

# An Overview of Topics for Exam 1

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The work of statisticians is intrinsically tied to data

- 1) We want to be able to *identify* and *describe* important trends
- 2) We want to be able to *explain* those trends

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  - ▶ *quantitative* variables (numeric variables) record a numeric measurement for each case (ie: enrollment, height, etc.)
- ▶ Being able to distinguish different types of variables allows us to determine the proper graphs and summary measures to use!
  - ▶ 1 categorical variable -> summarize with proportions; graph with bar charts
  - ▶ 1 quantitative variable -> summarize with mean, median, standard deviation, Q1, Q3, IQR; graph with histograms or box plots

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  - ▶ Tuition (quantitative) and Average Faculty Salary (quantitative) are associated if higher tuition corresponds with higher average faculty salaries

# Describing Associations

How we describe and display an association is also linked to the types of variables involved:

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  - ▶ Visualize using a *scatter plot*
- ▶ One categorical and one quantitative variable
  - ▶ Summarize by comparing *means, medians, Q1, Q3*, etc. across groups
  - ▶ Visualize using *side-by-side boxplots*

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  - ▶ Random chance
  - ▶ A real causal relationship

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  - ▶ Other biases -> using a **placebo**, **blinding**, and *careful measurement*
  - ▶ Random chance -> we'll explore this in greater detail soon, but think about coin flips in the infant toy choice example
  - ▶ A real causal relationship -> this is what we want!

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- 3) Identify possible associations between variables (either finding them yourself by studying bivariate graphs and descriptive summaries, or identifying explanatory and response variables based upon a pre-defined research question)
- 4) Evaluate the plausibility of these associations (by considering whether the explanations on the prior slide can be ruled out or not)