

Stat 394 Probability I

Lecture 5

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July 4, 2017

Example

- Toss a coin 10 times and let X be the number of Heads.
- Choose a random point in the unit square $\{(x, y) : 0 \leq x, y \leq 1\}$ and let X be its distance from the origin.
- Choose a random person in a class and let X be the height of the person, in inches.

Random variables

A **random variable** is a number whose value depends upon the outcome of a random experiment.



A random variable X is a real-valued function on sample space S :

$$\underline{X : S \rightarrow \mathcal{R}} .$$

Example

Toss 2 coins. $X = \# \text{ of } H's.$

$\{HH\}, \{HT\}, \{TH\}, \{TT\}.$

$\downarrow \quad \downarrow \quad \downarrow \quad \downarrow$
 $2 \quad 1 \quad 1 \quad 0$

$$P(X=0) = \frac{1}{4}$$

$$P(X=x) \xrightarrow{s \in S} P(s) \text{ where } X(s)=x$$



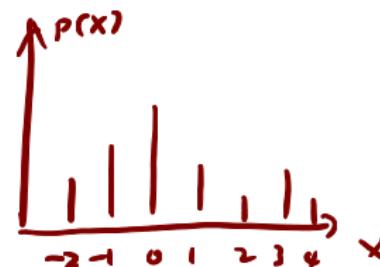
correspond to some events.

Discrete random variables

A **discrete random variable** X has finitely or countably many values $x_i, i = 1, 2, \dots$

$p(x_i) = P(X = x_i)$ with $i = 1, 2, \dots$ is called the **probability mass function** (or p.m.f) of X .

$$\sum_{l=1}^L p(x_l) = 1$$



Using r.v. to solve probability problems

4 balls are to be randomly selected without replacement from an urn containing 10 balls numbered 1 through 10. If we bet that at least one of the balls that are drawn has a number as large as or larger than 7, what is the probability that we win the bet?

X : largest # selected

$$P(X=i) = \frac{\binom{i-1}{3}}{\binom{10}{4}} \quad i = 4, 5, 6, \dots, 10$$

$$\begin{aligned} P(X \geq 7) &= P(X=7) + P(X=8) + P(X=9) + P(X=10) \\ &= \dots \end{aligned}$$

Bernoulli random variables

Suppose an experiment where the outcome can be considered as either a *success* or a *failure*, let $X = 1$ when outcome is success and $X = 0$ when outcome is failure, then X is said to be a **Bernoulli random variable** with parameter p .

p.m.f.

$$\begin{cases} p(0) = 1 - p \\ p(1) = p \end{cases}$$

Def 2.

a r.v. is a Bernoulli r.v if its p.m.f is as on the left.

And we can say $X \sim \text{Bern}(p)$

Binomial random variables

- $X \sim \text{Bern}(p) \Leftrightarrow X \sim \text{Bin}(n, p)$
- Example : Toss a fair coin 3 times, let X denote # of heads $\Rightarrow X \sim \text{Bin}(3, 1/2)$

Suppose for n independent trials, each of which results in a *success* with probability p and *failure* with probability $1 - p$, and let X represent the number of success that occur in the n trials, then X is said to be a **binomial random variable** with parameter (n, p) .

$$(X \sim \text{Bin}(n, p))$$

p.m.f : $p(i) = P(X=i) = \binom{n}{i} p^i (1-p)^{n-i}, i=0, 1, \dots, n$

We can check binomial formula.

$$\sum_{i=0}^n p(i) = \sum_{i=0}^n \binom{n}{i} p^i (1-p)^{n-i} = (p + (1-p))^n = 1$$

Bernoulli and Binomial distribution

$X \sim \text{Bin}(n, p)$

- X is # of successes out of n trials with success prob p

- $X = X_1 + X_2 + \dots + X_n$, $X_i = \begin{cases} 1 & \text{if trial } i \text{ is success,} \\ 0 & \text{otherwise.} \end{cases}$

i.e. $X_1, X_2, \dots, X_n \stackrel{i.i.d}{\sim} \text{Bern}(p)$

↳ independently and identically distributed

Cumulative distribution function (CDF)

Def.

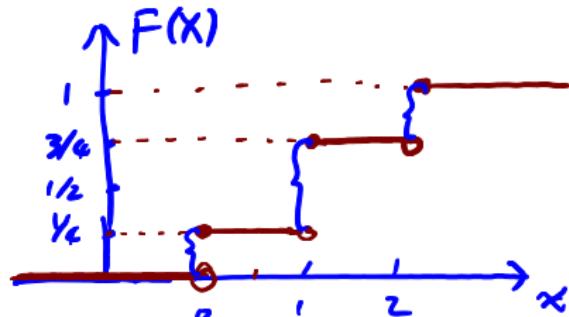
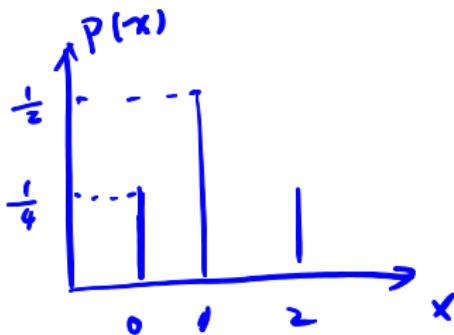
$$F(x) = P(X \leq x), \quad -\infty < x < \infty$$



event.

$$-\lim_{x \rightarrow -\infty} F(x) = 0$$

$$-\lim_{x \rightarrow \infty} F(x) = 1$$



$$F(0) = P(X \leq 0) = P(X=0) = \frac{1}{4}$$

$$F(0.5) = P(X \leq 0.5) = P(X=0)$$

Random variables
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Expectation and variance
oooooooooooooooooooo

Example

$$X \sim \text{Bin}(2k-1, p)$$

Example

Let X : # of functional comp. in the first $(2k-1)$ comp.

A communication system consists of n components, each of which will, independently, function with probability p . The total system will be able to operate effectively if at least one-half of its components function. In general, when is a $(2k+1)$ -component system better than a $(2k-1)$ -component system?

Ans: when $p > \frac{1}{2}$

$$P_1 = P(\text{2k+1 system is effective}) = P(X \geq k+1) = P(X=k) \cdot (1-(1-p)^2) + P(X=k-1) \cdot p^2$$

⇒ at least $k+1$ comp. function

$$P_2 = P(\text{2k-1 system is effective}) = P(X \geq k) = P(X=k) + P(X \geq k+1)$$

$$\begin{aligned} P_1 - P_2 &= P(X=k) (1-p)^2 + P(X=k-1) \cdot p^2 \\ &= \binom{2k-1}{k} p^k (1-p)^{k-1} \cdot (1-p)^2 + \binom{2k-1}{k-1} p^{k-1} (1-p)^k \cdot p^2 \\ &= \binom{2k-1}{k} [-p^k (1-p)^{k+1} + p^{k+1} (1-p)^k] - \binom{2k-1}{k} p^k (1-p)^k [-(1-p) + p] \end{aligned}$$

First, why do we want to study random variables

A quote:

"Once you get what a random variable is, it can be hard to explain.
Now that I understand what a random variable is, it is difficult to remember what was difficult to understand about it"

First, why do we want to study random variables

But seriously, the way I see the advantage of using random variables

- Describe more complicated events without cumbersome notations
- Focus on the quantities we care about
- Consider problems beyond calculating probabilities
- Handle continuity and infinity more easily

move on 395



Expectation

Assume X is a discrete random variable with possible values $x_i, i = 1, 2, \dots$. Then the **expected value**, or **expectation**, of X is

$$E(X) \text{ or } EX = \sum_i x_i p(x_i)$$

X : # H for 2 tosses

$$p(0) = 1/4$$

$$p(1) = 1/2$$

$$p(2) = 1/4$$

$$E(X) = 0 \times 1/4 + 1 \times 1/2 + 2 \times 1/4$$

$$= 1/2 + 2/4 = 1$$

Example

A box contains 5 red balls and 3 blue balls. Two balls are withdrawn randomly, and if they are of the same color, you win \$2, otherwise you lose \$1. What is the amount you expect to win?

X : amount of \$ you walk away with.

$$P(X=2) = \frac{\binom{5}{2} + \binom{3}{2}}{\binom{8}{2}} = \frac{13}{28}$$

$$P(X=-1) = 1 - \frac{13}{28} = \frac{15}{28}$$

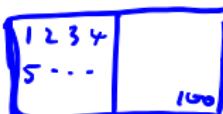
$$E(X) = 2 \times \frac{13}{28} + (-1) \times \frac{15}{28} = \frac{11}{28}$$

Chocolate store

A candy store has M boxes of chocolates. The number of chocolates in each box vary. Suppose for $k = 1, 2, \dots, N$, there are n_k boxes containing k chocolates, and $\sum_{k=1}^N n_k = M$. The store owner keeps a magic book where all the chocolates labeled $1, 2, \dots, \sum_{k=1}^N kn_k$ are listed.

   ...  → 100 chocolates

The store owner lets you to take one box of chocolate for free, and gives you two options:



1. Take one box without knowing how many chocolates are in the box.
 $X_1 : \# \text{ of } c \text{ you will get from opt 1.}$
2. Pick one chocolate from the book, and you can take the box containing that chocolate.
 $X_2 : \# \text{ you get from opt 2.}$

Chocolate store

$$P(X_1 = i) = \frac{n_i}{M}$$

$$P(X_2 = i) = \frac{\# \text{ chocolates in boxes containing } i \text{ chocolates}}{\# \text{ chocolates}}$$

$$= \frac{i \cdot n_i}{\sum_{k=1}^N k n_k}$$

$$E(X_1) = \sum_{i=1}^N i \cdot \frac{n_i}{M}$$

$$E(X_2) = \sum_{i=1}^N i \cdot \frac{i n_i}{\sum_{k=1}^N k n_k}$$

$$E(X_1)/E(X_2) < 1$$

pf omitted.

Expectation of function of a r.v.

$\{y_1, y_2, \dots\}$ denotes the possible values of $g(X)$

Assume X is a discrete random variable with possible values $x_i, i = 1, 2, \dots$. Then the **expected value**, or **expectation**, of $g(X)$, where g is some real-valued function, is

$$E(g(X)) \text{ or } Eg(X) = \sum_i g(x_i)p(x_i)$$

Pf.

$$\begin{aligned} \sum_i g(x_i)p(x_i) &= \sum_j \left(\sum_{i: g(x_i)=y_j} y_j p(x_i) \right) \\ &= \sum_j y_j \left(\sum_{i: g(x_i)=y_j} p(x_i) \right) \\ &= \sum_j y_j \underbrace{P(g(x)=y_j)}_{P(g(x)=y_j)} = E(g(x)) \end{aligned}$$



Example

Roll a die. When you roll k , you will win $(k^2 - 9)$ dollars.

X : value you roll

$$\begin{aligned} E(X^2 - 9) &= \sum_{i=1}^6 (i^2 - 9) \cdot p(i) \\ &= \frac{1}{6} ((1^2 - 9) + (2^2 - 9) + \dots + (6^2 - 9)) \\ &= \frac{37}{6} \end{aligned}$$

Variance

If X is a random variable, we say the **variance** of X is defined by

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

and the square root of the variance is usually called the **standard deviation** of X

Example

A box contains 5 red balls and 3 blue balls. There are two games offered to you. Both games ask you to take one ball out,

- Game 1: if it is a red ball, you win \$2, otherwise you lose \$2.
- Game 2: if it is a red ball, you win \$5, otherwise you lose \$7.

$$E(Y_1) = \frac{5}{8} \times 2 + \frac{3}{8} (-2) = \frac{9}{8} = \frac{1}{2}$$
$$E(Y_2) = \frac{5}{8} \times 5 + \frac{3}{8} (-7) = \frac{4}{8} = \frac{1}{2}$$

Return



$$\text{Var}(X_1) = E((X_1 - E(X_1))^2) = \frac{5}{8} \times (2 - \frac{1}{2})^2 + \frac{3}{8} \times (-2 - \frac{1}{2})^2 = 3.75$$

$$\text{Var}(X_2) = E((X_2 - E(X_2))^2) = \frac{5}{8} \times (5 - \frac{1}{2})^2 + \frac{3}{8} \times (-7 - \frac{1}{2})^2 = 33.75$$

Alternative formula for variance

$$\text{Var}(X) = E((X - E(X))^2) = E(X^2) - [E(X)]^2$$

But.

$$= \sum_x (x - E(X))^2 p(x)$$


$$= \sum_x (x^2 - 2x E(X) + E(X)^2) p(x)$$

$$= \sum_x x^2 p(x) - \sum_x 2x E(X) p(x) + \sum_x E(X)^2 p(x)$$

$$= \sum_x x^2 p(x) - 2E(X) \underbrace{\sum_x x p(x)}_{= 1} + E(X)^2 \sum_x p(x)$$

$$= E(X^2) - 2E(X) \cdot \underline{E(X)} + E(X)^2 \cdot 1$$

$$= E(X^2) - E(X)^2$$

Expectation and variance under linear transformation

$$X, aX + b$$

$$\underline{E(aX+b) = aE(X)+b}$$

Pf. LHS = $\sum_{x_i} (ax_i + b) p(x_i) = \sum_{x_i} ax_i p(x_i) + b p(x_i)$
 $= a \sum_{x_i} x_i p(x_i) + b \sum_{x_i} p(x_i)$
 $= a E(X) + b$

$$\underline{\text{Var}(aX+b) = a^2 \text{Var}(X)}$$

Pf. LHS = $E((aX+b - E(aX+b))^2)$

$= E((aX+b - (aE(X)+b))^2)$

$= E((a(X-E(X)))^2)$

$= E(\underline{a^2(X-E(X))^2})$

$= a^2 E((X-E(X))^2)$

$= a^2 \text{Var}(X)$

Binomial random variable revisit

$$X \sim \text{Bin}(n, p)$$

$$P(X) = \binom{n}{x} p^x (1-p)^{n-x}, \quad x=0, 1, 2, \dots, n$$

$$E(X) = np$$

$$\text{Pf. } E(X) = \sum_{x=0}^n x \binom{n}{x} p^x (1-p)^{n-x}$$

$$= \sum_{x=1}^n \underbrace{\alpha \binom{n}{x}}_{\text{Term 1}} p^x (1-p)^{n-x} + \underbrace{\alpha \cdot \binom{n}{0} p^0 (1-p)^{n-0}}_{\text{Term 2}}$$

Let $x' = x^{-1}$
 $n' = n-1$

$$= \sum_{x=1}^n n \binom{n-1}{x-1} p^x (1-p)^{n-x}$$

$$= \sum_{x=1}^{n'} (n'+1) \binom{n'}{x} p^{x'+1} (1-p)^{n'-x'}$$

$$= (n'+1) \sum_{x'=0}^{n'} \binom{n'}{x'} p^{x'+1} (1-p)^{n'-x'}$$

x from 1 to n
 x' from 0 to $n-1$
 x' from 0 to n'

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$$= (n+1) \sum_{k=0}^n \binom{n}{k} p^k (1-p)^{n-k}$$

$$= (n^t + 1) p$$

$$= np$$

Binomial random variable revisit

$$\text{Var}(X) = np(1-p)$$

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

$$E(X^2) = \sum_{x=0}^n \underbrace{x^2 \binom{n}{x}}_{\text{green}} p^x (1-p)^{n-x}$$

$$= \sum_{x=1}^n \underbrace{n x \binom{n-1}{x-1}}_{\text{green}} p^x (1-p)^{n-x}$$

$$= np \sum_{x=1}^n x \binom{n-1}{x-1} p^{x-1} (1-p)^{n-x} \quad \leftarrow \begin{cases} x' = x-1 \\ n' = n-1 \end{cases}$$

$$= np \sum_{x'=0}^{n'} (x'+1) \binom{n'}{x'} p^{x'} (1-p)^{n'-x'}$$

$$= np \sum_{x'=0}^{n'} x' \binom{n'}{x'} p^{x'} (1-p)^{n'-x'} + np \sum_{x'=0}^{n'} \binom{n'}{x'} p^{x'} (1-p)^{n'-x'}$$

$$= np((n-1)p) + np(1) = n^2 p^2 + np(1-p)$$

$$Y \sim \text{Bin}(n', p)$$

$$E(Y) = n'p$$

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

Example

Let X be the number of Heads in 50 tosses of a fair coin. Determine $E(X)$, $Var(X)$ and $P(X \leq 10)$.

$$X \sim \text{Bin}(50, 1/2)$$

$$E(X) = 50 \times \frac{1}{2} = 25$$

$$\text{Var}(X) = 50 \times \frac{1}{2} \times (1 - \frac{1}{2}) = 12.5$$

$$\begin{aligned} P(X \leq 10) &= \sum_{i=0}^{10} P(i) = \sum_{i=0}^{10} \binom{50}{i} \left(\frac{1}{2}\right)^i \left(\frac{1}{2}\right)^{50-i} \\ &\quad \text{↳ } F(10) \\ &= \sum_{i=0}^{10} \binom{50}{i} \left(\frac{1}{2}\right)^{50} \end{aligned}$$

Newsboy

Textbook 4b

A newsboy buy papers at 10 cents and sells them at 15 cents. However, he is not allowed to return unsold papers. If his daily demand is a binomial random variable with $n = 10$ and $p = \frac{1}{3}$, approximately how many papers should he purchase so as to maximize his expected profit?

Random variables
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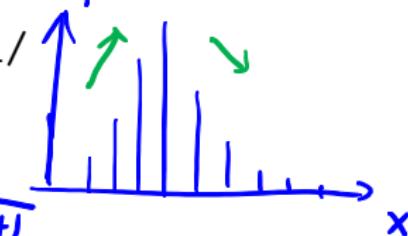
Expectation and variance
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Newsboy

If $X \sim \text{Bin}(n, p)$, which x maximizes $p(x)$
 Binomial p.m.f

<http://shiny.albany.edu/stat/binomial/>

$$\frac{P(X=k)}{P(X=k-1)} = \frac{\frac{n!}{k!(n-k)!} p^k (1-p)^{n-k}}{\frac{n!}{(k-1)!(n-k+1)!} p^{k-1} (1-p)^{n-k+1}}$$



$$= \frac{\frac{1}{k} \cdot p}{\frac{1}{n-k+1} \cdot (1-p)}$$

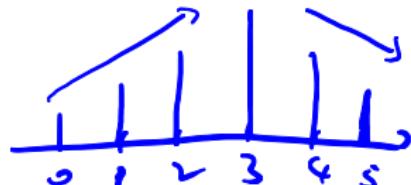
$$(*) > 1$$

$$\frac{p}{k} > \frac{1-p}{n-k+1}$$

$$\Rightarrow k < (n+1)p$$

e.g. $n=10, p=0.3$

$$k < 11 \times 0.3 = 3.3$$



$n=365$

Birthday problem revisit

X : the place in the line that's the first match.

$$P(X=x) = 1 \times \frac{n-1}{n} \times \frac{n-2}{n} \times \cdots \times \frac{n-(x-2)}{n} \times \frac{x-1}{n}$$
$$= \frac{x-1}{n^{x-1}} \times \frac{(n-1)!}{(n-(x-1))!}$$

$$\frac{P(X=x)}{P(X=x-1)} = \frac{\frac{x-1}{n^{x-1}} \times \frac{(n-1)!}{(n-(x-1))!}}{\frac{x-2}{n^{x-2}} \times \frac{(n-1)!}{(n-(x-2))!}}$$
$$= \cdots$$



$$\frac{P(X=x)}{P(X=x-1)} > 1 \Rightarrow x < \frac{19.61}{20.61}$$

Birthday problem revisit

