# Stat 201: Statistics I Chapter 5



date



# Chapter 5 Discrete Probability Distributions

# Section 5.1 Probability Distributions

A **random variable** is a variable that has a numeric value determined by chance from a range of possible values.

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## **Example**

- $\bullet$  X = the number of heads from three coin flips
- $\bullet$  Y =the sum of two dice
- $\bullet$  Z = the midterm score of a randomly selected student
  - A students grade (A, A-, B+, etc.) can not be used as a random variable because it is not numeric,
  - ullet ... Unless, the grade is coded as a number (i.e. A=4.0, A=3.7, etc.)

## Types of random variables

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- Number of defective insulin test strips in a box of 50
- Number of customers to enter a store in the next 10 minutes

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- Number of defective insulin test strips in a box of 50
- Number of customers to enter a store in the next 10 minutes

Continuous random variables:

- Height or weight of a test subject
- Survival time of a cancer patient
- Price of a company's stock at a particular moment

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- Often displayed in tables (if practical)

#### **Example**

A restaurant wants to track its taco sales. It records how many tacos customers order with each visit. The results are in the table,

Number of tacos	Probability
0	0.35
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The restaurant introduces a new "Super Taco" (beef, chicken and shrimp). It wonders if larger groups are more likely to order the new item. The results are in the table,

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What is the probability of a customer ordering less than two tacos?

$$P(X < 2) = P(X = 0 \text{ or } 1) = P(0) + P(1) = 0.35 + 0.2 = 0.55$$

## Weighted means

A **weighted mean** is the mean of values that are not considered equally, or do not have equal importance.

- Each value has an associated weight, which is its relative importance.
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$$\mu_w = \frac{\sum w_i \times x_i}{\sum w_i}$$

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#### **Example**

Fiona buys 4 lbs. of hamburger at 4.89 / lb. and 2 lbs. of steak at 11.99 / lb. What is the average price per pound she is paying?

$$\$/lb. = \frac{4 \times 4.89 + 2 \times 11.99}{6} = \frac{43.54}{6} = 7.26$$

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The mean of a probability distribution is also known as the **expected value** of the random variable.

Denoted with an "E", as in

$$\mathbb{E}(X) = \mu$$

# Mean, exmaple

## **Example**

Number of tacos	Probability
0	0.35
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What is the mean number of tacos ordered at the restaurant? That is, how many tacos should the restaurant expect each customer to order?

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$$\mathbb{E}(X) = \mu = \sum_{i=1}^{n} x_i \cdot P(x_i)$$

$$= 0 \cdot 0.35 + 1 \cdot 0.2 + 2 \cdot 0.3 + 3 \cdot 0.1 + 4 \cdot 0.05$$

$$= 0 + 0.2 + 0.6 + 0.3 + 0.2$$

$$= 1.3$$

# Standard deviation of probability distributions

Similarly, variance of a probability distribution is the weighted mean of difference from the mean squared and standard deviation is the square root of variance.

Thus,

$$\sigma^2 = \sum (x_i - \bar{x})^2 \cdot P(x_i)$$
$$\sigma = \sqrt{\sigma^2}$$

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$$= (0 - 1.3)^{2} \cdot 0.35 + \dots + (4 - 1.3)^{2} \cdot 0.05$$

$$= 0.5915 + \dots + 0.3645$$

$$= 1.41$$

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$$= 1.41$$

$$\sigma = \sqrt{\sigma^{2}} = 1.19$$

## Mean and var. of random variables in StatCrunch

- Stat  $\rightarrow$  Calculators  $\rightarrow$  Custom
- Under "Values in:" select the column which contains the random variable values
- Under "Weights in:" select the column which contains the random variable probabilities
- Click "Compute!"
- Mean and standard deviation will be displayed
- Variance can be calculated by squaring the standard deviation

Recall, the range rule of thumb for unusual values are those values more than 2 standard deviations away from the mean. In other words, x is unusual if

$$x < \mu - 2\sigma$$
 or  $x > \mu + 2\sigma$ 

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#### **Example**

From the probability distribution of tacos ordered,  $\mu = 1.3$  and  $\sigma = 1.19$ . What would be an unusual amount of tacos to order?

M. Shyne (Metro State)

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• The lower bound for unusual values is  $\mu-2\sigma=-1.08$ . Since you can't order negative tacos, there is not an unusually low number of tacos to order.

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From the probability distribution of tacos ordered,  $\mu=1.3$  and  $\sigma=1.19$ . What would be an unusual amount of tacos to order?

- The lower bound for unusual values is  $\mu-2\sigma=-1.08$ . Since you can't order negative tacos, there is not an unusually low number of tacos to order.
- The upper bound for unusual values if  $\mu+2\sigma=3.86$ . Thus, 4 (or more) tacos is an unusually high number of tacos to order.

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$$P(X \le x) < 0.025 \qquad \text{or} \qquad P(X \ge x) < 0.025$$

# Unusual events, example

#### **Example**

Tom and Barbara collect eggs from their chickens everyday. The number of eggs they collect follows the following distribution,

Is collecting no eggs unusual? Is collecting 7 eggs unusual?

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Eggs 
$$(x)$$
 0
 1
 2
 3
 4
 5
 6
 7
 8
 9

  $P(x)$ 
 0.01
 0.03
 0.1
 0.2
 0.3
 0.2
 0.1
 0.04
 0.02
 0

Is collecting no eggs unusual? Is collecting 7 eggs unusual?

- $P(X \le 0) = P(0) = 0.01 < 0.025$
- $P(X \ge 7) = P(7) + P(8) + P(9) = 0.06 < 0.025$

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Recall the example of flipping a coin 1000 times. Under the assumption the coin is fair (P(H) = P(T) = 1/2), the expected number of heads is 500.

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• Getting 523 heads is not unusual ( $P(X \ge 523) = 0.077$ ). There is no reason to think the coin is not fair.

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## **Example**

Recall the example of flipping a coin 1000 times. Under the assumption the coin is fair (P(H) = P(T) = 1/2), the expected number of heads is 500.

- Getting 523 heads is not unusual ( $P(X \ge 523) = 0.077$ ). There is no reason to think the coin is not fair.
- Getting 46 heads is unusual ( $P(X \le 46) = 6.23 \times 10^{-222}$ ). We would be justified in questioning the assumption that the coin is fair.

# **Group work**

• Work on question 1, all parts.

# Section 5.2 Binomial Probability Distributions

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- The number of students passing the midterm in class
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  - Passing the midterm is a "success", failing it is a "failure"
- The number of heads in three coin flips
  - Getting a head is a "success", getting a tail is a "failure"
- The number of of car crashes that result in fatalities
  - A fatality is a "success", no fatalities is a "failure"

There are four requirements to be considered a binomial distribution:

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- Each trial must have the same probability for success

The last two requirements are often summarized as "independent and identically distributed" and abbreviated as "iid".

#### **Example**

Recall the Metro State Statistics Club, with 6 male members and 4 females members. If the three officer positions are selected randomly, does the number of women selected follow a binomial distribution?

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• No. The probability of success changes (hopefully) with each test.

## Requirements for binomial distributions, example

#### **Example**

Recall the Youth Risk Behavior Survey (YRBS) which found the probability of a teenaged driver had texted or emailed while driving was 0.404. Suppose this is the probability for the population of teenaged drivers. Suppose 30 teenaged drivers are selected at random. Does the number of those drivers that had texted or emailed while driving follow a binomial distribution?

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Recall the Youth Risk Behavior Survey (YRBS) which found the probability of a teenaged driver had texted or emailed while driving was 0.404. Suppose this is the probability for the population of teenaged drivers. Suppose 30 teenaged drivers are selected at random. Does the number of those drivers that had texted or emailed while driving follow a binomial distribution?

- Yes.
  - There is a fixed number of trials (30).
  - Each trial has only two possible outcomes (had or had not texted).
  - Each trial is independent.
  - Each trial has the same probability of "success."

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

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#### **Example**

The probability a teenaged driver had texted or emailed while driving is 0.404. If the random variable Y is the number of teenaged drivers who had texted or emailed while driving out of a sample of 30,

$$Y \sim \text{Binom}(30, 0.404)$$

## Probabilities for binomial distributions

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- $p^xq^{n-x}$  is the probability of getting x successes and n-x failures in one particular order.

# Binomial probabilities in StatCrunch

- Stat  $\rightarrow$  Calculators  $\rightarrow$  Binomial
- Enter sample size ("n") and probability ("p")
- Select appropriate comparison symbol and numeric value
- Click "Compute" if necessary
- Probability will be displayed

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$$P(X=5)=0.165$$

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Standard deviation:

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$$\mathbb{E}(Y) = \mu = np = (30)(0.404) = 12.12$$

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- $\mathbb{E}(Y) = \mu = np = (30)(0.404) = 12.12$
- $\sigma^2 = npq = (30)(0.404)(1 0.404) = 7.22$
- $\sigma = \sqrt{\sigma^2} = \sqrt{7.22} = 2.69$

For a given binomial distribution, the boundaries for unusual values can be found. From the range rule of thumb, the lower boundary is  $\mu-2\sigma$  and the upper boundary is  $\mu+2\sigma$ .

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- The upper bound for unusual values is  $\mu + 2\sigma = 17.5$
- In a random sample of 30 teenaged drivers, it would be unusual to get 6 or fewer, or 18 or more, drivers who had texted or emailed while driving.

## **Group work**

• Work on question 2, all parts.

# Section 5.3 Poisson Probability Distributions

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### **Example**

- The number of fish caught in the next hour
- The number of customers to come into a store between 1 and 6 pm
- The number of insects found in a square foot of grassland

### **Poisson notation**

$$X \sim \text{Pois}(\lambda)$$

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- The random variable X has a Poisson distribution with an mean rate of events of  $\lambda$  (lambda).
- The mean rate ( $\lambda$ ) is scalable. That is, if  $\lambda$  is the mean number of events per hour,  $\operatorname{Pois}(\lambda/2)$  models the number of events in a half hour interval.

The probability of x events in an interval,

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

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Standard deviation:

$$\sigma = \sqrt{\lambda}$$

# Poisson probabilities in StatCrunch

- Stat  $\rightarrow$  Calculators  $\rightarrow$  Poisson
- Enter the rate value  $\lambda$  ("Mean")
- Select appropriate comparison symbol and numeric value
- Click "Compute" if necessary
- Probability will be displayed

#### **Example**

Paul fishes for an hour every morning. Over the past month (30 days) he has caught 62 fish. Paul is having friends over for dinner tonight, so he needs to catch 4 fish.

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What is the probability he catches 4 fish in an hour? Is that an unusually high number?

• Assuming the chances of catching a fish are the same every day, the mean rate of fish caught is  $\lambda=\frac{62}{30}=2.07$ 

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- Probability of catching four or more fish:  $P(X \ge 4) = 0.156$
- Rule of thumb boundary:  $\mu + 2\sigma = 2.07 + 2\sqrt{2.07} = 4.95$

## **Group work**

• Work on question 3, all parts.