Expectation Continued: Tail Sum, Coupon Collector, and Functions of RVs

CS 70, Summer 2019

Lecture 20, 7/29/19

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Last Time...

- Expectation describes the weighted average of a RV.
- ► For more complicated RVs, use linearity

Today:

- Proof of linearity of expectation
- The tail sum formula
- Expectations of Geometric and Poisson
- Expectation of a **function** of an RV

Sanity Check

Let X be a RV that takes on values in A. Let Y be a RV that takes on values in B. Let $c \in \mathbb{R}$ be a constant.

Both $c \cdot X$ and X + Y are also RVs! <u>cX</u> <u>X+Y</u> <u> $Yaiues</u>: {<math>c \cdot a, a \in A$ } <u>Yaives</u>: { $a+b, a \in A, b \in B$ } <u>Probs</u>:<u>P[cX=ca]=IP[X=a]</u> <u>IP[X+Y=a+b]=P[X=a, Y=b]</u></u></u>

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Proof of Linearity of Expectation I

Recall linearity of expectation:

 $\mathbb{E}[X_1 + \ldots + X_n] = \mathbb{E}[X_1] + \ldots + \mathbb{E}[X_n]$ For **constant** c, $\mathbb{E}[cX_i] = c \cdot \mathbb{E}[X_i]$ Xi values in A First, we show $\mathbb{E}[cX_i] = c \cdot \mathbb{E}[X_i]$: $\mathbb{E}[cX_i] = \sum_{a \in A} (ca) \mathbb{P}[x=a]$ $= C \sum_{i=1}^{n} a \cdot P[X=a] = C E[X_i]$ aeA HE[Xi]

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Proof of Linearity of Expectation II



The Tail Sum Formula

Let X be a RV with values in $\{0, 1, 2, ..., n\}$. integer We use **"tail"** to describe $\mathbb{P}[X \ge i]$.

What does $\sum_{i=1}^{\infty} \mathbb{P}[X \ge i]$ look like?

Small example: X only takes values $\{0, 1, 2\}$: $\sum_{i=1}^{2} \mathbb{P}[X \ge i] = \mathbb{P}[X \ge 1] + \mathbb{P}[X \ge 2]$ $= \mathbb{P}[X = i] + \mathbb{P}[X = 2] + \mathbb{P}[X = 2]$ $0 \cdot \mathbb{P}[X = 0] + 1 \cdot \mathbb{P}[X = 1] + 2 \cdot \mathbb{P}[X = 2]$ $\mathbb{E}[X]$

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The Tail Sum Formula

The **tail sum formula** states that: $\mathbb{E}[X] = \sum_{i=1}^{\infty} \mathbb{P}[X \ge i]$ integers

Proof: Let $p_i = \mathbb{P}[X = i]$. $\sum_{i=1}^{\infty} \mathbb{P}[X \ge i] = \mathbb{P}[X \ge i] + \mathbb{P}[X \ge 2] + \mathbb{P}[X \ge 3] + \mathbb{P}[X \ge 3] + \mathbb{P}[X \ge 4] + \mathbb{$ $0 \cdot P_{2} + 1 \cdot P_{1} + 2 \cdot P_{2} + 3 \cdot P_{3} + \dots$ = IE[X]イロト イヨト イヨト イヨト 三日

Expectation of a Geometric I

Let $X \sim \text{Geometric}(p)$. $\mathbb{P}[X \ge i] = (1 - p)^{i-1}$

Apply the tail sum formula:

$$\sum_{i=1}^{\infty} P[x \ge i] = 1 + (1-p) + (1-p)^{2} + \dots$$

First kerm Geometric series
 $a = 1, r = (1-p)$
 $= \frac{a}{1-r} = \frac{1}{1-(1-p)} = (\frac{1}{p})$

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Expectation of a Geometric II X~Geom(p) Use memorylessness: the fact that the geometric RV "resets" after each trial. I-P faul 1st trial. failed first trial. "Win in 1+1E[X] days" "respt" down start trials Two Cases: succeed "win in 1 day." "reset," days looks identical to day 1 E[X] = p(1) + (I-p)[1 + E[X]]Solve For IE[X] = pt I-p E[X]=占

Expectation of a Geometric III

Lastly, an intuitive but **non-rigorous** idea. $\chi \sim Glom(\rho)$

Let X_i be an indicator variable for success in a single trial. Recall trials are **i.i.d.**

 $X_i \sim \text{Ber}(\mathbf{p})$ $\mathbb{E}[X_1 + X_2 + \ldots + X_k] = \mathbb{E}[X_1] + \mathbb{E}[X_2] + \ldots + \mathbb{E}[X_n]$ = KE[X] = KPNeed K= = in order for IE[X,+...+Xk]=1 # of successes.

Coupon Collector I

(Note 19.) I'm out collecting trading cards. There are n types total. I get a random trading card every time I buy a cereal box.

What is the **expected number of boxes** I need to buy in order to get all *n* trading cards?



Coupon Collector II Let $X_i = \frac{1}{2} \frac{$ What is the dist. of X_2 ? $X_2 \sim \text{Geom}\left(\frac{n-1}{n}\right)$ What is the dist. of X_3 ? $\chi_3 \sim \operatorname{Geom}(\frac{n-2}{n})$ In general, what is the dist. of X_i ? $\chi_i \sim \operatorname{Geom}(\frac{n-(i-1)}{n})$

Coupon Collector III Let X = total # box co to get all m cards $X = \chi_1 + \chi_2 + \ldots + \chi_n$ linearity) $\mathbb{E}[X] = \mathbb{E}\left[X_1 + X_2 + \dots + X_n\right]$ = $IE[X_1] + IE[X_2] + ... + IE[X_n]$ $\chi_i \sim \text{Geom}\left(\frac{n-(i-1)}{n}\right)$ $= 1 + \frac{n}{n-1} + \frac{n}{n-2} + \dots + \frac{n}{1}$ $n = n \left(\sum_{i=1}^{n} \frac{1}{i} \right) e^{-??}$ イロト 不得 トイラト イラト 二日

Aside: (Partial) Harmonic Series Harmonic Series: $\sum_{k=1}^{\infty} \frac{1}{k}$ Diverges Approximation for $\sum_{k=1}^{n} \frac{1}{k}$ in terms of *n*? $\sum_{k=1}^{n} \frac{1}{k} \approx \int_{1}^{n} \frac{1}{x} dx$ = $\ln x \Big|_{1}^{n} = \ln n - \ln 1^{\circ}$ $r \simeq \ln n$ \Rightarrow coupon collector: $E[X] \approx n \log n$. 14/26

Break

A Bad Harmonic Series Joke...

A countably infinite number of mathematicians walk into a bar. The first one orders a pint of beer, the second one orders a half pint, the third one orders a third of a pint, the fourth one orders a fourth of a pint, and so on.

The bartender says ...

Expectation of a Poisson I

Recall the Poisson distribution: values 0, 1, 2, ..., $\chi \sim Poi(\lambda)$ $\mathbb{P}[X = i] = \frac{\lambda^{i}}{i!}e^{-\lambda}$

We can use the **definition** to find $\mathbb{E}[X]$! $\mathbb{E}[X] = \sum_{i=0}^{\infty} \dot{x} \left(\frac{\lambda^{i}}{V} e^{-\lambda} \right)$ Toy bornes for $\frac{\lambda^{i-1}}{(i-1)!}$

Expectation of a Poisson II

Optional but intuitive / non-rigorous approach:

Think of a Poisson(λ) as a Bin($n, \frac{\lambda}{n}$) distribution, taken as $n \to \infty$.

Let $X \sim Bin(n, \frac{\lambda}{n})$. $\mathbb{E}[X] = np$ $= \mathcal{H}(\frac{\lambda}{n}) = \lambda$

Rest of Today: Functions of RVs!

Recall X from Lecture 19:

$$X = \begin{cases} 1 & \text{wp } 0.4 \\ \frac{1}{2} & \text{wp } 0.25 \\ -\frac{1}{2} & \text{wp } 0.35 \end{cases}$$

Refresh your memory: What is X^2 ?

$$\chi^{2} = \begin{cases} 1 & \text{wp. 0.4} \\ \frac{1}{4} & \text{wp 0.25} = \begin{cases} 1 & \text{wp 0.4} \\ \frac{1}{4} & \text{wp 0.35} \end{cases}$$

Example: Functions of RVs



What is $\mathbb{E}[X^2]$? IE[x2]= 1·P[X2=1]+ 年1P[X2=日] = 1.0.4+ 五.0.6=0.55 What is $\mathbb{E}[3X^2 - 5]$? unearly of exp. $E[3X^2] - E(S] = 3IE[X^2] - 5$ = 3(055) - 5

In General: Functions of RVs

Let X be a RV with values in A. Distribution of f(X):

 $f(X) = \begin{cases} f(a) & \text{wp } F[X = a] \\ \vdots \\ g(x) &= \\ g(x) &$

Square of a Bernoulli

Let $X \sim \text{Bernoulli}(p)$. Write out the distribution of X.

 $X = \begin{cases} 1 & wp \ P \\ 0 & wp \ I-p \end{cases}$

What is X^2 ? $\mathbb{E}[X^2]$?

$$\chi^{2} = \begin{cases} 1 & wp \ P \\ 0 & wp \ I - p \end{cases} \quad \mathbb{E}[\chi^{2}] = p.$$

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Product of RVs Exercise

Let X be a RV with values in A. Let Y be a RV with values in B.

XY is also a RV! What is its distribution? (Use the **joint distribution!**)

Product of Two Bernoullis Exercise

Let $X \sim \text{Bernoulli}(p_1)$, and $Y \sim \text{Bernoulli}(p_2)$. X and Y are **independent**.

What is the distribution of XY?

What is $\mathbb{E}[XY]$?

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Square of a Binomial I

Let $X \sim Bin(n, p)$. Decompose into $X_i \sim Bernoulli(p)$.

 $X = \chi_{1} + \chi_{2} + \ldots + \chi_{n}$ $\mathbb{E}[X] = \mathbb{E}[\chi_{1} + \chi_{2} + \ldots + \chi_{n}]$ $= \mathbb{E}[\chi_{1} + \mathbb{E}[\chi_{2}] + \ldots + \mathbb{E}[\chi_{n}]$ = MP.

Square of a Binomial II

Recall,
$$\mathbb{E}[X_i^2] = P$$
 and $\mathbb{E}[X_iX_j] = p^2$.
 $\mathbb{E}[X^2] = \mathbb{E}[(X_1 + X_2 + ... + X_n)^2]$
 $= \mathbb{E}[(X_1^2 + X_2^2 + ... + X_n^2) + (X_1X_2 + X_1X_3 + ...)]$
Square terms cross terms
 $n \text{ of them}$ $n(n-1) \text{ of}$
 $= \mathbb{E}[Square \text{ terms}] + \mathbb{E}[cross \text{ terms}] \text{ them.}$
 $= n \mathbb{E}[X_i^2] + n(n-1) \mathbb{E}[X_1X_2]$
 $X_i^2 = \begin{cases} 1 & \text{MP P} \\ 0 & 1-p & X_1X_2 = \begin{cases} 1 & p^2 \\ 0 & 1-p^2 & \text{E}[X_1X_2] = p^2 \end{cases}$
 $= n P + n(n-1)P^2$

Summary

Today:

- Proof of linearity of expectation: did not use independence, but did use joint distribution
- Tail sum for non-negative int.-valued RVs!
- Coupon Collector: break problem down into a sum of geometrics.
- Expectation of a function of an RV: can apply definition and linearity of expectation (after expanding) as well!!