

Lecture 8: Random Variables and their distributions

Binomial Distribution $\text{Bin}(n, p)$

- (1) Story: X is # successes in n independent $\text{Bern}(p)$ trials
- n : any positive integer
 p : real number between 0 & 1

prob. of success

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

is a random variable that has this distribution

- (2) Sum of indicator random variables: $X = X_1 + X_2 + \dots + X_n$,

$$\text{where } X_j = \begin{cases} 1, & \text{if } j\text{th trial success} \\ 0, & \text{otherwise} \end{cases}$$

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

\hookrightarrow independent & identically distributed

Trials are independent and X_1, X_2, \dots, X_n are the indicators of success for each trial, so those should be independent.

identically — All of these X 's (i.e., X_1, X_2, \dots, X_n) have the same distribution. In other words, they are all $\text{Bern}(p)$.

- (3) PMF (What's the prob. that X takes on any particular value)

$$P(X=k) = \binom{n}{k} p^k q^{n-k}, \quad q = (1-p)$$

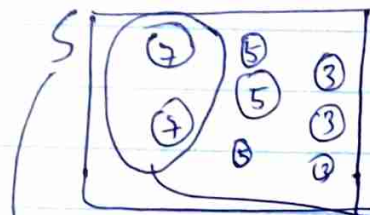
$k \in \{0, 1, \dots, n\}$

\downarrow
 n trials,
exactly k
successes

$X_1, X_2, \dots, X_n \rightarrow$ R.V.s
is mathematically
a function, but intuitively
 X_2 is 1 if first trial is
success, 0 otherwise that
depends on first trial.

Random Variables (R.V.s.)

different from distributions
Distribution — saying what is the prob. that X will behave in different ways. $X_1, X_2, \dots, X_n \rightarrow$ different random variables but all may have the same distribution.



→ Could be incredibly complicated possibly infinite space that we could never draw.

→ could be very high dimensional, infinite.

— Random variable is a function that assigns a number to each pebble (in pebble world case).

$X=7$ is an event

↳ subset of sample space

CDF (Cumulative distribution Function)

one way to describe a distribution

X is a function.

Here, $X=7$ is not a

$X \leq x$ is an event
 $F(x) = P(X \leq x)$,

telling prob. of different possible values of X

const. function, it's just a notation for an event that $X=7.6$

then F is called CDF of X .

PMF (for discrete random variables)

↳ possible values $a_1, a_2, a_3, \dots, a_n$ or a_1, a_2, \dots

$P(X=a_j)$, for all j
 $= P_j$

$$P_j \geq 0, \sum_j P_j = 1 \quad (\text{When does a PMF valid?})$$

$$P(X=k) = \binom{n}{k} p^k q^{n-k}, \quad q = 1-p$$

$$\left[\sum_{k=0}^n \binom{n}{k} p^k q^{n-k} = (p+q)^n = 1^n = 1 \right. \\ \left. \text{(By Binomial theorem)} \right]$$

$$\left. \begin{array}{l} X \sim \text{Bin}(n, p) \text{ [no. of successes in } n \text{ trials]} \\ Y \sim \text{Bin}(m, p) \text{ [No. of successes in } m \text{ trials]} \end{array} \right\} \begin{array}{l} \text{Independent} \\ \downarrow \\ \text{X, Y are} \\ \text{separate} \\ \text{set of trials} \end{array}$$

$$\Rightarrow \boxed{X+Y} \sim \text{Bin}(n+m, p)$$

Adding two functions.

$$(1) X = X_1 + X_2 + \dots + X_n$$

$$Y = Y_1 + Y_2 + \dots + Y_m, \quad X_j \text{ and } Y_j \text{ are indep. Bernoulli random variables}$$

$$X+Y = \sum_{j=1}^n X_j + \sum_{i=1}^m Y_i$$

$$\text{Sum of } (n+m) \text{ i.i.d. Bern}(p) \Rightarrow \text{Bin}(n+m, p)$$

$$(2) \text{ } \underbrace{P(X+Y=k)}_{\text{convolution}} = \sum_{j=0}^k P(X+Y=k | X=j) P(X=j) \quad \left[\begin{array}{l} \text{Condition on } X. \\ \text{Wishful thinking} \\ \text{It would be} \\ \text{easier if we} \\ \text{know the value of } X \end{array} \right]$$

check $X+Y$ is Binomial by computing its PMF

(Law of Total Prob.)

$$= \sum_{j=0}^k P(Y=k-j | X=j) \binom{n}{j} p^j q^{n-j}$$

we can also condition on Y

Binomial PMF

Since, X & Y are independent, so conditioning on X gives no information about Y . Since X is Binomial

$$\begin{aligned} &= \sum_{j=0}^k P(Y=k-j) \binom{n}{j} p^j q^{n-j} = \sum_{j=0}^k \binom{m}{k-j} p^{k-j} q^{m-k+j} \binom{n}{j} p^j q^{n-j} \\ &= p^k q^{m+n-k} \sum_{j=0}^k \binom{m}{k-j} \binom{n}{j} \quad \text{Vandermonde's identity} \\ &= p^k q^{m+n-k} \binom{m+n}{k} = \binom{m+n}{k} p^k q^{m+n-k} \end{aligned}$$

Example

5 card hand, find distribution of number of aces.

Solu:- Let $X = \# \text{ aces}$

find $P(X=K)$.

This is 0, except if

$K \in \{0, 1, 2, 3, 4\}$.

discrete problem as #aces
so, PMF (or, CDF) can be either
0, 1, 2, 3 or 4

- Distribution is not Binomial as trial is not independent, because we can think of each card as trial, but those trials are not independent because if first card is an ace, it is less likely that second card is an ace. The more aces we have in the earlier cards, it is less likely to have more aces.

PMF
just

by
thinking
about it)

$$P(X=K) = \frac{\binom{4}{K} \binom{48}{5-K}}{\binom{52}{5}}, \text{ for } K \in \{0, \dots, 4\}$$

(Like the elk problem)

↳ same as tagged elks & untagged elks

Example

~~the~~ Suppose we have a jar full of marbles. We have b black, w white marbles. Pick simple random sample of size n . Find distribution of number of white marbles in sample.

Solu:- $X = \# \text{ number of white marbles in sample}$

$$P(X=K) = \frac{\binom{w}{K} \binom{b}{n-K}}{\binom{w+b}{n}}$$

choose K white
marbles

$(w+b) \rightarrow \text{total population}$
 $n \rightarrow \text{sample}$

(Hyper
geometric
distribution)

$$0 \leq K \leq w, 0 \leq n-K \leq b$$

Not binomial (sampling without replacement)

Hypergeometric Distribution is different from

Binomial Distribution

↳ sampling with replacement

picking marble

without replacement

(sampling without replacement)

- When size of the population is very large, then both behave almost the same.

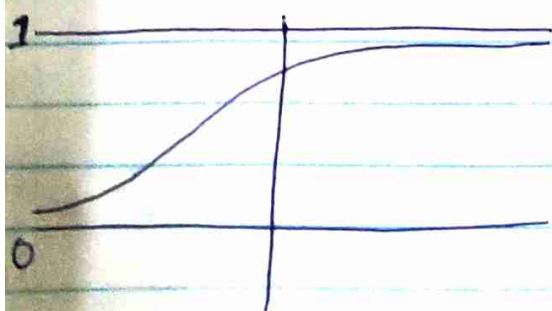
✓ - Suppose number of marbles is billion and our sample is very small compared to a billion (let's say 10). We are picking 10 marbles out of a billion, it is extremely unlikely that we pick the same marble more than 1. So, it must be that sampling with or without replacement should behave very similarly there.

$$\sum_{k=0}^W \frac{\binom{W}{k} \binom{b}{n-k}}{\binom{W+b}{n}} = \frac{1}{\binom{W+b}{n}} \sum_{k=0}^W \binom{W}{k} \binom{b}{n-k} = 1$$

VanderMonde's Identity

CDF $P(X \leq x)$

Continuous



discrete

