
INFO 2950: Intro to Data Science

Lecture 18
2023-10-30

Agenda

1. Geometric distributions
2. Negative Binomial distributions
3. Poisson distributions

How do you know if you can use a binomial distribution?

1. Each trial has only 2 possible outcomes: “success” (1) or “failure” (0)
2. You conduct each “trial” exactly the same way for a fixed number of times n
3. The probability of success p is the same for each trial
4. Trials are independent

Can you use a binomial distribution?

Sample students until you've found one who likes cats.
 X = number of students sampled.
Probability of a student liking cats is 28%.

Can you use a binomial distribution?

Nope, no n defined!

Sample students until you've found one who likes cats.

X = number of students sampled.

Probability of a student liking cats is 28%.

How do you know if you can use a geometric ~~binomial~~ distribution?

1. Each trial has only 2 possible outcomes:
“success” (1) or “failure” (0)
2. You conduct each “trial” exactly the same way
~~for a fixed number of times n~~
3. The probability of success p is the same for
each trial
4. Trials are independent

How do you know if you can use a geometric ~~binomial~~ distribution?

1. Each trial has only 2 possible outcomes:
“success” (1) or “failure” (0)
2. You conduct each “trial” exactly the same way
~~for a fixed number of times n~~
and count the number of trials until the first success
3. The probability of success p is the same for
each trial
4. Trials are independent

Geometric distribution

Event space: integers 0 to infinity

Parameters: chance of "success" p

Story: how many trials before the first success?

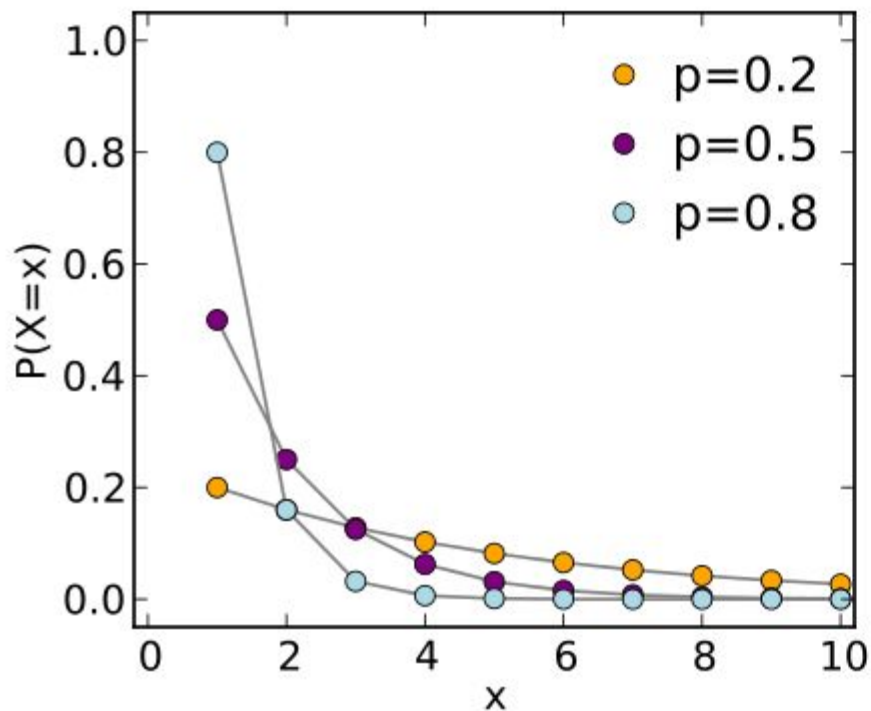
Geometric distribution (shifted)

Event space: integers **1** to infinity

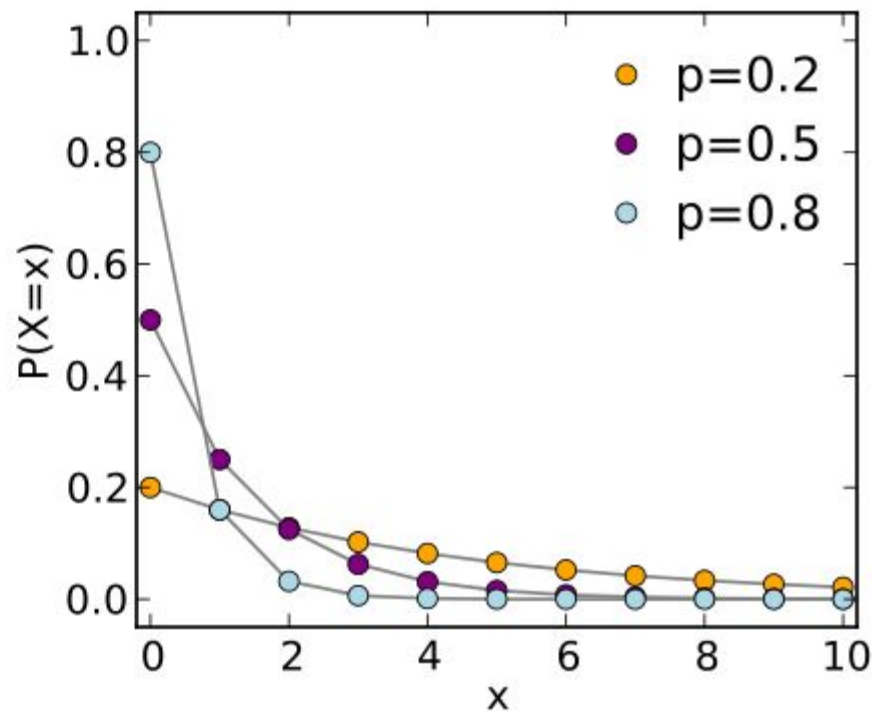
Parameters: chance of "success" p

Story: how many trials **to reach** the first success?

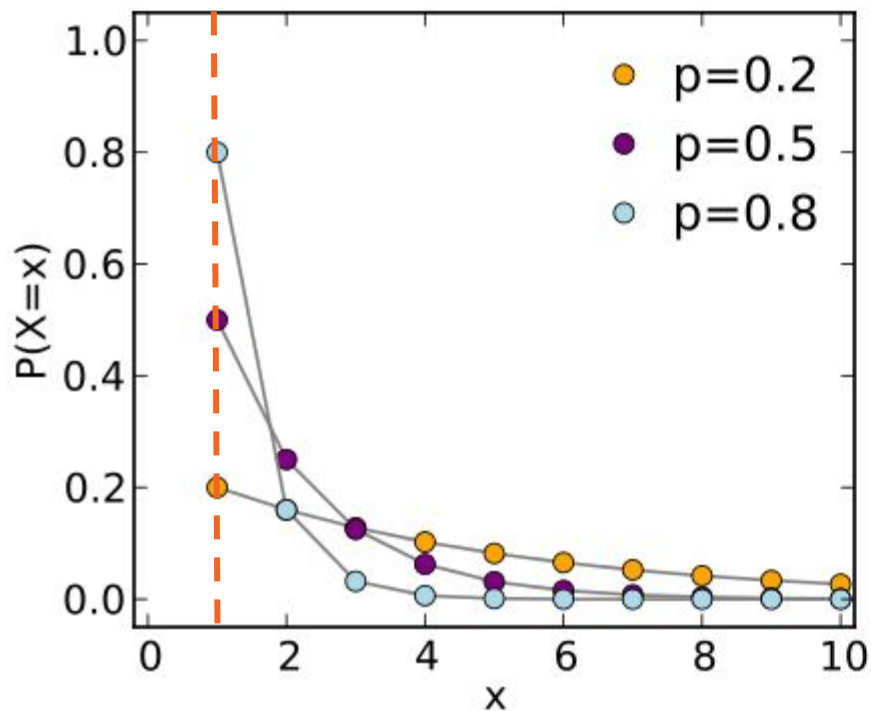
Shifted geometric distribution



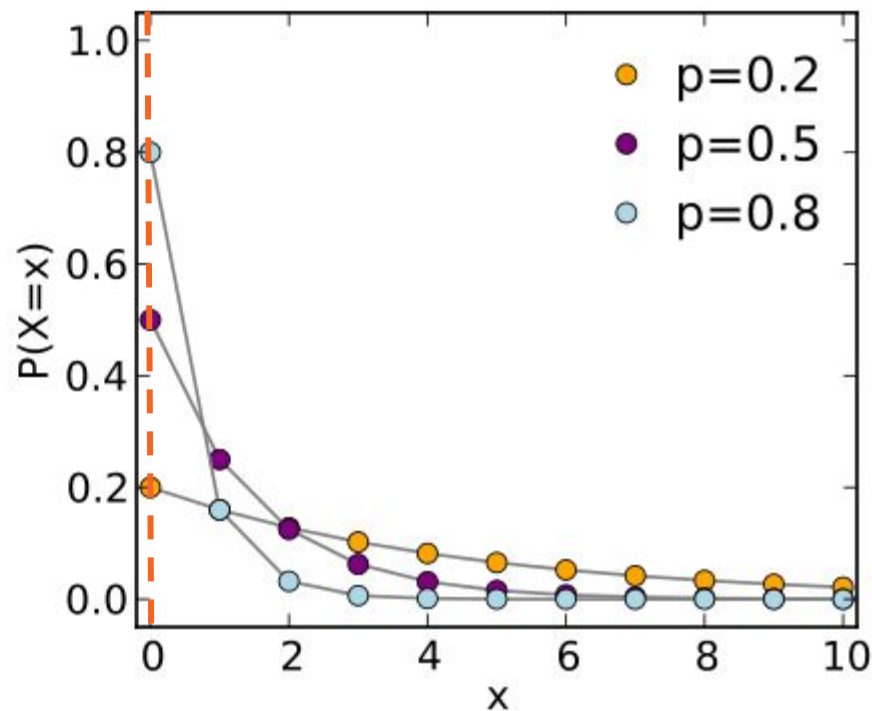
Geometric distribution



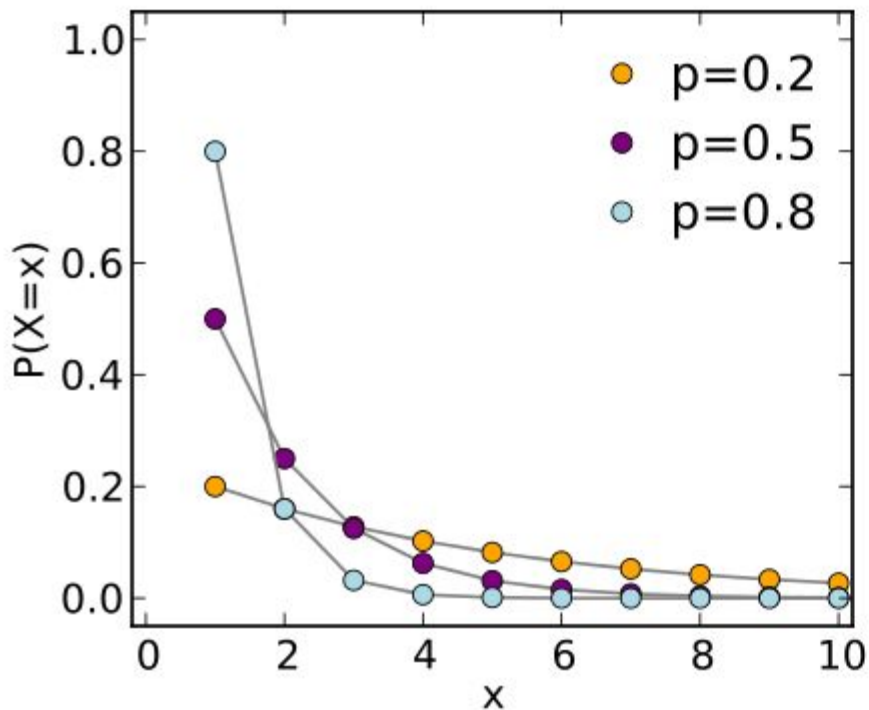
Shifted geometric distribution



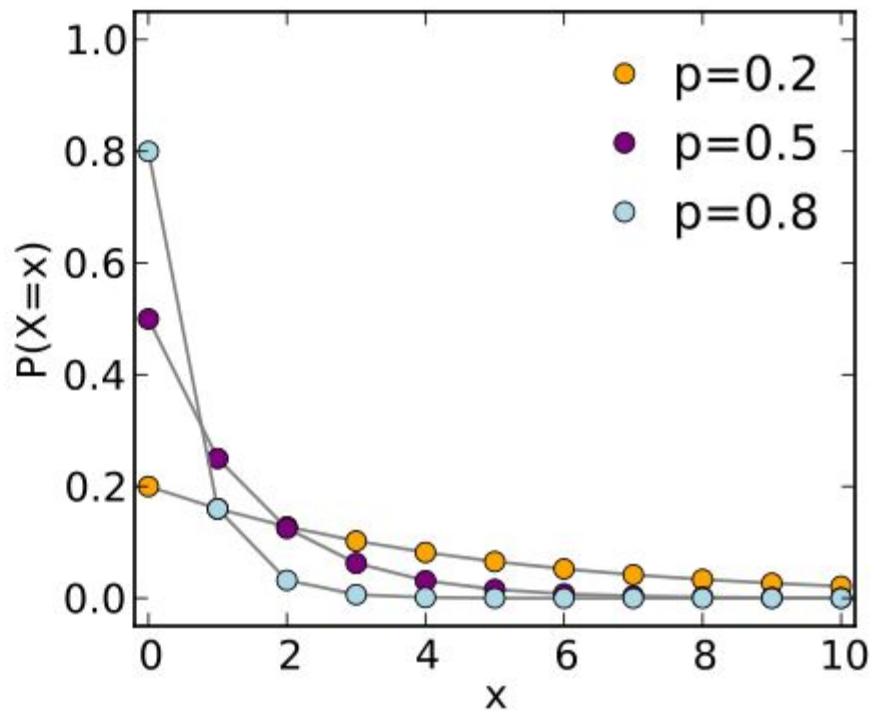
Geometric distribution



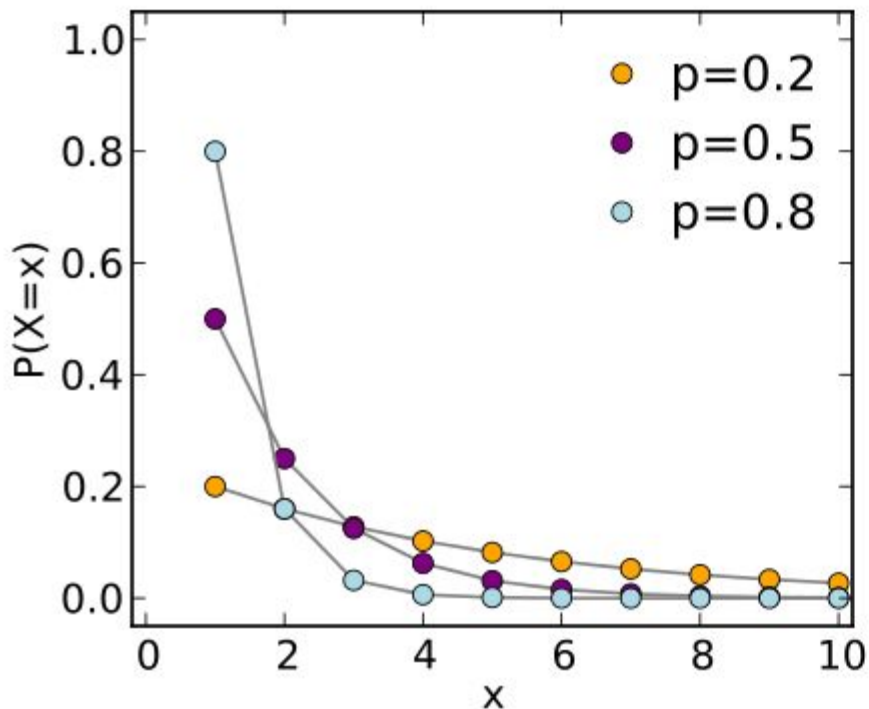
Shifted geometric distribution



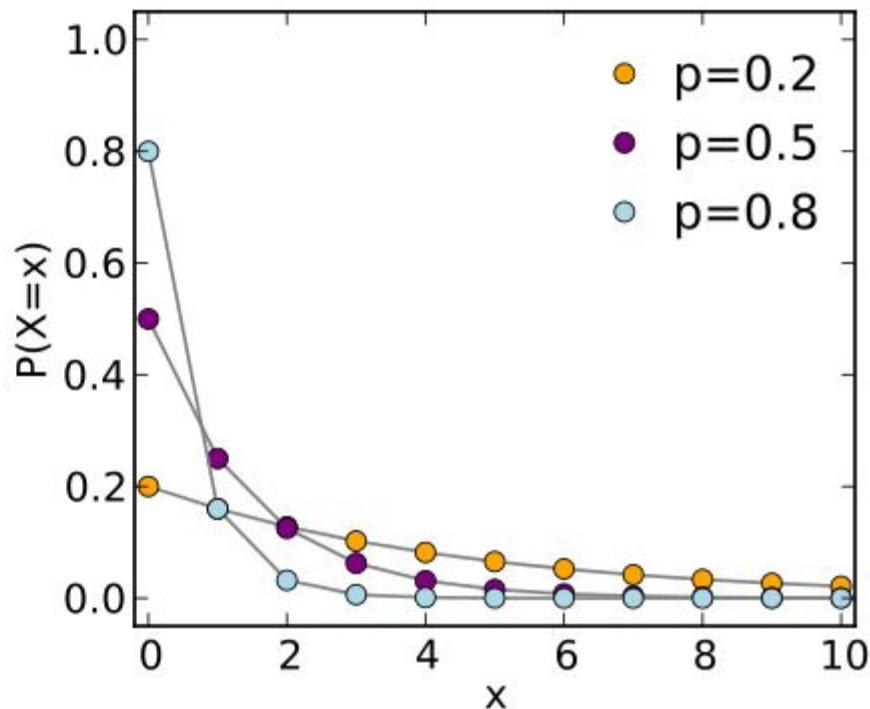
Geometric distribution



Shifted geometric distribution

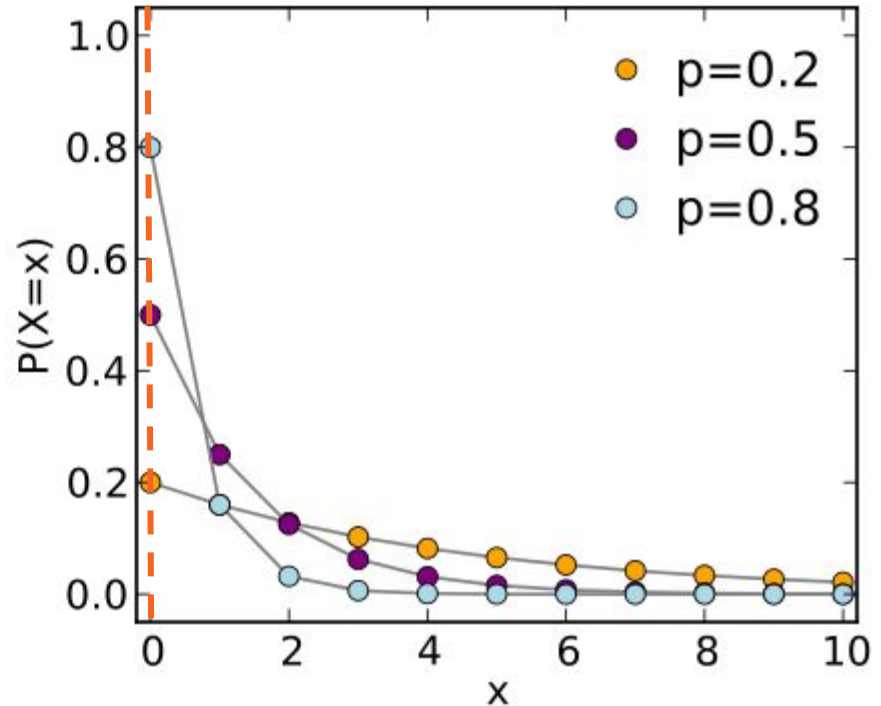


Geometric distribution

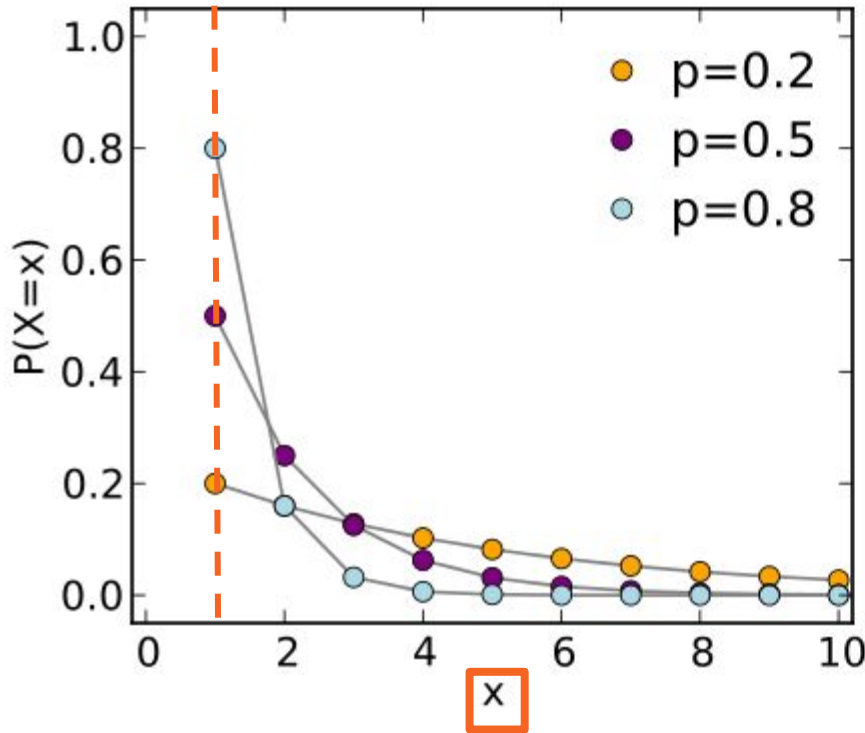


Geometric distribution

- x represents the number of failures before you get your first success
- $P(X=10)$ = the probability that you survey 11 people, the first 10 people prefer dogs, and only the 11th person prefers cats

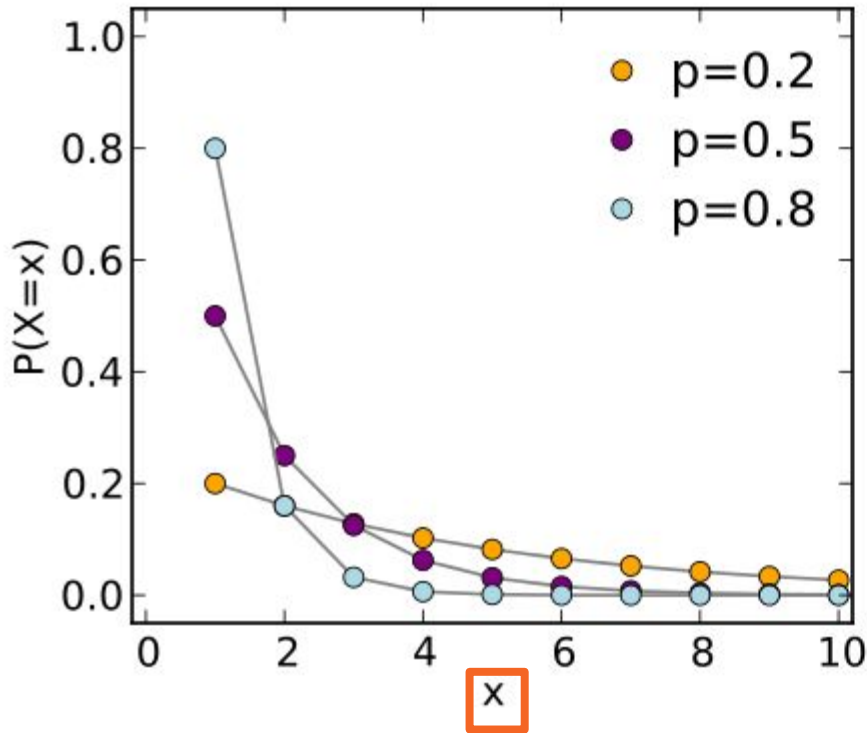


Shifted geometric distribution



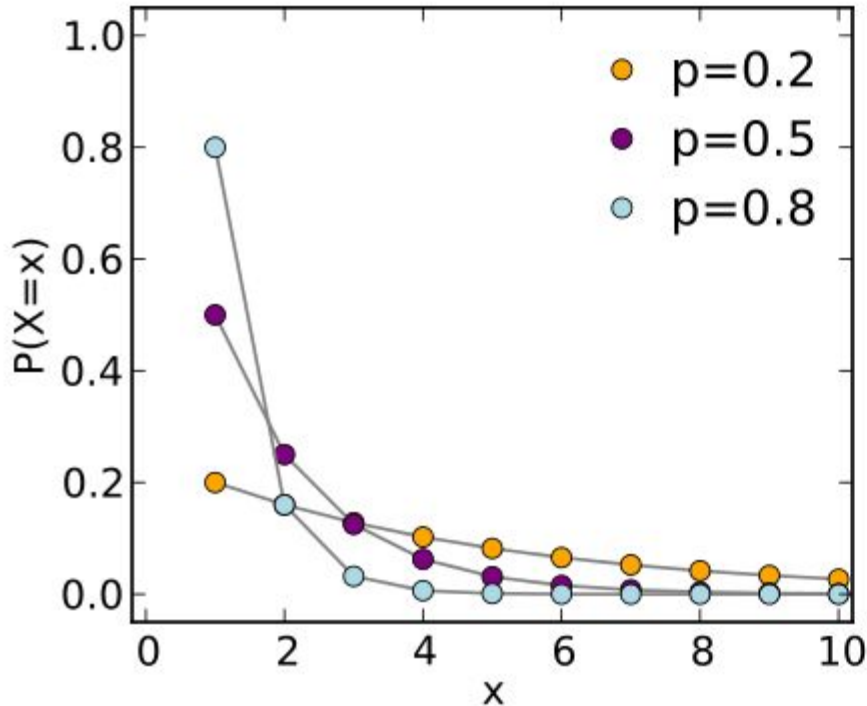
- x represents this number being the first time success happens
- $P(X=10)$ = the probability that you survey ____ people, the first ____ people prefer dogs, and only the ____th person prefers cats

Shifted geometric distribution



- x represents this number being the first time something happens
- $P(X=10)$ = the probability that you survey 10 people, the first 9 people prefer dogs, and only the 10th person prefers cats

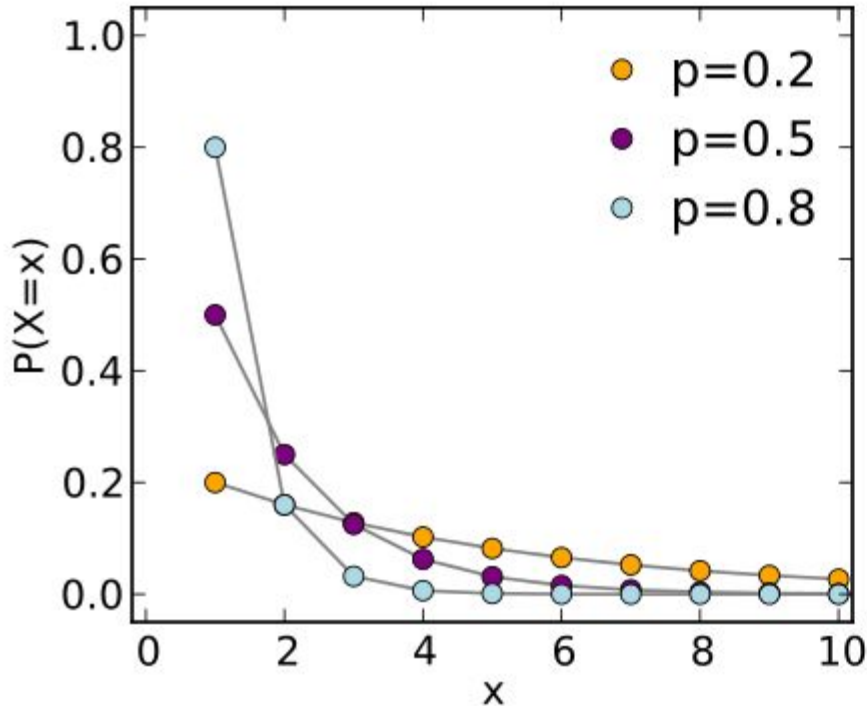
Shifted geometric distribution



Properties:

- Probability always decreases as x increases!

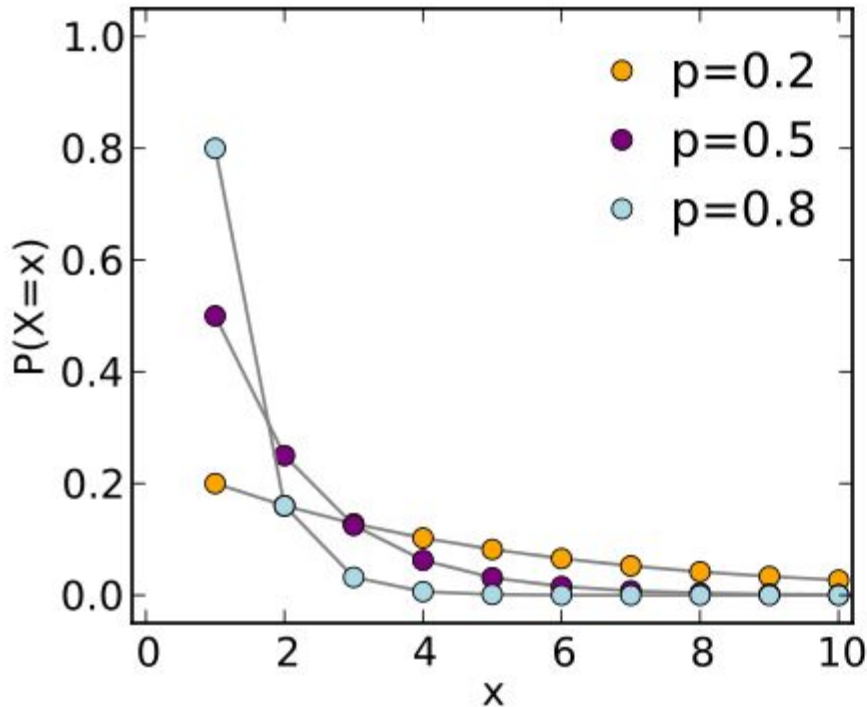
Shifted geometric distribution



Properties:

- Probability always decreases as x increases!
- **Memorylessness:** the distribution of “waiting time” X doesn't depend on how much time x has already elapsed.

Shifted geometric distribution



Properties:

- Probability always decreases as x increases!
- **Memorylessness: the coin doesn't remember its past flips!**

The Greed Game: each round



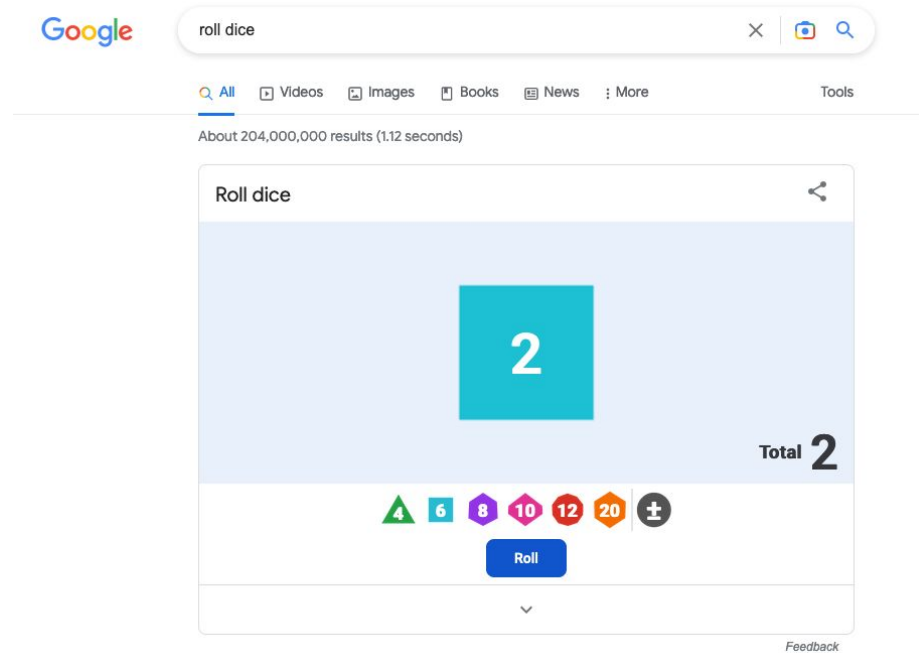
- If I roll a 1-5, you get +1 point
- If I roll a 6, you reset to 0 points and you're out of the game
- You can sit down at any time (if a 6 has not yet been rolled) and keep the points you've accumulated

The Greed Game: rules



- Everyone starts by standing
- I roll a dice simulator once each round
- **On your whiteboard:** track your points (update each round): +1 for any 1-5 roll
- Sit down **when you choose to stop** playing, or **when I roll a 6** (reset to 0 points)
- Game ends when a 6 is rolled; winner has most points

I will be using Google's unbiased 6-sided dice



Was there cheating?!



- How many points did the winner(s) have?
- How expected/unexpected is this?
- How can we tell?

(Shifted) Geometric Distribution

| | | |
|----------------------------------|--|---|
| Probability mass function | $P(X = k) = p(1-p)^{k-1}$ | $p = \text{probability of success}$ $k = \# \text{ of trials}$ |
| Cumulative Distribution Function | $P(X \leq k) = 1 - (1-p)^k$ $P(X \geq k) = (1-p)^{k-1}$ $P(X > k) = 1 - P(X \leq k) = (1-p)^k$ | |
| Mean: | $\mu = E(X) = \frac{1}{p}$ | |
| Variance: | $\sigma^2 = V(X) = \frac{(1-p)}{p^2}$ | |

(Shifted) Geometric Distribution

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$p = \text{probability of success}$

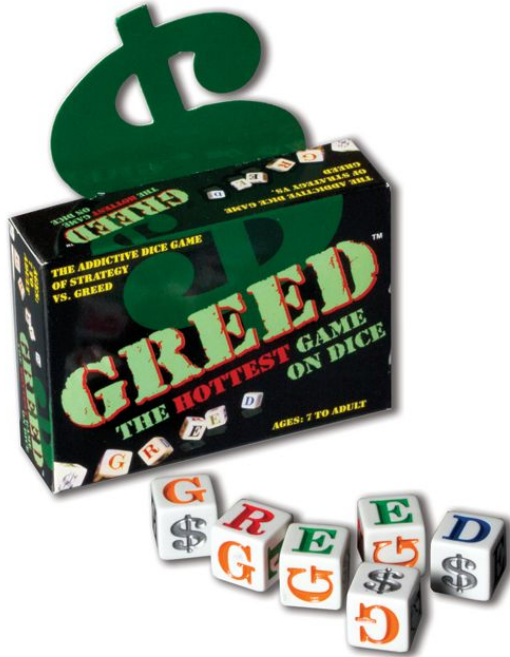
$k = \# \text{ of trials}$

PMF vs. CDF



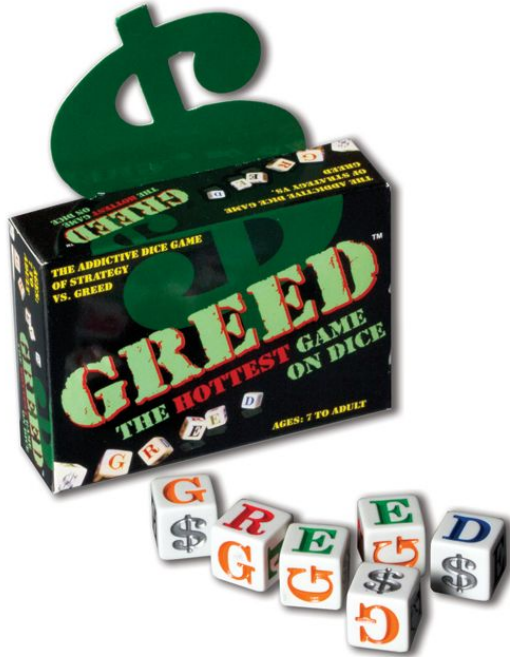
- How do we express the probability that we're looking to calculate?
- Let's assume our story X = the time to reach "success"
 - "Success" = # rolls until game end (which could be way after the winner of the game sits down)

PMF vs. CDF



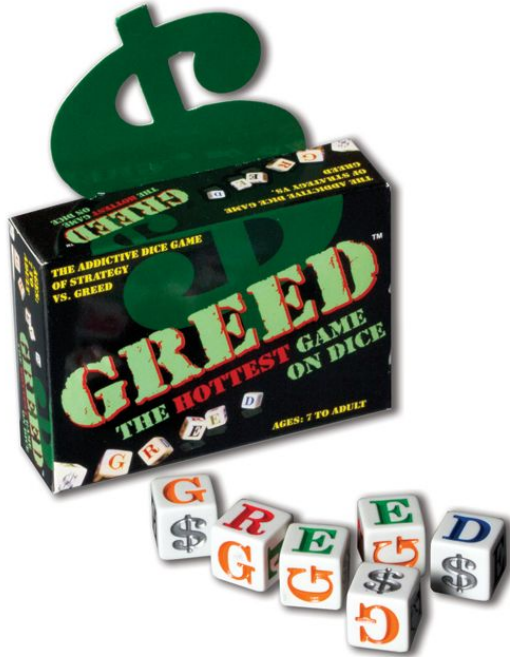
- X = the time to reach the end of the game
- Let k = some integer
- What is the difference between $P(X=k)$ and $P(X \geq k)$?

PMF vs. CDF



- X = the time to reach the end of the game
- Let k = some integer
- What is the difference between $P(X=k)$ and $P(X \geq k)$?
 - $P(X=k)$ is the probability that it takes k time exactly for the game to end (i.e. a 6 is rolled on the k^{th} round). $P(X \geq k)$ is the probability that it takes **at least** k time for the game to end (i.e., a 6 was definitely **not** rolled in the first $k-1$ rounds).

PMF vs. CDF



- X = the time to reach the end of the game
- To model the probability that someone wins the Greed Game with $k-1$ points, should we use $P(X=k)$ or $P(X \geq k)$?

PMF vs. CDF



- X = the time to reach the end of the game
- To model the probability that someone wins the Greed Game with $k-1$ points, should we use $P(X=k)$ or $P(X \geq k)$?

Someone wins by necessarily sitting down *before* a 6 is rolled. They must also be the last person to sit down, meaning no one else will be standing after the winner sits. So, if the winner sits at time k , the 6 could be rolled anytime after k , hence ending the game. This means we care more about $P(X \geq k)$.

How likely is a winning result?

Shifted Geometric Distribution
Cumulative
Distribution
Function

$$P(X \leq k) = 1 - (1 - p)^k$$

$$P(X \geq k) = (1 - p)^{k-1}$$

$$P(X > k) = 1 - P(X \leq k) = (1 - p)^k$$

How likely is a winning result?

Shifted Geometric Distribution
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- What is the probability that someone won with 20 points?
 - X = time until game ends, $k = ?$

How likely is a winning result?

Shifted Geometric Distribution
Cumulative
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- What is the probability that someone won with 20 points?
 - X = time until game ends, $k = 21$ (number of “independent trials”; we have to roll the 6 *after* the winner sits down in round 20, so $k = 21$)

How likely is a winning result?

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Cumulative
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$$P(X \leq k) = 1 - (1 - p)^k$$

$$P(X \geq k) = (1 - p)^{k-1}$$

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- What is the probability that someone won with 20 points?
 - X = time until game ends, $k = 21$ (number of “independent trials”; we have to roll the 6 *after* the winner sits down in round 20, so $k = 21$)
 - What is p ?

How likely is a winning result?

Shifted Geometric Distribution
Cumulative
Distribution
Function

$$P(X \leq k) = 1 - (1 - p)^k$$

$$P(X \geq k) = (1 - p)^{k-1}$$

$$P(X > k) = 1 - P(X \leq k) = (1 - p)^k$$

- What is the probability that someone won with 20 points?
 - X = time until game ends, $k = 21$
 - What is p ? p = probability of “success” (game ending) each trial = 1%

How likely is a winning result?

Shifted Geometric Distribution
Cumulative
Distribution
Function

$$P(X \leq k) = 1 - (1 - p)^k$$

$$P(X \geq k) = (1 - p)^{k-1}$$

$$P(X > k) = 1 - P(X \leq k) = (1 - p)^k$$

- What is the probability that someone won with 20 points?
 - X = time until game ends, $k = 21$
 - $p = \frac{1}{6}$

How likely is a winning result?

Shifted Geometric Distribution
Cumulative
Distribution
Function

$$P(X \leq k) = 1 - (1 - p)^k$$

$$P(X \geq k) = (1 - p)^{k-1}$$

$$P(X > k) = 1 - P(X \leq k) = (1 - p)^k$$

- What is the probability that someone won with 20 points?
 - The probability that the first “success” (i.e., the game ending) takes at least 21 trials = $P(X \geq 21) = (1 - p)^{k-1} = (1 - \frac{1}{6})^{20} = 0.026$. Seems unlikely (magic dice?)!

How likely is a winning result?

Shifted Geometric Distribution
Cumulative
Distribution
Function

$$P(X \leq k) = 1 - (1 - p)^k$$

$$P(X \geq k) = (1 - p)^{k-1}$$

$$P(X > k) = 1 - P(X \leq k) = (1 - p)^k$$

- What is the probability that someone won with [winning #] points?

How likely is a winning result?

Shifted Geometric Distribution
Cumulative
Distribution
Function

$$P(X \leq k) = 1 - (1 - p)^k$$

$$P(X \geq k) = (1 - p)^{k-1}$$

$$P(X > k) = 1 - P(X \leq k) = (1 - p)^k$$

- What is the probability that someone won with [winning # = n] points?
- The probability that the first “success” (i.e., the game ending) takes at least $n+1$ trials = $P(X \geq n+1) = (1-p)^n = (1-\frac{1}{6})^n = ?$

Using your intuition, on average,
how many rolls do we think it *should*
have taken to roll a 6?

Geometric distribution expectation

- We can model this with a geometric distribution, too!
- The **expected value** of the distribution = on average, how many trials do I have to experience before getting my first “success” (*here “success” = rolling a 6*)?

(Shifted) Geometric Distribution

| | |
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| Mean: | $\mu = E(X) = \frac{1}{p}$ |
| Variance: | $\sigma^2 = V(X) = \frac{(1-p)}{p^2}$ |

$p = \text{probability of success}$

$k = \# \text{ of trials}$

$1 / (1\%) = 6 \text{ rolls}$

**On average, how many rolls do we think
it should take to roll a 20 when using a
20-sided die?**

(Shifted) Geometric Distribution

| | |
|----------------------------------|--|
| Probability mass function | $P(X = k) = p(1-p)^{k-1}$ |
| Cumulative Distribution Function | $P(X \leq k) = 1 - (1-p)^k$ $P(X \geq k) = (1-p)^{k-1}$ $P(X > k) = 1 - P(X \leq k) = (1-p)^k$ |
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$p = \text{probability of success}$

$k = \# \text{ of trials}$

$1 / (1/20) = 20$
rolls

**On average, how many rolls do we think
it should take to roll a 6 when using a
20-sided die?**

(Shifted) Geometric Distribution

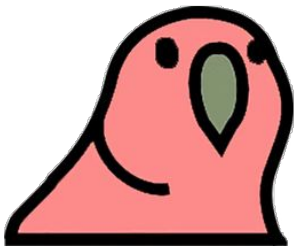
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| Mean: | $\mu = E(X) = \frac{1}{p}$ |
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$p = \text{probability of success}$

$k = \# \text{ of trials}$

Still $1 / (1/20) = 20$ rolls

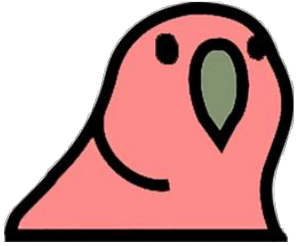
The probability of rolling a specific number (6) out of the total # sides of the die (20) is still 1 number / 20 options



Geometric distribution

We survey students, and we know that on average 3 out of every 75 students love parrots. What is the probability that we find the first parrot-loving student as the 6th student we ask?

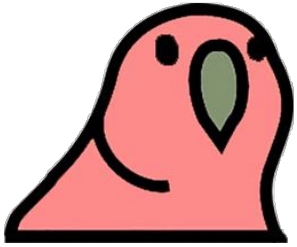
$P(X=k)$ or $P(X \geq k)$?



Geometric distribution

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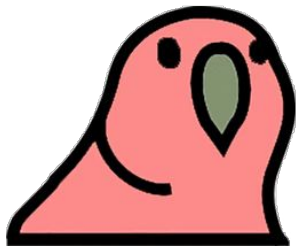
$$\begin{array}{l} p = \boxed{} \\ k = \boxed{} \end{array} \quad \left| \quad \begin{array}{l} \boxed{P(X = k)} = p(1 - p)^{k-1} \\ P(X = \boxed{}) = \boxed{} \end{array} \right.$$



Geometric distribution

We survey students, and we know that on average 3 out of every 75 students love parrots. What is the probability that we find the first parrot-loving student as the 6th student we ask?

$$\begin{array}{l} p = \boxed{} \\ k = \boxed{} \end{array} \left| \begin{array}{l} P(X = k) = p(1 - p)^{k-1} \\ P(X = \boxed{}) = \boxed{} \end{array} \right.$$



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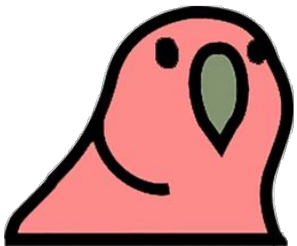
$$p = \frac{3}{75} = 0.04$$

$$k = 6$$

$$P(X = k) = p(1 - p)^{k-1}$$

$$P(X = 6) = 0.04(1 - 0.04)^{6-1}$$

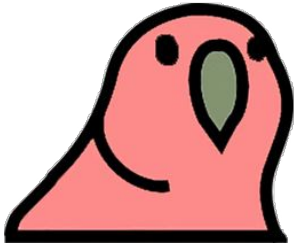
$$P(X = 6) = 0.04(0.96)^5 = 0.0326$$



Geometric distribution

We survey students, and we know that on average 3 out of every 75 students love parrots. What is the probability it takes asking **at least 6 students** to find the first parrot-loving student?

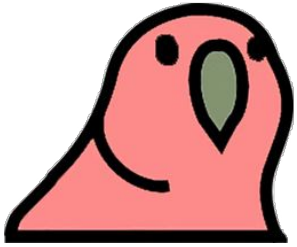
$P(X=k)$ or $P(X \geq k)$?



Geometric distribution

We survey students, and we know that on average 3 out of every 75 students love parrots. What is the probability it takes asking **at least 6 students** to find the first parrot-loving student?

$$\begin{array}{l|l} p = \boxed{} & P(X \geq k) = (1 - p)^{k-1} \\ k \geq \boxed{} & P(X \geq \boxed{}) = \boxed{} \end{array}$$



Geometric distribution

We survey students, and we know that on average 3 out of every 75 students love parrots. What is the probability it takes asking **at least 6 students** to find the first parrot-loving student?

$$p = \frac{3}{75} = 0.04$$

$$k \geq 6$$

$$P(X \geq k) = (1 - p)^{k-1}$$

$$P(X \geq 6) = (1 - 0.04)^{6-1}$$

$$P(X \geq 6) = (0.96)^5 = 0.815$$

Geometric Distribution in Python

1. **No packages:** calculate the pmf or cdf using the formulas we just used
2. **Numpy:** generate the geometric distribution
3. **Scipy:** use `scipy.stats.geom` to then call functions like `pmf()` and `cdf()`

No need for k
because this just
gives you a
random draw from
a geometric
distribution

Geometric Distribution in Python

`numpy.random.geometric`

`random.geometric(p, size=None)`

Draw samples from the geometric distribution.

Bernoulli trials are experiments with one of two outcomes: success or failure (an example of such an experiment is flipping a coin). The geometric distribution models the number of trials that must be run in order to achieve success. It is therefore supported on the positive integers, $k = 1, 2, \dots$

The probability mass function of the geometric distribution is

$$f(k) = (1 - p)^{k-1}p$$

where p is the probability of success of an individual trial.

How many draws you want
from the distribution

| | |
|--|--|
| <code>rvs(p, loc=0, size=1, random_state=None)</code> | Random variates. |
| <code>pmf(k, p, loc=0)</code> | Probability mass function. |
| <code>logpmf(k, p, loc=0)</code> | Log of the probability mass function. |
| <code>cdf(k, p, loc=0)</code> | Cumulative distribution function. |
| <code>logcdf(k, p, loc=0)</code> | Log of the cumulative distribution function. |
| <code>sf(k, p, loc=0)</code> | Survival function (also defined as <code>1 - cdf</code> , but <code>sf</code> is sometimes more accurate). |
| <code>logsf(k, p, loc=0)</code> | Log of the survival function. |
| <code>ppf(q, p, loc=0)</code> | Percent point function (inverse of <code>cdf</code> — percentiles). |
| <code>isf(q, p, loc=0)</code> | Inverse survival function (inverse of <code>sf</code>). |
| <code>stats(p, loc=0, moments='mv')</code> | Mean('m'), variance('v'), skew('s'), and/or kurtosis('k'). |
| <code>entropy(p, loc=0)</code> | (Differential) entropy of the RV. |
| <code>expect(func, args=(p,), loc=0, lb=None, ub=None, conditional=False)</code> | Expected value of a function (of one argument) with respect to the distribution. |
| <code>median(p, loc=0)</code> | Median of the distribution. |
| <code>mean(p, loc=0)</code> | Mean of the distribution. |
| <code>var(p, loc=0)</code> | Variance of the distribution. |
| <code>std(p, loc=0)</code> | Standard deviation of the distribution. |
| <code>interval(confidence, p, loc=0)</code> | Confidence interval with equal areas around the median. |

Geometric Distribution in Python

`scipy.stats.geom` #

`scipy.stats.geom = <scipy.stats._discrete_distns.geom_gen object>`

A geometric discrete random variable.

Recap on geometric distribution:

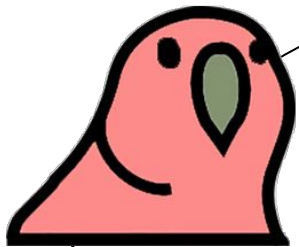
Use when...

1. Each trial has only 2 possible outcomes: “success” (1) or “failure” (0)
2. You conduct each “trial” exactly the same way and **count the # trials until the first success**
3. The probability of success p is the same for each trial
4. Trials are independent

1 min break

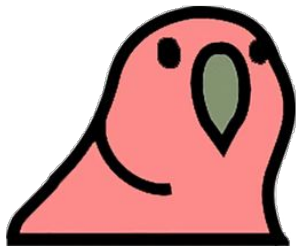
Draw your
favorite emoji,
if you'd like!





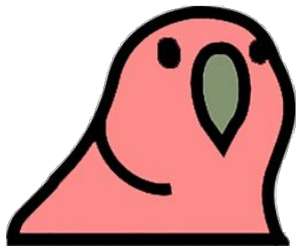
Geometric Distribution

We survey students, and we know that on average 3 out of every 75 students love parrots. What is the probability it takes asking **6 students** to find the **first** parrot-loving student?



Can we still use geometric distribution?

We survey students, and we know that on average 3 out of every 75 students love parrots. What is the probability it takes asking **6 students** to find the first **three** parrot-loving students?



Can we still use geometric distribution?

We survey students, and we know that on average 3 out of every 75 students love parrots. What is the probability it takes asking **6 students** to find the first **three** parrot-loving students?

No, we need to introduce a new distribution!

Negative Binomial

Event space: integers r to infinity

Parameters: chance of "success" p

Story: how many trials before the ~~first~~ r^{th} success?

Negative Binomial

Event space: integers **r** to infinity

Parameters: chance of "success" p

Story: how many trials before the **first** **rth** success?

$$f(x) = P(X = x) = \binom{x-1}{r-1} (1-p)^{x-r} p^r$$

Negative Binomial

Event space: integers **r** to infinity

Parameters: chance of "success" p

Story: how many trials before the **first** **rth** success?

$$f(x) = P(X = \textcircled{x}) = \binom{x-1}{r-1} (1-p)^{x-r} p^r$$

Notation? Where'd k go?

Negative Binomial

Event space: integers **r** to infinity

Parameters: chance of "success" p

Story: how many trials before the **first** **rth** success?

$$f(x) = P(X = \textcircled{x}) = \binom{x-1}{r-1} (1-p)^{x-r} p^r$$

little x represents the same thing as k

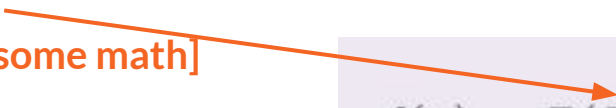
Negative Binomial

Event space: integers **r** to infinity

Parameters: chance of "success" p

Story: how many trials before the **first** r^{th} success?

```
def P(X, r, p):  
    return [some math]
```


$$f(x) = P(X = x) = \binom{x-1}{r-1} (1-p)^{x-r} p^r$$


Negative Binomial

Event space: integers **r** to infinity

Parameters: chance of "success" p

Story: how many trials before the **first** r^{th} success?

def P(X, r, p):
 return [some math]



The diagram shows two orange arrows. One arrow originates from the 'return [some math]' line in the code block and points to the $f(x)$ term in the equation. The other arrow originates from the 'P(X=4, r=3, p=0.5)' example and points to the x in the equation. The x is circled in orange.

$$f(x) = P(X = x) = \binom{x-1}{r-1} (1-p)^{x-r} p^r$$

P(X=4, r=3, p=0.5)

Negative Binomial

Event space: integers **r** to infinity

Parameters: chance of "success" p

Story: how many trials before the **first** r^{th} success?

The probability of running 4 trials before reaching the 3rd success, if each success occurs with probability 0.5

$$f(x) = P(X = \textcircled{x}) = \binom{x-1}{r-1} (1-p)^{x-r} p^r$$

$P(X=4, r=3, p=0.5)$

Negative Binomial

Geometric distributions are just a special case of negative binomial when $r = 1$!

Geometric PMF →

$$P(X = k) = p(1 - p)^{k-1}$$

Neg Bin PMF →

$$f(x) = P(X = x) = \binom{x-1}{r-1} (1-p)^{x-r} p^r$$

Negative Binomial

Geometric distributions are just a special case of negative binomial when $r = 1$!

Geometric PMF \rightarrow

$$P(X = k) = p(1 - p)^{k-1}$$

Neg Bin PMF
when $r = 1 \rightarrow$

$$f(x) = P(X = x) = \binom{x-1}{r-1} (1-p)^{x-r} p^r$$

($x-1$ choose 0) is 1 (there's only 1 way to pick nothing)

Negative Binomial

Geometric distributions are just a special case of negative binomial when $r = 1$!

Geometric PMF \rightarrow

$$P(X = k) = p(1 - p)^{k-1}$$

Neg Bin PMF
when $r = 1 \rightarrow$


$$f(x) = P(X = x) = (1 - p)^{x-1} p$$

Negative Binomial

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Negative Binomial

Event space: integers r to infinity

Parameters: chance of "success" p

Story: how many trials before the r^{th} success?

$$f(x) = P(X = x) = \binom{x-1}{r-1} (1-p)^{x-r} p^r$$

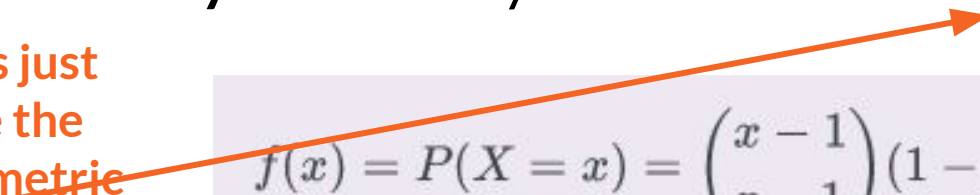
Negative Binomial

Event space: integers r to infinity

Parameters: chance of "success" p

Story: how many trials before the r^{th} success?

If $r = 1$, then this is just the # trials before the first success (geometric distribution)


$$f(x) = P(X = x) = \binom{x-1}{r-1} (1-p)^{x-r} p^r$$

Negative Binomial Properties

$$\mu = E(X) = \frac{r}{p}$$

$$\sigma^2 = Var(x) = \frac{r(1-p)}{p^2}$$

Negative Binomial Properties

At what value of r does the expected value of Negative Binomial dist equal the expected value of Geometric distribution?

$$\mu = E(X) = \frac{r}{p}$$

$$\sigma^2 = Var(x) = \frac{r(1-p)}{p^2}$$

Negative Binomial Properties

When $r=1$, we get $E(X) = 1/p$, which is what we had with the geometric distribution expected value!

$$\mu = E(X) = \frac{r}{p}$$

$$\sigma^2 = Var(x) = \frac{r(1-p)}{p^2}$$

Negative Binomial in Python

numpy.random.negative_binomial

`random.negative_binomial(n, p, size=None)`

Draw samples from a negative binomial distribution.

Samples are drawn from a negative binomial distribution with specified parameters, n successes and p probability of success where n is > 0 and p is in the interval $[0, 1]$.

Negative Binomial

Event space: integers r to infinity

Parameters: chance of "success" p

Story: how many trials before the ~~first~~ r^{th} success?

Be careful: lots of
different variants of
the NB distribution!

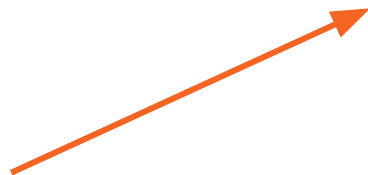
Negative Binomial(ish)

Event space: integers 0 to infinity

Parameters: chance of "success" p

Story: how many ~~trials~~ before the ~~first~~ r^{th} success?
failures

This is an 'alternative' formulation to the NB distribution, sometimes called the Pascal distribution



Neg Bin (Pascal) Distribution Properties

X = how many **failures** before the **r^{th}** success?

$$\Pr(X = k) = \binom{k + r - 1}{r - 1} (1 - p)^k p^r$$

the mean is $(1 - p)r/p$ and the variance is $(1 - p)r/p^2$

Neg Bin (Pascal) Distribution Properties

X = how many **failures** before the **r^{th}** success?

$$\Pr(X = k) = \binom{k + r - 1}{r - 1} (1 - p)^k p^r$$

the mean is $(1 - p)r/p$ and the variance is $(1 - p)r/p^2$

**Lots of programming languages (e.g. R)
default to the Pascal distribution when you
ask for a negative binomial distribution!**

Neg Bin vs. Neg Bin (Pascal)

Neg Bin PMF →

$$f(x) = P(X = x) = \binom{x-1}{r-1} (1-p)^{x-r} p^r$$

Pascal PMF →

$$\Pr(X = k) = \binom{k+r-1}{r-1} (1-p)^k p^r$$

Neg Bin vs. Neg Bin (Pascal)

Neg Bin PMF →

$$f(x) = P(X = x) = \binom{x-1}{r-1} (1-p)^{x-r} p^r$$

Pascal PMF →

$$\Pr(X = k) = \binom{k+r-1}{r-1} (1-p)^k p^r$$

If we let $x = k+r$, these are the exact same thing!

(Analogous to shifting by r , like between geometric dist and shifted geometric dist shifted by $r=1$)

Neg Bin vs. Neg Bin (Pascal)

Counting total trials x
to reach r successes \rightarrow

$$f(x) = P(X = x) = \binom{x-1}{r-1} (1-p)^{x-r} p^r$$

Counting total failures k
until r successes \rightarrow

$$\Pr(X = k) = \binom{k+r-1}{r-1} (1-p)^k p^r$$

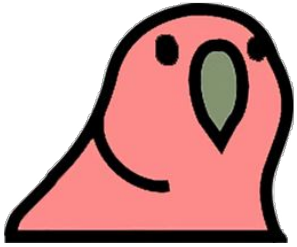
It is always true that $x = k+r$, since
total trials x = total failures k + total successes r

Tons of different formulations!

Be careful: always check documentation for distribution properties!

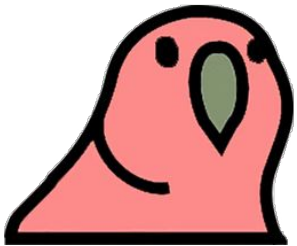
If you blindly copy numbers out of Python, you might be interpreting things incorrectly!

| | X is counting... | Probability mass function | Formula | Alternate formula (using equivalent binomial) | Alternate formula (simplified using: $n = k + r$) | Support |
|---|-------------------------------|----------------------------------|---|--|--|------------------------------|
| 1 | k failures, given r successes | $f(k; r, p) \equiv \Pr(X = k) =$ | $\binom{k+r-1}{k} p^r (1-p)^k$ [7][5][8] | $\binom{k+r-1}{r-1} p^r (1-p)^k$ [2] [9][10][11] | $\binom{n-1}{k} p^r (1-p)^k$ | for $k = 0, 1, 2, \dots$ |
| 2 | n trials, given r successes | $f(n; r, p) \equiv \Pr(X = n) =$ | $\binom{n-1}{r-1} p^r (1-p)^{n-r}$ [5][11][12][13][14] | $\binom{n-1}{n-r} p^r (1-p)^{n-r}$ | | for $n = r, r+1, r+2, \dots$ |
| 3 | n trials, given r failures | $f(n; r, p) \equiv \Pr(X = n) =$ | $\binom{n-1}{r-1} p^{n-r} (1-p)^r$ | $\binom{n-1}{n-r} p^{n-r} (1-p)^r$ | | |
| 4 | k successes, given r failures | $f(k; r, p) \equiv \Pr(X = k) =$ | $\binom{k+r-1}{k} p^k (1-p)^r$ | $\binom{k+r-1}{r-1} p^k (1-p)^r$ | $\binom{n-1}{k} p^k (1-p)^r$ | for $k = 0, 1, 2, \dots$ |
| - | k successes, given n trials | $f(k; n, p) \equiv \Pr(X = k) =$ | This is the binomial distribution not the negative binomial: $\binom{n}{k} p^k (1-p)^{n-k}$ | | | for $k = 0, 1, 2, \dots, n$ |



Negative Binomial example

We survey students, and we know that on average 3 out of every 75 students love parrots. What is the probability it takes asking **6 students** to find the first **3** parrot-loving students?



Negative Binomial example

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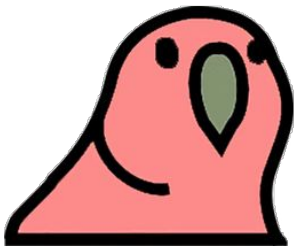
X = number of students to survey

$x = ?$

$r = ?$

$p = ?$

Want to calculate $P(?)$



Negative Binomial example

We survey students, and we know that on average 3 out of every 75 students love parrots. What is the probability it takes asking **6 students** to find the first **3** parrot-loving students?

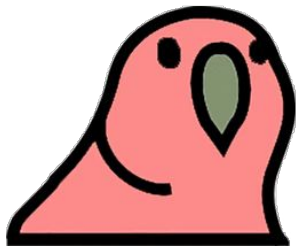
X = number of students to survey

$x = 6 (?)$

$r = 3$

$p = 3/75 = 0.04$

Want to calculate $P(X=x)$

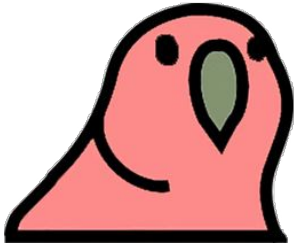


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We survey students, and we know that on average 3 out of every 75 students love parrots. What is the probability it takes asking **6 students** to find the first **3** parrot-loving students?

X = number of students to survey
 $x = 6$, $r = 3$, $p = 3/75$, want to find $P(X = x)$

Formulation: do we want to count
total # trials or total # failures?

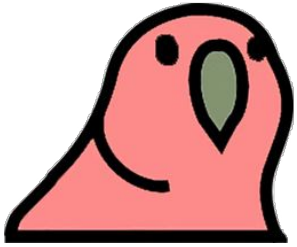


Negative Binomial example

We survey students, and we know that on average 3 out of every 75 students love parrots. What is the probability it takes asking **6 students** to find the first **3** parrot-loving students?

X = number of students to survey
 $x = 6, r = 3, p = 3/75$, want to find $P(X = x)$
where $x =$ **total number of trials**

$$f(x) = P(X = x) = \binom{x-1}{r-1} (1-p)^{x-r} p^r$$



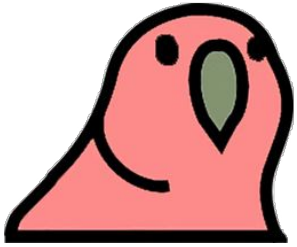
Negative Binomial example

We survey students, and we know that on average 3 out of every 75 students love parrots. What is the probability it takes asking **6 students** to find the first **3** parrot-loving students?

If x = total # failures, then we'd *instead* want to study $P(X=3)$ using the Pascal dist.

X = number of students to survey
 $x = 6, r = 3, p = 3/75$, want to find $P(X = x)$
where x = **total number of trials**

$$f(x) = P(X = x) = \binom{x-1}{r-1} (1-p)^{x-r} p^r$$

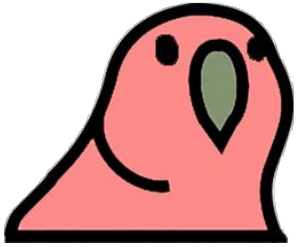


Negative Binomial example

We survey students, and we know that on average 3 out of every 75 students love parrots. **What is the probability it takes asking 6 students to find the first 3 parrot-loving students?**

X = number of students to survey
 $x = 6, r = 3, p = 3/75$, want to find $P(X = x)$
where x = total number of trials

$$f(x) = P(X = x) = \binom{x-1}{r-1} (1-p)^{x-r} p^r$$

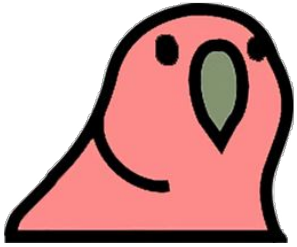


Negative Binomial example

We survey students, and we know that on average 3 out of every 75 students love parrots. What is the probability it takes asking **6 students** to find the first **3** parrot-loving students?

X = total number of trials = total number of students to survey (i.e. trials until 3 successes)

$$P(X = k) = \binom{5}{2} * (1 - (3/75))^{6-3} * (3/75)^3 = 10 * 0.96^3 * 0.04^3 = 0.000566$$



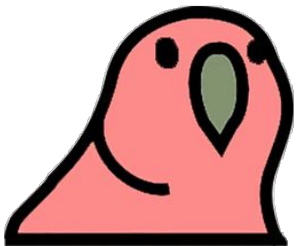
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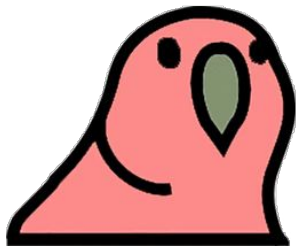
$$P(X = k) = \binom{5}{2} * (1 - (3/75))^{6-3} * (3/75)^3 = 10 * 0.96^3 * 0.04^3 = 0.000566$$

Unlikely it'd take only 6 trials to find 3 parrot lovers!



Negative Binomial example

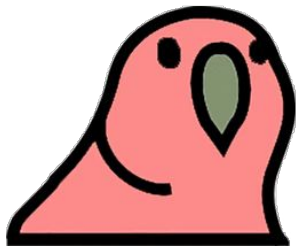
We survey students, and we know that on average 3 out of every 75 students love parrots. What is the probability it takes asking **6 not-parrot-loving students** to find the first **3** parrot-loving students?



Negative Binomial example

We survey students, and we know that on average 3 out of every 75 students love parrots. What is the probability it takes asking **6 not-parrot-loving students** to find the first **3** parrot-loving students?

Formulation: do we want to count
total # trials or total # failures?

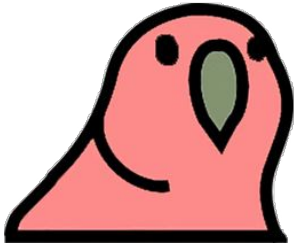


Negative Binomial example

We survey students, and we know that on average 3 out of every 75 students love parrots. What is the probability it takes asking **6 not-parrot-loving students** to find the first **3** parrot-loving students?

$$\Pr(X = k) = \binom{k + r - 1}{r - 1} (1 - p)^k p^r$$

(Now, k is counting **total # failures**)

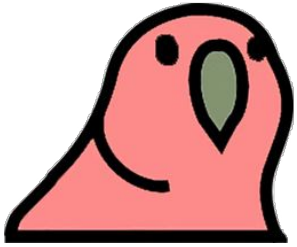


Negative Binomial example

We survey students, and we know that on average 3 out of every 75 students love parrots. What is the probability it takes asking **6 not-parrot-loving students** to find the first **3** parrot-loving students?

X = number of students to fail in survey
 $k = 6, r = 3, p = 3/75$, want to find $P(X = k)$

$$\Pr(X = k) = \binom{k + r - 1}{r - 1} (1 - p)^k p^r$$



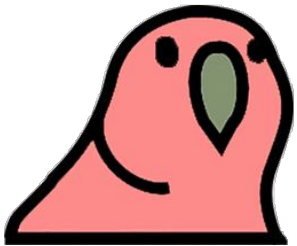
Negative Binomial example

We survey students, and we know that on average 3 out of every 75 students love parrots. What is the probability it takes asking **6 not-parrot-loving students** to find the first **3** parrot-loving students?

Alternatively, we could have X = total # surveyed using our Neg Bin formulation from before, but use $\Pr(X=9)$

X = number of students to fail in survey
 $k = 6$, $r = 3$, $p = 3/75$, want to find $P(X = k)$

$$\Pr(X = k) = \binom{k + r - 1}{r - 1} (1 - p)^k p^r$$

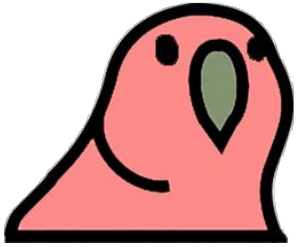


Negative Binomial example

We survey students, and we know that on average 3 out of every 75 students love parrots. What is the probability it takes asking **6 not-parrot-loving students** to find the first **3** parrot-loving students?

$X =$ **number of students to fail in survey**

$$P(X = k) = (8 \text{ choose } 2) * (1 - (3/75))^6 * (3/75)^3 =$$
$$\mathbf{28 * 0.96^6 * 0.04^3 = 0.0014}$$



Negative Binomial example

We survey students, and we know that on average 3 out of every 75 students love parrots. What is the probability it takes asking **6 not-parrot-loving students** to find the first **3** parrot-loving students?

$X =$ **number of students to fail in survey**

$$P(X = k) = (8 \text{ choose } 2) * (1 - (3/75))^6 * (3/75)^3 = \\ \mathbf{28 * 0.96^6 * 0.04^3} = 0.0014$$

More likely to find 3 parrot lovers when you have 6 failures as compared to having 6 total trials (0.000566)



Which NB
formulation is
this? (**k = # trials**
or **# failures?**)

Negative Binomial Distribution

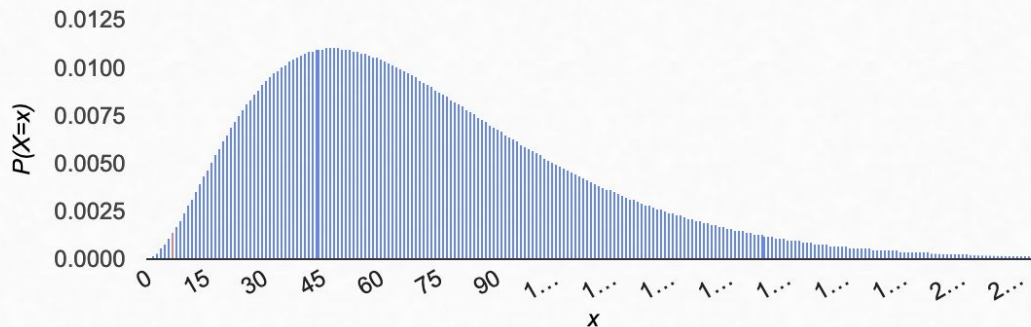
$$X \sim NB(r, p) (I)$$

$r =$ 3

$p =$ 0.04

$x =$ 6

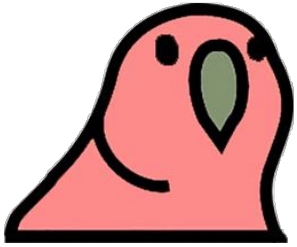
$P(X=x) =$ 0.0014



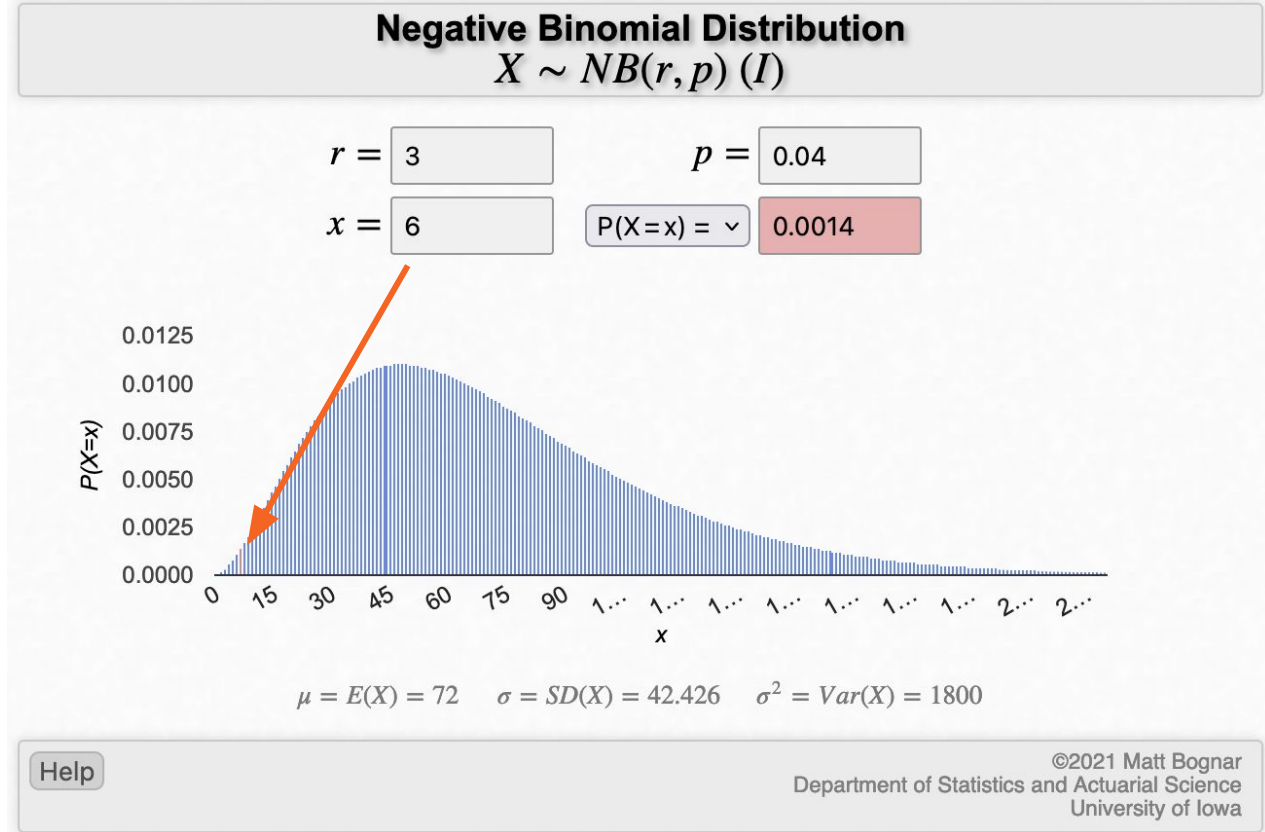
$$\mu = E(X) = 72 \quad \sigma = SD(X) = 42.426 \quad \sigma^2 = Var(X) = 1800$$

Help

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Department of Statistics and Actuarial Science
University of Iowa



$\Pr(X=6) = 0.0014$
for **X = # failures**



1 min break & attendance



tinyurl.com/22wmzshh

Admin

- Phase 3 due Thursday 11/3
- HW5 due Tuesday 11/7
 - Problem C2 typo: should say “tick_label” instead of “tick_labels”

A new distribution!

Story: number of events that happen during an interval

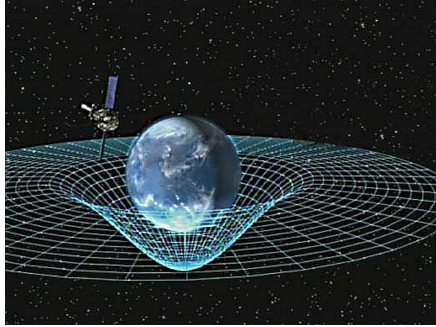
Poisson Distribution

Event space: integers 0 to infinity

Parameters: rate $\lambda > 0$

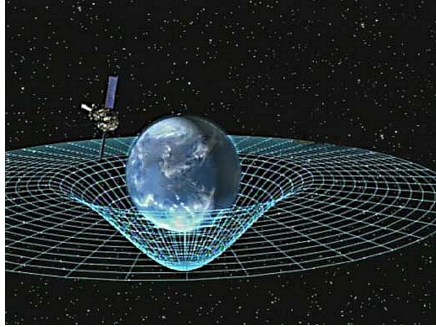
Story: number of events that happen during an interval

Poisson Examples



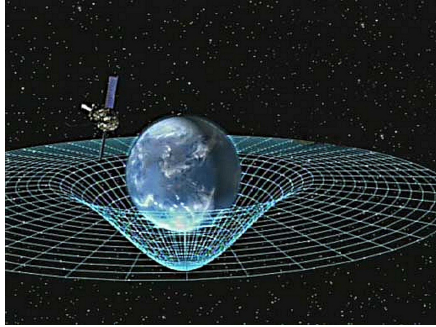
- “Intervals” can occur in time or **space**
- Our story X can represent...
 - The # students arriving to Instructor Thalken’s OH
 - The # typos on a textbook page

Poisson Examples



- “Intervals” can occur in **time** or space
- Our story X can represent...
 - The # students arriving to Instructor Thalken’s OH
 - The # typos on a textbook page
 - The # mobile users seeing an ad each minute
 - The # Steph Curry 3pt shots per 36 minutes


Poisson Examples



- “Intervals” can occur in **time or space**
- Our story X can represent...
 - The # students arriving to Instructor Thalken’s OH
 - The # typos on a textbook page
 - The # mobile users seeing an ad each minute
 - The # Steph Curry 3pt shots per 36 minutes
 - The # customers at the Gates Gimme in 10 minute intervals

Poisson Properties

Probability that k
events happened in
an interval



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

$$\lambda = E(X) = \text{Var}(X)$$

Poisson Properties

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

"rate"
lambda


$$\lambda = E(X) = \text{Var}(X)$$

How many 100-year floods?

If massive floods occur randomly with average 1 in 100 years, how many "100-year" floods do you expect within a 100 year period?

How many 100-year floods?

If massive floods occur randomly with average 1 in 100 years, how many "100-year" floods do you expect within a 100 year period?

Binomial model: **What is N?** ____ **What is p?** ____

(This isn't asking about *negative binomials*.
Remember binomial models like coin flips?)

How many 100-year floods?

If massive floods occur randomly with **average 1 in 100 years**, how many "100-year" floods do you expect within a **100 year period**?

Binomial model: What is N ? **100** What is p ? **1/100**

How many 100-year floods?

If massive floods occur randomly with average 1 in 100 years, how many "100-year" floods do you expect within a 100 year period?

Binomial model: What is N ? **100** What is p ? **1/100**

In Python (scipy.stats): For each number of floods k , the probability is **`binom.pmf(k, 100, 0.01)`**

How many 100-year floods?

If massive floods occur randomly with average 1 in 100 years, how many "100-year" floods do you expect within a 100 year period?

Binomial model: What is N? **100** What is p? **1/100**

| | 0 floods | 1 | 2 | 3 | 4 |
|-----------------|-----------------|----------|----------|----------|----------|
| Binomial | 36.6% | 36.9% | 18.4% | 6.0% | 1.5% |

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`binom.pmf(2, 100, 0.01)`

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If you wanted a cumulative probability, **`binom.cdf(2, 100, 0.01)`**

What if we use Poisson instead?

If massive floods occur randomly with **average 1** in 100 years, how many "100-year" floods do you expect within a 100 year period?

Poisson model: Rate $\lambda =$

| | | | | | |
|--|-----------------|----------|----------|----------|----------|
| | 0 floods | 1 | 2 | 3 | 4 |
|--|-----------------|----------|----------|----------|----------|

What if we use Poisson instead?

If massive floods occur randomly with **average 1** in 100 years, how many "100-year" floods do you expect within a 100 year period?

Poisson model: Rate $\lambda = 1$

| | 0 floods | 1 | 2 | 3 | 4 |
|---------|----------|-------|-------|------|------|
| Poisson | 36.8% | 36.8% | 18.3% | 6.3% | 1.5% |

What if we use Poisson instead?

If massive floods occur randomly with **average 1** in 100 years, how many "100-year" floods do you expect within a 100 year period?

Poisson model: Rate $\lambda = 1$

| | 0 floods | 1 | 2 | 3 | 4 |
|---------|----------|-------|-------|------|------|
| Poisson | 36.8% | 36.8% | 18.3% | 6.3% | 1.5% |

In Python (scipy.stats): For each number of floods k , the probability is `poisson.pmf(k,1)`

What if we use Poisson instead?

If massive floods occur randomly with **average 1** in 100 years, how many "100-year" floods do you expect within a 100 year period?

Poisson model: Rate $\lambda = 1$

| | 0 floods | 1 | 2 | 3 | 4 |
|---------|----------|-------|-------|------|------|
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`poisson.pmf()`

What if we use Poisson instead?

If massive floods occur randomly with **average 1** in 100 years, how many "100-year" floods do you expect within a 100 year period?

Poisson model: Rate $\lambda = 1$

| | 0 floods | 1 | 2 | 3 | 4 |
|---------|----------|-------|-------|------|------|
| Poisson | 36.8% | 36.8% | 18.3% | 6.3% | 1.5% |

`poisson.pmf(2,1)`

What if we use Poisson instead?

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Poisson model: Rate $\lambda = 1$

| | 0 floods | 1 | 2 | 3 | 4 | |
|---------|----------|-------|-------|------|------|--|
| Poisson | 36.8% | 36.8% | 18.3% | 6.3% | 1.5% | |

If you want to go cumulative in the other direction, instead of using cdf, you can use **poisson.sf(2,1)**

What if we use Poisson instead?

If massive floods occur randomly with average 1 in 100 years, how many "100-year" floods do you expect within a 100 year period?

Poisson model: Rate $\lambda = 1$

These results are pretty similar, even though they use different distributions!

| | 0 floods | 1 | 2 | 3 | 4 |
|----------|----------|-------|-------|------|------|
| Binomial | 36.6% | 36.9% | 18.4% | 6.0% | 1.5% |
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Why would you ever use Poisson instead of Binomial?

If massive floods occur randomly with average 1 in 100 years, how many "100-year" floods do you expect within a 100 year period?

Poisson model: Rate $\lambda = 1$

| | 0 floods | 1 | 2 | 3 | 4 |
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Why would you ever use Poisson instead of Binomial?

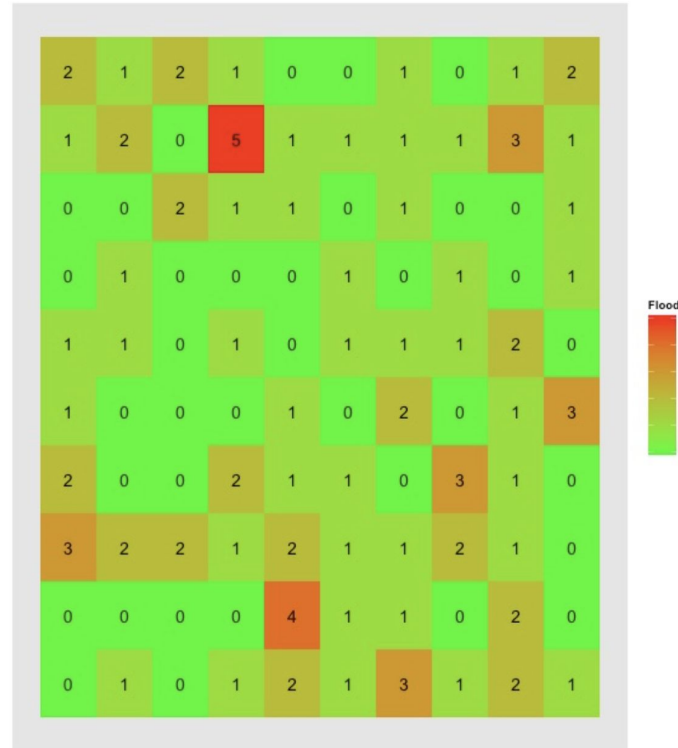
If massive floods occur randomly with average 1 in 100 years, how many "100-year" floods do you expect within a 100 year period?

Use Poisson when you aren't sure what N and p are, but you can guess the mean # of occurrences

Poisson model: Rate $\lambda = 1$

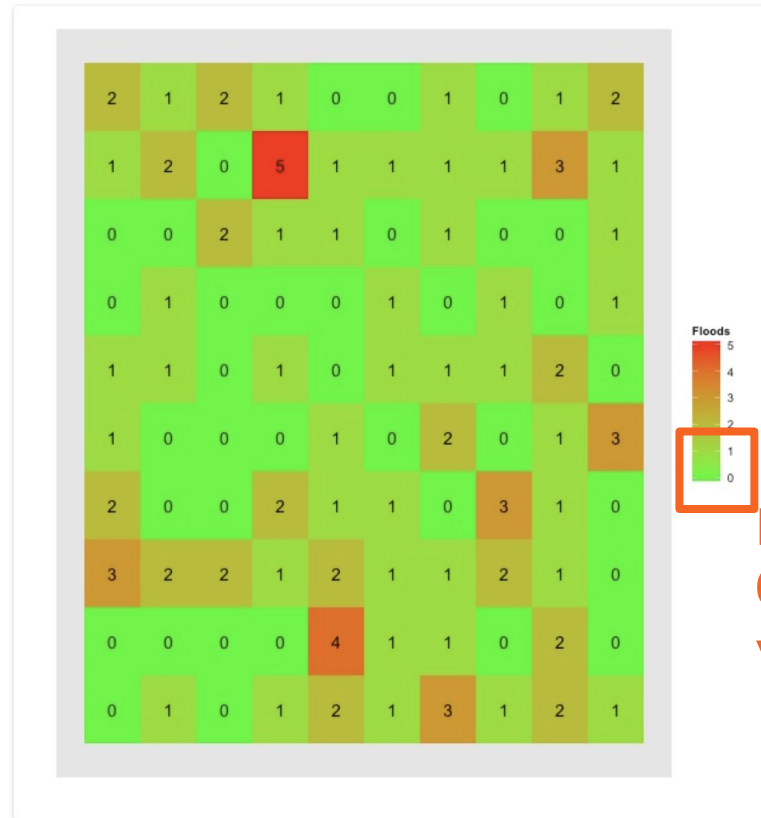
| | 0 floods | 1 | 2 | 3 | 4 |
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| Binomial | 36.6% | 36.9% | 18.4% | 6.0% | 1.5% |
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100 simulations from a Poisson distribution



Simulation of the number of 100-year floods occurring in 100 (10 x 10), 100-year sequences

100 simulations from a Poisson distribution

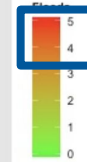
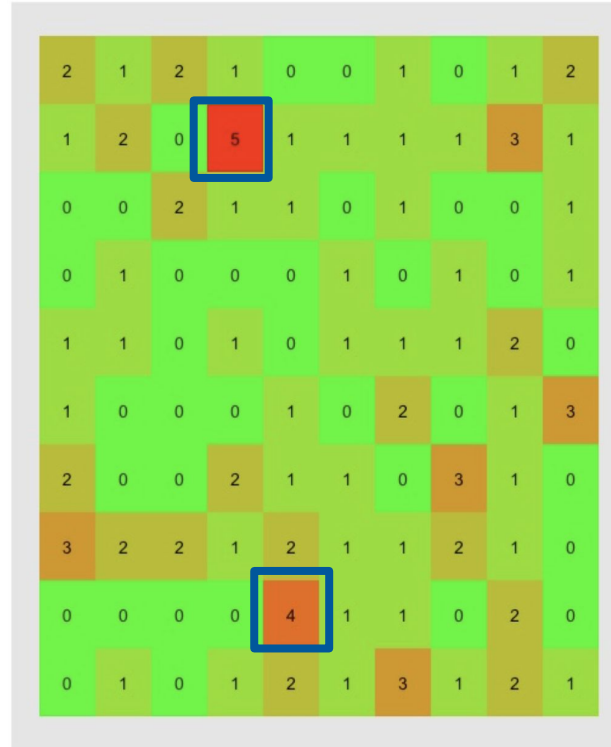


Simulation of the number of 100-year floods occurring in 100 (10 x 10), 100-year sequences

Mostly expecting
0-1 floods in 100
years

100 simulations
from a Poisson
distribution

In 2/100
simulations we
predict 4-5 floods
in a 100 year span



Simulation of the number of 100-year floods occurring in 100 (10 x 10), 100-year sequences



Poisson example

of Steph Curry 3P shots per game $\sim \text{Poisson}(4.5)$

of Andrew Wiggins 3P per game $\sim \text{Poisson}(2.5)$



Poisson example

of Steph Curry 3P shots per game $\sim \text{Poisson}(4.5)$

of Andrew Wiggins 3P per game $\sim \text{Poisson}(2.5)$

In Python, to get a random draw from these distributions:

`np.random.poisson(4.5, size=None)`

`np.random.poisson(2.5, size=None)`



Poisson is additive and divisible

of Steph Curry 3P shots per game $\sim \text{Poisson}(4.5)$

of Steph Curry 3P shots per **1/2 game** $\sim \text{Poisson}(?)$



Poisson is additive and divisible

of Steph Curry 3P shots per game \sim Poisson(4.5)

Divide 4.5 by 2!

of Steph Curry 3P shots per 1/2 game \sim Poisson(2.25)



Poisson is additive and divisible

of Steph Curry 3P shots per game $\sim \text{Poisson}(4.5)$

of Andrew Wiggins 3P per game $\sim \text{Poisson}(2.5)$

Curry OR Wiggins 3P shots per game $\sim \text{Poisson}(?)$



Poisson is additive and divisible

of Steph Curry 3P shots per game \sim Poisson(4.5)

of Andrew Wiggins 3P per game \sim Poisson(2.5)

We can just sum these!

Curry OR Wiggins 3P shots per game \sim Poisson(7)

Poisson Properties

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

$$\lambda = E(X) = \text{Var}(X)$$

If rate $\lambda = 4$, what is the standard deviation of the Poisson distribution?

Poisson Properties

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

$$\lambda = E(X) = \text{Var}(X)$$

$$\text{Stdev} = \text{sqrt}(\text{Var}(X)) = \text{sqrt}(\lambda) = \text{sqrt}(4) = 2$$

Poisson Properties

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

$$\lambda = E(X) = \text{Var}(X)$$

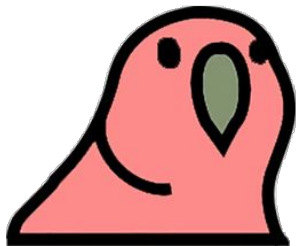
If rate $\lambda = 400$, what is the standard deviation of the Poisson distribution?

Poisson Properties

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

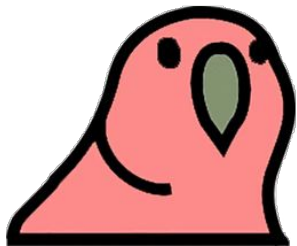
$$\lambda = E(X) = \text{Var}(X)$$

$$\text{Stdev} = \text{sqrt}(\text{Var}(X)) = \text{sqrt}(\lambda) = \text{sqrt}(400) = 20$$



Poisson Example

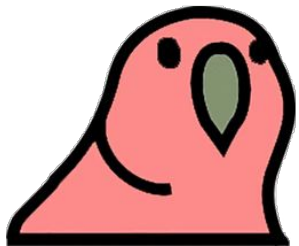
There is a mean of 3 parrot-lovers per discussion section.
What is the probability that a randomly selected discussion section has one parrot-lover?



Poisson Example

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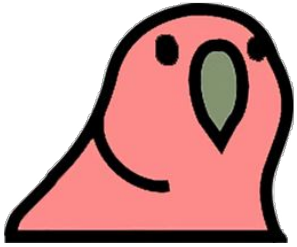
X = number of parrot lovers in a discussion section
 $\lambda = ?$



Poisson Example

There is a mean of 3 parrot-lovers per discussion section.
What is the probability that a randomly selected discussion section has one parrot-lover?

X = number of parrot lovers in a discussion section
 λ = **mean = 3**

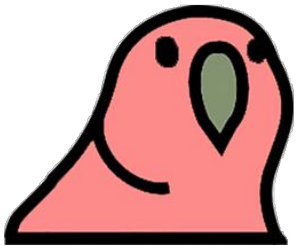


Poisson Example

There is a mean of 3 parrot-lovers per discussion section.
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We are trying to find: $P(?)$

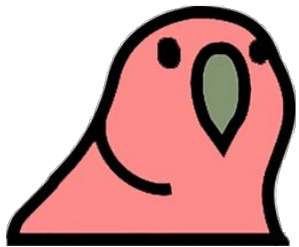


Poisson Example

There is a mean of 3 parrot-lovers per discussion section.
What is the probability that a randomly selected discussion section has **one parrot-lover**?

X = number of parrot lovers in a discussion section
 λ = mean = 3

We are trying to find: $P(X=1)$

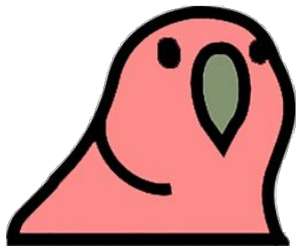


Poisson Example

There is a mean of 3 parrot-lovers per discussion section. What is the probability that a randomly selected discussion section has **one** parrot-lover?

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

$$P(X=1) = 3^1 * e^{-3} / 1! = 3e^{-3} = 0.15$$

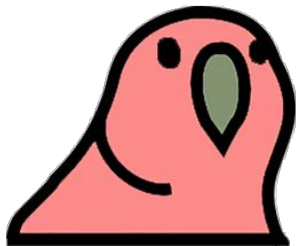


Poisson Example

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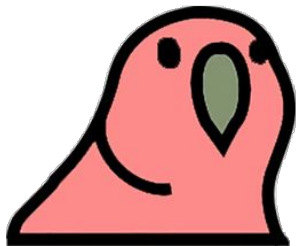


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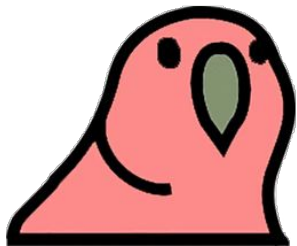


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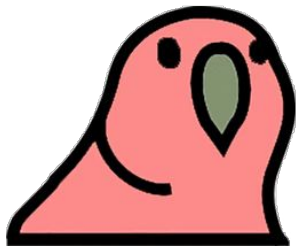


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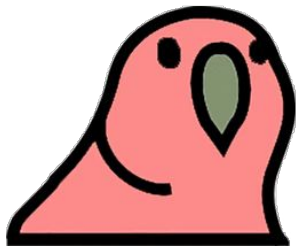
Poisson Example

There is a mean of 3 parrot-lovers per discussion section. What is the **probability** that a randomly selected discussion section has one parrot-lover?

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

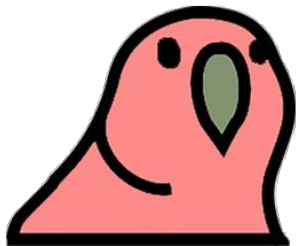
$$P(X=1) = 3^1 * e^{-3} / 1! = 3e^{-3} = \mathbf{0.15}$$

**Final
answer:
15%**



Poisson Example

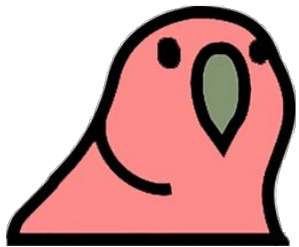
There is a mean of 3 parrot-lovers per discussion section. What is the probability that a randomly selected discussion section has **at least one** parrot-lover?



Poisson Example

There is a mean of 3 parrot-lovers per discussion section. What is the probability that a randomly selected discussion section has **at least one** parrot-lover?

We are trying to find: $P(?)$



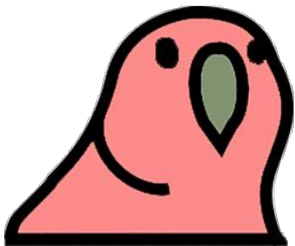
Poisson Example

There is a mean of 3 parrot-lovers per discussion section. What is the probability that a randomly selected discussion section has **at least one** parrot-lover?

We are trying to find: $P(X \geq 1)$

But, we only showed you the PMF and not the CDF

Can we still solve this?

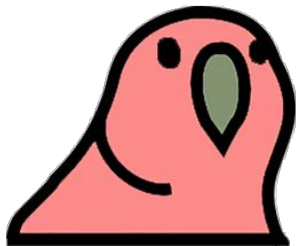


Poisson Example

There is a mean of 3 parrot-lovers per discussion section. What is the probability that a randomly selected discussion section has **at least one** parrot-lover?

Fill in the blank:

$$P(X \geq 1) = 1 - \underline{\hspace{2cm}}$$

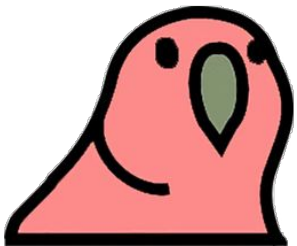


Poisson Example

There is a mean of 3 parrot-lovers per discussion section. What is the probability that a randomly selected discussion section has **at least one** parrot-lover?

$$P(X \geq 1) = 1 - P(X < 1) = 1 - P(X = 0)$$

Discrete numbers mean that we can add up our PMF values to get cumulative probabilities!

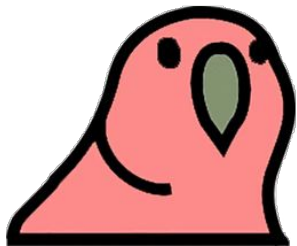


Poisson Example

There is a mean of 3 parrot-lovers per discussion section. What is the probability that a randomly selected discussion section has **at least one** parrot-lover?

$$P(X \geq 1) = 1 - P(X=0) = ?$$

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



Poisson Example

There is a mean of 3 parrot-lovers per discussion section. What is the probability that a randomly selected discussion section has **at least one** parrot-lover?

$$P(X \geq 1) = 1 - P(X=0) = 1 - [3^0 * e^{-3} / 0!] = \mathbf{1 - e^{-3}} = 0.95$$

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

Poisson Distribution

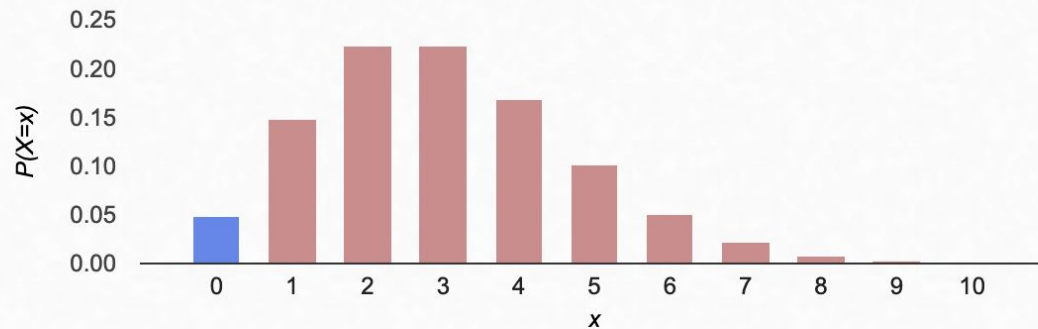
$$X \sim \text{Pois}(\lambda)$$

$$\lambda = 3$$

$$x = 1$$

$P(X \geq x) = \downarrow$

0.95021



$$\mu = E(X) = 3 \quad \sigma = SD(X) = 1.732 \quad \sigma^2 = Var(X) = 3$$

Help

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Department of Statistics and Actuarial Science
University of Iowa

Poisson Distribution

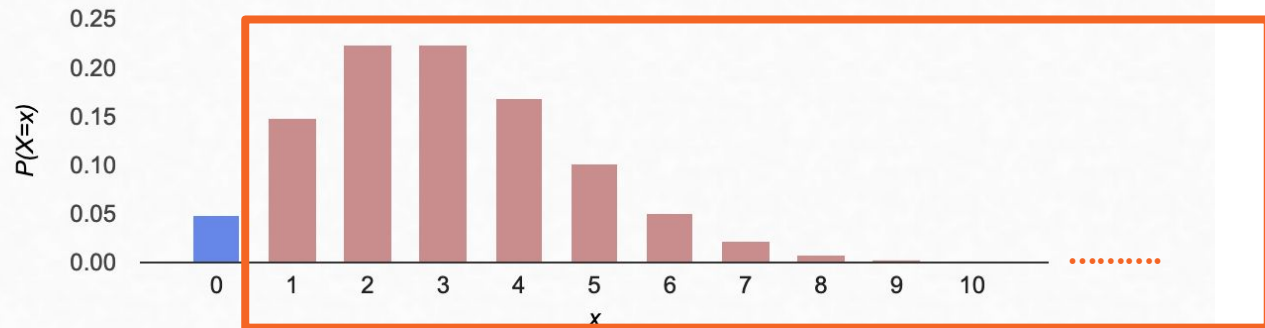
$X \sim \text{Pois}(\lambda)$

$\lambda =$

$x =$

$P(X \geq x) =$

0.95021



$$\mu = E(X) = 3 \quad \sigma = SD(X) = 1.732 \quad \sigma^2 = Var(X) = 3$$

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