

Stat 111 Week 1: Review of Discrete Random Variables and Distributions

1. Bernoulli Variables

The Bernoulli distribution is the simplest probability distribution, and in some ways, a building block for all other discrete distributions.

- Define $I_{(A)}$ to be an indicator variable for the event A , meaning $I_{(A)} = 1$ if A occurs and $I_{(A)} = 0$ if A^c occurs. Relate this to the Bernoulli random variable. Explain how an indicator variable represents the *fundamental bridge* between probability and expected value (see Blitzstein 4.4).
- Use indicator variables to prove Boole's inequality: $P(A_1 \cup A_2 \cup \dots \cup A_n) \leq P(A_1) + P(A_2) + \dots + P(A_n)$. Consider the special case where the events are all independent with the same probability.
- Suppose n graduates all throw their caps in the air and then retrieve a cap at random. Find an expression for the probability that none of the students retrieve their own cap (a *derangement*). Find the limit of this probability if the number of caps $n \rightarrow \infty$. Hint: For $i = 1, \dots, n$, let A_i represent the event that person i retrieves their own cap. Then $(A_1 \cup A_2 \cup \dots \cup A_n)^c$ is the event that nobody ends up with their own cap. See the *Useful Facts* at the end of this document.
- A $\text{Poisson}(\lambda)$ variable may be represented as the limit of the sum of n iid $\text{Bernoulli}(p)$ variables as $n \rightarrow \infty$ and $np \rightarrow \lambda$. Explain why, for large n , the count X of graduates who retrieve their own cap is approximately $\text{Poisson}(1)$. Compute the exact and approximate probabilities $P(X = 0)$ and $P(X = 1)$ for $n = 6$.

2. Binomial and Hypergeometric

A count X of successes in n trials may be expressed as the sum of Bernoulli variables. With iid Bernoulli trials, the sum follows a Binomial distribution. The Binomial distribution may arise as the limit of a Hypergeometric distribution. The Hypergeometric distribution may arise as a conditional distribution for Binomial counts.

- For n independent trials, each with success probability p , the distribution of X is $\text{Binomial}(n, p)$. Write out the probability mass function for X . Explain why the sum of two independent Binomial variables is also Binomial, if and only if their probabilities are equal.
- If sampling is done without replacement from a population with r successes and $N - r$ failures, then X is a hypergeometric variable. Write out the probability mass function for X . Be careful to designate the appropriate support for X (i.e., what values have non-zero probability?). Could the sum of two Hypergeometric variables also be Hypergeometric?
- For a hypergeometric random variable, show that, as $N \rightarrow \infty$ with $r/N \rightarrow p$, the distribution of X converges to $\text{Binomial}(n, p)$. See the useful facts at the end of this document.
- I have carried out experimental surveys to determine the effect of wording of a question of the response. Suppose I give out n_1 surveys with wording 1 and n_2 surveys with wording 2. Suppose students answer independently and will agree to either wording with probability p (i.e., the wording does not matter). Let r be the total number of students who agree (to either wording), the distribution of X , the number of students who agree to wording 1 (e.g.) is a Hypergeometric variable. This is the basis for the Fisher Exact Test (Blitzstein 3.9.1).
- For the situation in part d, show that the marginal distribution of X is $\text{Binomial}(n_1, p)$.

3. Binomial and Poisson

For a Poisson process in time, events (e.g., text messages received) occur at a constant rate of λ events per unit time (on average), and the counts of events in non-overlapping time intervals of lengths t_1 and t_2 are independent Poisson random variables with rates λt_1 and λt_2 . The probability mass function (pmf) for $X \sim \text{Poisson}(\lambda)$, the count of events in a unit time interval ($t = 1$), is $P(X = x) = \lambda^x e^{-\lambda} / x!$, for $x = 0, 1, \dots$. The Poisson pmf arises as the limit of the Binomial pmf.

- Find the probability of at least one event occurring in a time interval of length t . As $t \rightarrow 0$, show that this probability divided by t converges to λ , meaning the probability behaves like λt for t close to 0. Also consider the probability of exactly 1 event occurring in an interval of length t .
- Imagine partitioning a unit time interval into n non-overlapping subintervals, each of length $t = 1/n$. Let X be the count of intervals that contain at least one event. Show, in the limit as $n \rightarrow \infty$ that X represents the total count of events, and has the $\text{Poisson}(\lambda)$ pmf.
- Let $N \sim \text{Poisson}(\lambda)$ be the number of scratch-off lottery tickets sold in a day in a particular store. Each ticket has probability p of being a winner, independent of any other ticket outcomes. Let X_1 be the number of winning tickets, and $X_2 = N - X_1$ the number of losing tickets sold in a day. Show that X_1 and X_2 are independent Poisson variables by finding

$$P(X_1 = x_1, X_2 = x_2) = P(N = x_1 + x_2)P(X_1 = x_1 | N = x_1 + x_2)$$

- If $X_1 \sim \text{Poisson}(\lambda_1)$ is independent of $X_2 \sim \text{Poisson}(\lambda_2)$, show that

$$N = X_1 + X_2 \sim \text{Poisson}(\lambda_1 + \lambda_2)$$

4. Poisson and Negative Binomial

A Negative Binomial variable arises as the count of failures before a specified number of successes, and as a Poisson variable with a rate parameter generated according to a Gamma distribution.

- Suppose $X_1 \sim \text{Poisson}(\lambda_1)$ is independent of $X_2 \sim \text{Poisson}(\lambda_2)$, and let $N = X_1 + X_2 \sim \text{Poisson}(\lambda_1 + \lambda_2)$. Show $X_1 | N = n \sim \text{Binom}(n, p)$, for $p = \frac{\lambda_1}{\lambda_1 + \lambda_2}$.
- For a sequence of iid Bernoulli(p) variables, let Y be the number of failures before the r th success. Write out the pmf for Y .
- Suppose two soccer matches are played simultaneously and that the goals scored in match 1 and in match 2 represent independent Poisson processes with rates λ and 1, respectively. Let Y be the number of goals scored in match 2 at the time of the r th goal in match 1. Explain how Y follows the same distribution as Y in part b (what are the Bernoulli variables and what is p ?).
- The time θ of the r th event in a Poisson process with rate λ is a continuous random variable that follows a $\text{Gamma}(r, \lambda)$ distribution. For the situation in part c, what is the distribution of Y conditional on θ ? What is the marginal distribution of Y ?

Useful Facts:

$$(1) \quad P(A_1 \cup \dots \cup A_n) = \sum_{i=1}^n P(A_i) - \sum_{i < j} P(A_i \cap A_j) + \sum_{i < j < k} P(A_i \cap A_j \cap A_k) - \dots$$

$$(2) \quad (a + b)^n = \sum_{k=0}^n \binom{n}{k} a^k b^{n-k}$$

$$(3) \quad \lim_{n \rightarrow \infty} \frac{n!/(n-k)!}{n^k} = \lim_{n \rightarrow \infty} \frac{n(n-1)\dots(n-k+1)}{n^k} = 1, \quad k = 0, 1, \dots$$

$$(4) \quad e^x = \sum_{k=0}^{\infty} \frac{x^k}{k!} = \lim_{n \rightarrow \infty} (1 + x/n)^n$$