

STAT 88: Lecture 22

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Section 6.3: Markov's Inequality

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Warm up:

(a) State Markov's inequality.

(b) Is it possible that half of US flights have delay times at least 3 times the national average?

Last time

Upper bounds for tail probability:

Markov's inequality For a non-negative random variable X and a positive constant $c > 0$,

$$P(X \geq c) \leq \frac{E(X)}{c}.$$

Chebyshev's inequality For a random variable X with mean μ and SD σ and a positive constant $c > 0$,

$$P(|X - \mu| \geq c) \leq \frac{\sigma^2}{c^2} = \frac{\text{Var}(X)}{c^2}.$$

One tail bound:

$$P(X - \mu \geq c) \leq \frac{\sigma^2}{c^2} = \frac{\text{Var}(X)}{c^2}.$$

Alternative formula for $\text{Var}(X)$:

- $\text{Var}(X) = E((X - \mu)^2)$.
- $\text{Var}(X) = E(X^2) - E(X)^2$. This implies $E(X^2) = E(X)^2 + \text{Var}(X)$.

Example: Suppose that each year, Berkeley admits 15,000 students on average, with an SD of 5,000 students. Assuming that the application pool is roughly the same across years, find the smallest upper bound you can on the probability that Berkeley will admit at least 22,500 students in 2019.

Example: Suppose a list of numbers $x = \{x_1, \dots, x_n\}$ has mean μ and standard deviation σ . Let k be the smallest number of standard deviations away from μ we must go to ensure the range $(\mu - k\sigma, \mu + k\sigma)$ contains at least 50% of the data in x . What is k ?

Example: A list of non negative numbers has an average of 1 and an SD of 2. Let p be the proportion of numbers over 4. To get an upper bound for p , you should:

- (a) Assume a binomial distribution.
- (b) Use Markov's inequality.
- (c) Use Chebyshev's inequality.
- (d) None of the above.

Example: Let X be a non negative random variable such that $E(X) = 100 = \text{Var}(X)$.

(a) Can you find $E(X^2)$ exactly? If not what can you say?

(b) Can you find $P(70 < X < 130)$ exactly? If not what can you say?

7.1. Sums of Independent Random Variables

We know that $E(X + Y) = E(X) + E(Y)$ but does $\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y)$?

Let X be the number of hours a student is awake a day and let Y be the number of hours a student is asleep a day. Then $X + Y = 24$, so trivially

$$\text{Var}(X + Y) = \text{Var}(24) = 0 \neq \text{Var}(X) + \text{Var}(Y).$$

So when X and Y are dependent, it is possible that $\text{Var}(X + Y) \neq \text{Var}(X) + \text{Var}(Y)$.

Theorem If X and Y are independent,

$$\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y).$$

Ex: Let X_1, X_2, \dots, X_n be a i.i.d. random sample with mean μ and SD σ . Let $S_n = \sum_{i=1}^n X_i$. Then

$$\text{Var}(S_n) = \text{Var}(X_1) + \text{Var}(X_2) + \dots + \text{Var}(X_n) = n\sigma^2.$$

So, $\text{SD}(S_n) = \sigma\sqrt{n}$.

SD of Binomial Let $X \sim \text{Bernoulli}(p)$. Then

$$\text{Var}(X) = E(X^2) - E(X)^2 = p - p^2 = p(1 - p).$$

Now let $X \sim \text{Binomial}(n, p)$. Write X as a sum of n Bernoulli random variables and find $\text{SD}(X)$.

SD of Poisson Recall that Binomial(n, p) can be approximated by Poisson(np) for large n and small p .

$$\text{Binomial}(n, p) \xrightarrow[\substack{n \rightarrow \infty \\ p \rightarrow 0 \\ np \rightarrow \mu}]{\quad} \text{Poisson}(\mu = np)$$

So we can find SD of Poisson(μ) from limit of SD of Binomial(n, p) as $n \rightarrow \infty, p \rightarrow 0, np \rightarrow \mu$.

$$\sqrt{npq-p} \xrightarrow[\substack{n \rightarrow \infty \\ p \rightarrow 0 \\ np \rightarrow \mu}]{\quad} \sqrt{\mu}$$

$$\Rightarrow \text{If } X \sim \text{pois}(\mu), \\ \text{SD}(X) = \sqrt{\mu}$$

SD of Geometric Fact: If $X \sim \text{Geom}(p)$,

$$\text{SD}(X) = \frac{\sqrt{1-p}}{p}.$$

Ex: (Waiting till the 10th success) Suppose you roll a die until the 10th success. Let R be the number of rolls required. Find $\text{SD}(R)$.