

Chapter 7 Sampling Distributions of Estimates

We have studied probability, we have discussed random variables and their probability distributions. Now we turn to *Statistical Inference*: process of using data to draw conclusions about some wider population.

Section 7.1 Parameters and Estimates

Some distinctions to keep in mind ...

- Population versus Sample
- Parameter versus Statistic and Estimate

Definitions:

- **Parameter**: A number that describes the **population**.
- **Statistic (or estimator)**: An abstraction that describes a **sample**.
- **Estimate**: A number **calculated from the data** to estimate an unknown parameter

Notation:

- Population proportion p versus sample proportion $\hat{P}$ or $\hat{p}$
- Population mean μ versus sample mean $\bar{X}$ or $\bar{x}$

Since we hardly ever know the true population parameter value, we take a sample $X_1, X_2, \dots, X_n$ and use the sample statistic to **estimate** the parameter. When we do this, the sample statistic may not be equal to the population parameter, in fact, it would change every time we take a new sample (that is why we call this process *ESTIMATION*). Will the observed sample statistic value be a reasonable estimate? If the sample is a **random sample**, then we will be able to say something about the accuracy of the estimation process. What do we mean by a **random sample**?

Recall that $X_1, X_2, \dots, X_n$ is a random sample if ...

- 1.
- 2.

Example:

A poll was conducted by The Heldrich Center for Workforce Development (at Rutgers University).
 A random sample of 1000 workers resulted in 460 (for 46%) stating they work more than 40 hours per week.

Identify:

Population =

Parameter =

Sample =

Estimate =

Can anyone say how close this observed sample proportion $\hat{p}$ is to the true population proportion p ? If we were to take another random sample of the same size $n = 1000$, would we get the same value for the sample proportion $\hat{p}$?

The value of a statistic from a random sample, like $\hat{p}$, will vary from sample to sample. So a statistic (also called **estimator**) is a random variable and it will have a probability distribution. This probability distribution is called the **sampling distribution** of the statistic. We will study in this chapter the sampling distribution of the following statistics:

Sample Mean: $\bar{X}$ = the mean of a sample of size n

Sample Proportion: $\hat{P} = \frac{X}{n}$ = the proportion of *successes* in a sample of size n

Section 7.2 Sampling Distribution of the Sample Mean**Scenario:**

A poll was conducted by The Heldrich Center for Workforce Development (at Rutgers University).
 A **sample of 1000 workers** resulted in a **mean number of hours worked per week** of 43.

Population =

Parameter =

Sample =

Estimate =

Can anyone say how close this observed sample mean $\bar{x}$ is to the true population mean μ ?

If we were to take another random sample of the same size $n = 1000$, would we get the same value for the sample mean $\bar{x}$?

Picture Model:

The value of the sample mean from a random sample will vary from sample to sample. So the sample mean (a statistic) is a random variable and it will have a probability distribution. This probability distribution is called the **sampling distribution** of the statistic. In Section 7.2 we learn about the sampling distribution for a sample mean, $\bar{X}$ = the mean of a sample of size n . What can we say about the $\bar{X}$ basket (population)?

Notation:

- population mean μ
- population standard deviation σ
- random sample of size n : $X_1, X_2, \dots, X_n$
- sample mean $\bar{X} = \frac{X_1 + X_2 + \dots + X_n}{n}$

Main Results:

1. The average of all of the possible sample mean values (from all possible samples of the same size n) is equal to _____

That is, the mean of the sample means is equal to the population mean.

Thus the sample mean is an _____ estimator of the population mean.

2. The standard deviation of all of the possible sample mean values is equal to the original population standard deviation divided by $\sqrt{n}$.

This is **approximately the average distance of the possible sample mean values from the population mean**. It is a measure of the *accuracy of the process* of using a sample mean to estimate the population mean.

Would this quantity $\frac{\sigma}{\sqrt{n}}$ tell me how far away a particular $\bar{x}$ value is from μ ?

Note: If the sample size increases, the standard deviation decreases, which says the possible sample mean values will be closer to the true population mean (on average).

3. If the **population has a normal distribution**,
then the distribution of the sample mean is _____

Note: Sums and differences of independent normal r.v. also have a normal distribution (Section 6.4.3).

4. If the **population is not necessarily normally distributed**,
but the **sample size n is large**,
then the distribution of the sample mean is _____

This result is called the _____

See Figures 7.2.3- 7.2.5 on pages 285-287.

Rule of Thumb: In general, the approximation is good if _____

Example

ACT scores: **Normally distributed** with mean of 18.6 and standard deviation of 5.9.

- What is the probability that a **single student** randomly selected will score 21 or higher?
- A random sample of 50 students is obtained. What is the probability that the **mean score** for these 50 students will be 21 or higher?
- What if the normal distribution assumption were not given? How would your answer to part b change? Explain.

Example: Uniform Distribution

Suppose the random variable X = actual flight time (in minutes) for Delta Airlines flights from Cincinnati to Tampa follows a *uniform distribution over the range of 110 minutes to 130 minutes*.

- a. Sketch the distribution for X (include labels for the axes and some values on the axes).
- b. Suppose we were to repeatedly take a random sample of size 100 from this distribution and compute the sample mean for each sample. What would the *histogram of the sample mean values* look like? Provide a sketch and any details about the distribution of the sample mean that you can.

Example:

Grinding Machine prepares axles: target diameter = 40.125 mm. The standard deviation of diameters is 0.002 mm. A (random) sample of 4 axles taken each hour and the sample mean diameter is recorded and plotted. What will be the mean and standard deviation of the numbers recorded?

Note:

When $n = 1$ we have:

When $n = 4$ we have:

So to reduce the standard deviation in half, we must _____

Question: What sample size would be needed for the standard deviation of the sample mean to be 0.0005 mm?

Standard Error of Sample Mean

Recall that the s.d. of sample mean $\bar{X}$ is _____

In practice, we don't know s.d. σ , so we have to estimate σ and $sd(\bar{X})$.

We can estimate σ by _____

and estimate $sd(\bar{X})$ by _____

The Standard Error of the Sample Mean is

The Standard Error is a measure of the _____ of an estimate. You should always report an estimate and its standard error. One way to do this is to report a **confidence intervals**.

A two-standard-error interval for μ is