STAT 2593

Lecture 030 - The Z Test for Hypotheses about a Population Mean

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The Z Test for Hypotheses about a Population Mean

Learning Objectives

1. Understand how we test population means in (approximately) normally distributed populations.

See my past discussion of introductory statistics education.

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- How do we find our p-value?

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 - Note here we do not take the absolute value.

Rejection Regions for Hypothesis Tests - Critical Values



Two Sided Hypothesis Test – Rejection Region



If the sampling distribution is approximately normally distributed, can use a N(0, 1) to run hypothesis tests.

The rejection region depends on the alternative being considered.