

Expectations and Variances

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StatR 101 - Lecture 6b
November 5, 2012

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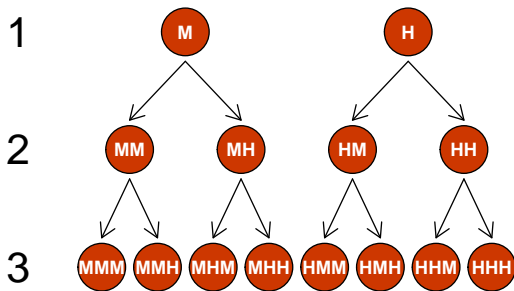
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Some sample spaces are not “numerical”

Example: Sequence of three free throws



- $S = \{MMM, MMH, MHM, MHH, HMM, HMH, HHM, HHH\}$

There's no natural way to order this state space.

Some sample spaces are numerical

Example: Sum of points after three free throws.

- Question II: How many baskets will the basketball player make total?

try 0:		0
try 1:		0 1
try 2:		0 1 2
try 3:		0 1 2 3

- $S = \{0, 1, 2, 3\}$
- This is a naturally “numerical” sample space.
- Every outcome can be assigned a value

Definition

A Random Variable ...

... is a variable whose value is a numerical outcome of a random phenomenon.

Or (more technically) a **random variable** X is a function that takes each element of a sample space S and assigns it to a real number.

Discrete random variable

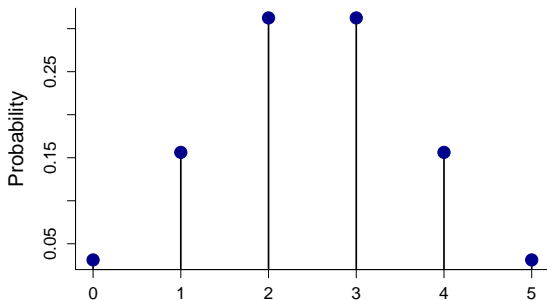
- Every value that a random variable can take is associated with a probability
- The probabilities sum to 1

Value of X	Probability
x_1	p_1
x_2	p_2
x_3	p_3
\dots	\dots
x_k	p_k

Example: Number of heads after 5 coin flips

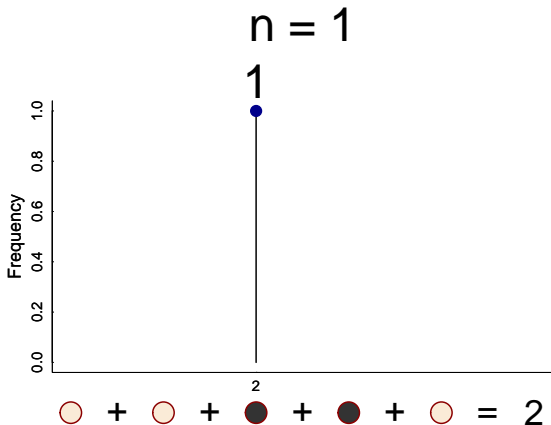
- X is the total number of heads after 5 coin flips
- Possible values of X are: $\{0,1,2,3,4,5\}$
- Probability distribution of X is:
 $P(X = k) = \text{Binomial}(k|n = 5, p = 1/2)$

x	$P(X = x)$
0	0.03125
1	0.15625
2	0.31250
3	0.31250
4	0.15625
5	0.03125



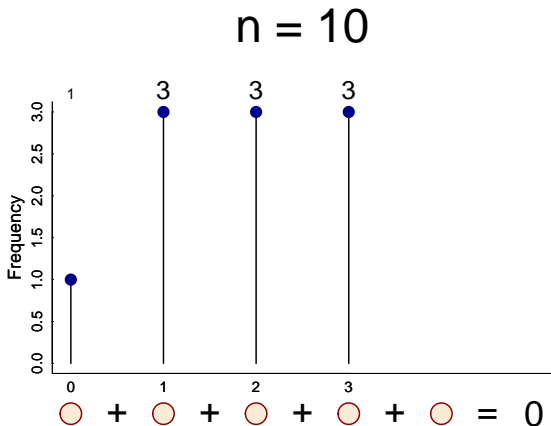
Question: How many heads do we *expect* to get?

- One way to think of this problem is that if we repeated the experiment many many times - what would the average score be?



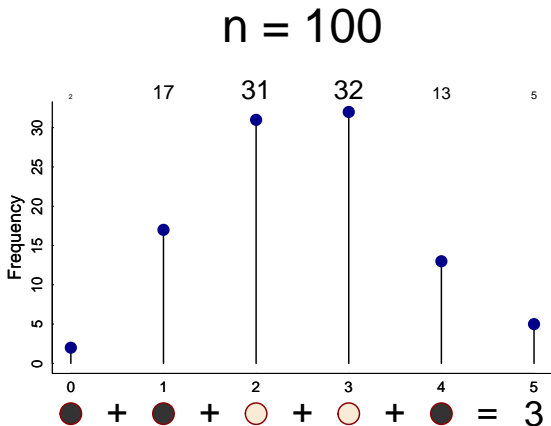
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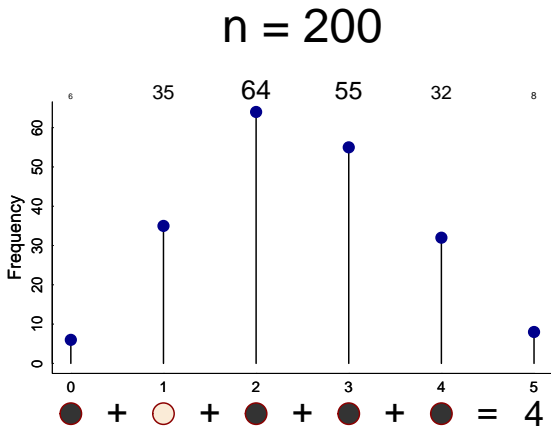
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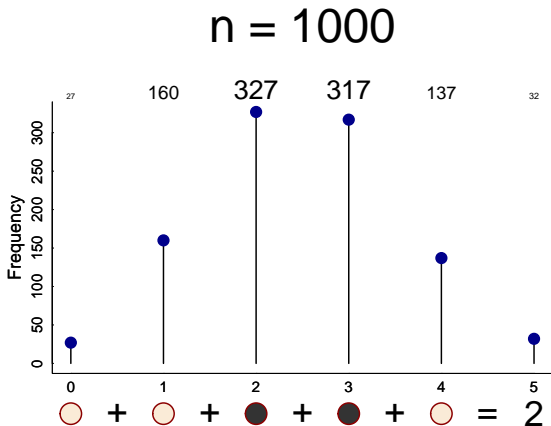
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R code

```
> rbinom(100000,5,.5)
```

```
0 1 2 3 4 5
```

```
3200 15651 31116 31223 15664 3146
```

The mean of these realizations is:

x	$P(X = x)$	$E(N_i)$	N_i
0	0.03125	3,125	3,200
1	0.15625	15,625	15,651
2	0.31250	31,250	31,116
3	0.31250	31,250	32,223
4	0.15625	15,625	15,664
5	0.03125	3,125	3,146
Sum	1	100,000	100,000

$$\begin{aligned}\bar{X} &= (3,200 \times 0 + 15,651 \times 1 + \dots \\ &\quad \dots + 3,146 \times 5) / 100000 = 2.499 \\ &\approx \sum_{i=1}^6 \frac{x_i (100000 P(X = x_i))}{100000} \\ &= \sum_{i=1}^6 x_i P(X = x_i)\end{aligned}$$

Definition

Expectation

The **expected value** or **expectation** of a discrete random variable X with probability function $f(x)$ is

$$E(X) = \sum_{i=1}^n x_i f(x_i)$$

where $\{x_1, x_2, \dots, x_n\}$ is the set of all values that X can take.

It is essentially the mean of *possible values* weighted by their *probability*.

In statistics, $E(X)$ it is often denoted μ or μ_X .

Question: How many heads do we *expect* to get?

$$E(X) = \sum_{i=1}^n x_i p(x_i)$$

x	$f(x)$	$xf(x)$
0	0.03125	
1	0.15625	
2	0.31250	
3	0.31250	
4	0.15625	
5	0.03125	
Sum		

Question: How many heads do we *expect* to get?

$$E(X) = \sum_{i=1}^n x_i p(x_i)$$

x	$f(x)$	$xf(x)$
0	0.03125	0
1	0.15625	0.15625
2	0.31250	0.6250
3	0.31250	0.9375
4	0.15625	0.6250
5	0.03125	0.15625
Sum		

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4	0.15625	0.6250
5	0.03125	0.15625
Sum	1	2.5

Question: How many heads do we *expect* to get?

$$E(X) = \sum_{i=1}^n x_i p(x_i)$$

x	$f(x)$	$xf(x)$
0	0.03125	0
1	0.15625	0.15625
2	0.31250	0.6250
3	0.31250	0.9375
4	0.15625	0.6250
5	0.03125	0.15625
Sum	1	2.5

So, the **expected value** of X is **2.5**.

Expectation of the binomial distribution

Binomial distribution:

$$f(x|n, p) = \Pr(X = x) = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x}$$

Solving for the expectation of the binomial distribution:

$$\begin{aligned} E(X) &= \sum_{i=0}^n f(x|n, p) \\ &= \sum_{i=0}^n x \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x} \\ &= \sum_{i=0}^n \frac{n!}{(x-1)!(n-x)!} p^x (1-p)^{n-x} \\ &= np \sum_{i=0}^n \frac{(n-1)!}{(x-1)!(n-x)!} p^{x-1} (1-p)^{n-x} \\ &= np \sum_{i=0}^n \text{Binomial}(x-1|n-1, p) = np \end{aligned}$$

Expectation of the binomial distribution

Binomial distribution:

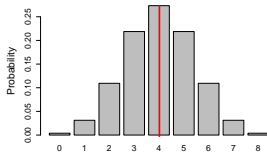
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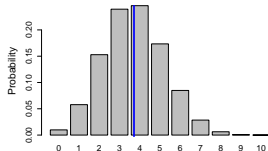
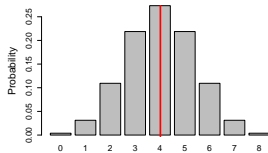
Example of the binomial expectation

- How many heads do we expect after 8 tosses?
 - $p = 0.5, n = 8, E(X) = np = 4$
- How many baskets do we expect Shaq to make in 10 attempts?
 - $p = 0.37, n = 10, E(X) = 3.7$
- How many baskets do we expect Ray Allen to make in 20 attempts?
 - $p = 0.96, n = 20, E(X) = 19.2$



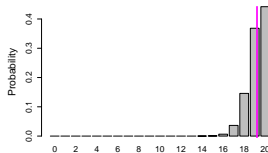
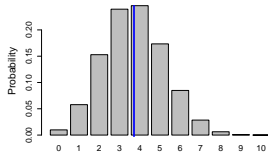
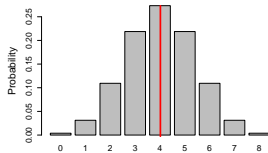
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The **Expectation**

is often denoted μ and called the **mean** of a distribution.

- The expectation tells you the **true mean** of any known, theoretical, distribution
- It is not quite the same as the **sample mean** which we obtain empirically for data.

Some basic arithmetical property of expectations

$$E(A + B) = E(A) + E(B)$$

$$E(kA) = kE(A); \text{ where } k \text{ is a constant}$$

$$E(AB) = E(A)E(B); \text{ only if } A \text{ and } B \text{ are independent}$$

Variance

- Another very important quantity is the **variance** of a distribution.
- It is the **Expected Squared Deviation from the Mean**
 - Convert that to math notation:

$$\text{Var}(X) = E\left((X - E(X))^2\right)$$

- Use the properties of expectation to simplify:

$$\begin{aligned}\text{Var}(X) &= E(X^2 - 2XE(X) + E(X)^2) \\ &= E(X^2) - 2E(XE(X)) + E(X)^2 \\ &= E(X^2) - 2E(X)E(X) + E(X)^2 \\ &= E(X^2) - E(X)^2\end{aligned}$$

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Example: Variance of 4 coin flips

- X is the total number of heads after 4 coin flips
- Possible values of X are: $x = \{0,1,2,3,4\}$
- Probability distribution of X is:
 $P(X = x) = \text{Binomial}(k|n = 4, p = 1/2)$
- Expected value of X is: $E(X) = \mu = np = 2$

x	$P(X = x)$	$x - \mu$	$(x - \mu)^2$	$(x - \mu)^2 P(x)$
0	1/16	-2	4	1/4
1	1/4	-1	1	1/4
2	3/8	0	0	0
3	1/4	1	1	1/4
4	1/16	2	4	1/4
Σ	1			1

$$\text{Var}(X) = 1$$

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Σ	1			1

$$\text{Var}(X) = 1$$

The **Variance**

- The **variance** of a random variable X is defined by the following expressions:

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

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

- For **discrete random variables**, $X \in \{x_1, x_2, x_3 \dots x_n\}$, with known probability function $P(X = x) = f(x)$:

$$\begin{aligned}\text{Var}(X) &= \sum_{i=1}^n \left((x_i - \sum_{i=1}^n x_i f(x_i))^2 f(x_i) \right) \\ &= \sum_{i=1}^n x_i^2 f(x_i) - \left(\sum_{i=1}^n x_i f(x_i) \right)^2\end{aligned}$$

Variance of the binomial distribution

Binomial distribution:

$$f(x|n, p) = \Pr(X = x) = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x}$$

Solving for the variance of the binomial distribution:

$$\begin{aligned} \text{Var}(X^2) &= \sum_{i=0}^n x^2 f(x|n, p) - E(X)^2 \\ &= \dots \\ &= \text{lots of algebra similar to last time} \\ &= \dots \\ &= np(1-p) \end{aligned}$$

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The **Variance**

...is often denoted σ^2 .

- It tells you something quantitative about the amount of spread in a distribution from the **mean**.
 - The square root of the variance, σ is the standard deviation - which is the units of the random variable X .
-
- This quantity is the **true variance** of any known, theoretical, distribution
 - It is not exactly the same as the **sample variance** which we obtain empirically for data.

Example of the binomial variance

- What's the variance of heads after 8 coin tosses?

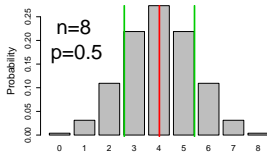
- $p = 0.5$, $n = 8$, $E(X) = 4$,
 $\text{Var}(X) = np(1 - p) = 2$

- What's the variance of Shaq's 10 FT attempts?

- $p = 0.37$, $n = 10$, $E(X) = 3.7$,
 $\text{Var}(X) = 2.331$

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Example of the binomial variance

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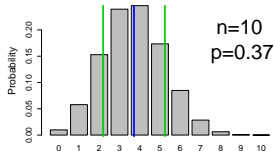
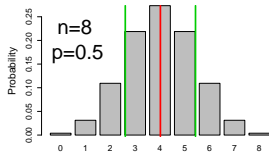
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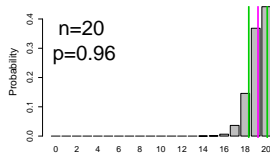
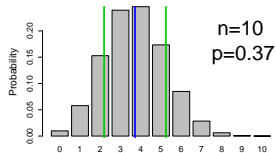
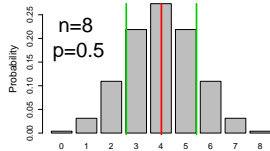
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 $Var(X) = 0.768$



Basic arithmetical property of variances

- $\text{Var}(kA) = k^2\text{Var}(A)$
- If A and B are independent,

$$\text{Var}(A + B) = \text{Var}(A) + \text{Var}(B)$$

Components of a probability distribution

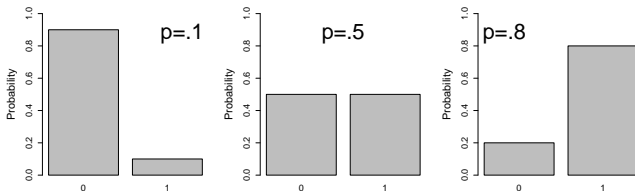
- X : The **random variable** (r.v.)
- x : The possible values (or **support**) of X
 - $X \in \{x_1, x_2, x_3 \dots x_n\}$
- $P(X = x|\theta) = f(x, \theta)$: The **probability mass functions**
 - Often contracted to “p.m.f.” or “pmf”
 - for continuous r.v.’s, called “**probability density function**” (p.d.f.)
- θ : The **parameters** of the pdf.
- $E(X) = \mu$: The theoretical **mean** of the pdf.
- $\text{Var}(X) = \sigma^2$: The theoretical **variance** of the pdf.

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

Name:	Binomial Distribution	Total number of successes with probability p after n tries.
Support:	x	$0, 1, 2, 3, \dots, n$
Parameters:	$n \in \mathbf{N}$ (positive integers) $p \in [0, 1]$	number of trials (positive integer) probability of success
pmf	$P(X = x n, p)$	$f(x) = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x}$
mean	$E(X)$	np
variance	$\text{Var}(X)$	$np(1-p)$

Bernoulli distribution

- The Bernoulli distribution is the distribution of a **single** event with probability p .
- $$P(X = x) = \begin{cases} 1 - p & \text{for } x = 0 \\ p & \text{for } x = 1 \end{cases}$$
- Examples: a single coin flip, a single FT.



Question: What is the expected value and variance of the Bernoulli distribution?

$X \sim \text{Bernoulli}(p)$

Name:	Bernoulli Distribution	Models number of successes after one trial
Support:	x	0,1
Parameters:	$p \in [0, 1]$	probability of success
pmf	$P(X = x p)$	$\begin{cases} 1 - p & \text{for } x = 0 \\ p & \text{for } x = 1 \end{cases}$
mean	$E(X)$	p
variance	$\text{Var}(X)$	$p(1 - p)$

Relationship between Binomial and Bernoulli distribution

- A Binomial(n, p) is the **sum** of n Bernoulli(p) trials
- Formally, if $X_1, X_2, X_3 \dots X_n$ are each independent variables with:

$$X_i \sim \text{Bernoulli}(p)$$

then

$$Y = \sum_{i=1}^n X_i \sim \text{Binomial}(n, p)$$

Using arithmetic properties of Expectation and Variance

Question: What is the expectation and variance of $Y = \sum_{i=1}^n X_i$ where X_i are Bernoulli(p) trials?

- Recall

$$E(A + B) = E(A) + E(B)$$

And (if A and B are independent)

$$\text{Var}(A + B) = \text{Var}(A) + \text{Var}(B)$$

- Then:

$$E(Y) = E\left(\sum_{i=1}^n X_i\right) = \sum_{i=1}^n E(X_i) = \sum_{i=1}^n p = np$$

- And (very similarly):

$$\text{Var}(Y) = \text{Var}\left(\sum_{i=1}^n X_i\right) = \sum_{i=1}^n \text{Var}(X_i) = \sum_{i=1}^n p(1-p) = np(1-p)$$

- These are much easier ways to calculate the mean and variance of the Binomial distribution!

Example Problem: The Lottery

Rules of Washington State Powerball

- Buy a ticket for \$1
 - Pick 6 numbers from 1, ..., 53
 - Win \$1,000,000 if your numbers are identical to winning numbers (in any order).
-
- Q1: What is the probability of winning the jackpot?
 - Uniform sample space: Choose 6 from 53
 - $\binom{53}{6} = 22,957,480$
 - $P(\text{Winning}) = 1/22957480$

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Example Problem: The Lottery

$$P(\text{Winning}) = 1/22957480$$

- Q2: What are your expected winnings?
- Call X the dollar amount of your winnings
 - $X \in \{-1, 999999\}$
 - $P(X = x) \begin{cases} 22957479/22957480 & \text{for } x = 0 \\ 1/22957480 & \text{for } x = 10000000 \end{cases}$
 - $E(X) = -1 \times \frac{22957479}{22957480} + 999,999 \times \frac{1}{22957480} \approx -0.956$
- So: You will lose 0.96 cents, on average, every time you play.
- **Good luck!**

R code: Generic Powerball

```
> p <- 1/choose(53, 6)
> cost <- 1
> reward <- 1e6
> X <- c(0,reward)-cost
> pmf <- c(1-p,p)
> sum(pmf * X)
[1] -0.9564412
```

Example Problem: The Lottery

$$P(\text{Winning}) = 1/22957480$$

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 - $X \in \{-1, 999999\}$
 - $P(X = x) \begin{cases} 22957479/22957480 & \text{for } x = 0 \\ 1/22957480 & \text{for } x = 10000000 \end{cases}$
 - $E(X) = -1 \times \frac{22957479}{22957480} + 999,999 \times \frac{1}{22957480} \approx -0.956$
- So: You will lose 0.96 cents, on average, every time you play.
- **Good luck!**

R code: Generic Powerball

```
> p <- 1/choose(53, 6)
> cost <- 1
> reward <- 1e6
> X <- c(0, reward) - cost
> pmf <- c(1-p, p)
> sum(pmf * X)
[1] -0.9564412
```

Example Problem: The Lottery

$$P(\text{Winning}) = 1/22957480$$

- Q2: What are your expected winnings?
- Call X the dollar amount of your winnings
 - $X \in \{-1, 999999\}$
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Geometric distribution

Questions:

- How long (how many rolls) will it take me to roll a 6 on average?
- How long will it take Shaq to make a free throw?
- If I bought a lottery ticket every day, how long would it take me to win?

The Geometric Distribution

- The geometric distribution describes the number of Bernoulli trials with probability p before a success - the *waiting time* of a distribution.
- $P(X = k) = p(1 - p)^k$ where $k = \{0, 1, 2, 3, \dots\}$
- Note: one p is for success, k $(1 - p)$'s for failure.

Geometric distribution

Questions:

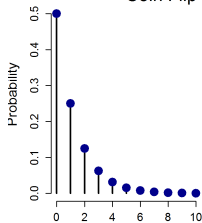
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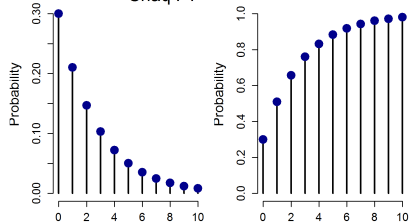
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Geometric distribution: examples

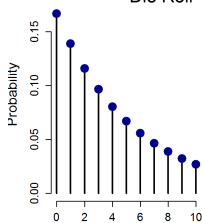
Coin Flip



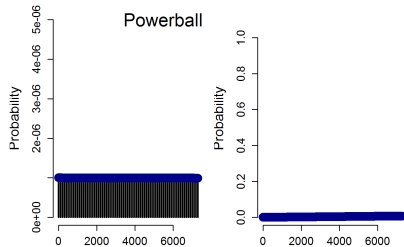
Shaq FT



Die Roll



Powerball



Geometric distribution: examples

- What is the probability that Shaq ($p=0.3$) will miss exactly three times before making it?

$$p(1-p)^3 = \text{dgeom}(3, 0.3) = 10.3\%$$

- What is the probability that Shaq ($p=0.3$) will miss less than three times in a row?

$$\sum_{i=0}^2 p(1-p)^i = \text{pgeom}(2, 0.3) = 65\%$$

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$$\sum_{i=4}^{\infty} p(1-p)^i = 1 - \text{pgeom}(4, 0.3) = 17\%$$

- What is the probability that you will have won SuperLotto at least once after 20 years of playing daily?

$$\sum_{i=0}^{365*20} p(1-p)^i = \text{pgeom}(365*20, p = 1/22e6) = 0.03\%$$

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Geometric distribution: Memorylessness

- After 3 misses, what is the probability Shaq will miss 3 more times?
ALSO 10.3%!
- After 20 years of trying, what is the probability you might win after another 20 years?
ALSO 0.03%!
- It does not matter how long you have been trying to get a success, the waiting time will always have the same distribution.
- This is called “memorylessness” and is very special.

$$P(X > m + n | X > m) = P(X > n)$$

Geometric distribution

$$E(X) = \sum_{i=0}^{\infty} k p (1-p)^k = \frac{1-p}{p}$$

$$\text{Var}(X) = \sum_{i=0}^{\infty} k^2 p (1-p)^k - E(X)^2 = \frac{1-p}{p^2}$$

- How long (how many flips) will it take me before I get a head from a fair coin on average?

Answer: $\mu_x = (1 - 1/2)/1/2 = 1$ flip, $\sigma_x = \sqrt{2}$

- How long will it take me to roll a 6 on average?

Answer: $\mu_x = (1 - 1/6)/1/6 = 5$ rolls, $\sigma_x = \sqrt{30}$

- How long will it take Shaq to make a free throw?

Answer: $\mu_x = (1 - 0.3)/0.3 = 2.333$ attempts, $\sigma_x = 2.789$

- How long would it take me to win Powerball?

Answer: $\mu_x = (1 - 1/23e6)/23e6 = 22e6$ days = 63,013 years,
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$X \sim \text{Geometric}(p)$

Name:	Geometric Distribution	Waiting time of success for Bernoulli trials
Support:	x	$0, 1, 2, 3 \dots$
Parameters:	$p \in [0, 1]$	probability of success
pmf	$P(X = x p)$	$p(1 - p)^x$
mean	$E(X)$	$\frac{1-p}{p}$
variance	$\text{Var}(X)$	$\frac{1-p}{p^2}$
Special Feature:	Memorylessness!	