

# Gov 2001: Problem Set 6

Spring 2026

## Instructions:

- The Problem set is due on **March 31, 11:59 PM Eastern Time**.
- Please upload a PDF of your solutions to Gradescope. Make sure to assign to each question all the pages with your work on that question.
- **Do not use AI assistants (ChatGPT, Claude, Copilot, etc.) on this problem set.** Work with each other instead. The struggle is where learning happens.
- Remember: 70% of your grade comes from in-class exams. Use problem sets to *learn*, not just to get answers.

## Short Questions

1. Suppose  $X_1, \dots, X_n \stackrel{iid}{\sim} \text{Exp}(\lambda)$ . Write down the likelihood function  $L(\lambda)$  and the log-likelihood function  $\ell(\lambda)$ .

## Long Questions

2. Let  $X$  be a nonnegative random variable with  $\mathbb{E}[X] = 10$  and  $\text{Var}(X) = 36$ .

- Use **Markov's inequality** to give an upper bound for  $\mathbb{P}(X \geq 25)$ .
- Use **Chebyshev's inequality** to give an upper bound for  $\mathbb{P}(|X - 10| \geq 15)$ .
- Use part (b) to obtain an upper bound for  $\mathbb{P}(X \geq 25)$  and compare it with part (a).

3. Let  $X_1, \dots, X_n \stackrel{iid}{\sim} \text{Pois}(\lambda)$  with  $\lambda > 0$ . Since  $\mathbb{E}[X_i] = \text{Var}(X_i) = \lambda$ , we can construct two plug-in estimators of  $\lambda$ :

$$\hat{\lambda}_1 = \bar{X}, \quad \hat{\lambda}_2 = \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^2$$

- Show that  $\hat{\lambda}_2 = \overline{X^2} - (\bar{X})^2$ .
- Show that both estimators are consistent for  $\lambda$ .
- Use CLT to derive the asymptotic distribution of  $\sqrt{n}(\hat{\lambda}_1 - \lambda)$ .
- Use CLT and Delta method to derive the asymptotic distribution of  $\sqrt{n}(\hat{\lambda}_2 - \lambda)$ .
- We already know that  $\hat{\lambda}_2$  is biased and the bias-corrected estimator is  $\hat{\lambda}_3 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2 = \frac{n}{n-1} \hat{\lambda}_2$ . Does  $\hat{\lambda}_3$  have the same asymptotic distribution with  $\hat{\lambda}_2$ ?

4. Let  $X_1, \dots, X_n \stackrel{iid}{\sim} \mathcal{N}(\mu, \sigma^2)$ , where  $\sigma^2$  is known. The estimand is  $\theta = \mu^2$ .

- Write down the likelihood function and log-likelihood function for  $\mu$ .
- Take the first-order condition and solve for the MLE  $\hat{\mu}_{MLE}$  step by step.
- Define the MLE for  $\theta$  as  $\hat{\theta}_{MLE} = (\hat{\mu}_{MLE})^2$ . Use CLT and Delta method to derive the asymptotic distribution of  $\sqrt{n}(\hat{\theta}_{MLE} - \theta)$ .

5. Let  $Y_1, \dots, Y_n$  be i.i.d. random variables such that  $\mathbb{E}[Y_i^2] = 1$ ,  $\text{Var}(Y_i^2) < \infty$ , and define

$$S_n = \frac{1}{n} \sum_{i=1}^n Y_i^2, \quad V_n = \frac{1}{n} \sum_{i=1}^n (Y_i^2 - S_n)^2.$$

Let's see how CLT, LLN, CMT, Slutsky, and Delta method are used in practice.

- What is the approximate distribution of  $S_n$  as  $n \rightarrow \infty$ ? Your answer can include  $\text{Var}(Y_i^2)$  but should not include  $\mathbb{E}[Y_i^2]$ .
- Show that  $\sqrt{V_n} \xrightarrow{p} \sqrt{\text{Var}(Y_i^2)}$ . (Hint: Use LLN and CMT.)
- Given parts (a) and (b), prove

$$\frac{\sqrt{n}(S_n - 1)}{\sqrt{V_n}} \xrightarrow{d} N(0, 1).$$

(Hint: Use Slutsky.)

- Find the constant  $c$  such that

$$c \cdot \frac{\sqrt{n}(\exp(S_n) - e)}{\sqrt{V_n}} \xrightarrow{d} N(0, 1).$$

(Hint: Use the Delta method.)

6. Let  $X_1, \dots, X_n \stackrel{iid}{\sim} \text{Bern}(p)$ , where  $p \in (0, 1)$ , and let

$$\hat{p} = \bar{X} = \frac{1}{n} \sum_{i=1}^n X_i.$$

We will construct confidence intervals for  $p$  using two different methods.

- Compute  $\mathbb{E}[\hat{p}]$  and  $\text{Var}(\hat{p})$ .
- Use Chebyshev's inequality to construct a conservative  $(1 - \alpha)$  confidence interval for  $p$ .
- Write the asymptotic distribution of  $\sqrt{n}(\hat{p} - p)$  and  $\hat{p}$ .
- Let  $\sigma$  be the standard deviation of  $\hat{p}$ 's asymptotic distribution. Express  $\sigma$  in the form of  $p$  and construct  $\hat{\sigma}$  by plugging in  $\hat{p}$  for  $p$ .
- Use your result in (c) and (d) to construct an approximate  $(1 - \alpha)$  confidence interval for  $p$ . Notice that your confidence interval shouldn't contain the unknown estimand  $p$ .