



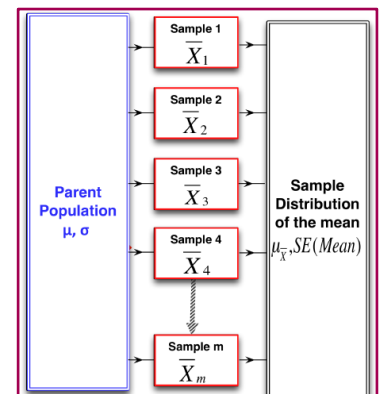
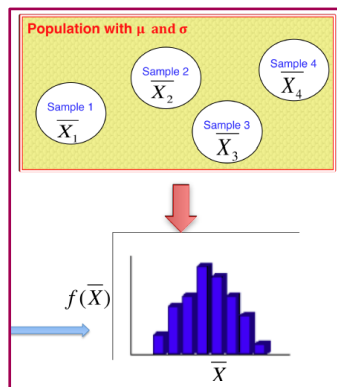
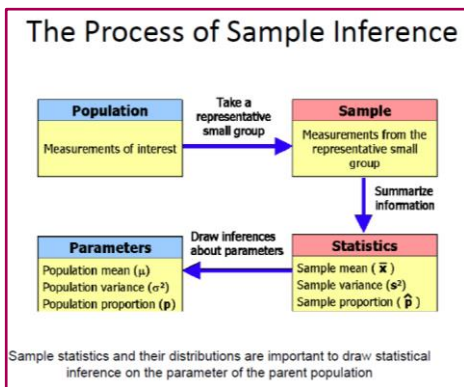
Pharmaceutical statistics

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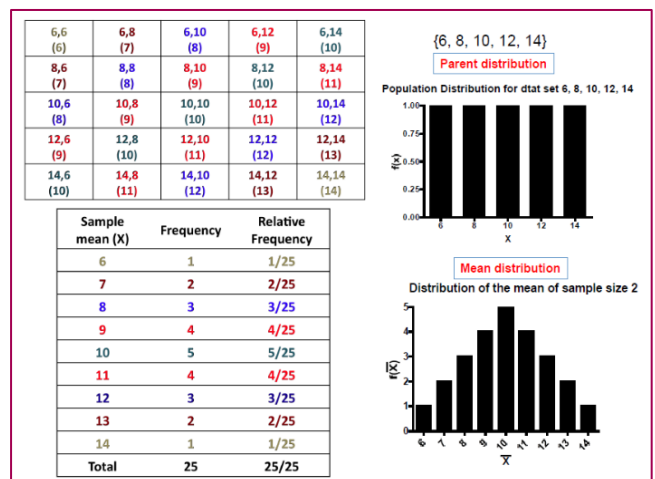
Sampling Distribution of the mean

- **Sampling Distribution of a Sample Statistic:** The distribution of values for a sample statistic obtained from repeated samples, all of the same size and all drawn from the same population.
 - **Population** is all the elements that are included in our study.
 - The sample is a subset that is chosen randomly from that population
 - To make inferences about a population from a repeated number of samples.
 - I will take all possible samples, calculate their mean (\bar{x}), then make distribution graph (\bar{x}) and $f(\bar{x})$ we call it *sampling distribution of the mean*.
 - Samples mean can vary from sample to sample. Since \bar{X} is a number and it can change from each time, it can be thought of as a random variable.
- **How are statistics (means) distributed over many samples?**
 - The sampling distribution of a statistic is the probability distribution for the values of the statistic that results when random samples of size n are repeatedly drawn from the population.
 - Sampling distribution is just a *probability distribution* that involves a statistic.
 - **The standard error** is the standard deviation of the sampling distribution of a statistic and is represented by the symbol SE (mean).

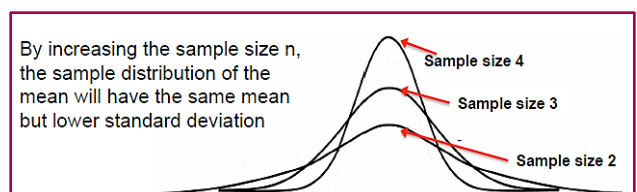


Sampling With Replacement

- For a population set of data: 6, 8, 10, 12, 14
- $\mu = 10, \sigma^2 = 8$
- We want to construct a sampling distribution of the mean using *sampling size of 2*.
- The number of different samples (all possible sample size) is $(N^n) = 5^2 = 25$
- $\mu_{\bar{X}} = \mu = 10$
- $\sigma_{\bar{X}}^2 = \frac{\sigma^2}{n} = 8/2 = 4$
- $\sigma_{\bar{X}} = SE = \sqrt{\frac{\sigma^2}{n}} = \frac{\sigma}{\sqrt{n}} = 2$ Note that: $(\sigma_{\bar{X}} < \sigma)$
- By increasing the sample size n , the sample distribution of the mean will have the same mean but **lower** standard deviation.

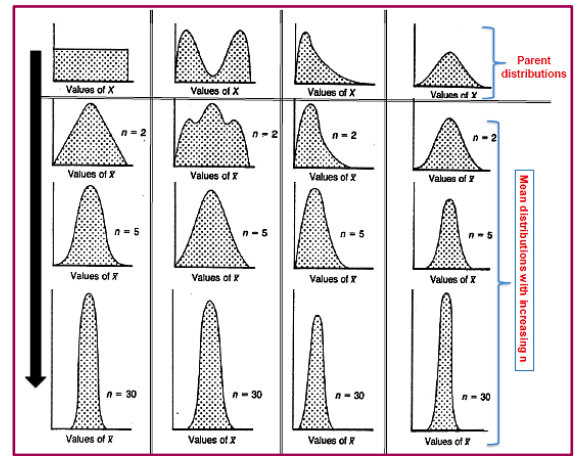


Sample size	# possible samples (N^n)	Mean of samples ($\mu_{\bar{X}}$)	Standard deviation of samples $\sigma_{\bar{X}}$
2	25	10	2
3	125	10	1.63
4	625	10	1.41



• **Sampling from Nonnormally Distributed Populations**

- For the case where sampling is from a *nonnormally distributed population*, we refer to an important mathematical theorem called the *central limit theorem*.
- Given a population of any nonnormal functional form with a mean μ and a variance σ^2 , the sampling distribution computed from a sample size n from this population, will have mean μ and a variance σ^2/n and will be approximately normally distributed when the sample size is large. As sample size is increased the distribution becomes more normal and its standard deviation is decreased.
- If the population is not normal distributed, then to get normal sample distribution you should take a sample Size ≥ 30 .



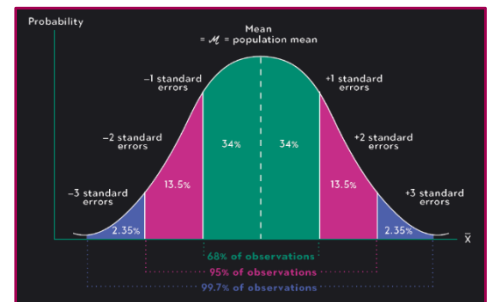
• **Summary**

- The mean of the sampling distribution of has the same value as the mean of the original population μ
- The variance of the sampling distribution of the mean is *not equal* to the population variance, instead it is equal to the population variance divided by the size of the sample used to obtain the sampling distribution.
- The square root of the variance of the sampling distribution is called the *standard error of the mean* or, simply, the standard error. SE(Mean).

• **Why do we want the sampling distribution of the mean to be normal??**

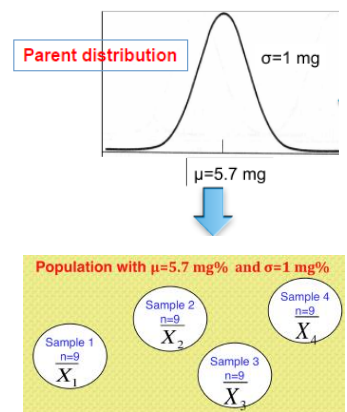
➤ **Easier to Make Inferences:**

- ✓ The normal distribution has well-known properties, such as symmetry and predictable proportions (e.g., 68%, 95%, 99.7% for 1, 2, and 3 standard deviations, respectively).
- ✓ This allows us to calculate probabilities.
- ✓ We can apply all the rules of normal distribution on the sampling distribution of the mean.



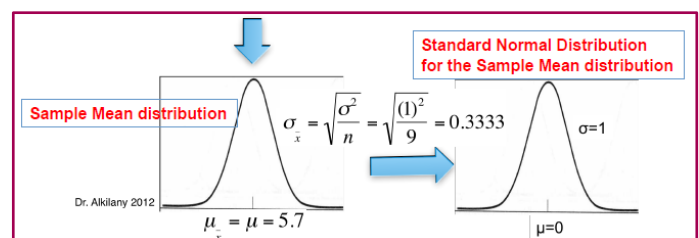
★ **Example 1:**

- ✓ If uric acid values in normal adult males are approximately *normally distributed* with a mean and standard deviation of 5.7 & 1 mg%, respectively.
- ✓ Find the probability that a *sample of size 9* will yield a *mean*:
 - 1) Greater than 6 mg %
 - 2) Between 5 and 6 mg %
 - 3) Less than 5.2 mg%



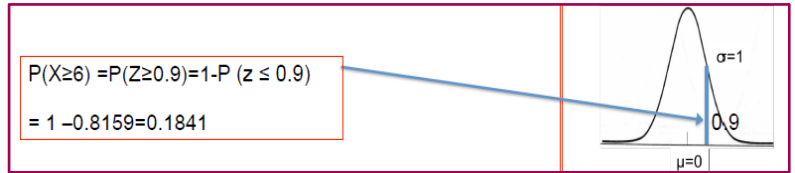
- ✓ The mean and the standard deviation of the sampling distribution (n=9) is: $\mu_{\bar{x}} = \mu = 5.7$
- ✓ Now we deal with the sample mean distribution and we will find the z-score for mean=6 and use the table as we did before.....

✓
$$Z_{\bar{x}} = \frac{\bar{x} - \mu_{\bar{x}}}{\sigma_{\bar{x}}}$$



- ✓ **Q1:** Find the probability that a *sample of size 9* will yield a *mean* greater than 6 mg %

$$Z_{\bar{x}} = \frac{6 - 5.7}{0.3333} = 0.90$$



- ✓ **Q2:** Find the probability that a *sample of size 9* will yield a *mean* between 5 and 6 mg %

$$Z_6 = \frac{6 - 5.7}{0.3333} = 0.90$$

$$Z_5 = \frac{5 - 5.7}{0.3333} = -2.1$$

$$P(-2.1 < z < 0.9) = P(Z < 0.9) - P(Z < -2.1) = 0.8159 - 0.0179 = 0.798$$

- ✓ Find the probability that a *sample of size 9* will yield a *mean* less than 5.2 mg%

$$Z_{5.2} = \frac{5.2 - 5.7}{0.3333} = -1.5$$

$$P(Z < 1.5) = 0.0668$$

★ **Example 2:**

- ✓ If the mean and standard deviation of serum iron values for healthy men are 120 μg/100 ml and 15 μg/100 mL, respectively, what is the probability that a random sample of 50 normal men will yield a mean between 115 and 125 μg/100 mL?

- ✓ $\mu = 120 \mu\text{g}/100 \text{ mL}$

- ✓ $\sigma = 15 \mu\text{g}/100 \text{ mL}$

- ✓ Sample size: $n = 50$

- ✓ $SE = 2.121 \mu\text{g}/100 \text{ mL}$.

- ✓ Find $P(115 < \bar{x} < 125)$:

$$Z_{115} = -2.36$$

$$Z_{125} = 2.36$$

$$P(Z < 2.36) - P(Z < -2.36) = 0.9909 - 0.0091 = 0.9818.$$

★ **Example 3:**

- ✓ Consider a normal population with mean $\mu = 50$ and $\sigma = 15$. Suppose a sample of size 9 is selected at random. $SE = 5$

- ✓ Find:

- ✓ $P(45 \leq x \leq 60)$

$$P(Z \leq 2) - P(Z \leq 1) = 0.97 - 0.16 = 0.81.$$

- ✓ $P(x \leq 47.5)$

$$Z = -0.5$$

$$P(Z \leq -0.5) = 0.3085$$



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