

Test Chapter 6 and 7 Formulas

Chapter 6 Formulas

Conversion from x to Z and Z to x with a Normal Distribution

Convert an x value to a Z score

$$z = \frac{(x - \text{mean})}{\text{standard deviation}}$$

$$z = \frac{(x - \mu)}{\sigma} \text{ or } z = \frac{(x - \bar{x})}{s_x}$$

Convert a Z score to an x value

$$x = \text{mean} + z \text{ score (standard deviation)}$$

$$x = \mu_x + (z)(\sigma_x) \text{ or } x = \bar{x} + (z)(s_x)$$

Central Limit Theorem:

Every possible **sample of size n** is taken from a population of values x_1, x_2, x_3, \dots and the average of each sample is recorded. This creates a population of sample means.

This population is called the Distribution of Sample Means and is labeled \bar{X}

The Distribution of Sample Means \bar{X} is Normal
if the **distribution of x values is normal** or **$n > 30$**

$$\mu_{\bar{x}} = \mu_x \text{ and } \sigma_{\bar{x}} = \frac{\sigma_x}{\sqrt{n}}$$

Convert an \bar{x} value to a Z score

$$z = \frac{(\bar{x} - \mu_{\bar{x}})}{\left(\frac{\sigma_x}{\sqrt{n}}\right)}$$

Using a Continuous Normal Distribution to Approximate a Discrete Binomial Probability Distribution

If both $n \cdot p \geq 5$ **AND** $n \cdot q \geq 5$ or if $p = .5$ (where p is the probability of a success in 1 trial , and q is the probability of a failure in 1 trial and n is the numbers of trails) then the Discrete Binomial Probability Distribution can be **approximated** by a Continuous Normal Distribution of x values

with a **mean** of $\mu_x = n \cdot p$ and a **standard deviation** of $\sigma_x = \sqrt{n \cdot p \cdot q}$

Convert an adjusted x value to a Z score

You must use a continuity correction for the x value in the binomial distribution.

$$z = \frac{(\text{adjusted } x - \mu_x)}{\sqrt{n \cdot p \cdot q}}$$

Chapter 7 Formulas

Confidence Intervals

$$\hat{p} - E < p < \hat{p} + E \quad \text{where} \quad E = z_{\alpha/2} \sqrt{\frac{\hat{p}\hat{q}}{n}} \quad \text{If } n \cdot \hat{p} \geq 5 \text{ and } n \cdot \hat{q} \geq 5$$

$$\bar{x} - E < \mu < \bar{x} + E \quad \text{where} \quad E = z_{\alpha/2} \cdot \frac{\sigma}{\sqrt{n}} \quad \text{If the population is normal or } n > 30$$

$$\bar{x} - E < \mu < \bar{x} + E \quad \text{where} \quad E = t_{\alpha/2} \cdot \frac{s_x}{\sqrt{n}} \quad \text{If the population is normal or } n > 30$$

$$\sqrt{\frac{(n-1)(s_x)^2}{\chi_R^2}} < \sigma < \sqrt{\frac{(n-1)(s_x)^2}{\chi_L^2}} \quad \text{If the population is normal}$$

Sample Size for estimating p

$$n = (z_{\alpha/2})^2 \cdot \frac{.25}{E^2}$$

Sample Size for estimating μ_x

$$n = \left[\frac{z_{\alpha/2} \cdot \sigma}{E} \right]^2 \quad \text{if } \sigma \text{ IS known}$$

Sample Size for estimating σ

Sample size n to estimate σ	
To be 95 % confident that the estimate is within	of the value of σ the sample size n should be at least
1 %	19, 204
5 %	767
10 %	191
20 %	47
30 %	20
40 %	11
50 %	7

Sample size n to estimate σ	
To be 99 % confident that the estimate is within	of the value of σ the sample size n should be at least
1 %	33,218
5 %	1,335
10 %	335
20 %	84
30 %	37
40 %	21
50 %	13