

# The Poisson Distribution

August 15, 2019

# Midterm Results

<b>Statistic</b>	<b>Raw Score</b>	<b>Percentage</b>
Mean	32.2	71.5
Median	32	71.1
Standard Deviation	5.0	10.8
Maximum	42	93.3
Minimum	19	42.2

Note: Raw scores are out of 45 possible points.

Any extra credit has already been added to the midterm scores.

# Calculating Your Current Grade

You can use a little bit of statistics to calculate your current grade!

- Labs are worth 10%
- Quizzes (homeworks) are worth 20%
- The midterm is worth 30%
- The final will be worth 40%

# Calculating Your Current Grade

- You have grades for  $10\% + 20\% = 30\% = 60\%$  of the class.
- We can use this to scale these values and calculate your current overall grade.
  - For your current grade, labs are worth  $0.1/0.6 = 0.1667$  or  $16.67\%$
  - Quizzes (homeworks) are worth  $.2/.6 = 0.3333$  or  $33.33\%$
  - And the midterm is worth  $.3/.6 = 0.50$  or  $50\%$

# Calculating Your Current Grade

We can use these to calculate a weighted average. This will give you your current grade.

$$0.167 \times (\text{Lab } \%) + 0.333 \times (\text{Quiz } \%) + 0.5 \times (\text{Midterm } \%)$$

This is just like calculating an expected value!

# Calculating Your Current Grade

So if I have 100% in lab, an 85% on quizzes, and got a 70% on the midterm, my current grade is

$$0.167 \times (1) + 0.333 \times (0.85) + 0.5 \times (0.7) = 0.80$$

Note: Remember we drop your lowest lab score! There is also still time to get your quiz grade up a bit.

# Midterm Results

- The labs and quizzes are designed to help pad your grade.
- If you're showing up to lab and doing well on your quizzes, I am not worried about your ability to excel in this course.
- If you calculate your grade and it's lower than you want it to be, let's talk!
- If overall grades at the end of the term are low, **I will "curve" the class by adding a set number of points to everyone's grade.**

# Working With Cumulative Probabilities

Practice rewriting the following probabilities in terms of  $P(X \leq x)$  and  $P(X = x_1) + P(X = x_2) + \dots$

- $P(X < 4)$
- $P(X > 4)$
- $P(X \leq 4)$
- $P(X \geq 4)$

# Working With Cumulative Probabilities

Practice rewriting the following probabilities in terms of  $P(X \leq x)$  and  $P(X = x_1) + P(X = x_2) + \dots$

- $P(6 \leq X \leq 8)$
- $P(6 \leq X < 8)$
- $P(6 < X \leq 8)$
- $P(6 < X < 8)$

# Motivating Example: Poisson Distribution

- There are about 8 million people in New York City.
- How many New Yorkers would be expect to be hospitalized due to heat attack, each day?
- Historical records suggest the average is 4.4
- But what about the distribution?
- What might a histogram of daily counts look like?

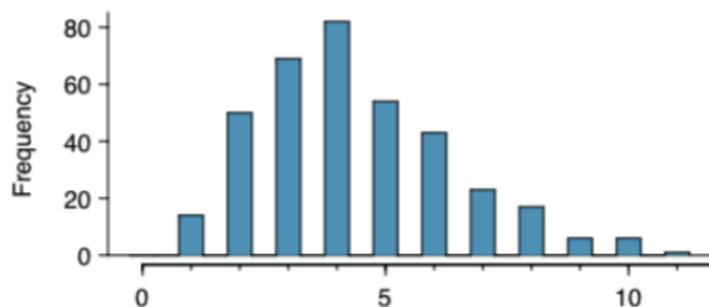
# Example

Intuitively, we might think that

- The average is 4.4.
- We don't know the standard deviation.
- The minimum is 0.
- The (theoretical) maximum is about 800 million or so. It's so far away from the mean as to be meaningless.

Clearly the maximum is a lot further from the average than the minimum is, so we might guess that this distribution is skewed to the right.

# Example



The number of heart attack hospitalizations were recorded every day for a year.

- The sample mean is 4.38, similar to the historical average of 4.4.
- The sample standard deviation is about 2.
- The distribution is unimodal and right-skewed.

# The Poisson Distribution

The **Poisson distribution** is used to describe the number of events that occur in a large population over some period of time. We might measure

- Marriages
- Births
- Heart attacks
- Lightning strikes

In each case, we can count the number of times that event occurs during a period of time.

# The Poisson Distribution

- The average number of occurrences per period of time is called the **rate**.
- In the heart attack example, we had a rate of 4.4 *heart attacks* (the event) per *day* (the period of time).
- We denote the rate by the Greek letters  $\lambda$  (lambda) or  $\mu$ .
- We can use this information to find the probability of observing exactly  $k$  events during a particular period of time.

# The Poisson Distribution

Suppose we are interested in some events and the number of observed events follows a Poisson distribution with rate  $\lambda$ . Let  $X$  be the number of events observed. Then

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

where  $k$  can take on any whole number greater than or equal to 0. The letter  $e$  is a constant:  $e \approx 2.719$ .

# The Poisson Distribution

The Poisson distribution has an interesting property: its mean and variance are the same!

$$E(X) = Var(X) = \lambda$$

and the standard deviation is then  $\sqrt{\lambda}$ .

# Is It Poisson?

Guidelines for determining if the Poisson distribution is appropriate:

- ① We are interested in the number of events that occur.
- ② There is a set period of time that we are interested in.
- ③ Events occur independently of each other.
- ④ The population that generates the events is quite large.

## Example: Coffee Shop Customers

A coffee shop serves an average of 75 customers per hour during the morning rush.

- Which distribution is appropriate for working with the probability of a given number of customers arriving within one hour during the morning rush?

## Example: Coffee Shop Customers

- ① We are interested in the number of customers served.
  - "Customer served" is the event.
- ② We are interested in this event over the course of one hour (during morning rush), so there is a set period of time.
- ③ We assume that customers are served (more or less) independently of one another.
- ④ The population that generates these events is anyone who could possibly walk in and be served during the morning rush. That's quite a lot of people, so can be confident that the population is large.

So the Poisson distribution is appropriate.

## Example: Coffee Shop Customers

What are the mean and standard deviation of the number of customers this coffee shop serves in one hour during the morning rush?

## Example: Coffee Shop Customers

Would it be considered unusually low if only 60 customers showed up to the coffee shop in one hour during the morning rush?

## Example: Coffee Shop Customers

Find the probability that the coffee shop serves 70 people in one hour during the morning rush.

## Example: Coffee Shop Customers

What is the probability that the coffee shop serves between (and including) 73 and 76 customers?

# Poisson Approximation to Binomial

This is another binomial approximation method that will help us avoid difficult factorial expressions.

We can use the Poisson approximation to the binomial distribution when

- $n$  is large
- $np < 7$

# Poisson Approximation to Binomial

Suppose we have a lot size of 1000 and the proportion of defective items is 0.001. What is the probability of exactly 3 defective items?

- Items are either defective or not.
- Defective status is independent between items.
- There is a fixed lot size of  $n = 1000$  items.
- The probability of success (a defect) is 0.001.

The binomial probability for exactly 3 defectives is

$$\begin{aligned}P(X = 3) &= \binom{1000}{3} (0.001)^3 (0.999)^{1000-3} \\ &= 0.0613\end{aligned}$$

OR, we can let  $\lambda = np = 1$  and use the Poisson distribution:

$$\begin{aligned}P(X = 3) &= \frac{e^{-1} 1^3}{3!} \\ &= 0.0613\end{aligned}$$