

- Each of the 200 ENGG2430 students shows up to class independently with probability 0.9 and asks Poisson(0.05) questions in there. Let  $S$  be the number of students in class and  $Q$  the total number of questions asked. Find (a)  $E[S]$ , (b)  $E[Q|S]$ , (c)  $E[Q]$ , (d)  $\text{Var}[E[Q|S]]$ , (e)  $\text{Var}[Q|S]$ , (f)  $E[\text{Var}[Q|S]]$ , (g)  $\text{Var}[Q]$ .

**Solution:** Let  $Q_i$  be the number of question asked by the  $i$ -th student present in class;  $Q = Q_1 + \cdots + Q_S$ .

- $E[S] = 200 \cdot 0.9 = 180$ .
  - $E[Q|S] = \sum_{i=1}^S E[Q_i] = S \cdot 0.05 = 0.05S$  by linearity of expectation.
  - $E[Q] = E[E[Q|S]] = E[0.05S] = 0.05 \cdot 180 = 9$  by (b).
  - $\text{Var}[E[Q|S]] = \text{Var}[0.05S] = 0.05^2 \text{Var}[S] = 0.05^2 \cdot (200 \cdot 0.09 \cdot 0.01) = 0.00045$  by (b).
  - $\text{Var}[Q|S] = \sum_{i=1}^S \text{Var}[Q_i] = S \cdot 0.05 = 0.05S$  by independence of  $Q_i$ 's.
  - $E[\text{Var}[Q|S]] = E[0.05S] = 9$  by (e).
  - $\text{Var}[Q] = \text{Var}[E[Q|S]] + E[\text{Var}[Q|S]] = 9.00045$  by (d) and (f).
- You flip a coin with unknown probability of heads  $p$ . You want to learn the value of  $p$ .
    - Alice suggests the following estimator  $\hat{P}_A$ : Keep flipping the coin until you see the first head in the  $N$ -th flip. Set  $\hat{P}_A = 1/N$ .
    - Bob suggests another estimator  $\hat{P}_B$ : Flip the coin 10 times, count the number of heads  $Y$  and set  $\hat{P}_B = Y/10$ .

What is the expectation of each estimator in terms of  $p$ ? Which one is better?

**Solution:**  $\hat{P}_A$  has expectation

$$E[\hat{P}_A] = \sum_{n=1}^{\infty} \frac{1}{n} \cdot (1-p)^{n-1} p = \frac{p}{1-p} \sum_{n=1}^{\infty} \frac{(1-p)^n}{n} \approx \frac{p}{1-p} \cdot (-\log p),$$

since the infinite series is the Taylor expansion of  $-\log p$ .  $\hat{P}_B$  has expectation

$$E[\hat{P}_B] = E\left[\frac{Y}{10}\right] = \frac{1}{10} E[Y] = \frac{10p}{10} = p.$$

The expectation of  $\hat{P}_A \neq p$  in general while  $\hat{P}_B$  does;  $\hat{P}_A$  is said to be biased and  $\hat{P}_B$  unbiased.

In the next two questions, estimate the quantity of your interest using the method of your choice: Markov's inequality, Chebyshev's inequality, or the Central Limit Theorem. Justify why the method is applicable and discuss the quality of the estimate.

Summary of the assumption and result of the three methods:

- Markov's inequality:  $P(X \geq a) \leq E[X]/a$ .
  - Applies for any **non-negative** random variable  $X$ , and any  $a > 0$  ( $a > E[X]$  for a meaningful bound).
  - Requires only  $E[X]$ , useful when it is small.
  - Is an upper bound to a one-sided (right tail) probability.

- Chebyshev's inequality:  $P(|X - \mu| \geq t\sigma) \leq 1/t^2$ , where  $\mu = E[X]$ ,  $\sigma = \sqrt{\text{Var}[X]}$ .
  - Applies for any random variable  $X$  (with finite  $\mu, \sigma$ ), and any  $t$  ( $t > 1$  for a meaningful bound).
  - Requires both expectation and variance of  $X$ .
  - Is an upper bound to a two-sided probability.
- Central Limit Theorem:  $(X - \mu_X)/\sigma_X \approx \text{Normal}(0, 1)$ , where  $X = X_1 + \dots + X_n$  are independent and have the same PDF/PMF,  $\mu_X = E[X] = n E[X_i]$ ,  $\sigma_X = \sqrt{\text{Var}[X]} = \sqrt{n \text{Var}[X_i]}$ .
  - Applies for  $X$  being sum of many (usually  $n \geq 30$ ) independent random variables; no restriction on distribution of  $X_i$ 's.
  - Requires both  $E[X_i]$  and  $\text{Var}[X_i]$ .
  - Approximates the CDF of  $X$ . Using the axioms of probability, we can use it to approximate other events of interest (e.g.  $P(X < -5 \text{ or } X > 7)$ ).

Roughly speaking, in terms of generality, Markov's inequality  $>$  Chebyshev's inequality  $>$  CLT; and in terms of tightness, Markov's inequality  $<$  Chebyshev's inequality  $<$  CLT (if  $n$  is large enough).

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3. The following exam statistics are posted on the course website:

section	no. students	average grade	std. dev.
A	30	65	5
B	20	70	10

what can you say about the number of students whose exam grade was 30 or below?

**Solution:** Let  $X_A$  and  $X_B$  be the grade of a random student in section A and section B, respectively. The table tells us that  $\mu_A = E[X_A] = 65$ ,  $\sigma_A = \sqrt{\text{Var}[X_A]} = 5$ ,  $\mu_B = E[X_B] = 70$ ,  $\sigma_B = \sqrt{\text{Var}[X_B]} = 10$ . By Chebyshev's inequality, for a random student in section A,

$$P(X_A \leq 30) = P(X_A \leq \mu_A - 7 \cdot \sigma_A) \leq P(|X_A - \mu_A| \geq 7\sigma_A) \leq 1/49 \approx 0.0204.$$

Although we are only interested in the probability that  $X_A$  is 7 standard deviations smaller than its mean, Chebyshev's inequality only tells us the probability of the possibly larger event that  $X_A$  is either 7 standard deviations smaller or 7 standard deviations larger than its mean. This is already a tremendously small probability – about 2%.

Similarly, for a random student in section B,

$$P(X_B \leq 30) = P(X_B \leq \mu_B - 4 \cdot \sigma_B) \leq P(|X_B - \mu_B| \geq 4\sigma_B) \leq 1/16 \approx 0.00625.$$

Since there are 30 students in section A, at most  $1/49 \cdot 30$  students must have received 30 or below, so nobody got that kind of grade. In section B, at most  $1/16 \cdot 20$  students got 30 or below, so at most one student in the whole class could have received 30 or below on the exam.

**Alternative Solution:** Alternatively, we can first calculate the statistics for a random student  $X$  in the whole course and then apply Chebyshev's inequality to  $X$ . Let  $X$  be a random student and  $Y$  their section (using 1 and 2 for sections A and B). Then  $E[X|Y]$  takes value 65 with probability  $3/5$  and 70 with probability  $2/5$ .

By total expectation theorem,

$$\mu = E[X] = E[E[X|Y]] = 65 \cdot 3/5 + 70 \cdot 2/5 = 67.$$

By total variance theorem  $\text{Var}[X] = \text{Var}[\text{E}[X|Y]] + \text{E}[\text{Var}[X|Y]]$ , where

$$\text{Var}[\text{E}[X|Y]] = (65 - 67)^2 \cdot 3/5 + (70 - 67)^2 \cdot 2/5 = 6,$$

$$\text{E}[\text{Var}[X|Y]] = 5^2 \cdot 3/5 + 10^2 \cdot 2/5 = 55,$$

hence standard deviation of  $X$  is  $\sigma = \sqrt{61} \approx 7.8103$ . Chebyshev's inequality says

$$\text{P}(X \leq 30) = \text{P}(X \leq \mu - (37/\sqrt{61}) \cdot \sigma) \leq \text{P}(|X - \mu| \geq (37/\sqrt{61}) \cdot \sigma) \leq 61/37^2 \approx 0.0445,$$

so the number of students who got 30 or below is at most  $0.0445 \cdot 50 = 2.2225$ , so at most 2.

4. You are collecting donations for a charity. Each donor gives you \$10 with probability half and \$20 with probability half. Assuming donors are independent, estimate the probability that you have collected at least \$1200 after taking in 100 donations.

**Solution:** Let  $X$  be the total amount of money collected. We want to estimate  $\text{P}(X \geq 1200)$ .  $X$  is the sum of 100 random variables with the same PMF so we can use the Central Limit Theorem. We have

$$\mu = \text{E}[X] = 100 \cdot (10 \cdot 1/2 + 20 \cdot 1/2) = 1500$$

$$\sigma = \sqrt{\text{Var } X} = \sqrt{100 \cdot ((10 - 15)^2 \cdot 1/2 + (20 - 15)^2 \cdot 1/2)} = \sqrt{100 \cdot 25} = 50.$$

Therefore,

$$\text{P}(X \geq 1200) \approx \text{P}(X \geq \mu - 6\sigma) \approx \text{P}(N \geq -6) \approx 1 - 9.86 \cdot 10^{-10},$$

where  $N$  is a  $\text{Normal}(0, 1)$  random variable.

5. You randomly divide 48 boys and 48 girls into teams of equal size. Show that if you divide them into 12 teams of 8 then there are no same-sex teams with probability at least 90%.

**Solution:** For  $12 \geq i \geq 1$ , let  $X_i$  be the indicator variable that the  $i$ -th team consists of all boys or all girls, then  $X = \sum_{i=1}^{12} X_i$ . The  $X_i$ 's are not independent, so the Central Limit Theorem doesn't apply.

The probability that any given team is all-boys is

$$p = \frac{48}{96} \cdot \frac{47}{95} \cdots \frac{41}{89}$$

using the formula for conditional probabilities (the first member is a boy, the second member is a boy given the first one is etc.). As boys and girls are symmetric, the probability the team is same-sex is  $2p$ . By linearity of expectation,

$$\text{E}[X] = \text{E}[X_1] + \cdots + \text{E}[X_{12}] = 12 \cdot (2p) \approx 0.068.$$

At this point we can proceed in two ways. We can use Markov's inequality to conclude that  $\text{P}(X \geq 1) \leq \text{E}[X]/1 \approx 0.068$ , so the probability of having no same-sex teams is  $\text{P}(X = 0) \approx 1 - 0.068 = 0.932$ . This meets the requirement and we are done.

Alternatively, we can calculate  $\text{Var}[X]$  and apply Chebyshev's inequality, which could result in a better bound. This is feasible but a bit difficult since  $X_1, \dots, X_{12}$  are not independent so we need their covariances.