## SAMPLING DISTRIBUTIONS



The sampling distribution of a statistic is the distribution of all values of the statistic when all possible samples of the same size $n$ are taken from the same population.

The sampling distribution of the mean is the distribution of sample means when all possible samples of the same size $n$ are taken from the same population.
$\bar{x}=$ sample mean
$\mu=$ population mean

The sampling distribution of the variance is the distribution of sample variances when all possible samples of the same size $n$ are taken from the same population.

$$
\begin{aligned}
& s^{2}=\text { sample variance } \\
& \sigma^{2}=\text { population variance }
\end{aligned}
$$

The sampling distribution of the proportion is the distribution of sample proportions when all possible samples of the same size $n$ are taken from the same population.
$\hat{p}=$ sample proportion
$p=$ population proportion

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We also say that each of these is an unbiased estimator of the corresponding population parameter.

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The median and the range are also not unbiased estimators. They do not target the corresponding population parameters as the sample size increases.

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On the other hand, the sampling distribution of the sample variances tends toward a distribution that is skewed to the right as the sample size increases.

The Central Limit Theorem: Let $X$ be a random variable with mean $\mu$ and standard deviation $\sigma$, and consider the sampling distribution of the sample means for all simple random samples of size $n$. We will denote the mean of this distribution by $\mu_{\bar{x}}$ and the standard deviation by $\sigma_{\bar{x}}$. Then,

1. $\mu_{\bar{x}}=\mu$
2. $\sigma_{\bar{x}}=\frac{\sigma}{\sqrt{n}} \quad$ (the standard error of the mean)
3. The sampling distribution of the sample means approaches a normal distribution as the sample size increases.

## Additionally, we have following two results:

1. If the parent population is normally distributed, then the sampling distribution of the sample means will always be normally distributed, regardless of the sample size $n$.
2. If the parent population is not normally distributed, then we do not expect the sampling distribution of the sample means to reasonably approximate a normal distribution unless $n>30$.

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$$
z=\frac{x-\mu}{\sigma}=\frac{185-172}{29} \approx 0.448
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What is the probability that 20 randomly selected men have a mean weight greater than 185 lb ?

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z=\frac{\bar{x}-\mu_{\bar{x}}}{\sigma_{\bar{x}}}=\frac{\bar{x}-\mu}{\sigma / \sqrt{n}}=\frac{185-172}{29 / \sqrt{20}} \approx 2.005
$$



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Does this difference have practical significance to you?

