

# Probability and Statistics Notes

## Chapter Two

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- 1 Section 2.1: Discrete Random Variables
- 2 Section 2.2: Mathematical Expectation
- 3 Section 2.3: The Mean, Variance, and Standard Deviation
- 4 Section 2.4: Bernoulli Trials and the Binomial Distribution
- 5 Section 2.5: The Moment-Generating Function
- 6 Section 2.6: The Poisson Distribution

# Random Variables

## Definition

If  $A$  and  $B$  are sets, and  $f$  is a function from  $A$  to  $B$ , we write

$$f : A \rightarrow B.$$

## Definition

A *random variable*  $X$  is a function from the sample space  $S$  to the real numbers.

$$X : S \rightarrow \mathbb{R}$$

## Example

Two coins are flipped and the resulting sequence of heads/tails is noted. Let  $X$  be the number of heads in the sequence.

## Example

Assuming the coins are fair and independent, calculate  $P(X = 1)$  and  $P(X \geq 1)$ .

- $(X = 1)$  is shorthand for  $\{s \in S \mid X(s) = 1\} = \{HT, TH\}$
- $(X \geq 1)$  is shorthand for  $\{s \in S \mid X(s) \geq 1\} = \{HT, TH, HH\}$

## Definition

If  $A \subseteq \mathbb{R}$ , define the event

$$(X \in A) = \{s \in S \mid X(s) \in A\}.$$

## Example

Roll two independent fair dice, and let  $X$  be the sum of the rolls. Calculate  $P(X = x)$ , for  $x = 5, 6, 7, 8, 9$ .

## Definition

The *support* of a random variable  $X$  is the set of possible values of  $X$ ,

$$\text{supp}(X) = \{X(s) \mid s \in S\}$$

## Definition

A random variable is called *discrete* if its support is countable (is finite or can be put in one-to-one correspondence with the positive integers).

## Example

A fair coin is flipped until the result is heads, and  $X$  is the number of flips that occur.

- What is the support of  $X$ ?
- Is  $X$  a discrete random variable?

## Definition

The *probability mass function*  $f$  of a discrete random variable  $X$  is

$$f : \mathbb{R} \rightarrow [0, 1]$$

$$f(x) = P(X = x)$$

- abbreviated p.m.f.
- also called the probability *distribution* function or probability *density* function (p.d.f.)

## Proposition

A function  $f : \mathbb{R} \rightarrow [0, 1]$  is the p.m.f. of some random variable if and only if

- $f(x) \geq 0$ , for  $x \in \mathbb{R}$ , and
- 

$$\sum_{x \in \mathbb{R}} f(x) = 1.$$

## Example

Let  $X$  be a random variable with p.m.f.

$$f(x) = \begin{cases} cx^2, & x = 1, 2, 3, 4, 5 \\ 0 & \text{otherwise.} \end{cases}$$

Find  $c$ .

## Example

Let  $X$  be the number of aces in a five-card poker hand.

- Find the p.m.f. of  $X$ .
- Draw a probability histogram for  $X$ .
- The number of aces in each of ten poker hands is listed below:

0, 0, 0, 1, 0, 0, 2, 0, 1, 1

Draw a relative frequency histogram for this data on the same set of axes as the probability histogram.



## Definition (Hypergeometric Distribution)

Setting:

- Set of objects of two types
- $N$  = total number of objects
- $N_1$  = number of objects of the 1st type
- $N_2$  = number of objects of the 2nd type
- Select  $n$  objects randomly without replacement
- $X$  = number of objects in sample of 1st type
- The p.m.f. of  $X$  is

$$P(X = x) = \frac{\binom{N_1}{x} \binom{N_2}{n-x}}{\binom{N}{n}}.$$

- $X$  is said to have a *hypergeometric distribution*.

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## Example

When you buy a scratch-off lottery ticket, you have an 80% chance of winning nothing, a 15% chance of winning \$2, and a 5% chance of winning \$10. If the ticket costs \$1, should you buy one?

## Definition

Suppose  $X$  is a discrete random variable with p.m.f.  $f$ . Then the *expected value* of  $X$  is

$$E(X) = \sum_{x \in \mathbb{R}} xf(x),$$

assuming the series converges absolutely. Otherwise, the expected value does not exist.

## Example

Let  $X$  be the number of heads occurring when a fair coin is flipped 3 times.

- What is the expected value of  $X$ ?
- Find  $E(X^2 + 7X)$
- Find  $E(5X + 4)$

Suppose  $u : \mathbb{R} \rightarrow \mathbb{R}$ .

$$E(u(X)) = \sum_{x \in \mathbb{R}} u(x)f(x)$$

For random variables  $X$  and  $Y$  and a constant  $c$ ,

- $E(X + Y) = E(X) + E(Y)$
- $E(cX) = cE(X)$
- $E(c) = c$

# An Insurance Policy

## Example

An automobile insurance policy has a deductible of \$500. Let  $X$  be the cost of damages to a vehicle in an accident, and assume  $X$  has the following p.m.f.

$x$	0	250	500	1000	2000
$f(x)$	0.1	0.2	0.4	0.2	0.1

If an accident occurs, what is the expected value of the payment made by the insurance company?

# Expected Value for the Hypergeometric Distribution

## Example

In a club with a 100 members, 60 members approve of the president. In a random sample of size 5, let  $X$  be the number of people who approve of the mayor. Find the expected value of  $X$ .

Let  $X$  have a hypergeometric distribution where

- $N_1$  = number of objects of type 1
- $N$  = total number of objects
- $n$  = sample size

$$E(X) = \frac{N_1}{N}n$$

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$$E(X) = \frac{N_1}{N}n$$

### Example

If you have 10 red pens and 4 blue pens, and you select 6 pens at random, what is the expected value of the number of blue pens in your sample?

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- $E(X)$  = expected value of  $X$
- $E(X)$  = “average value” of  $X$
- $E(X)$  also called the *mean* of  $X$
- Alternative notation:

$$\mu = E(X) \text{ or } \mu_X = E(X)$$

## Definition

The *variance* of  $X$  is

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

The *standard deviation* of  $X$  is

$$\sigma = \sqrt{\text{Var}(X)}.$$

## Definition

The  $r$ th moment of  $X$  about  $b$  is

$$E[(X - b)^r].$$

The  $r$ th moment of  $X$  about the origin,  $E(X^r)$ , is usually just called the  $r$ th moment of  $X$ .

## Definition

Let  $x_1, x_2, \dots, x_n$  be a sample.

- The *sample mean* is

$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i.$$

- The *sample variance* is

$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2.$$

- The sample variance can be computed more easily as follows:

$$s^2 = \frac{\sum_{i=1}^n x_i^2 - \frac{1}{n} (\sum_{i=1}^n x_i)^2}{n-1}.$$

- The *sample standard deviation* is  $s = \sqrt{s^2}$ .

The variance of a hypergeometric random variable is

$$\text{Var}(X) = np(1 - p)\frac{N - n}{N - 1},$$

where  $p = \frac{N_1}{N}$ .

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## Definition

- A *Bernoulli trial* is a random experiment that only has 2 possible outcomes.
- Sample space:  $S = \{\text{success, failure}\}$
- Suppose

$$X(\text{success}) = 1 \text{ and } X(\text{failure}) = 0.$$

- p.m.f. for  $X$ :

$$f(x) = \begin{cases} p, & x = 1 \\ 1 - p, & x = 0 \end{cases}$$

- $X$  has a *Bernoulli distribution* with parameter  $p$ .
- $E(X) = p$
- $\text{Var}(X) = p(1 - p)$
- $\sigma_X = \sqrt{p(1 - p)}$
- Alternative notation:  $q = 1 - p$

## Definition

- Consider a sequence of Bernoulli trials such that
  - ▶  $n$  = the number of trials
  - ▶  $p$  = the probability of success on each trial
  - ▶ the trials are independent
  - ▶  $X$  = number of success that occur
- $X$  has a *binomial distribution* with parameters  $n$  and  $p$ .
- $X \sim b(n, p)$
- p.m.f. for  $X$ :

$$f(x) = \binom{n}{x} p^x (1 - p)^{n-x}, x = 0, 1, \dots, n.$$

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

## Definition

The *cumulative distribution function* of  $X$  is

$$F(x) = P(X \leq x).$$

Often, it is simply called the *distribution function* of  $X$ .

## Example

There is a 15% chance that items produced in a certain factory are defective. Assuming that 9 items are produced, and assuming that they are statistically independent, what is the probability that

- at most 4 are defective?
- at least 6 are defective?
- more than 6 are defective?
- the number of defective items is between 2 and 5 inclusive?



# Connection Between the Hypergeometric and Binomial Distributions

## Random Sampling

- Without replacement: hypergeometric
- With replacement: binomial

## Example

In a university organization with 200 members, 60 are seniors. In a random sample of size 10, what is the probability that 4 are seniors, if the sampling is done

- without replacement?
- with replacement?
- Find the expected value, variance, and standard deviation of the number of seniors in the sample under both types of sampling.

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## Definition

The *moment-generating function* of  $X$  is

$$M(t) = E(e^{tX}),$$

assuming  $E(e^{tX})$  is finite on some open interval  $-h < t < h$ .

## Example

Let  $X$  be a random variable with p.m.f.

$$f(x) = \frac{1}{14}x^2, \text{ for } x = 1, 2, 3.$$

Find the moment generating function of  $X$ .

## Example

- If the p.m.f. of  $X$  is  $f(x) = \frac{6}{\pi^2 x^2}$ , for  $x = 1, 2, \dots$
- then  $X$  does not have a moment generating function.

## Example

- Suppose the m.g.f. of  $X$  is  $M(t) = \frac{3}{6}e^t + \frac{2}{6}e^{2t} + \frac{1}{6}e^{3t}$ .
- Find the p.m.f. of  $X$ .

## Example

Find the p.m.f. of  $X$  if the m.g.f. is

$$M(t) = \frac{e^{t/2}}{1 - e^{t/2}}, \quad t < \ln(2).$$

## Theorem

$X$  and  $Y$  have the same m.g.f. if and only if they have the same p.m.f.

- $E(X) = M'(0)$
- $E(X^2) = M''(0)$
- $E(X^r) = M^{(r)}(0)$
- $\text{Var}(X) = M''(0) - M'(0)^2$

- Discrete Random Variables

- ▶ Definitions of random variables, discrete random variables, p.m.f., and support.
- ▶ Probabilities involving random variables
- ▶ Properties of a p.m.f.
- ▶ Hypergeometric distribution

- Mathematical Expectation

- ▶ Definition
- ▶ Calculating  $E(X)$
- ▶ Properties of  $E$
- ▶ Expected value hypergeometric random variable:  $E(X) = np$

- The Mean, Variance, and Standard Deviation
  - ▶ Definitions/notation for mean, variance, and standard deviation
  - ▶ for random variables and samples
  - ▶ Shortcut formulas for variance of a random variable/sample
  - ▶ Be able to compute everything “by hand”.
  - ▶ Variance of a hypergeometric random variable:  
$$\text{Var}(X) = np(1 - p)\frac{N-n}{N-1}.$$
  - ▶ Property of variance:  $\text{Var}(aX + b) = a^2\text{Var}(X)$ , if  $a$  and  $b$  are constants.
- Bernoulli Trials and the Binomial Distribution
  - ▶ Probabilities
  - ▶ p.m.f.
  - ▶ c.d.f. table/computer/calculator
  - ▶  $E(X) = np$ ,  $\text{Var}(X) = np(1 - p)$

## ● Moment-Generating Functions

- ▶ p.m.f.  $\rightarrow$  m.g.f.
- ▶ m.g.f.  $\rightarrow$  p.m.f.
- ▶  $E(X^r) = M^{(r)}(0)$
- ▶  $\text{Var}(X) = M''(0) - M'(0)^2$



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# Approximate Poisson Process with Parameter $\lambda > 0$

## Setting

- Measuring occurrences of some event on a continuous interval.
- Examples:
  - ▶ Number of phone calls received in 1 hour
  - ▶ Number of defects in 1 meter of wire

## Assumptions

- Occurrences in non-overlapping intervals are independent.
- In a sufficiently short interval of length  $h$ , the probability of 1 occurrence is approximately  $\lambda h$ .
- In a sufficiently short interval, the probability of 2 or more occurrences is essentially zero.

# Poisson Distribution

- Let  $X = \#$  of occurrences in an interval of length 1
- Then  $X$  has a *Poisson distribution* with parameter  $\lambda$ .
- 

$$f(x) = P(X = x) = \frac{\lambda^x e^{-\lambda}}{x!}$$

## Example

Phone calls received by a company are a Poisson process with parameter  $\lambda = 4$ . In a 1 minute period, find the probability of receiving

- 2 calls?
- 5 calls?
- at most 3 calls?
- at least 7 calls?

- If  $X$  has a Poisson distribution with parameter  $\lambda$ , then
- $E(X) = \text{Var}(X) = \lambda$
- For a Poisson process with parameter  $\lambda$ ,
- $\lambda$  is the average # of occurrences in an interval of length 1.

### Example

Phone calls received by a company are a Poisson process, and the company receives an average of 4 calls per minute. In a 3 minute period, find the probability of the company receiving

- 10 calls?
- at most 15 calls?

## Interval of Length $t$

- Consider a Poisson process with parameter  $\lambda$
- If  $X = \#$  of occurrences in an interval of length  $t$ ,
- then  $X$  has a Poisson distribution with mean  $\lambda t$ .

## Example

On average, there are 3 flaws in 8 meters of copper wire. For a piece of wire 20 meters long, find the probability of observing

- 5 flaws.
- fewer than 9 flaws.
- Find the expected value, variance, and standard deviation of the number of flaws on a 20 meter piece of wire.

## Example

Let  $X$  equal the number of green m&m's in a package of size 22. Forty-five observations of  $X$  yielded the following frequencies for the possible outcomes of  $X$ :

Outcome (x):	0	1	2	3	4	5	6	7	8	9
Frequency:	0	2	4	5	7	9	8	5	3	2

- Calculate  $\bar{x}$  and  $s^2$ . Are they close?
- Compare the relative frequency histogram to the probability histogram of a Poisson random variable with mean  $\lambda = 5$ .
- Do these data appear to be observations from a Poisson random variable?

# Poisson Approximation to the Binomial

- If  $n$  is large and  $p$  is small,
- $\text{bin}(n, p) \approx \text{Poisson with } \lambda = np.$
- $n \geq 100$  and  $p \leq 0.1$
- $n \geq 20$  and  $p \leq 0.05$

## Example

In a shipment of 2000 items, 4% are defective. In a random sample of size 100, find the approximate probability that more than 10 items are defective.