

# STAT 100 Lecture 13: Sampling Distributions

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- 1 Calculations with Normal Distributions
- 2 The Normal Approximation to the Binomial Distribution
  - The  $(n, p)$  Binomial is approximated by  $N(np, \sqrt{np(1-p)})$ .
  - $P[a \leq X_{\text{binomial}} \leq b] \approx P[a - 0.5 \leq X_{\text{normal}} \leq b + 0.5]$ .

# Today's Agenda

- 1 Sampling Distribution of a Statistic (Section 7.1, 7.2)
- 2 The Central Limit Theorem (Section 7.3)
- 3 Examples

# Parameters versus Statistics

**Parameters** are intrinsic features of a population, whereas **Statistics** are quantities inferred from samples from the population. For example,  $\mu$  and  $\sigma$  are Parameters of  $N(\mu, \sigma)$ . If  $x_1, x_2, \dots, x_N$  are all generated from  $N(\mu, \sigma)$  then

$$\bar{x} = \frac{x_1 + x_2 + \dots + x_N}{N}$$

and

$$s = \sqrt{\frac{(x_1 - \bar{x})^2 + (x_2 - \bar{x})^2 + \dots + (x_N - \bar{x})^2}{N - 1}}$$

are Statistics.

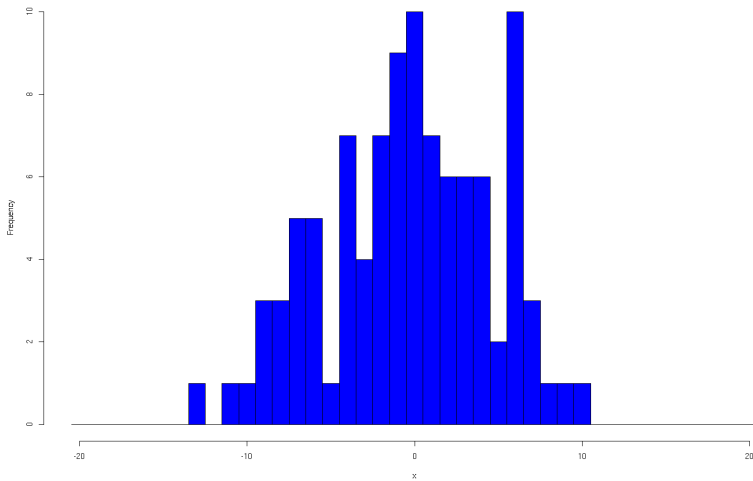
# Parameter or Statistic?

## Example

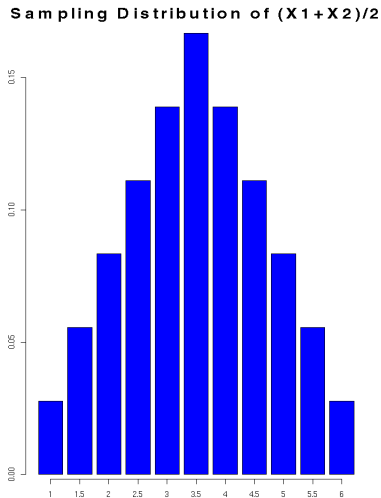
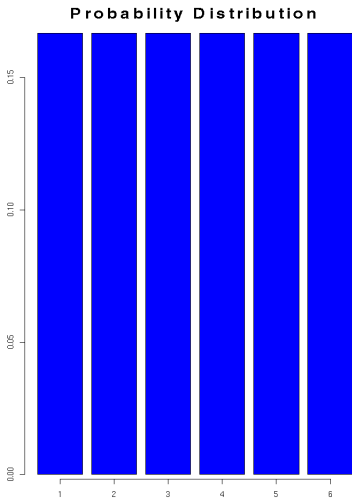
- 1 *Median of a population.*
- 2 *Mean of an empirical distribution.*
- 3 *Sample variation.*

# Variation of a Statistic

Means of 100 Samples of 25 from  $N(0,1)$



# Sampling Distribution



# Conditions for Randomness

Observations  $X_1, X_2, \dots, X_N$  are a **Random Sample** from a population if they are independent and they always have the same distribution.

Now, if we sample without replacement, then the population changes for each  $X_i$  and they are also dependent. However, the sample can be considered Random if the population is much larger than  $N$ .

# Example: Sampling Distribution

## Example

Consider the set  $S = \{1, 2, 3, 4\}$ . Two samples will be drawn from  $S$  without replacement.

- 1 List all possible samples and evaluate  $\bar{x}$  for each sample.
- 2 Determine the sampling distribution of  $\bar{X}$ .

# The Sampling Distribution of $\bar{X}$ for a Normal Population

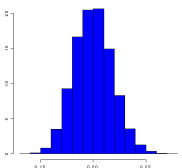
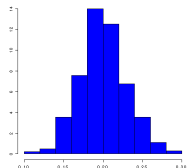
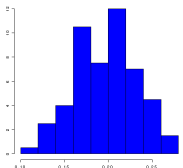
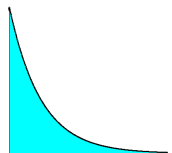
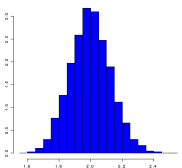
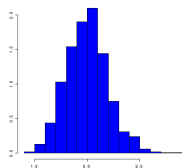
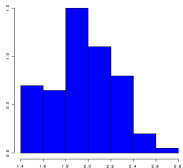
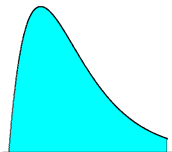
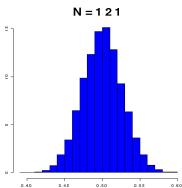
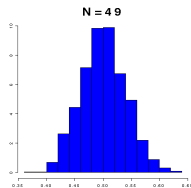
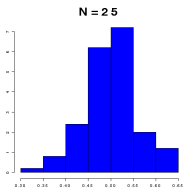
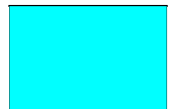
## Fact

If the population is distributed as  $N(\mu, \sigma)$ , then  $\bar{X}$  is distributed as  $N\left(\mu, \frac{\sigma}{\sqrt{N}}\right)$ .

## Example

The number of Skittles in a bag has a binomial distribution with  $n = 1000$  and  $p = 0.9$ . What is the probability that the average number of Skittles in 9 bags is greater than 800?

# Sample Distributions for $\bar{X}$ as $N$ gets Large



# Parameters of the Sampling Distribution of $\bar{X}$

## Fact

*For a population with mean  $\mu$  and standard deviation  $\sigma$ , the sampling distribution of  $\bar{X}$  based on samples of size  $N$  satisfies*

$$E(\bar{X}) = \mu,$$

$$\text{Var}(\bar{X}) = \frac{\sigma^2}{N},$$

*and*

$$\text{sd}(\bar{X}) = \frac{\sigma}{\sqrt{N}}.$$

# The Central Limit Theorem

## The Central Limit Theorem

*For ANY population, the distribution of  $\bar{X}$  is approximately normal when  $N$  is large. Moreover, the distribution of*

$$Z = \frac{\bar{X} - \mu}{\sigma/\sqrt{N}}$$

*is well approximated by  $N(0, 1)$  if  $\mu$  and  $\sigma$  are the population mean and population standard deviation, respectively.*

# Example: Application of the Central Limit Theorem

## Example

Consider a population with  $\mu = 5$  and  $\sigma = 2$ .

- 1 If a random sample of size  $N = 49$  is selected, what is the probability that the sample mean is between 4.5 and 5.5?
- 2 If a random sample of size  $N = 25$  is selected, what is the probability that the sample mean is less than 7?

# For Next Time

- MINITAB Project 03! Due Monday!
- Group Problems:

|         |      |      |      |      |      |
|---------|------|------|------|------|------|
| Group   | 1    | 2    | 3    | 4    | 5    |
| Problem | 7.1  | 7.3  | 7.5  | 7.7  | 7.9  |
| Group   | 6    | 7    | 8    | 9    | 10   |
| Problem | 7.11 | 7.13 | 7.15 | 7.21 | 7.23 |