n.$$

In other words, the posterior mean of α combines prior information (with weight w) and data (with weight $1 - w$).

- (c) [2 points]** What happens to the weight w if $n = 0$ and as $n \rightarrow \infty$? Interpret these two cases for the insurance problem at hand.

2. **[2 points]** State one benefit and one limitation regarding conjugate distributions. Please provide your answer in no more than two sentences. All correct and reasonable answers will be accepted.

3. [7 points] **Normal Distributions** Assume

$$Y_i \mid \theta \stackrel{iid}{\sim} \text{Normal}(\theta, \sigma^2), \quad i = 1, \dots, n. \quad (0.1)$$

For now, assume that θ is unknown and $\sigma^2 > 0$ is known.

(a) [2 points]

- i. What is a conjugate prior for θ ? State the distribution and parameters in terms of μ_o , and τ_o^2 .

- ii. Provide the posterior distribution of $\theta \mid y_{1:n}$. **You may state it without proof.**

- (b) [2 points] Let's now consider predicting a new observation \tilde{Y} from the population after having observed the data $y_{1:n}$ and assuming the model in equation 0.1. Recall that $\sigma^2 > 0$ is known. Find the posterior predictive distribution $p(\tilde{y} \mid y_{1:n})$. **You must provide step by step details to receive full credit.**

Hint:

$$\tilde{Y} \mid \theta, \sigma^2 \sim \text{Normal}(\theta, \sigma^2) \iff \tilde{Y} = \theta + \epsilon, \quad \text{where} \quad \epsilon \mid \theta, \sigma^2 \sim \text{Normal}(0, \sigma^2).$$

(c) i. **[1 point]** Provide an example, where it would make sense to fit the model in equation 0.1 (give a concrete example). **Limit yourself to two–three sentences.**

ii. **[1 point]** Now give a concrete example where it seems reasonable that we do not know θ or $\sigma^2 > 0$. **Limit yourself to two–three sentences.**

(d) **[1 point]** What model would you utilize/propose using in (c), part ii? **You may either state the name of the model or give its full parameterization.**

4. [11 points] **Understanding the aging process of laptops** Suppose you are a statistical consultant working with a warranty company. You are modeling the time-to-failure of n products (denoted as Y_1, \dots, Y_n). Based upon the empirical distribution of the data, you adopt the following Weibull sampling model:

$$Y_i | \gamma \stackrel{i.i.d.}{\sim} \text{Weibull}(\gamma, k), \quad i = 1, \dots, n.$$

Here, $\gamma = \frac{1}{\lambda}$ is an unknown rate parameter, and k is a shape parameter (assumed fixed).

The Weibull distribution for a single $y_i | \gamma = \frac{1}{\lambda}$ can be written as the following:

$$p(y_i | \gamma = \frac{1}{\lambda}, k) = k\lambda(y_i\lambda)^{k-1} \exp\{-(y_i\lambda)^k\}, \quad \text{where } y \geq 0.$$

You may find that https://en.wikipedia.org/wiki/Weibull_distribution may help you on part a.

For this entire problem, you must show all of your step by step derivations in order to receive full/partial credit.

- (a) [2 points] Suppose the product under warranty is a laptop computer. What should you set k as in the Weibull model? Explain briefly.

- (b) [1 point] Show (step by step) the following:

$$p(y_{1:n} | \lambda) \propto \lambda^n \lambda^{n(k-1)} \exp \left\{ - \sum_{i=1}^n (y_i \lambda)^k \right\}.$$

- (c) [**1 point**] Suppose $k = 2$. Is $\lambda \sim \text{Gamma}(a = \text{shape}, b = \text{rate})$ conjugate for the likelihood function? Specifically, consider

$$p(\lambda) \propto \lambda^{a-1} \exp\{-b\lambda\}.$$

- (d) [**2 points**] Let's define a new parameter such that $\lambda = \theta^{1/k}$. Show (step by step) that $\theta \sim \text{Gamma}(a, b)$ is conjugate under this re-parametrized model for any k . That is, derive the posterior distribution

$$p(\theta|y_{1:n}).$$

- (e) **[3 points]** From part (d), what is the posterior mean $\mathbb{E}(\theta|y_{1:n})$? Suppose a good “data-based” estimate of θ is $\hat{\theta} = n / \sum_{i=1}^n y_i^k$. Write the posterior mean of θ as a weighted average of $\hat{\theta}$ and the prior mean $\mathbb{E}(\theta)$. Give two scenarios in which one would place higher weight on the prior mean.

- (f) **[2 points]** Suppose you set $k = 1$ in part (d). Let Z be the time-to-failure of a new product, following $Z|\theta \sim Weibull(\theta^{1/k}, k) = Weibull(\theta, 1)$. What is the predictive distribution $p(Z|y_{1:n})$?

Hint: Let A be a random variable with probability density function $p_A(x)$. Then $B = A - c$ is a r.v. with p.d.f. $p_B(x) = p_A(x + c)$.