# Probability Theory

Applications for Data Science Module 3 Discrete Random Variables

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Discrete Random Variables

### Random Variables

At the end of this module, students should be able to

- Define a discrete random variable and give examples of a probability mass function and a cumulative distribution function.
- Calculate probabilities of Bernoulli, Binomial, Geometric, and Negative Binomial random variables.
- Calculate the expectation and variance of a discrete rv.

## Binomial random variable

# Key ideas: -n Bernoulli trials - same prob of success in each trial - indep. Bernoulli trials

### Examples:

- ► Suppose you toss a fair coin 12 times. What is the probability that you'll get 5 heads? ► = 1/2
- Suppose you pick a random sample of 25 circuit boards used in the manufacture of a particular cell phone. You know that the long run percentage of defective boards is 5%. What is the probability that 3 or more boards are defective? P=.o≤

These three situations, and many more, can be modeled by a binomial random variable.

Properties of a binomial random variable;

- $\triangleright$  Experiment is n trials (n is fixed in advance)
- Trials are identical and result in a success or a failure (i.e. Bernoulli trials) with P(success) = p and P(failure) = 1 p.
- ► Trials are independent (outcome of one trial does not influence any other)

If X is the number of successes in the n independent and identical trials, X is a binomial random variable. Notation:  $X \sim Bin(n, p)$  Find the pmf, expectation, and variance for a binomial random variable,  $X \sim Bin(n, p)$ .

What is the sample space for a binomial experiment?
$$S = \{(x_1, x_2, ..., x_n) \mid x_i = \} \text{ if successor the first } \} \text{ ISI = a}$$
but each clement decs not have equal prob
$$P(X = 0) = P(\{10 - 0, 010 - 0, ..., 000 - 01\}) = n P(1-p)^n$$

$$P(X = 1) = P(\{10 - 0, 010 - 0, ..., 000 - 01\}) = n P(1-p)^n$$

$$P(X = 2) = P(\{10 - 0, 010 - 0, ..., 000 - 01\}) = n P(1-p)^{n-2}$$

$$P(X = k) = P(\{10 - 0, 010 - 0, ..., 000 - 01\}) = n P(1-p)^{n-2}$$

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**Definition**: The expected value of a discrete random variable, E(X), is given by

$$E(X) = \sum_{k} kP(X = k)$$

$$X \sim Bin(n,p)$$

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

Recall: Bern (p), expected value is p.

**Definition**: The **variance** of a random variable is given by  $\sigma_X^2 = V(X) = E[(X - E(X))^2].$ 

Computational formula:  $V(X) = E(X^2) - (E(X))^2$ .

For 
$$X \sim Bin(n,p)$$
  
 $V(X) = \sum_{k} (k - E(X))^2 P(X=k)$   
 $= \sum_{k=0}^{\infty} (k - np)^2 (np) p^k (1-p)^{n-k}$   
 $= np (1-p)$   
variance of Bern(p)  
 $= \sum_{k=0}^{\infty} (1-p)^n - k$   
 $= \sum_{k=0}^{\infty} (1-p)^n - k$ 

## Negative binomial random variable

indep Bernoulli trials
until r successes
count & of fadures until r
successes.

#### Examples:

- ► Suppose you toss a fair coin until you obtain <u>5</u> heads. How many tails before the fifth head?
- ► Suppose you randomly choose circuit boards until you find 3 defectives. You know that the long run percentage of defective boards is 5%. How many must you examine?
- ➤ Suppose 40% of online purchasers of a particular book would like a new copy and 60% want a used copy. How many new books are sold before the fiftieth used book?

These three situations can be modeled by a **negative** binomial random variable.

Definition: Repeat independent Bernoulli trials until a total of r successes is obtained. The negative binomial random variable Y counts the number of failures before the  $r^{th}$  success. Notation:  $Y \sim NB(r, p)$ .

- The number of successes r is fixed in advance.
- Trials are identical and result in a success or a failure (i.e. Bernoulli trials) with P(success) = p and P(failure) = 1 p.
- ► Trials are independent (outcome of one trial does not influence any other)

Compare to  $X \sim Bin(n, p)$ : X is the number of successes in the n independent and identical trials and n is fixed in advance.

Example: A physician wishes to recruit 5 people to participate in a medical study. Let p=.2 be the probability that a randomly selected person agrees to participate. What is the probability that 15 people must be asked before 5 are found who agree to participate.

Y is the number of failures before the 5 people are found.  

$$S = \{(x, x_{2}, x_{3}, ...) \mid x_{i} = \{(x, x_{$$

pmf P(Y=k) = (k+r-1) pr(1-p), k=0,1,2,... Y~NB(r,p) Show: \$ P(Y=k)=1 E(Y)= r(1-12) V(Y)= r(1-p) Kelationship between geometrie v.v. + NB (r.p.) X~ (geom (p) = repeat indep, identical Bernoulli'
trials until the first success Yw NB(I,p) < counts the number of failures
before 1st success
Note: Yn X-1 Mote:  $Y \sim X - 1$   $E(Y) = E(X - 1) = E(X) - 1 = \frac{1 - P}{P}$ NB(r,p) \_\_\_\_\_\_\_i\_tsuccess