40.017 PROBABILITY & STATISTICS

Homework 1

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Section 2

February 6, 2024

(a)

Let A be the random variable denoting the number of questions chosen from section A. Since the student 'samples' the question without replacement, A follows a hypergeometric distribution $A \sim \text{hypgeo}(12, 9, 9)$. The pmf f(x) is given by:

$$f(x) = \mathbb{P}(X = x) = \frac{\binom{j}{x} \binom{k}{n-x}}{\binom{j+k}{n}}$$

for 6 questions from section A, we have $\mathbb{P}(X=6)$:

$$\mathbb{P}(X=6) = \frac{\binom{9}{6}\binom{9}{6}}{\binom{18}{12}}$$
$$= \frac{84}{221}$$
$$\approx 0.38$$

(b)

Now we need to consider two cases, one where 5 questions come from Section A and the other 7 from Section B, and vice versa. i.e.

$$\mathbb{P}(A=5) + \mathbb{P}(A=7) = \frac{\binom{9}{5}\binom{9}{7}}{\binom{18}{12}} + \frac{\binom{9}{7}\binom{9}{5}}{\binom{18}{12}}$$
$$= \frac{108}{221}$$
$$\approx 0.49$$

(a)

To find the probability that at least 1 man receives his own hat, we consider the complementary case of no man receiving his own hat, given by D_6 . Thus:

$$\begin{split} \mathbb{P}(\text{`at least 1 man receives his own hat'}) &= 1 - \frac{D_6}{6!} \\ &= 1 - \frac{6!(1 - \frac{1}{1!} + \frac{1}{2!} - \frac{1}{3!} + \frac{1}{4!} - \frac{1}{5!} + \frac{1}{6!})}{6!} \\ &= \left\lceil \frac{91}{144} \right\rceil \end{split}$$

(b)

Again we consider the complementary case that no man receives his own hat, or only 1 man receives his own hat. Thus:

$$\mathbb{P}(\text{'at least 2 men receives their own hats'}) = 1 - \underbrace{\frac{D_6}{6!}}_{\text{no man receives their own hat}} - \underbrace{\frac{6 \times D_5}{6!}}_{\text{1 man receives his own hat}}$$

$$= \frac{91}{144} - 6 \times \frac{5!(1 - \frac{1}{1!} + \frac{1}{2!} - \frac{1}{3!} + \frac{1}{4!} - \frac{1}{5!})}{6!}$$

$$= \boxed{\frac{191}{720}}$$

(c)

The probability that at least 5 men receive their own hats can be interpreted as every man receiving their own hats, since if 5 men receive their own hats, the last man will automatically also receive his own hat. There is only 1 way for all 6 men to receive their own hats, so we have:

$$\mathbb{P}(\text{'at least 5 men receives their own hats'}) = \frac{1}{6!} \\ = \boxed{\frac{1}{720}}$$

(a)

Since X is the number of coin tosses upon completion, we model X as a geometric distribution with the "success" being obtaining a T: $X \sim \text{geometric}(\frac{1}{2}), X \in \{1, 2, ..., n\}$. For $x \in \{1, 2, ..., n-1\}$, we thus have: $\mathbb{P}(X = x) = (1 - \frac{1}{2})^{x-1}(\frac{1}{2}) = (\frac{1}{2})^x$. As for x = n, we take the complement of all the other probabilities (since they sum to 1):

$$\mathbb{P}(X = n) = 1 - \left(\sum_{i=1}^{n-1} \mathbb{P}(X = i)\right)$$

$$= 1 - \left(\sum_{i=1}^{n-1} \left(\frac{1}{2}\right)^{i}\right)$$

$$= 1 - \frac{\frac{1}{2}(1 - (\frac{1}{2})^{n-1-1+1})}{1 - \frac{1}{2}}$$

$$= \left(\frac{1}{2}\right)^{n-1}$$

The p.m.f. table is shown below:

| X | 1 | 2 | 3 | n |
|-------------------|---------------|--|---|--------------------------------------|
| $\mathbb{P}(X=x)$ | $\frac{1}{2}$ | $\left(\frac{1}{2}\right)\left(\frac{1}{2}\right)$ | $(\frac{1}{2})(\frac{1}{2})(\frac{1}{2})$ | $\left(\frac{1}{2}\right)^{n-1}$ |

and in piecewise notation:

$$\mathbb{P}(X=x) = \begin{cases} \left(\frac{1}{2}\right)^x, & x \in \{1, 2, \dots, n-1\} \\ \left(\frac{1}{2}\right)^{x-1}, & x = n \\ 0, & \text{otherwise} \end{cases}$$

(b)

To find $\mathbb{E}(X)$, we sum over both cases of the p.m.f:

$$\mathbb{E}(X) = \sum_{i=1}^{n-1} i \cdot \mathbb{P}(X = i) + n \cdot \left(\frac{1}{2}\right)^{n-1}$$

$$= \sum_{i=1}^{n-1} i \cdot \left(\frac{1}{2}\right)^{i} + n \cdot \left(\frac{1}{2}\right)^{n-1}$$

$$= \frac{1}{2} \sum_{i=1}^{n-1} i \cdot \left(\frac{1}{2}\right)^{i-1} + n \cdot \left(\frac{1}{2}\right)^{n-1}$$
(1)

Using the hint, we differentiate the finite geometric series from the fact that $\frac{d}{dx}x^a = ax^{a-1}$. Letting $r = \frac{1}{2}$:

$$\frac{d}{dr} \left[\sum_{i=0}^{n-1} r^i \right] = \frac{d}{dr} \left[\frac{1 - r^n}{1 - r} \right]$$

$$\sum_{i=1}^{n-1} i \cdot r^{i-1} = \frac{1 - r^n}{(1 - r)^2} - \frac{nr^{n-1}}{1 - r}$$

Now, substituting back into Equation (1), we have:

$$\mathbb{E}(X) = \frac{1}{2} \left[\frac{1 - (\frac{1}{2})^n}{(1 - \frac{1}{2})^2} - \frac{n(\frac{1}{2})^{n-1}}{1 - \frac{1}{2}} \right] + n \cdot \left(\frac{1}{2}\right)^{n-1}$$

$$= \frac{1}{2} \left[4\left(1 - \left(\frac{1}{2}\right)^n\right) - 2n\left(\frac{1}{2}\right)^{n-1} \right] + n \cdot \left(\frac{1}{2}\right)^{n-1}$$

$$= 2 - \left(\frac{1}{2}\right)^{n-1} - n \cdot \left(\frac{1}{2}\right)^{n-1} + n \cdot \left(\frac{1}{2}\right)^{n-1}$$

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

Chebyshev's inequality in the standard form gives the **upper bound** for the probability of being k standard deviations ($\sigma = 5$) away from the mean $\mu = 50$.

$$\mathbb{P}(|X - \mu| \ge k\sigma) \le \frac{1}{k^2} \tag{2}$$

Let X be the weekly production of the factory (in tons). We want to find the probability $\mathbb{P}(41 \le X \le 61)$. Since the bounds are not symmetric about μ , we need the tighter bound $\mathbb{P}(41 \le X \le 59)$. Then, the value of k is now $k = \frac{9}{5}$ standard deviations away from the mean.

From Figure 1, we can see that 9 is indeed the "tighter bound":

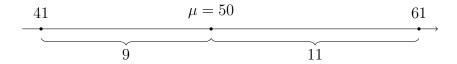


Figure 1: Illustration of bounds

From Equation (2), to find a *lower* bound, we need to take the complement:

$$\mathbb{P}(|X - \mu| < k\sigma) \ge 1 - \frac{1}{k^2} \tag{3}$$

Thus the lower bound for the probability that the weekly production is between 41 and 61 tons is:

$$\mathbb{P}(41 \le X \le 61) \ge \mathbb{P}\left(|X - \mu| < \frac{9}{5}\sigma\right)$$
$$\ge 1 - \frac{1}{\left(\frac{9}{5}\right)^2}$$
$$= \boxed{\frac{56}{81}}$$

Let $X \sim \text{negbin}(n, p)$. In the recitation, to find the MGF of X, since the negative binomial distribution is the sum of n independent geometric distributions, let $Y \sim \text{Geometric}(p)$, and q = 1 - p.

$$M_Y(t) = \frac{pe^t}{1 - qe^t} \tag{4}$$

We have $Y_1 + Y_2 + \cdots + Y_n \sim \operatorname{negbin}(n, p)$. Since each Y_i 's are independent, by the sum property, the MGF of X is given by:

$$M_{Y_1+Y_2+\dots+Y_n}(t) = M_{Y_1}(t) \cdot M_{Y_2}(t) \dots M_{Y_n}(t)$$

$$= [M_Y(t)]^n$$

$$= M_X(t)$$

$$= \left[\frac{pe^t}{1 - qe^t}\right]^n$$

Proof. To find $\mathbb{E}(X)$, we need to find the first moment, which requires the derivative:

$$M_X(t) = \left[\frac{pe^t}{1 - qe^t}\right]^n$$

$$M_X'(t) = n \left[\frac{pe^t}{1 - qe^t}\right]^{n-1} \cdot \frac{d}{dt} \left[\frac{pe^t}{1 - qe^t}\right]$$

$$= n \left[\frac{pe^t}{1 - qe^t}\right]^{n-1} \cdot \left[\frac{pe^t(qe^t + (1 - qe^t))}{(1 - qe^t)^2}\right]$$

$$= n \left[\frac{pe^t}{1 - qe^t}\right]^{n-1} \cdot \left[\frac{pe^t}{(1 - qe^t)^2}\right]$$

$$= \left[\frac{pe^t}{1 - qe^t}\right] \cdot \frac{n}{(1 - qe^t)}$$

Now, to find $\mathbb{E}(X)$, we set t=0:

$$M_X'(0) = \left[\frac{p}{1-q}\right]^n \cdot \frac{n}{1-q}$$

$$= \left[\frac{p}{1-(1-p)}\right]^n \cdot \frac{n}{1-(1-p)}$$

$$= \left[\frac{n}{p}\right] = \mathbb{E}(X)$$

(a)

To find the MGF of X, since X is continuous, we need to integrate over all X, taking note that the MGF is only defined on -1 < t < 1:

$$M_X(t) = \int_{-\infty}^{\infty} e^{tx} f(x) dx$$

$$= \int_{-\infty}^{\infty} e^{tx} \frac{1}{2} e^{-|x|} dx$$

$$= \int_{-\infty}^{0} e^{tx} \cdot \frac{1}{2} e^{x} dx + \int_{0}^{\infty} e^{tx} \cdot \frac{1}{2} e^{-x}$$

$$= \frac{1}{2} \left[\int_{-\infty}^{0} e^{(t+1)x} dx + \int_{0}^{\infty} e^{(t-1)x} dx \right]$$

$$= \frac{1}{2} \left[\frac{1}{t+1} e^{(t+1)x} \Big|_{-\infty}^{0} + \frac{1}{t-1} e^{(t-1)x} \Big|_{0}^{\infty} \right]$$

$$= \frac{1}{2} \left[\frac{1}{t+1} - \frac{1}{t-1} \right] \quad \because (-1 < t < 1)$$

$$= \frac{1}{2} \left[\frac{t-1-t-1}{(t+1)(t-1)} \right]$$

$$= \frac{1}{2} \left[\frac{-2}{t^2-1} \right]$$

$$= \left[\frac{1}{1-t^2} \right]$$

(b)

Note that $Var(X) = \mathbb{E}(X^2) - \mathbb{E}(X)^2$, so we need to find the first and second moments. For the first moment we need the derivative:

$$\mathbb{E}(X) = M_X'(0)$$

$$= \frac{d}{dt} \left(\frac{1}{1 - t^2} \right) \Big|_{t=0}$$

$$= -(1 - t^2)^{-2} (-2t) \Big|_{t=0}$$

$$= \frac{2t}{(1 - t^2)^2} \Big|_{t=0}$$

$$= 0$$

For the second moment:

$$\begin{split} \mathbb{E}(X^2) &= M_X''(0) \\ &= \frac{\mathrm{d}}{\mathrm{d}t} \left(\frac{2t}{(1-t^2)^2} \right) \bigg|_{t=0} \\ &= 2t(-2(1-t^2)^{-3}(-2t)) + (1-t^2)^{-2}(2) \bigg|_{t=0} \\ &= \left(\frac{8t^3}{(1-t^2)^3} \right) + \frac{2}{(1-t^2)^2} \bigg|_{t=0} \\ &= 2 \end{split}$$

Thus, we have:

$$Var(X) = \mathbb{E}(X^2) - \mathbb{E}(X)^2$$
$$= 2 - 0^2$$
$$= \boxed{2}$$

(a)

Manipulating the right tail probability of a standard normal r.v. Z, we have:

$$\mathbb{P}(Z \ge t) = \mathbb{P}(tZ \ge t^2) = \mathbb{P}(e^{tZ} \ge e^{t^2}) \tag{5}$$

Markov's Inequality gives an upper bound on the tail probability, subject to any constant a > 0:

$$\mathbb{P}(X \ge a) \le \frac{\mathbb{E}(X)}{a} \tag{6}$$

Proof. Applying Equation (6) on Equation (5), we have $a = e^{tZ}$.

$$\mathbb{P}(e^{tZ} \ge e^{t^2}) \le \frac{\mathbb{E}(e^{tZ})}{e^{t^2}}$$

$$= \frac{M_Z(t)}{e^{t^2}}$$

$$= \frac{e^{\frac{t^2}{2}}}{e^{t^2}}$$

$$= e^{-\frac{t^2}{2}}$$

Where $M_Z(t) = e^{t^2}$ was obtained from Week 2 Class 2 Example 3 (too lazy to typeset). From here, we have:

$$\mathbb{P}(Z \ge t) = \mathbb{P}(e^{tZ} \ge e^{t^2}) \le e^{-\frac{t^2}{2}}$$
 (7)

(b)

From Equation (7), we substitute t=-7, and the fact that the standard normal Z is symmetric about $\mu=0$, to get:

$$\mathbb{P}(Z \le -7) = \mathbb{P}(Z \ge 7)$$

$$\le e^{-\frac{7^2}{2}}$$

$$\le e^{-\frac{49}{2}}$$