

Discrete Distributions¹

October 18, 2017

¹HMS, 2017, v1.0

Chapter References

- ▶ Diez: Chapter 3.3, 3.4 (not 3.4.2), 3.5.2
- ▶ Navidi, Chapter 4.1, 4.2, 4.3

Discrete Distributions

- ▶ Bernoulli – Yes/No Responses
- ▶ Binomial Distribution – Sums of Bernoulli Responses
- ▶ Geometric – Number of trials until first success
- ▶ Poisson – Points in time or space

Bernoulli Distribution

- ▶ A Bernoulli event/trial is one where the probability the event occurs is p and the probability the event does not occur is $1 - p$; i.e., the event has two possible outcomes usually viewed as **success** or **failure**.
- ▶ A Bernoulli distribution describes how a Bernoulli event is distributed between success and failure.
- ▶ The Bernoulli distribution for a random variable X is often denoted by:

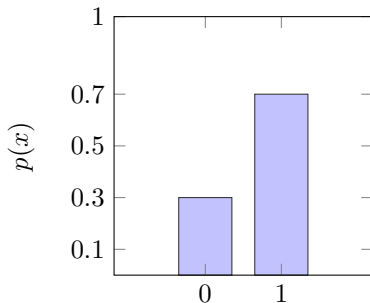
$$X \sim \text{Bernoulli}(p)$$

Bernoulli Distribution

- ▶ Examples of Bernoulli Processes:
 - Flipping a coin (again....)
 - Political opinion poll (yes or no to support a candidate)
 - Birth of boy or girl
 - Cell culture grows or does not
 - Ace or King in a pack of cards (with replacement!)
 - Left or right-handedness
 - Whether a drug works or does not

Bernoulli Distribution

- ▶ **example:** $X \sim \text{Bernoulli}(0.7)$



Failure and success will often have the values 0 and 1 respectively.

Mean and Variance of a Bernoulli Random Variable

- ▶ Recall that the mean is given by $\sum x_i p(x_i)$. Therefore the mean of a Bernoulli random variables is:

$$\mu_X = (0)(1 - p) + (1)p = p$$

- ▶ The variance can be computed similarly using the variance equation

$$\sigma_X^2 = \sum (x_i - \mu_X)^2 p(x_i)$$

Class exercise: Derive the variance

Class exercise: Derive variance

$$\sigma_X^2 = \sum (x_i - \mu_X)^2 p(x_i)$$

- ▶ The variance can be computed similarly using the variance equation:

$$\sigma_X^2 = (0 - p)^2(1 - p) + (1 - p)^2(p) = p(1 - p)$$

Mean and Variance of a Bernoulli Random Variable

If $X \sim \text{Bernoulli}(p)$, then:

▶ Mean:

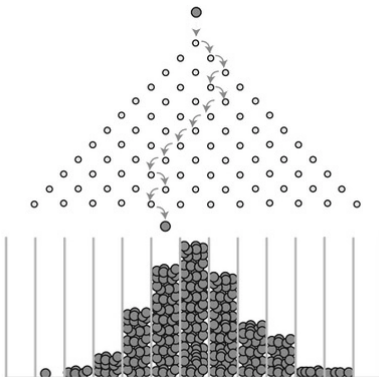
$$\mu_X = p$$

▶ Variance:

$$\sigma_X^2 = p(1 - p)$$

Galton Board

- ▶ Example of something where each trial (row on the board) is a Bernoulli trial.



The Binomial Distribution

- ▶ Carry out a Bernoulli trial, and repeat it n times.
- ▶ Let the number of successes be a random variable, X
- ▶ Then X will have a **Binomial Distribution**.

A random variable X that has a binomial distribution with n trials, and probability p for each trial, is denoted by:

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

The Binomial Distribution

- ▶ A fair coin is tossed 10 times. Let X be the number of heads that appear. What is the distribution of X ?
- ▶ Since each trial has two outcomes with probability $p = 0.5$, and each trial is independent of the other, then:

$$X \sim \text{Bin}(10, 0.5)$$

Assumptions in a Binomial Distribution

Formal Definition:

- ▶ There are only two possible outcomes for each trial, arbitrarily labeled "success" or "failure".
- ▶ The trials are **independent** of each other.
- ▶ The probability of success is the **same** for each trial.
- ▶ There are a **fixed number** n of repeated trials of a simple experiment.

Binomial Distribution

▶ Toss a coin three times and record the result.

▶ Question:

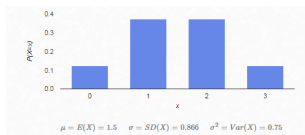
How many possible arrangements of the coin could there be?

Binomial Distribution

- ▶ There will be $2^3 = 8$ possible arrangements:
- ▶ HHH, TTT, HHT, HTH, THH, THT, TTH, HTT
- ▶ Plot the distribution of the number of heads:

Number of heads, X :	0	1	2	3
Number of heads observed:	1	3	3	1

- ▶ The binomial distribution will predict the above table



Deriving the PMF for the Binomial Distribution

- ▶ The distribution of the number heads (or successes), $p(k)$ can be shown to be:

$$p(k) = \binom{n}{k} q^k (1 - q)^{n-k}$$

- ▶ Where

q = probability of success

n = number of trials

k = The probability we seek for k number of heads

Binomial Distribution: Example

Assumptions:

1. n identical trials are performed.
2. There are two outcomes, success and failure in each trial.
3. The trials are independent.
4. The success probability, q , remains constant from trial to trial.

Steps:

1. Identify the success.
2. Determine the probability of success, q .
3. Determine, n , the number of trials.
4. Apply the binomial formula for the number of success we seek.

Binomial Distribution: Example

- ▶ **example:** Given five coins what is the probability of throwing 2 heads?

$$p(k) = \binom{n}{k} q^k (1 - q)^{n-k}$$

- ▶ Set

$$q = 0.5$$

$$n = 5$$

$$k = 2$$

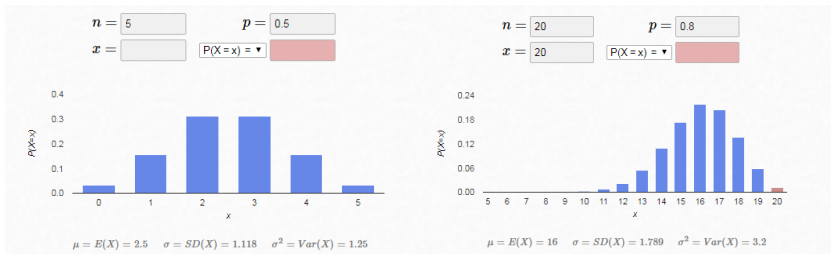
$$p(2) = \binom{5}{2} 0.5^2 (1 - 0.5)^{5-2} = 10 \times 0.25 \times 0.125 = 0.3125$$

Binomial Distribution: Example

Given the answer of 0.3125 from the last slide, what does it mean?

Deriving the PMF for the Binomial Distribution

► examples:



<http://homepage.divms.uiowa.edu/~mbognar/applets/bin.html>

Deriving the PMF for the Binomial Distribution

- ▶ Throw three coins
- ▶ Consider $p(0)$, i.e the probability of **not seeing** a single head.
- ▶ Let q equal the probability of throwing a head in a single trial.
- ▶ Probability of not throwing a head = $(1 - q)$
- ▶ Therefore probability of throwing three tails (not three heads):

$$p(0) = (1 - q)^3$$

Deriving the PMF for the Binomial Distribution

- ▶ Consider $p(1)$, ie the probability of seeing a single head?
- ▶ For example what is the probability of throwing HTT?
- ▶ Let q equal the probability of throwing a head in a single trial.
- ▶ Probability of **not** throwing a head = $(1 - q)$
- ▶ Therefore probability of throwing one head and two tails:

$$p(1) = q(1 - q)^2$$

Deriving the PMF for the Binomial Distribution

- ▶ Pattern:

$$p(0) = q^0(1 - q)^3$$

$$p(1) = q^1(1 - q)^2$$

$$p(2) = q^2(1 - q)^1$$

$$p(3) = q^3(1 - q)^0$$

- ▶ In general, if we have n trials and k is the number of heads then:

$$p(k) = q^k(1 - q)^{n-k}$$

- ▶ Or is it that simple.....

Deriving the PMF for the Binomial Distribution

- ▶ Consider $p(1)$ more closely:
- ▶ $p(1)$ is the probability of seeing one head in a toss of three coins and it was $q(1 - q)^2$, e.g HTT
- ▶ However we know that there are three possible ways to get one head and two tails: HTT, THT, TTH.
- ▶ Therefore the probability of seeing any of the three patterns (HTT or THT or TTH) is the sum of the three corresponding probabilities:

$$p(1) = 3q(1 - q)^2$$

Deriving the PMF for the Binomial Distribution

- ▶ Likewise for the other three $p(0)$, $p(2)$, and $p(3)$.
- ▶ In each of these cases there are 1, 3 and 1 way respectively.

Deriving the PMF for the Binomial Distribution

- ▶ How can we count the number of configurations?
- ▶ We've seen previously that for a word such as MISSISSIPPI, the number of variations is given by:

$$\frac{n!}{n_1! n_2! \dots}$$

- ▶ The same applies to a sequence such as HHTT, here the number of arrangements is:

$$\frac{4!}{2! 2!} = 6$$

Deriving the PMF for the Binomial Distribution

- ▶ Assume we have two types of symbol, eg H and T, of which there are n in total.
- ▶ We can define r to be the number of heads (H), therefore the number of tails must be $(n - r)$.
- ▶ The arrangement formula can now be written as:

$$\frac{n!}{r!(n - r)!}$$

- ▶ The result is of course the number of combinations:

$$\binom{n}{r}$$

Deriving the PMF for the Binomial Distribution

- ▶ **example:** How many ways can we arrange two heads using four coins, HHTT, HTHT, etc
- ▶ $n = 4$, and $r = 2$ (number of heads)
- ▶ Compute the combinations:

$$\binom{n}{r} = \frac{n!}{r!(n-r)!} = \frac{4!}{2!(4-2)!} = \frac{24}{4} = 6$$

- ▶ Or two heads and six coins, HHTTTT, THHTTT, etc

$$\frac{n!}{r!(n-r)!} = \frac{6!}{2!(6-2)!} = \frac{720}{3(24)} = \frac{720}{72} = 100$$

Why Combinations?

H_1	H_2		
	H_2	H_3	
H_1		H_3	
	H_2		H_4
		H_3	H_4
H_1			H_4

Deriving the PMF for the Binomial Distribution

- ▶ To compute the probability of the number of heads $p(k)$ we write:

$$p(k) = \binom{n}{k} q^k (1 - q)^{n-k}$$

Binomial Distribution: Example

- ▶ Consider a roulette wheel with 38 slots. You win if the ball lands on a number you picked.
- ▶ What is the probability of winning twice in 50 spins?
- ▶ Each spin is independent, we have a failure or success, therefore we can describe this using a binomial distribution.
- ▶ $n = 50, k = 2, q = 1/38$, therefore:

$$p(2) = \binom{50}{2} (1/38)^2 (1 - 1/38)^{50-2} = 0.278$$

Binomial Example

- ▶ Each year one in three adults aged 65 years and older have severe falls.
- ▶ Find the probability that 3 among 11 adults aged 65 years and older will fall in the following year.
- ▶ Fall or not fall, independent but finite number of trials.

$$p(k) = \binom{n}{k} q^k (1 - q)^{n-k}$$

$$p(3) = \binom{11}{3} \left(\frac{1}{3}\right)^3 \left(\frac{2}{3}\right)^{11-3} = 0.238$$

Binomial Distribution: Properties

- ▶ Mean and Variance

$$\mu = nq$$

$$\sigma^2 = nq(1 - q)$$

Binomial Theorem

$$(x + y)^2 = x^2 + 2xy + y^2,$$

$$(x + y)^3 = x^3 + 3x^2y + 3xy^2 + y^3,$$

$$(x + y)^4 = x^4 + 4x^3y + 6x^2y^2 + 4xy^3 + y^4,$$

$$(x + y)^5 = x^5 + 5x^4y + 10x^3y^2 + 10x^2y^3 + 5xy^4 + y^5,$$

$$(x + y)^6 = x^6 + 6x^5y + 15x^4y^2 + 20x^3y^3 + 15x^2y^4 + 6xy^5 + y^6,$$

$$(x + y)^7 = x^7 + 7x^6y + 21x^5y^2 + 35x^4y^3 + 35x^3y^4 + 21x^2y^5 + 7xy^6 + y^7$$

