STAT 2593 Lecture 036 - Analysis of Paired Data

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Learning Objectives

1. Construct hypothesis tests and confidence intervals for two sample data, when the data are paired.



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 - \triangleright $X_j \not\perp Y_j$, for the same j.

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 - σ_{12} is the **covariance** between X and Y.

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▶ The mean difference, \overline{D} will have variance

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- Paired data should be used with caution. The results are less conservative.



When the data are paired, we can consider the differences between observations directly.

This results in a single sample, for which all of the previous procedures apply.

The variance of the samples depends on the individual variances as well as the covariance between the two samples.