# STAT165/265 HW 8

March 14, 2025

### Due Friday, March 21, 2025 at 11:59pm

# **Deliberate Practice: Invalidating Considerations**

#### Expected completion time: 120 minutes

Graded on accuracy

This exercise follows the activity in last week's discussion. We will use the following simplified model for invalidating considerations:

- Suppose you want an 80% confidence interval around some quantity
- You have an initial distribution  $p_0$  for what happens in a "normal" world. For example, maybe this is the distribution that you get by looking at a reference class.
- After brainstorming invalidating considerations, you realize that there is a small probability  $\epsilon$  that you are instead in a "crazy" world, in which case your distribution is instead  $p_1$ .
- Your new probability distribution is therefore the mixture  $(1 \epsilon)p_0 + \epsilon p_1$ .

For each of the following examples of  $p_0, p_1$ , and  $\epsilon$ , do the following:

- Give an 80% confidence interval when accounting for the  $\epsilon$  probability of a "crazy" world. Your interval should be centered, i.e. your lower and upper bounds should be the 10th and 90th percentiles of the mixture distribution.
- Confirm your reasoning by simulation: using Python, draw 1000 samples from  $(1 \epsilon)p_0 + \epsilon p_1$ , and use them to estimate the 10th and 90th percentiles of this distribution. Sampling from a mixture happens in two steps:
  - 1. Sample from a Bernoulli( $\epsilon$ ) to choose between  $p_0$  and  $p_1$ .
  - 2. Sample from the chosen distribution (in this exercise the distributions will be uniform).

- 1. Non-overlapping uniform distributions
  - $p_0 = \text{Uniform}(0, 2)$
  - $p_1 = \text{Uniform}(2.5, 3)$
  - (a)  $\epsilon = 0.05$ , (b)  $\epsilon = 0.1$ , (c)  $\epsilon = 0.2$

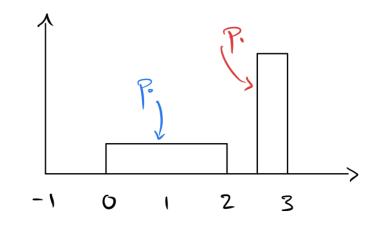


Figure 1: Non-overlapping uniform distributions

- 2. Overlapping uniform distributions
  - $p_0 = \text{Uniform}(0, 2)$
  - $p_1 = \text{Uniform}(1,3)$
  - (a)  $\epsilon = 0.05$ , (b)  $\epsilon = 0.1$ , (c)  $\epsilon = 0.2$

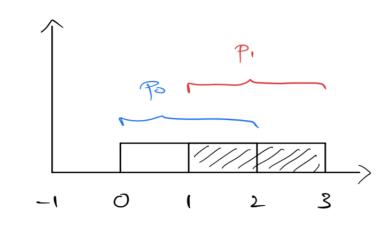


Figure 2: Overlapping uniform distributions

On Gradescope, please also submit the time it took to complete this exercise. Please note that we have a separate assignment set up for this, which is worth 1 point for this question.

## **Deliberate Practice: Numerical Sensitivity**

Expected completion time: 180 minutes Graded on accuracy

Consider the formula for the number of days T until the peak of Omicron that we used in the *Turning Considerations into Probabilities* lecture:

$$T = \log_2 \left( N/2N_0 \right) \cdot t + \Delta_0 + \Delta_1,$$

where:

- N is the total number of future UK Omicron cases
- $N_0$  is the current number of UK Omicron cases
- t is the Omicron doubling time
- $\Delta_0$  is the lag between case peak and hospital peak
- $\Delta_1$  is the lag between single-day hospital peak and 7-day average hospital peak

Using simulations, we will assess the sensitivity of this formula to variations of its five inputs.

- 1. Suppose the inputs are sampled independently from Normal and LogNormal distributions. For this, sample  $\alpha, \beta, \gamma$ , and  $\delta$  independently from Normal(0, 1), and define the inputs N,  $N_0, t$ , and  $\Delta$  as follows:
  - $N = \exp(15.57 + 0.30 \cdot \alpha)$
  - This means that N follows a LogNormal distribution with mean  $6.7 \times 10^6$  and standard deviation  $4 \times 10^6$ .
  - $N_0 = \exp(12.18 + 0.06 \cdot \beta)$
  - This means that  $N_0$  follows a LogNormal distribution with mean  $0.2 \times 10^6$  and standard deviation  $0.05 \times 10^6$ .
  - $t = 2.4 + 0.5 \cdot \gamma$
  - $\Delta = \Delta_0 + \Delta_1 = 12 + 3 \cdot \delta.$

Sample the inputs in this way 1000 times, computing each time the corresponding value of T using the formula above. Plot the histogram of the distribution of T: what are its 10th, 25th, 50th, 75th, and 90th percentiles?

2. Increase the standard deviation of N while leaving the other distributions constant: for this, sample  $\alpha$  from Normal(0, 4) rather than Normal(0, 1). Do the same but for  $\Delta$  instead: double the standard deviation of  $\delta$ , while leaving the other distributions constant. Which of these leads to the greatest increase in the variance of T?

- 3. Replace the Normal distributions by Student's t distributions: instead of sampling  $\alpha, \beta, \gamma$  and  $\delta$  from Normal(0, 1), sample them from Student( $\nu = 5$ ). Plot the histogram of the distribution of T: how did using a t distribution instead of a Normal change the quantiles and variance of T?
- 4. Repeat your simulation with  $\nu = 3, \nu = 7$  and  $\nu = 9$ : how does the degrees of freedom parameter  $\nu$  affect the distance between the 10th and 90th percentiles of T?
- 5. Suppose now that instead of being independent, the inputs are correlated. That is, sample  $[\alpha, \beta, \gamma, \delta]^{\top}$  from a multivariate Gaussian distribution:

$$\begin{array}{c} \alpha \\ \beta \\ \gamma \\ \delta \end{array} \right] \sim \operatorname{Normal}(\mathbf{0}, \mathbf{\Sigma}), \text{ with } \mathbf{\Sigma} = \begin{bmatrix} 1 & \rho & \rho & \rho \\ \rho & 1 & \rho & \rho \\ \rho & \rho & 1 & \rho \\ \rho & \rho & \rho & 1 \end{bmatrix}, \text{ and } \rho \in \left(-\frac{1}{3}, \frac{1}{3}\right).$$

Draw 1000 samples, and plot a histogram of the values of T. Do the 10th and 90th quantiles get closer together or farther apart as the correlation parameter  $\rho$  increases from 0 to  $\frac{1}{3}$ ? as  $\rho$  decreases from 0 to  $-\frac{1}{3}$ ?

On Gradescope, please also submit the time it took to complete this exercise. Please note that we have a separate assignment set up for this, which is worth 1 point for this question.

## Predictions

Expected completion time: 90 minutes Graded on accuracy as part of the class forecasting competition

Make and submit predictions to the questions on this Google Form: https://forms.gle/k868FmJVhVckH23s6.

### [STAT 265 only] Sums of correlated variables

Expected completion time: 60 min Graded on accuracy

Consider random variables  $X_1, ..., X_k$  each with standard deviation  $\sigma$ . Let  $X^*$  denote their sum. That is,

$$X^* = X_1 + \ldots + X_k$$

In this problem we will explore how different levels of correlation between each  $X_i$  change the standard deviation of  $X^*$ , which we denote as  $\sigma^*$ .

Let

$$\vec{X} = \begin{bmatrix} X_1 \\ X_2 \\ \vdots \\ X_n \end{bmatrix}$$

Suppose each  $X_i$  has correlation  $\rho$  with all others, so we have the following correlation matrix:

$$\operatorname{Corr}(\vec{X}) = \begin{bmatrix} 1 & \rho & \rho & \dots & \rho \\ \rho & 1 & \rho & \dots & \rho \\ \rho & \rho & 1 & \dots & \rho \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \rho & \rho & \rho & \dots & 1 \end{bmatrix}$$

- Derive a formula for σ\* in terms of k, σ, ρ.
  You may use the property Var(v<sup>T</sup>X) = v<sup>T</sup>Cov(X)v without proof.
  Hint: You can express X\* as 1X where 1 is a vector of ones.
- 2. For each correlation  $\rho$ , calculate the value of  $\sigma^*$  in terms of  $k, \sigma$ .
  - a.  $\rho = 0$  (the  $X_i$  are uncorrelated)
  - b.  $\rho = 1$  (the  $X_i$  are perfectly positively correlated)
  - c.  $\rho = \frac{1}{2}$  (the  $X_i$  are halfway to being perfectly positively correlated)
  - d.  $\rho = \frac{-1}{k-1}$  (the  $X_i$  are as negatively correlated as possible)
  - e.  $\rho = \frac{-1}{2(k-1)}$  (the X<sub>i</sub> are halfway to being as negatively correlated as possible)
- 3. Reflect on what happened to  $\sigma^*$  for different values of  $\rho$ . When is  $\sigma^*$  the largest? Smallest? How does  $\sigma^*$  scale as a function of k in each of the cases above?
- 4. In this next part, we will explore a creatively modified and simplified version of what led to the 2008 Financial Crisis.

Suppose each  $X_i$  represents a home mortgage and  $X^*$  is a financial instrument whose value is the sum of all the individual mortgage values. Suppose  $X_i = 0.9Y_i + 0.1Y_0$  where  $Y_1, \ldots, Y_n$ are independent (for  $i \neq j, Y_i \perp Y_j$ ), but  $Y_0$  is the same for all of them. In other words,  $Y_i$  is different for each home mortgage, but  $Y_0$  is a global, variable, say, one that depends on the national interest rate. Let k = 10000, and assume for simplicity that  $Var(Y_i) = 1$ .

- a. Compute  $\operatorname{Corr}(\vec{X})$ .
- b. Compute the standard deviation of  $X^*$ .
- c. Compute the standard deviation if people had failed to model the  $Y_0$  component (i.e. let  $X_i = Y_i$ ).
- d. What do you observe about the difference in standard deviations (1-2 sentences)?