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# Sampling Variability and Confidence Intervals

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#### **Lecture Topics**

- Sampling distribution of a sample mean
- Variability in the sampling distribution
- Standard error of the mean
- Standard error vs. standard deviation
- Confidence intervals for the population mean μ
- Sampling distribution of a sample proportion
- Standard error for a proportion
- Confidence intervals for a proportion



#### Section A

The Random Sampling Behavior of a Sample Mean Across Multiple Random Samples

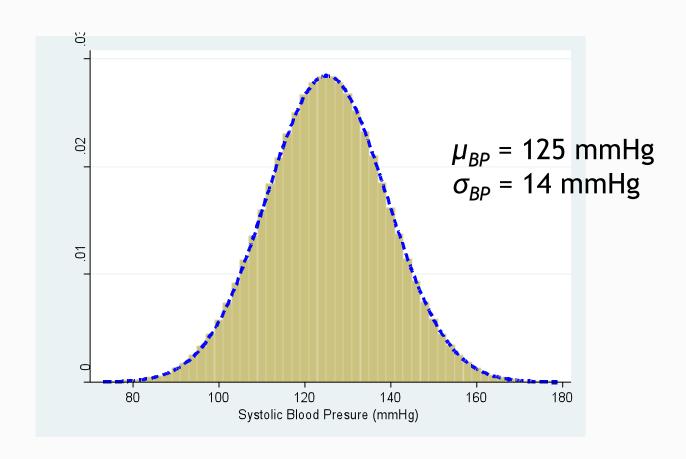
#### Random Sample

- When a sample is randomly selected from a population, it is called a random sample
  - Technically speaking values in a random sample are representative of the distribution of the values in the population sample, regardless of size
- In a simple random sample, each individual in the population has an equal chance of being chosen for the sample
- Random sampling helps control systematic bias
- But even with random sampling, there is still sampling variability or error

#### Sampling Variability of a Sample Statistic

- If we repeatedly choose samples from the same population, a statistic will take different values in different samples
- If the statistic does not change much if you repeated the study (you get similar answers each time), then it is fairly reliable (not a lot of variability)

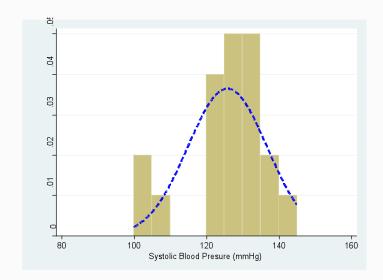
- Recall, we had worked with data on blood pressures using a random sample of 113 men taken from the population of all men
- Assume the population distribution is given by the following:



- Suppose we had all the time in the world
- We decide to do an experiment
- We are going to take 500 separate random samples from this population of men, each with 20 subjects
- For each of the 500 samples, we will plot a histogram of the sample BP values, and record the sample mean and sample standard deviation
- Ready, set, go . . .

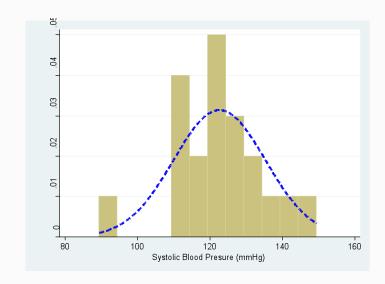
## Random Samples

■ Sample 1: *n* = 20



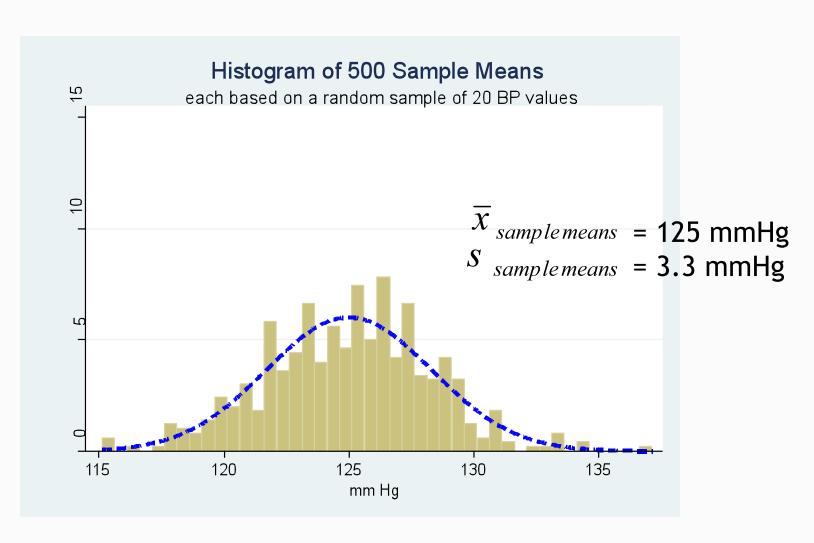
$$\overline{X}_{BP}$$
 = 125.7 mmHg  
 $S_{BP}$  = 10.9 mmHg

■ Sample 2: *n* = 20



$$\overline{x}_{BP}$$
 = 122.6 mmHg  
 $S_{BP}$  = 12.7 mmHg

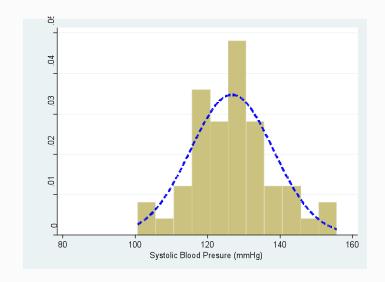
 So we did this 500 times: now let's look at a histogram of the 500 sample means



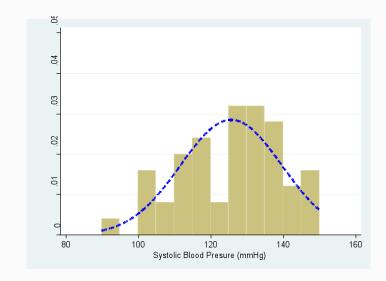
- We decide to do another experiment
- We are going to take 500 separate random samples from this population of me, each with 50 subjects
- For each of the 500 samples, we will plot a histogram of the sample BP values, and record the sample mean and sample standard deviation
- Ready, set, go . . .

## Random Samples

■ Sample 1: *n* = 50



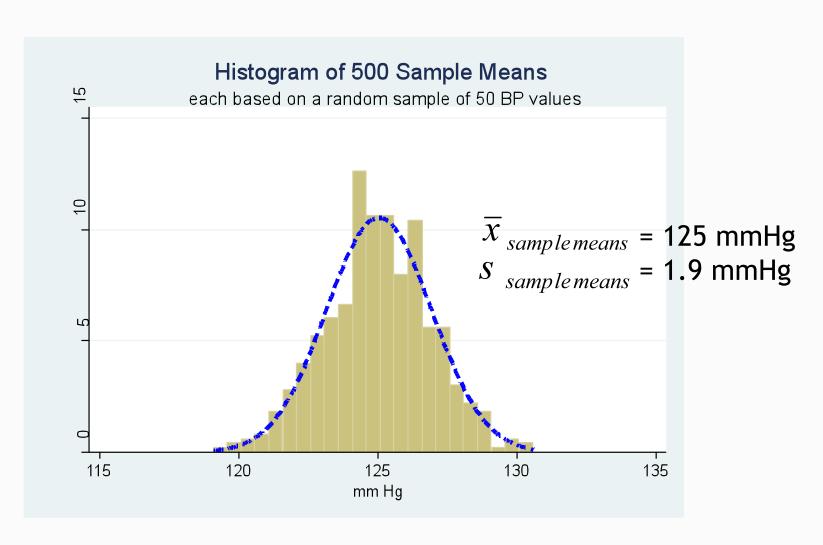
■ Sample 2: *n* = 50



$$\overline{X}_{BP}$$
 = 126.7 mmHg  $S_{BP}$  = 11.5 mmHg

$$\overline{X}_{BP}$$
 = 125.5 mmHg  $S_{BP}$  = 14.0 mmHg

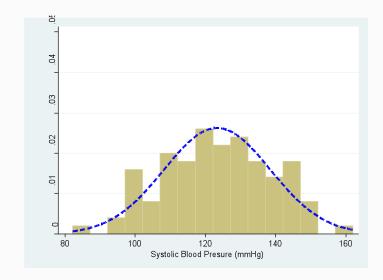
 So we did this 500 times: now let's look at a histogram of the 500 sample means



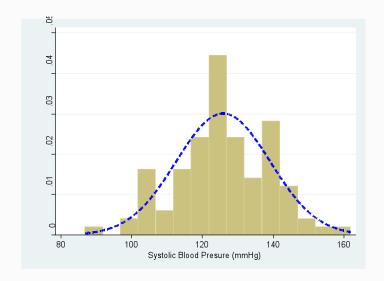
- We decide to do one more experiment
- We are going to take 500 separate random samples from this population of men, each with 100 subjects
- For each of the 500 samples, we will plot a histogram of the sample BP values, and record the sample mean, and sample standard deviation
- Ready, set, go . . .

## Random Samples

■ Sample 1: *n* = 100



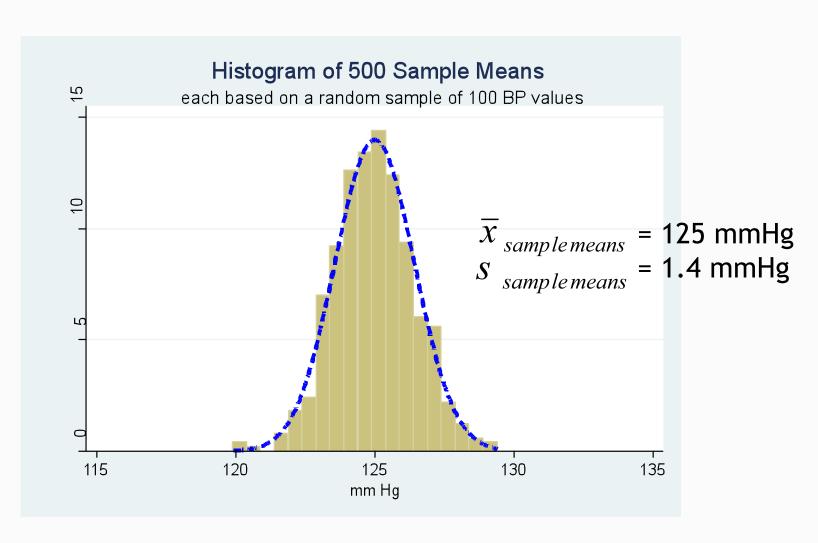
■ Sample 2: *n* = 100



$$\overline{X}_{BP}$$
 = 123.3 mmHg  $S_{BP}$  = 15.2 mmHg

$$\overline{X}_{BP}$$
 = 125.7 mmHg  $S_{BP}$  = 13.2 mmHg

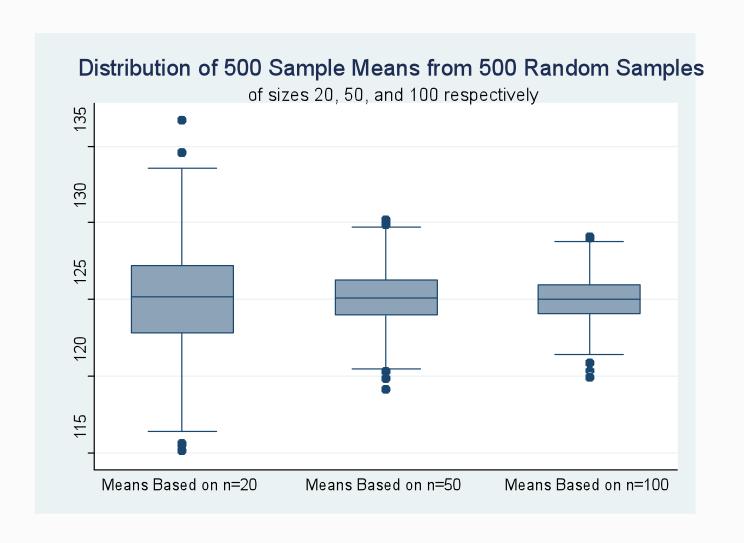
 So we did this 500 times: now let's look at a histogram of the 500 sample means



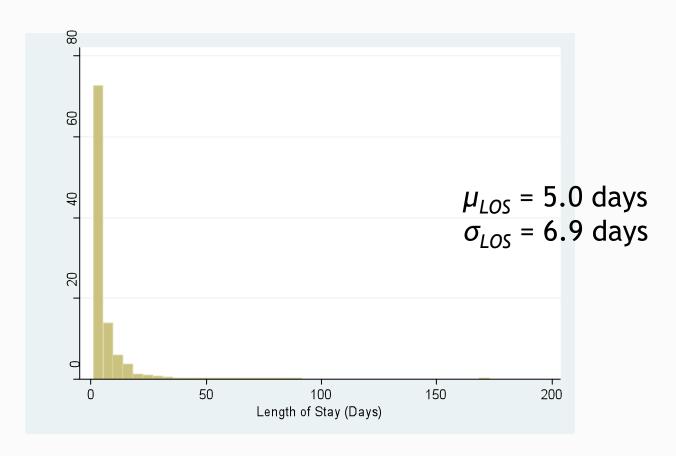
- Let's review the results
  - Population distribution of individual BP measurements for males: normal
  - True mean  $\mu$  = 125 mmHg:  $\sigma$  = 14 mmHg
  - Results from 500 random samples:

Sample Sizes	Means of 500 Sample Means	SD of 500 Sample Means	Shape of Distribution of 500 sample means
n = 20	125 mmHg	3.3 mm Hg	Approx normal
n = 50	125 mmHg	1.9 mm Hg	Approx normal
n = 100	125 mmHg	1.4 mm Hg	Approx normal

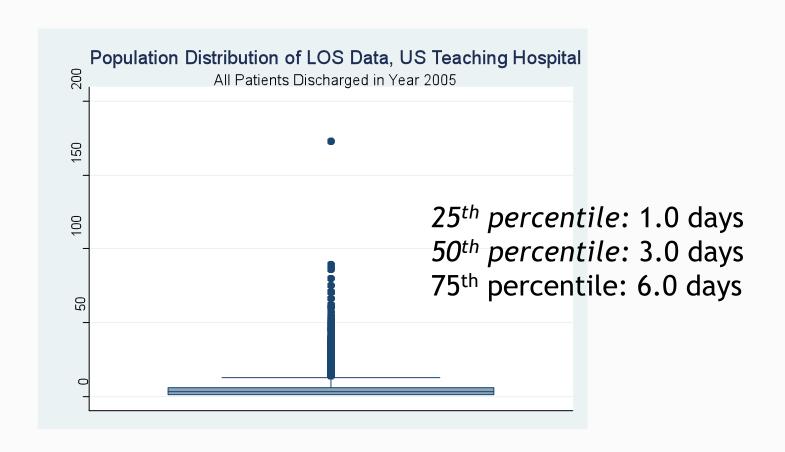
Let's review the results



- Recall, we had worked with data on length of stay (LOS) using a random sample of 500 patients taken from sample of all patients discharged in year 2005
- Assume the population distribution is given by the following:



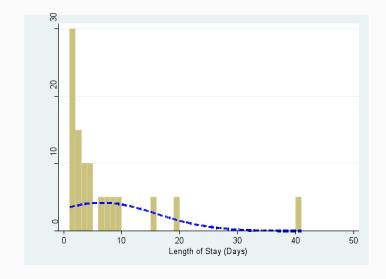
Boxplot presentation



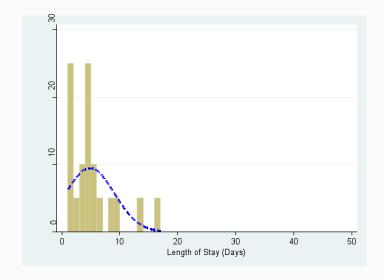
- Suppose we had all the time in the world again
- We decide to do another set of experiments
- We are going to take 500 separate random samples from this population of patients, each with 20 subjects
- For each of the 500 samples, we will plot a histogram of the sample LOS values, and record the sample mean and sample standard deviation
- Ready, set, go . . .

## Random Samples

■ Sample 1: *n* = 20



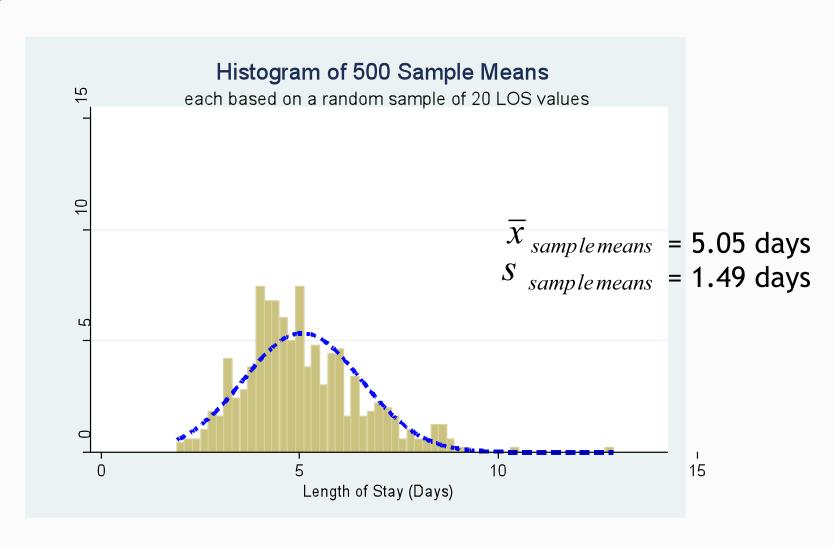
■ Sample 2: *n* = 20



$$\overline{x}_{LOS}$$
 = 6.6 days  $S_{LOS}$  = 9.5 days

$$\overline{X}_{LOS}$$
 = 4.8 days  $S_{LOS}$  = 4.2 days

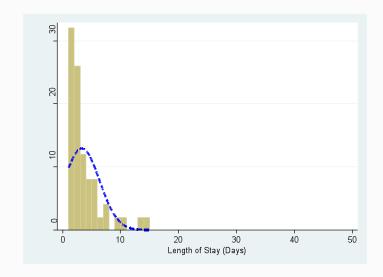
 So we did this 500 times: now let's look at a histogram of the 500 sample means



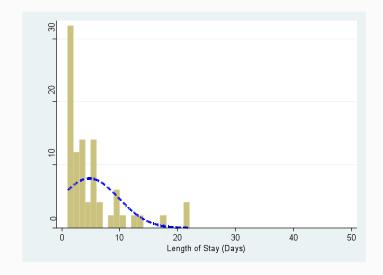
- Suppose we had all the time in the world again
- We decide to do one more experiment
- We are going to take 500 separate random samples from this population of me, each with 50 subjects
- For each of the 500 samples, we will plot a histogram of the sample LOS values, and record the sample mean and sample standard deviation
- Ready, set, go . . .

## Random Samples

■ Sample 1: *n* = 50



■ Sample 2: *n* = 50

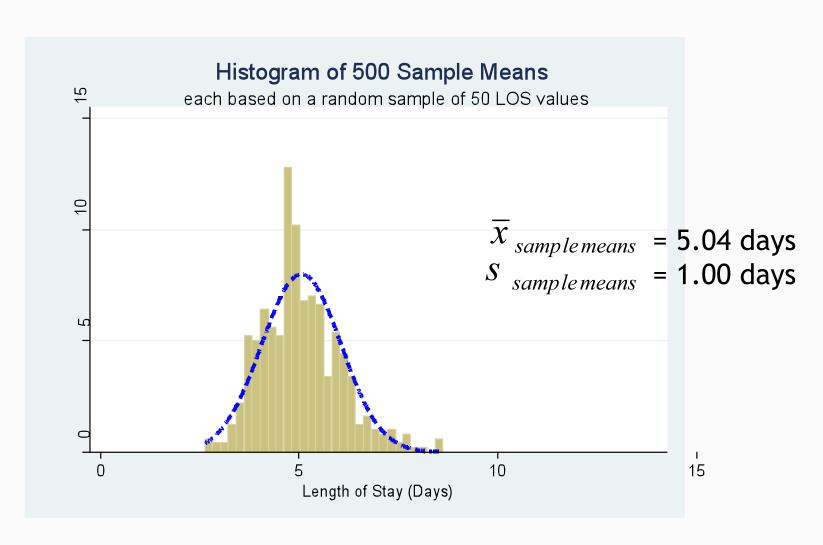


$$\overline{x}_{LOS}$$
 = 3.3 days  $S_{LOS}$  = 3.1 days

$$\overline{x}_{LOS}$$
 = 4.7 days  
 $s_{LOS}$  = 5.1 days

#### Distribution of Sample Means

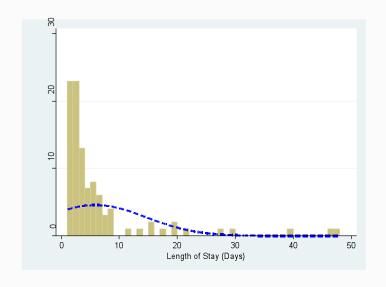
 So we did this 500 times: now let's look at a histogram of the 500 sample means



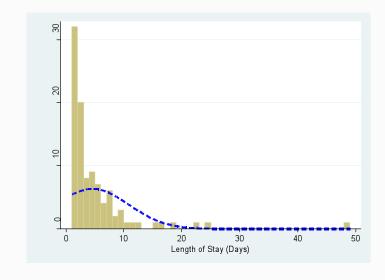
- Suppose we had all the time in the world again
- We decide to do one more experiment
- We are going to take 500 separate random samples from this population of me, each with 100 subjects
- For each of the 500 samples, we will plot a histogram of the sample BP values, and record the sample mean and sample standard deviation
- Ready, set, go . . .

## Random Samples

■ Sample 1: *n* = 100



■ Sample 2: *n* = 100

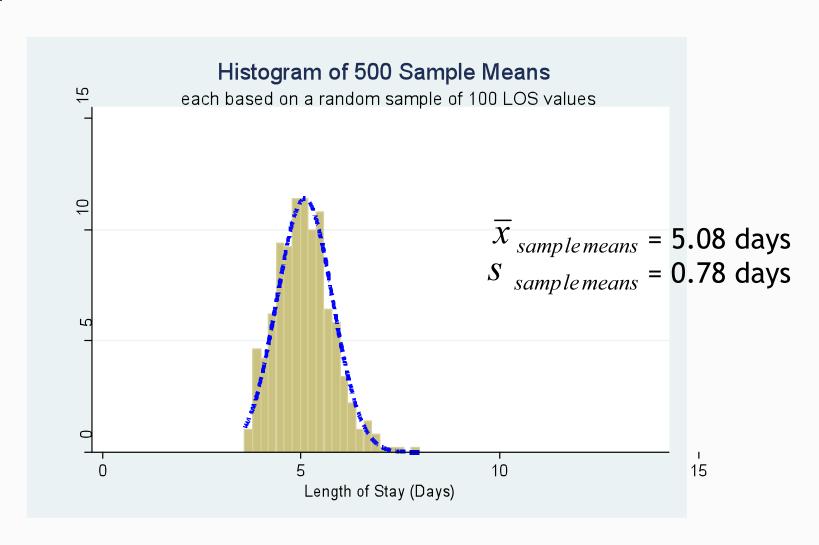


$$\overline{x}_{LOS}$$
 = 5.8 days  
 $s_{LOS}$  = 9.7 days

$$\overline{x}_{LOS}$$
 = 4.5 days  
 $s_{LOS}$  = 6.5 days

#### Distribution of Sample Means

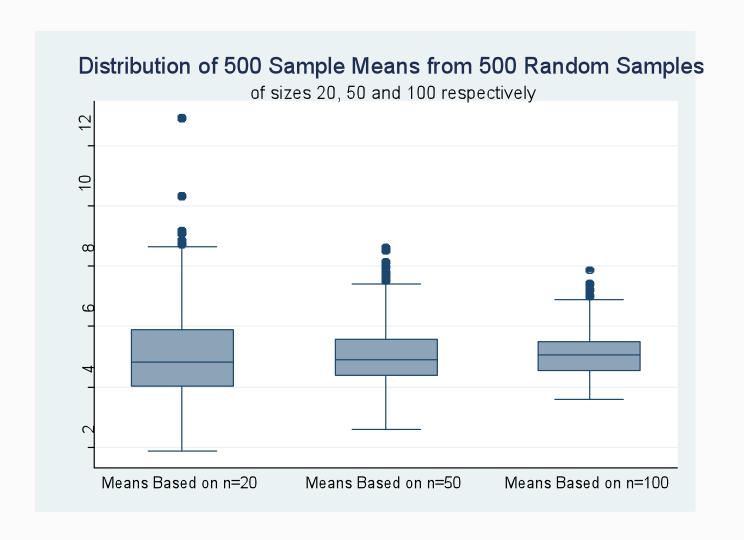
 So we did this 500 times: now let's look at a histogram of the 500 sample means



- Let's review the results
  - Population distribution of individual LOS values for population of patients: right skewed
  - True mean  $\mu$  = 5.05 days:  $\sigma$  = 6.90 days
  - Results from 500 random samples:

Sample Sizes	Means of 500 Sample Means	SD of 500 Sample Means	Shape of Distribution of 500 Sample Means
n = 20	5.05 days	1.49 days	Approx normal
n = 50	5.04 days	1.00 days	Approx normal
n = 100	5.08 days	0.70 days	Approx normal

Let's review the results



#### Summary

- What did we see across the two examples (BP of men, LOS for teaching hospital patients)?
- A couple of trends:
  - Distributions of sample means tended to be approximately normal even when original, individual level data was not (LOS)
  - Variability in sample mean values decreased as size of sample of each mean based upon increased
  - Distributions of sample means centered at true, population mean

#### Clarification

 Variation in sample mean values tied to size of each sample selected in our exercise: NOT the number of samples

