

Chapter 7: Sampling Distributions

Section 7.2 Sample Proportions

> The Practice of Statistics, 4th edition – For AP* STARNES, YATES, MOORE

Chapter 7 Sampling Distributions

7.1 What is a Sampling Distribution?

- **7.2** Sample Proportions
- **7.3** Sample Means



Learning Objectives

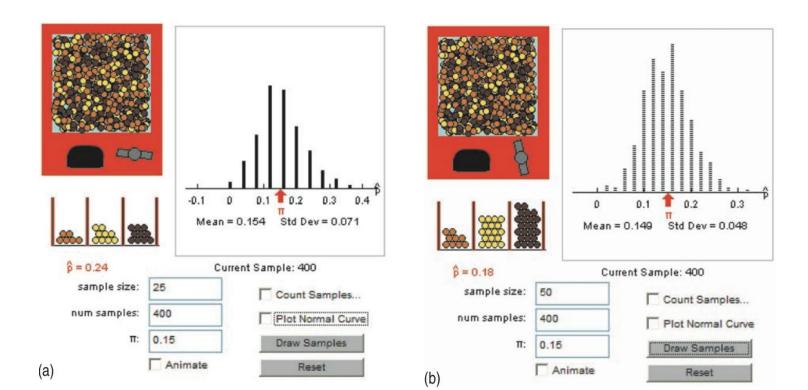
After this section, you should be able to...

- FIND the mean and standard deviation of the sampling distribution of a sample proportion
- DETERMINE whether or not it is appropriate to use the Normal approximation to calculate probabilities involving the sample proportion
- CALCULATE probabilities involving the sample proportion
- EVALUATE a claim about a population proportion using the sampling distribution of the sample proportion

How good is the statistic \hat{p} as an estimate of the parameter p? The sampling distribution of \hat{p} answers this question.

Consider the approximate sampling distributions generated by a simulation in which SRSs of *Reese's Pieces* are drawn from a population whose proportion of orange candies is either 0.45 or 0.15.

What do you notice about the shape, center, and spread of each?



What did you notice about the shape, center, and spread of each sampling distribution?

Shape: In some cases, the sampling distribution of \hat{p} can be approximated by a Normal curve. This seems to depend on both th sample size n and the population proportion p.

Center : The mean of the distribution $is\mu_{\hat{p}} = p$. This makes sense because the sample proportion \hat{p} is an unbiased estimator of p.

Spread: For a specific value of p, the standard deviation $\sigma_{\hat{p}}$ gets smaller as n gets larger. The value of $\sigma_{\hat{p}}$ depends on both n and p.

There is and important connection between the sample proportio \hat{p} and the number of "successes" *X* in the sample.

$$\hat{p} = \frac{\text{count of successes in sample}}{\text{size of sample}} = \frac{X}{n}$$

In Chapter 6, we learned that the mean and standard deviation of a binomial random variable *X* are

$$\mu_X = np$$
 $\sigma_X = \sqrt{np(1-p)}$

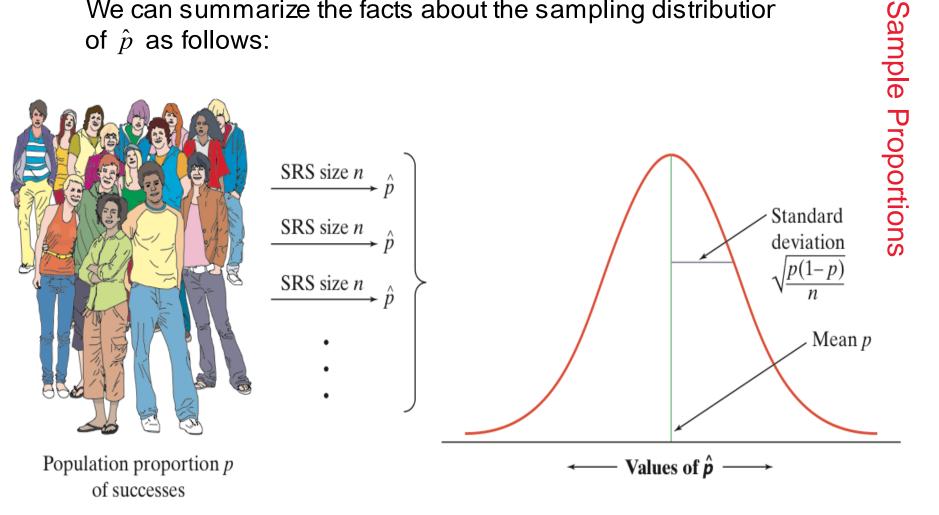
Since $\hat{p} = X/n = (1/n) \cdot X$, we are just multiplying the random variable X by a constant (1/n) to get the random variable \hat{p} . Therefore,

$$\mu_{\hat{p}} = \frac{1}{n}(np) = p \qquad \qquad \hat{p} \text{ is an unbiased estimator op}$$

$$\sigma_{\hat{p}} = \frac{1}{n}\sqrt{np(1-p)} = \sqrt{\frac{np(1-p)}{n^2}} = \sqrt{\frac{p(1-p)}{n}}$$

As sample size increases, the spread decreases.

We can summarize the facts about the sampling distribution of \hat{p} as follows:



Using the Normal Approximation for \hat{p}

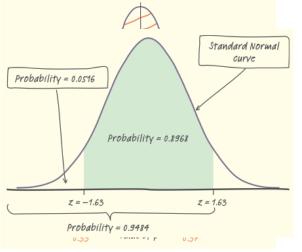
Inference about a population proportion p is based on the sampling distribution of \hat{p} . When the sample size is large enough for np and n(1-p) to both be at least 10 (the Normal condition) the sampling distribution of \hat{p} is approximately Normal.



A polling organization asks an SRS of 1500 first-year college students how far away their home is. Suppose that 35% of all first-year students actually attend college within 50 miles of home. What is the probability that the random sample of 1500 students will give a result within 2 percentage points of this true value?

STATE: We want to find the probability that the sample proportion falls between 0.33 and 0.37 (within 2 percentage points, or 0.02, of 0.35).

PLAN: We have an SRS of size n = 1500 drawn from a population in which the proportion p = 0.35 attend college within 50 miles of home.



 $\mu_{\hat{p}} = 0.35 \qquad \qquad \sigma_{\hat{p}} = \sqrt{\frac{(0.35)(0.65)}{1500}} = 0.0123$ **DO**: Since np = 1500(0.35) = 525 and n(1 - p) = 1500(0.65) = 975 are both greater than 10, we'll standardize and then use Table A to find the desired probability. $z = \frac{0.33 - 0.35}{0.123} = -1.63 \qquad z = \frac{0.37 - 0.35}{0.123} = 1.63$ $P(0.33 \le \hat{p} \le 0.37) = P(-1.63 \le Z \le 1.63) = 0.9484 - 0.0516 = 0.8968$

CONCLUDE: About 90% of all SRSs of size 1500 will give a result within 2 percentage points of the truth about the population.

Alternate Example – Planning for College

The superintendent of a large school district wants to know what proportion of middle school students in her district are planning to attend a four-year college or university. Suppose that 80% of all middle school students in her district are planning to attend a four-year college or university. What is the probability that a SRS of size 125 will give a result within 7 percentage points of the true value?

STATE: We want to find the probability that the proportion of middle school students who plan to attend a four-year college or university falls between 73% and 87%. That is, $P(0.73 \le p \le 0.87)$.

PLAN: Because the school district is large, we can assume that there are more than 10(125) = 1250 middle school students so

$$\mu_{\hat{p}} = 0.80$$
 $\sigma_{\hat{p}} = \sqrt{\frac{(0.80)(0.20)}{125}} = 0.036$

We can consider the distribution of \hat{p} to be approximately Normal

since $np = 125(0.80) = 100 \ge 10$ and $n(1-p) = 125(0.20) = 25 \ge 10$.

DO: $P(0.73 \le \hat{p} \le 0.87) = normalcdf(0.73, 0.87, 0.80, 0.036) = 0.948$

(Note: To get full credit when using normalcdf on an AP exam question, students must explicitly state the mean and standard deviation of the distribution as in the Plan step above.)

CONCLUDE: About 95% of all SRSs of size 125 will give a sample proportion within 7 percentage points of the true proportion of middle school students who are planning to attend a four-year college or university.

Section 7.2 Sample Proportions

Summary

In this section, we learned that...

When we want information about the population proportion p of successes, we \checkmark often take an SRS and use the sample proportion \hat{p} to estimate the unknown parameter p. The **sampling distribution** of \hat{p} describes how the statistic varies in all possible samples from the population.

The **mean** of the sampling distribution of \hat{p} is equal to the population proportion p. That is, \hat{p} is an unbiased estimator of p.

The standard deviation of the sampling distribution of \hat{p} is $\sigma_{\hat{p}} = \sqrt{\frac{p(1-p)}{n}}$ for

an SRS of size *n*. This formula can be used if the population is at least 10 time as large as the sample (the 10% condition). The standard deviation ϕ^{\ddagger} gets smaller as the sample size*n* gets larger.

When the sample size *n* is larger, the sampling distribution of \hat{p} is close to a

Normal distribution with mean *p* and standard deviation $\sigma_{\hat{p}} = \sqrt{\frac{p(1-p)}{n}}$.

✓ In practice, use this Normal approximation when both $np \ge 10$ and $n(1 - p) \ge 10$ (the Normal condition).



In the next Section...

We'll learn how to describe and use the sampling distribution of sample means.

We'll learn about

- \checkmark The sampling distribution of $\overline{\chi}$
- Sampling from a Normal population
- The central limit theorem