

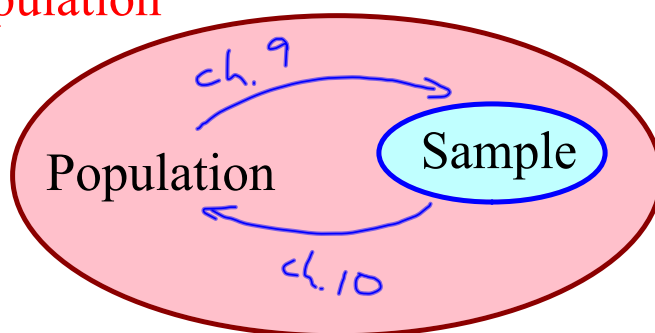
9.1 Sampling distributions



One day some papers catch fire in a wastebasket in the Dean's office. Luckily, a physicist, a chemist, and a statistician happen to be nearby. Naturally, they rush in to help. The physicist whips out a notebook and starts to work on how much energy would have to be removed from the fire in order to stop the combustion. The chemist works on determining which reagent would have to be added to the fire to prevent oxidation. While they are doing this, the statistician is setting fires to all the other wastebaskets in the adjacent offices. "What are you doing?" the Dean demands. To which the statistician replies, "To solve a problem of this magnitude, you need a large sample size."

Recall these terms:

parameter: a fixed, often unknown #
describing a **population**



statistic: a number known for a particular sample that describes that **sample**, but the value likely changes from one sample to another sample

We estimate parameters 2 ways:
point estimates (ch. 9) and
interval estimates (ch. 10).

These statistics are the best point estimates for these parameters:

	<i>mean</i>	<i>variance</i>	<i>standard deviation</i>	<i>proportion</i>
statistics	\bar{x}	s^2	s	\hat{p}
parameters	μ	σ^2	σ	p

\approx

Common Sense Thing #1 Random samples are good.

A random sample should represent the population well, so a statistic from a random sample should provide a reasonable estimate of a population parameter.

Common Sense Thing #2 Statistics have error.

All statistics have some error in estimating population parameters.

Common Sense Thing #3 Statistics have distributions.

If repeated samples are taken from a population and some statistic is calculated from each sample, the statistics will vary. That is, the statistic will have a distribution with a mean and standard deviation.

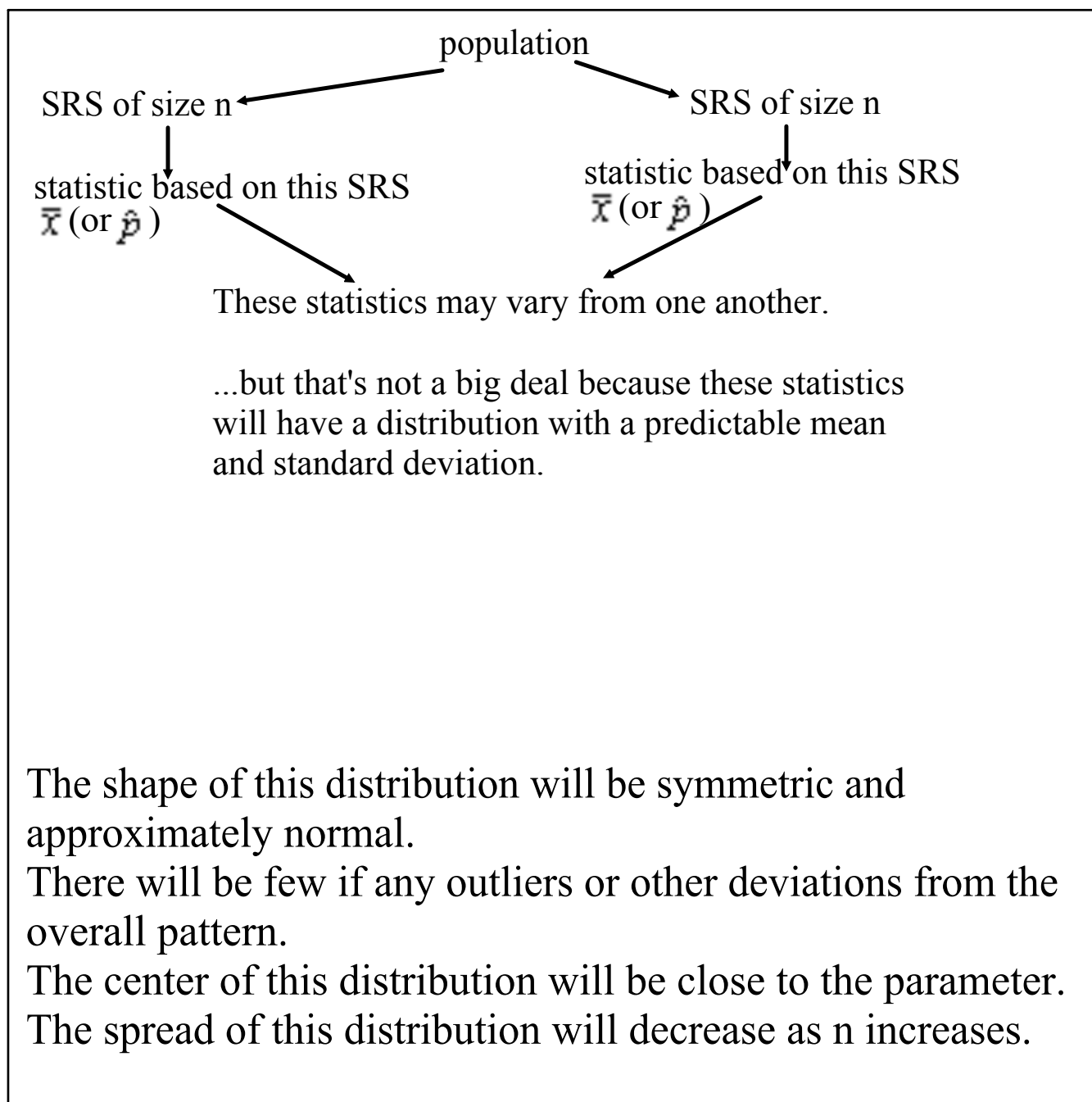
Common Sense Thing #4 Larger sample size is better: less error.

A larger sample provides more information than a smaller sample so a statistic from a large sample should provide have less error than a statistic from a small sample.

Fact: Different samples yield different values of statistics.
That's called sampling variability.

The sampling distribution of a statistic is

- the probability distribution of the statistic.
- the distribution of values taken by the statistic in all possible samples of the same size from the same population.



A statistic (used to estimate a parameter) is **unbiased** if the mean of its sampling distribution is equal to the value of the parameter being estimated.

The **variability of a statistic** is described by the spread of its sampling distribution.

Larger samples give smaller spread. However, as long as the population is much larger than the sample (say, about 10 times larger, $N > 10n$), the spread of the sampling distribution is approximately the same for any population size.

Drag these displays to the appropriate places!

Low Variability

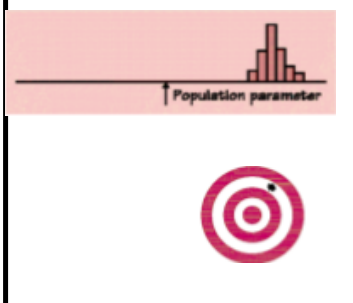
High Variability

Drag from here.

Low Bias



High Bias



When a SRS is taken, the observed sample statistic is likely to differ from the expected population parameter. This difference is due to:

- Bias from the sampling procedure (we do everything we can to reduce or eliminate this)
and/or
- Chance error, since every possible subset of the population has the same probability of being chosen
and/or
- Something significant, leading us to believe that the expected population parameter differs from what we assumed.