Chapter 5
Probability Distributions
5-1 Overview
5-2 Random Variables
5-3 Binomial Probability Distributions
5-4 Mean, Variance, and Standard Deviation for the Binomial Distribution
5-5 The Poisson Distribution

Overview
This chapter will deal with the construction of probability distributions by combining the methods of descriptive statistics presented in Chapter 2 and those of probability presented in Chapter 3.
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.

Definitions
A random variable is a variable (typically represented by X) that has a single numerical value, determined by chance, for each outcome of a procedure.

A probability distribution is a graph, table, or formula that gives the probability for each value of the random variable.

Definitions
A discrete random variable has 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.

A continuous random variable has 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

\[ 0 \leq P(x) \leq 1 \]

for every individual value of \( x \)

Mean, Variance and Standard Deviation of a Probability Distribution

\[
\mu = \sum (x \cdot P(x))
\]

\[
\sigma^2 = \sum x^2 \cdot P(x) - \mu^2
\]

\[
\sigma = \sqrt{\sigma^2}
\]

Using TI Probability Distribution

1) Enter \( x \)-values into L1,
2) Enter corresponding \( P(x) \) into L2,
3) stat, calc, 1-var stats, L1, L2, enter

Note: \( n=1 \) and \( S_x=\text{blank} \)

Using TI: Probability Histogram

1) Enter these numbers into L1 and L2.
2) Then perform: stat, calc, 1-var stats, L1, L2, enter.

Using TI: Probability Histogram

3) Select Window, and enter the following setting.

4) Then perform:
   2nd, STATPLOT, ENTER to select PLOT1
   Make sure other plots are turned off.

Using TI: Probability Histogram

5) Now enter the following setting for PLOT1

6) Now select GRAPH
Roundoff Rule for $\mu$, $\sigma$, and $\sigma^2$

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 $\mu$, $\sigma$, and $\sigma^2$ and 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 = $\mu + 2\sigma$
- Minimum usual value = $\mu - 2\sigma$

Identifying Unusual Results

Probabilities

Rare Event Rule

If, under a given assumption (such as the assumption that boys and girls are equally likely), the probability of a particular observed event (such as 13 girls in 14 births) 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) is very small (such as 0.05 or less).
- **Unusually low**: $x$ successes among $n$ trials is an unusually low number of successes if $P(x$ or fewer) is very small (such as 0.05 or less).

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 $\sum [x \cdot P(x)]$.

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

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**.
4. The probabilities must remain **constant** for each trial.

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

- $P(S) = p$  ($p$ = probability of success)
- $P(F) = 1 - p = q$  ($q$ = 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, the events can be treated as if they were independent if the sample size is no more than 5% of the population size. (That is \( n \) is less than or equal to 0.05\( N \).)

## Methods for Finding Probabilities

We will now present three methods for finding the probabilities corresponding to the random variable \( x \) in a binomial distribution.

## Method 1: Using the Binomial Probability Formula

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

for \( x = 0, 1, 2, \ldots, n \)

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 (\( q = 1 - p \))
Using TI: Binomial Distribution

**Evaluation of the formula**

**Example:** Find \( P(x=2) \) when \( n=6 \), and \( p=0.7 \).

1) Enter 6, MATH, PRB, nCr, 2, X
2) \( 0.7 \times 2 \times 0.3 \times 4 \), then enter
   \( \binom{6}{2} \times 0.7^2 \times 0.3^4 \).

**Final Answer:** 

\[ \binom{6}{2} \times 0.7^2 \times 0.3^4 = 0.2025 \]

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**Method 2: Using Table A-1 in Appendix A**

Part of Table A-1 is shown below. With \( n = 4 \) and \( p = 0.2 \) in the binomial distribution, the probabilities of 0, 1, 2, 3, and 4 successes are 0.410, 0.410, 0.154, 0.026, and 0.002 respectively.

<table>
<thead>
<tr>
<th>( n )</th>
<th>( x )</th>
<th>( P(x) )</th>
</tr>
</thead>
<tbody>
<tr>
<td>4</td>
<td>0</td>
<td>0.410</td>
</tr>
<tr>
<td>4</td>
<td>1</td>
<td>0.410</td>
</tr>
<tr>
<td>4</td>
<td>2</td>
<td>0.154</td>
</tr>
<tr>
<td>4</td>
<td>3</td>
<td>0.026</td>
</tr>
<tr>
<td>4</td>
<td>4</td>
<td>0.002</td>
</tr>
</tbody>
</table>

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**Method 3: Using Technology**

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

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**Using TI: Binomial Distribution**

**Example:** Find \( P(x=2) \) when \( n=6 \), and \( p=0.7 \).

1) Select 2nd, VARS, arrow down to get to 0:binompdf( enter to select
2) Enter \( 6, 0.7, 2 \) )

This result was obtained earlier by directly using the binomial distribution formula.
Using TI Binomial Distribution

\[ P(x \leq a) \]

1) 2nd VARS( DISTR )
2) Arrow down to binomcdf(
3) enter
4) n, , p , a ) enter

Using TI Binomial Distribution

**Example:** Find \( P(x \leq 2) \) when \( n=6, p=0.7 \).

1) Select 2nd, VARS, arrow down to get to 0:binomcdf( enter to select
2) Enter 6,0.7, 2 )

Using TI Binomial Distribution

3) Enter to execute this operation and get the final answer.

\[ P(x \leq 2) = P(x=0) + P(x=1) + P(x=2) \]

Using TI: Binomial Distribution

**Example:** Find \( P(x \geq 2) \) when \( n=6, p=0.7 \).

1) Enter 1 – Select 2nd, VARS, arrow down to get to 0:binomcdf( enter to select, then type 6,0.7, 2–1 )
2) Enter to get the final answer

\[ P(x \geq 2) = P(x=2) + P(x=3) + \cdots + P(x=6) \]

Using TI: Binomial Distribution

**Example:** Find \( P(x \geq 2) \) when \( n=6, p=0.7 \).

1) Enter 1 – 2nd VARS( DISTR )
2) Arrow down to binomcdf( enter to select
3) enter
4) n, , p , a–1 ) enter

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
Example

Find the mean and standard deviation for the number of girls in groups of 14 births.

This scenario is a binomial distribution where:

\[ n = 14 \quad p = 0.5 \quad q = 0.5 \]

Using the binomial distribution formulas:

Example (cont)

This scenario is a binomial distribution where

\[ n = 14 \quad p = 0.5 \quad q = 0.5 \]

Using the binomial distribution formulas:

\[ \mu = (14)(0.5) = 7 \text{ girls} \]
\[ \sigma = \sqrt{(14)(0.5)(0.5)} = 1.9 \text{ girls (rounded)} \]

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 = \( \mu + 2\sigma \)
Minimum usual values = \( \mu - 2\sigma \)

Example

Determine whether 68 girls among 100 babies could easily occur by chance.

For this binomial distribution,

\[ \mu = 50 \text{ girls} \]
\[ \sigma = 5 \text{ girls} \]

\[ \mu + 2\sigma = 50 + 2(5) = 60 \]
\[ \mu - 2\sigma = 50 - 2(5) = 40 \]

The usual number girls among 100 births would be from 40 to 60. So 68 girls in 100 births is an unusual result.

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

\[ P(x) = \frac{\mu^x e^{-\mu}}{x!} \quad \text{where} \quad e \approx 2.71828 \]

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 \( \mu \).
- The standard deviation is \( \sigma = \sqrt{\mu} \).
Using TI: Poisson Distribution
Evaluation of the formula

Example: Find \( P(x=2) \) when \( \mu=3 \).

1) Enter \( 3^2 \times 2 \text{nd LN } e^{(-3)} \) 
2) then enter for 
\[ 3^2 \times e^{(-3)} = 2 \]
Final Answer: 
\[ \frac{e^{-3}}{2} \approx 0.224 \]

Using TI: Poisson Distribution

\[ P(x = a) \]
1) 2nd VARS ( \text{DISTR} )
2) Arrow down to \text{poissonpdf}( 
3) enter \( \mu, , , a \) enter

Using TI: Poisson Distribution

Example: Find \( P(x=2) \) when \( \mu=3 \).

1) 2nd VARS arrow down to \text{poissonpdf}( 
2) enter to select, then type \( 3, 2 \)
Final Answer: 
\[ P(x=2) \approx 0.224 \]
This result was obtained earlier by directly using the Poisson distribution formula.

Using TI: Poisson Distribution

Example: Find \( P(x\leq3) \) when \( \mu=3 \).

1) 2nd VARS arrow down to \text{poissoncdf}( 
2) enter to select, then type \( 3, 3 \)
Final Answer: 
\[ P(x\leq3) = P(x=0) + P(x=1) + P(x=2) + P(x=3) \]
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 $\mu$.
- In a binomial distribution the possible values of the random variable are $x = 0, 1, \ldots, n$, but a Poisson distribution has possible $x$ values of $0, 1, \ldots$, with no upper limit.

Example

World War II Bombs In analyzing hits by V-1 buzz bombs in World War II, South London was subdivided into 576 regions, each with an area of 0.25 km$^2$. A total of 535 bombs hit the combined area of 576 regions.

If a region is randomly selected, find the probability that it was hit exactly twice.

The Poisson distribution applies because we are dealing with occurrences of an event (bomb hits) over some interval (a region with area of 0.25 km$^2$).

Example

The mean number of hits per region is

$$\mu = \frac{\text{number of bomb hits}}{\text{number of regions}} = \frac{535}{576} = 0.929$$

$$P(x) = \frac{\mu^x \cdot e^{-\mu}}{x!} = \frac{0.929^2 \cdot 0.71828 \cdot 0.929}{2!} = 0.170$$

The probability of a particular region being hit exactly twice is $P(2) = 0.170$.

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

- $n \geq 100$
- $np \leq 10$

Poisson as Approximation to Binomial - $\mu$

- $n \geq 100$
- $np \leq 10$

Value for $\mu$

$$\mu = n \cdot p$$

Monterey Park Police department issues 5 J-walking citations in average during one week of school to ELAC students. Find the probability that at least 4 J-walking tickets will be issued this week by MPPD?

Solution: $P(x \geq 4) = 1 - P(x < 4)$

1) Enter 1–2nd VARS arrow down to poissoncdf

2) ENTER to select
Monterey Park Police department issues 5 J-walking citations in average during one week of school to ELAC students. Find the probability that at least 4 J-walking tickets will be issued this week by MPPD?

3) Type \(7, 4 - 1\)

\[
1 - \text{poisscdf}(7, 4) = 0.916234356
\]

4) Enter to execute this command and get the **final answer**.