Lecture Slides

ELEMENTARY STATISTICS

TENTH EDITION

Elementary Statistics Tenth Edition

and the Triola Statistics Series

by Mario F. Triola

MARIO F. TRIOLA 🚫

Chapter 5 Probability Distributions

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- 5-2 Random Variables
- **5-3 Binomial Probability Distributions**
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Section 5-1 Overview

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This chapter will deal with the construction of

discrete probability distributions

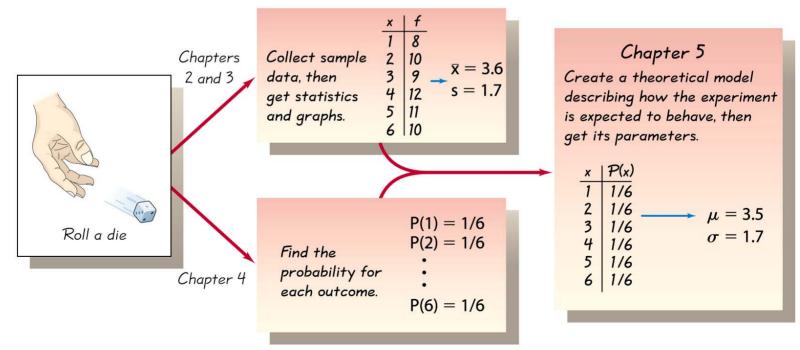
by combining the methods of descriptive statistics presented in Chapter 2 and 3 and those of probability presented in Chapter 4.

Probability Distributions will describe what will probably happen instead of what actually did happen.



Combining Descriptive Methods and Probabilities

In this chapter we will construct probability distributions by presenting possible outcomes along with the relative frequencies we expect.





Section 5-2 Random Variables



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Key Concept

This section introduces the important concept of a probability distribution, which gives the probability for each value of a variable that is determined by chance.

Give consideration to distinguishing between outcomes that are likely to occur by chance and outcomes that are "unusual" in the sense they are not likely to occur by chance.



Definitions

Random variable

a variable (typically represented by **x**) that has a single numerical value, determined by chance, for each outcome of a procedure

Probability distribution

a description that gives the probability for each value of the random variable; often expressed in the format of a graph, table, or formula



Definitions

Discrete random variable

either a finite number of values or countable number of values, where "countable" refers to the fact that there might be infinitely many values, but they result from a counting process

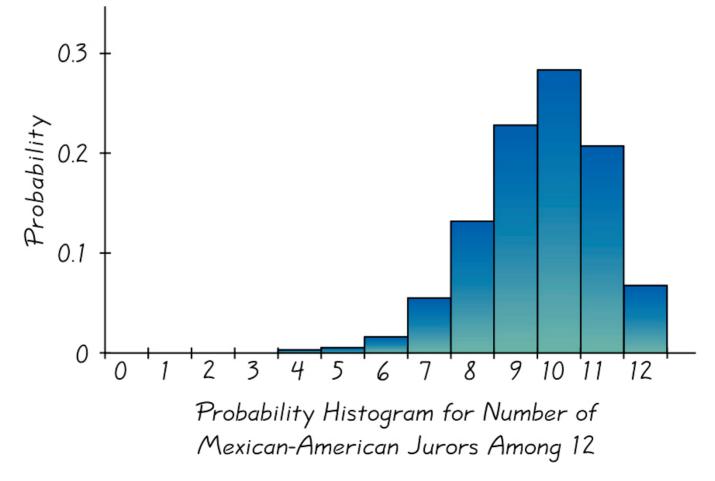
Continuous random variable

infinitely many values, and those values can be associated with measurements on a continuous scale in such a way that there are no gaps or interruptions



Graphs

The probability histogram is very similar to a relative frequency histogram, but the vertical scale shows probabilities.





Requirements for Probability Distribution

$\sum P(x) = 1$ where x assumes all possible values.

$0 \le P(x) \le 1$ for every individual value of *x*.



Mean, Variance and Standard Deviation of a Probability Distribution

 $\boldsymbol{\mu} = \boldsymbol{\Sigma} \left[\boldsymbol{x} \cdot \boldsymbol{P}(\boldsymbol{x}) \right] \qquad \text{Mean}$

 $\sigma^2 = \Sigma \left[(\mathbf{x} - \boldsymbol{\mu})^2 \cdot \boldsymbol{P}(\mathbf{x}) \right] \qquad \text{Variance}$

 $\sigma^2 = [\Sigma x^2 \cdot P(x)] - \mu^2 \qquad \text{Variance (shortcut)}$

$$\sigma = \sqrt{\Sigma [x^2 \cdot P(x)]} - \mu^2$$
 Standard Deviation



Roundoff Rule for μ , σ , and σ^2

Round results by carrying one more decimal place than the number of decimal places used for the random variable **x**. If the values of **x** are integers, round μ , σ , and σ^2 to one decimal place.



Identifying Unusual Results Range Rule of Thumb

According to the range rule of thumb, most values should lie within 2 standard deviations of the mean.

We can therefore identify "unusual" values by determining if they lie outside these limits:

Maximum usual value = μ + 2 σ Minimum usual value = μ - 2 σ



Identifying Unusual Results Probabilities

Rare Event Rule

If, under a given assumption (such as the assumption that a coin is fair), the probability of a particular observed event (such as 992 heads in 1000 tosses of a coin) is extremely small, we conclude that the assumption is probably not correct.

◆ Unusually high: x successes among n trials is an unusually high number of successes if P(x or more) ≤ 0.05.

◆ Unusually low: x successes among n trials is an unusually low number of successes if P(x or fewer) ≤ 0.05.



Definition

The expected value of a discrete random variable is denoted by E, and it represents the average value of the outcomes. It is obtained by finding the value of $\Sigma [x \cdot P(x)]$.

$E = \sum [x \cdot P(x)]$



Recap

In this section we have discussed:

- Combining methods of descriptive statistics with probability.
- Random variables and probability distributions.
- Probability histograms.
- Requirements for a probability distribution.
- Mean, variance and standard deviation of a probability distribution.
- Identifying unusual results.
- Expected value.



Section 5-3 Binomial Probability Distributions

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This section presents a basic definition of a binomial distribution along with notation, and it presents methods for finding probability values.

Binomial probability distributions allow us to deal with circumstances in which the outcomes belong to two relevant categories such as acceptable/defective or survived/died.



Definitions

A binomial probability distribution results from a procedure that meets all the following requirements:

- 1. The procedure has a fixed number of trials.
- 2. The trials must be independent. (The outcome of any individual trial doesn't affect the probabilities in the other trials.)
- 3. Each trial must have all outcomes classified into two categories (commonly referred to as success and failure).
- 4. The probability of a success remains the same in all trials.



Notation for Binomial Probability Distributions

S and F (success and failure) denote two possible categories of all outcomes; *p* and *q* will denote the probabilities of S and F, respectively, so

 $P(S) = p \qquad (p = \text{probability of success})$ $P(F) = 1 - p = q \quad (q = \text{probability of failure})$



Notation (cont)

- *n* denotes the number of fixed trials.
- *x* denotes a specific number of successes in *n* trials, so *x* can be any whole number between 0 and *n*, inclusive.
- *p* denotes the probability of success in one of the *n* trials.
- *q* denotes the probability of failure in one of the *n* trials.
- P(x) denotes the probability of getting exactly x successes among the *n* trials.



Important Hints

Be sure that x and p both refer to the same category being called a success.

♦ When sampling without replacement, consider events to be independent if $n \le 0.05N$.



Methods for Finding Probabilities

We will now discuss three methods for finding the probabilities corresponding to the random variable **x** in a binomial distribution.



Method 1: Using the Binomial Probability Formula

$$P(x) = \frac{n!}{(n-x)!x!} \cdot p^x \cdot q^{n-x}$$

where

- **n** = number of trials
- **x** = number of successes among *n* trials
- p = probability of success in any one trial
- q = probability of failure in any one trial (q = 1 p)



Method 2: Using Table A-1 in Appendix A

Part of Table A-1 is shown below. With n = 12 and p = 0.80 in the binomial distribution, the probabilities of 4, 5, 6, and 7 successes are 0.001, 0.003, 0.016, and 0.053 respectively.

n x	<i>p</i>		
	0.80	<i>x</i>	p
4	0.001	4	0.001
5	0.003	5	0.003
6	0.016	6	0.016
7	0.053	7	0.053



Method 3: Using Technology

STATDISK, Minitab, Excel and the TI-83 Plus calculator can all be used to find binomial probabilities.

STATDISK

🚮 Binomia	l Probability		
Num Trials,	n: 12		Evaluate
Success Pr	ob, p: 0.8		
Mean:	9.6	000	
St Dev:	1.3		
Variance:	1.9		
variance.	1.5.	200	
x	P(x)	P(x or fewer)	P(x or greater)
0	0.0000000	0.0000000	1.0000000 🔺
1	0.0000002	0.0000002	1.0000000
2	0.0000043	0.0000045	0.9999998
3	0.0000577	0.0000622	0.9999955
4	0.0005190	0.0005812	0.9999378
5	0.0033219	0.0039031	0.9994188
6	0.0155021	0.0194053	0.9960969
7	0.0531502	0.0725555	0.9805947
8	0.1328756	0.2054311	0.9274445
9	0.2362232	0.4416543	0.7945689
10	0.2834678	0.7251221	0.5583457
11	0.2061584	0.9312805	0.2748779
10	0.0607106	1 000000	0 0807105
Help ?)	Clear Cop	ру

Minitab

Binomial with $n = 12$ and $p = 0.8$			
х	P(X = x)		
0	0.00000		
1	0.00000		
2	0.00004		
3	0.000058		
4	0.000519		
5	0.003322		
6	0.015502		
7	0.053150		
8	0.132876		
9	0.236223		
10	0.283468		
11	0.206158		
12	0.068719		



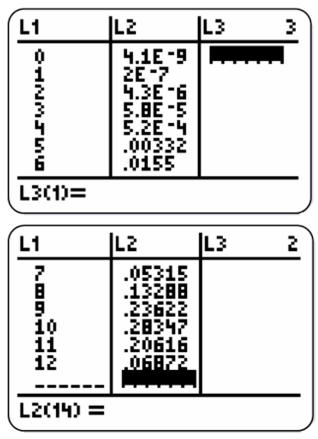
Method 3: Using Technology

STATDISK, Minitab, Excel and the TI-83 Plus calculator can all be used to find binomial probabilities.

Excel

	А	В
1	0	4.096E-09
2	1	1.966E-07
3	2	4.325E-06
4	3	5.767E-05
5	4	0.000519
6	5	0.0033219
7	6	0.0155021
8	7	0.0531502
9	8	0.1328756
10	9	0.2362232
11	10	0.2834678
12	11	0.2061584
13	12	0.0687195

TI-83 Plus calculator





Strategy for Finding Binomial Probabilities

- Use computer software or a TI-83 Plus calculator if available.
- If neither software nor the TI-83 Plus calculator is available, use Table A-1, if possible.
- If neither software nor the TI-83 Plus calculator is available and the probabilities can't be found using Table A-1, use the binomial probability formula.



Rationale for the Binomial Probability Formula

$$P(x) = \frac{n!}{(n-x)!x!} \cdot p^{x} \cdot q^{n-x}$$
The number of
outcomes with
exactly x
successes
among n trials



Binomial Probability Formula

$$P(\mathbf{x}) = \frac{n!}{(n-x)!x!} \cdot p^{x} \cdot q^{n-x}$$

Number of outcomes with exactly *x* successes among *n* trials The probability of x successes among n trials for any one particular order



Recap

In this section we have discussed:

The definition of the binomial probability distribution.

Notation.

Important hints.

Three computational methods.

Rationale for the formula.



Section 5-4 Mean, Variance, and Standard Deviation for the Binomial Distribution

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Key Concept

In this section we consider important characteristics of a binomial distribution including center, variation and distribution. That is, we will present methods for finding its mean, variance and standard deviation.

As before, the objective is not to simply find those values, but to interpret them and understand them.



For Any Discrete Probability Distribution: Formulas $\boldsymbol{\mu} = \boldsymbol{\Sigma} [\boldsymbol{x} \boldsymbol{\cdot} \boldsymbol{P}(\boldsymbol{x})]$ Mean $\sigma^{2} = [\Sigma \mathbf{x}^{2} \cdot \mathbf{P}(\mathbf{x})] - \mu^{2}$ Variance $\sigma = \sqrt{\left[\sum x^2 \cdot P(x)\right] - \mu^2}$ Std. Dev



Binomial Distribution: Formulas

Mean $\mu = n \cdot p$

Variance $\sigma^2 = n \cdot p \cdot q$

Std. Dev.
$$\sigma = \sqrt{n \cdot p \cdot q}$$

Where

- **n** = number of fixed trials
- *p* = probability of success in one of the *n* trials
 q = probability of failure in one of the *n* trials

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Interpretation of Results

It is especially important to interpret results. The range rule of thumb suggests that values are unusual if they lie outside of these limits:

Maximum usual values = μ + 2 σ Minimum usual values = μ - 2 σ



Recap

In this section we have discussed:

- Mean, variance and standard deviation formulas for the any discrete probability distribution.
- Mean, variance and standard deviation formulas for the binomial probability distribution.
- Interpreting results.



Section 5-5 The Poisson Distribution

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Key Concept

The Poisson distribution is important because it is often used for describing the behavior of rare events (with small probabilities).



Definition

The Poisson distribution is a discrete probability distribution that applies to occurrences of some event over a specified interval. The random variable *x* is the number of occurrences of the event in an interval. The interval can be time, distance, area, volume, or some similar unit.

Formula $P(x) = \frac{\mu^{x} \cdot e}{x!}^{\mu} \text{ where } e \approx 2.71828$



Poisson Distribution Requirements

- The random variable x is the number of occurrences of an event over some interval.
- The occurrences must be random.
- The occurrences must be independent of each other.
- The occurrences must be uniformly distributed over the interval being used.

Parameters

- ✤ The mean is µ.
- The standard deviation is $\sigma = \sqrt{\mu}$.



Difference from a Binomial Distribution

The Poisson distribution differs from the binomial distribution in these fundamental ways:

The binomial distribution is affected by the sample size *n* and the probability *p*, whereas the Poisson distribution is affected only by the mean *μ*.

In a binomial distribution the possible values of the random variable x are 0, 1, ..., n, but a Poisson distribution has possible x values of 0, 1, ..., with no upper limit.



Poisson as Approximation to Binomial

The Poisson distribution is sometimes used to approximate the binomial distribution when *n* is large and *p* is small.

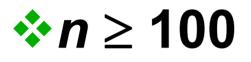
Rule of Thumb







Poisson as Approximation to Binomial - μ





Value for μ

 $\mu = n \cdot p$



Recap

In this section we have discussed:

- Definition of the Poisson distribution.
- Requirements for the Poisson distribution.
- Difference between a Poisson distribution and a binomial distribution.
- Poisson approximation to the binomial.

