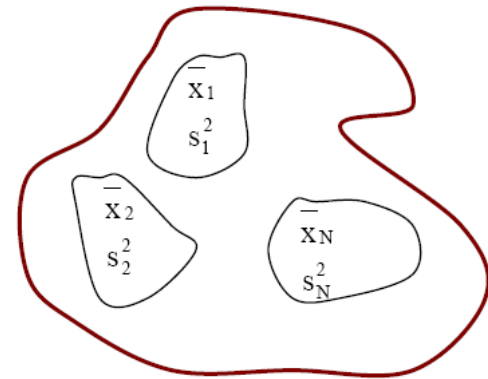


## Chapter 8 (Fundamental Sampling Distributions and Data Descriptions)

### Population with mean $\mu$ and variance $\sigma^2$

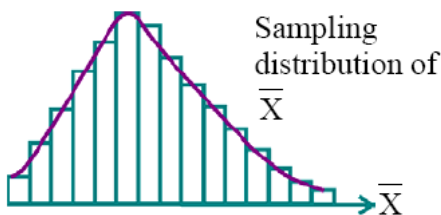


$\bar{X}_1, \bar{X}_2, \dots, \bar{X}_N$  are observed means of samples of size  $n$ , or alternatively, are observations of the random variable  $\bar{X}$

$$\bar{X} = \frac{\sum_{i=1}^n X_i}{n}$$

$s_1^2, s_2^2, \dots, s_N^2$  are the variances of samples of size  $n$ , or alternatively, are observations of the random variable  $S^2$

$$S^2 = \frac{\sum_{i=1}^n (X_i - \bar{X})^2}{n-1} = \frac{n \sum_{i=1}^n X_i^2 - \left( \sum_{i=1}^n X_i \right)^2}{n(n-1)}$$



- ❖ A sample is a subset of a population.
- ❖ A population consists of the totality of the observations with which are concerned.
- ❖ Statistics are measures of a random sample.

### Draw Conclusions About $\mu$

If population of  $X$  is normal, then the sampling distribution of the mean  $\bar{X}$  will be normal and  $\mu = \mu_{\bar{X}}$  and

$$\sigma_{\bar{X}} = \frac{\sigma}{\sqrt{n}}$$

Central Limit Theorem  $\rightarrow$  Regardless of original population distribution, if  $n \geq 30$ , and  $\sigma$  is known, then

$$Z = \frac{\bar{X} - \mu_{\bar{X}}}{\sigma_{\bar{X}}} \text{ follows a standard normal distribution. Also applies to, } Z = \frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{\sqrt{\left(\frac{\sigma_1^2}{n_1}\right) + \left(\frac{\sigma_2^2}{n_2}\right)}} \text{ and}$$

$$\mu_{\bar{X}_1 - \bar{X}_2} = \mu_1 - \mu_2 \quad \text{and} \quad \sigma_{\bar{X}_1 - \bar{X}_2}^2 = \frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}$$

If  $\sigma$  is unknown, then use the **t-distribution** with  $v = n-1$  degrees of freedom and  $T = \frac{\bar{X} - \mu}{S/\sqrt{n}}$

and the original population must be normal.

### Draw Conclusions About $\sigma$

Need to describe the sampling distribution of the variance  $S^2$ . Note that if original population of  $X$  is

normal, then  $\chi^2 = \frac{(n-1)s^2}{\sigma^2}$  follows a **chi-squared distribution** with  $v = n-1$  degrees of freedom.

If you want to compare two sample variances, then use the **f-distribution** where  $F = \frac{(S_1^2/\sigma_1^2)}{(S_2^2/\sigma_2^2)}$  with

$v_1 = n_1 - 1$ ,  $v_2 = n_2 - 1$  and original population must be normal. Note that  $f_{1-\alpha}(v_1, v_2) = 1/[f_{\alpha}(v_2, v_1)]$