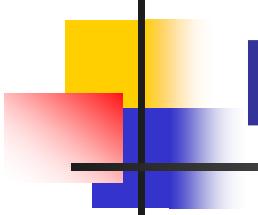


# Research Methods & Experimental Design

16.422 Human Supervisory Control

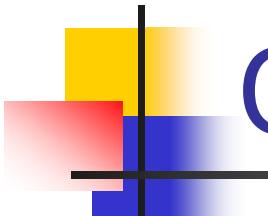
April 2004



# Research Methods

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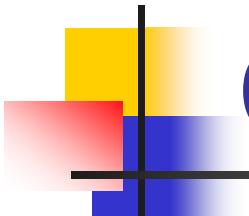
- Qualitative vs. quantitative
- Understanding the relationship between objectives (research question) and variables is critical
- Information  $\neq$  Data
  - Information = data + analysis
- Planning in advance is a must
  - To include how data will be analyzed



# Qualitative Research Methods

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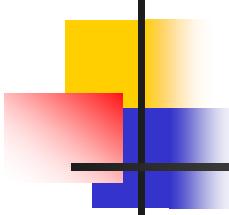
- Social & cultural phenomenon
- Case studies
- Focus groups
- Observations
- Usability testing
  - Can be quantitative
- Interviews
- Questionnaires



# Quantitative Research Methods

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- Natural phenomenon
- Mathematical modeling
- Experiments
- Optimization
- Game theory
- Surveys
- Bottom line – statistics are a must

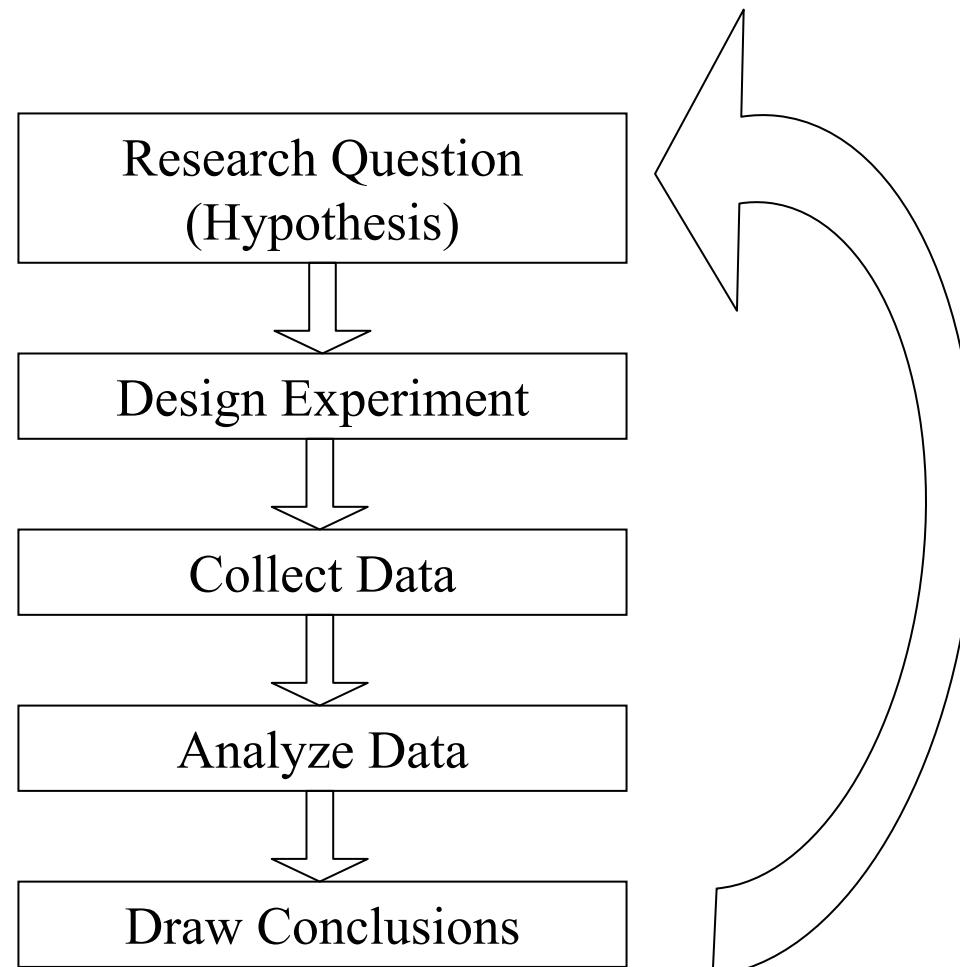


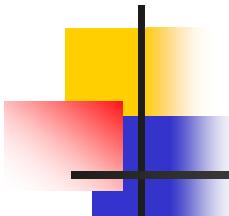
# Project Assignment

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- Design and conduct an experiment in which you explore some measure of human performance through testing, analyze the results, and discuss the broader implications.
- Design an actual display that uses automation for decision support... While formal experimental testing is not required, a small group of users should be used to identify problems with the design to include functionality evaluation as well as recommendation for future improvements and systems integration.

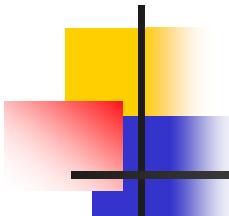
# The Experimental Design Process





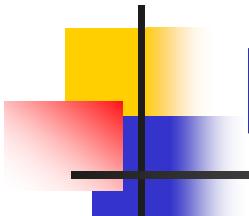
# Experimental Design

- Design of Experiments (DOE) defined:
  - A theory concerning the minimum number of experiments necessary to develop an *empirical* model of a research question and a methodology for setting up the necessary experiments.
  - A parsimony model
- Human subject vs. object experimentation
- Other DOE Constraints
  - Time
  - Money



# Experimental Design Basics

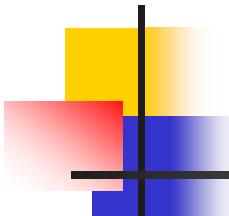
- Two kinds of data gathering methodologies
  - Observation
    - Can't prove cause & effect but can establish associations.
    - Hawthorne effect, social facilitation
  - Experimental
    - Cause & effect
    - Variables of interest – factors vs. treatments
      - Independent variable
        - Treatment – manipulations of variables of interest
        - Treatment vs. control group
      - Dependent variable is what you are measuring



# More Basics

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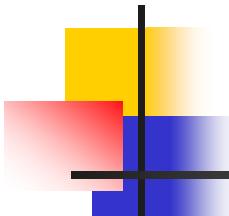
- Confounds
- Randomization Concerns
  - Randomization prevents experimental bias
    - Assignment by experimenter
      - Counterbalancing
    - Statistical assumptions
      - A requirement for statistical tests of significance
  - Why would you use the observation methodology instead of experiments?



# DOE Terminology

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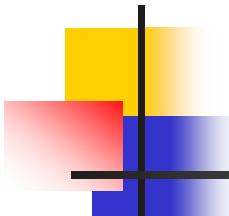
- Replications
  - Independent observations of a single treatment.
- Variance
  - The measuring stick that compares different treatments.
- Internal validity
  - The extent to which an experiment accomplishes its goal(s).
- Reproducibility
  - Given the appropriate information, the ability of others to replicate the experiment.



# DOE Terminology (cont.)

- External validity
  - How representative of the target population is the sample?
    - Can the results be generalized?
    - Generalizations for field experiments are easier to justify than lab experiments because of artificialities.
- Medical Trials
  - Placebo
  - Double Blind

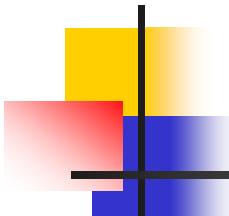
- If so, what is the population to which it can be generalized?
- Can the results be generalized to the real world?



# Data Analysis

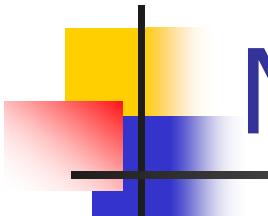
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- Data Types
  - Variables
    - Categorical
    - Numerical
  - Scales of Measurement
    - Nominal
    - Ordinal
    - Interval
- Computer Programs
  - Excel, SAS, S+, SPSS



# Basic Statistical Tests

- Assumptions for comparison of means
  - Independent & random
  - Normality
  - Variances roughly equal
- t-tests
  - One or two samples
- Chi-square tests
  - NID(0,1)
  - Categorical data, non-parametric



# Null Hypothesis: $H_0$

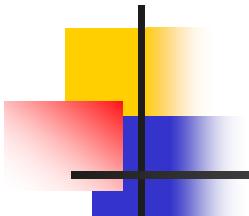
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- Defined: The difference in two different populations parameters is 0.

$$H_0: \mu_1 = \mu_2$$

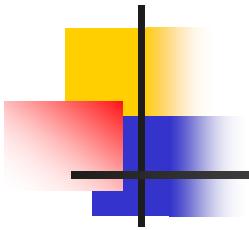
$$H_a: \mu_1 \neq \mu_2$$

- $H_0$ : Always predicts absence of a relationship & assumed to be true.
- If the null hypothesis is NOT rejected, we CANNOT conclude that there is no difference, only that the method did not detect any difference.
- $p < .05$  ????



# A Very Important Research Question

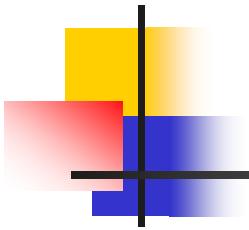
- Does drinking cappuccino one hour before a test improve results?
  - What is the metric (dependent variable)?
- Experimental Design
  - Treatment group vs. control group
    - A single comparison
  - Experimental efficiency
    - Perhaps we want to look at who makes the cappuccino (Seattle's, Starbucks, Pete's) as well as the difference between coffee and cappuccino.
    - 2X3 Factorial
    - Interaction effects



# Caffeine/Performance Experiment

	GB	SB	ER
Capp			
Coffee			

We now know the general layout of the experiment – but what is missing?



# Caffeine/Performance Experiment

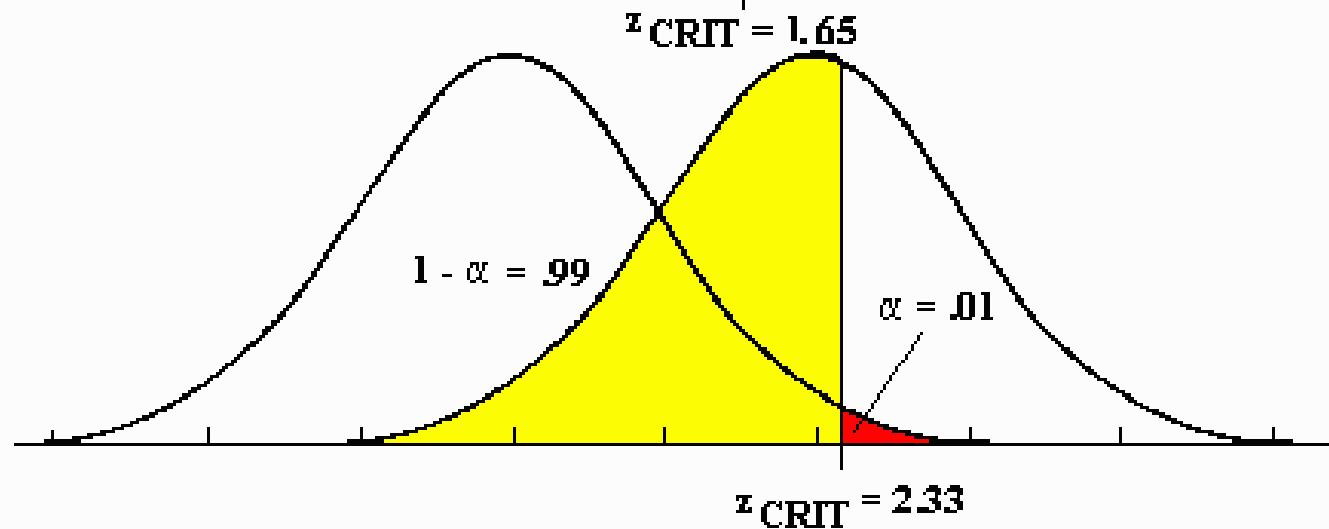
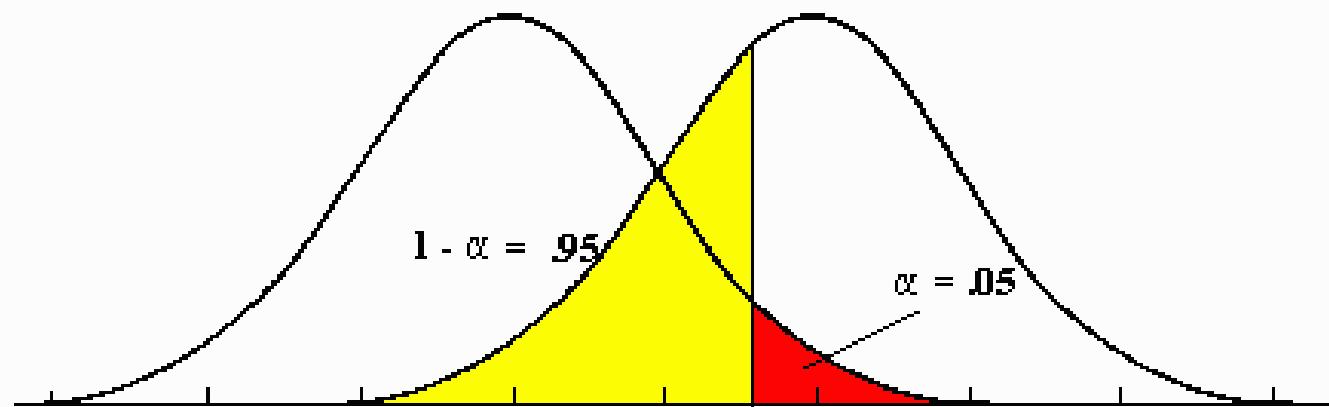
- How many subjects do we need?
  - Sample Size
    - Related to power – the complement of a Type II error...

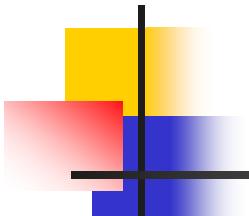
Decision	$H_0$ True	$H_0$ False
Reject $H_0$	Type I error $p = \alpha$	Correct decision $p = 1 - \beta = \text{POWER}$
Fail to reject $H_0$	Correct decision $p = 1 - \alpha$	Type II error $p = \beta$

Ask what  $H_0$  is?

Null hypothesis – no significant difference exists between experimental groups.

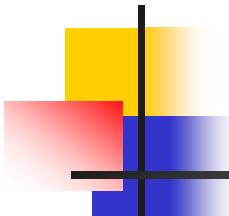
# Don't Panic...





# Caffeine/Performance Experiment

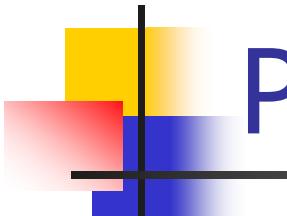
- So how do you determine sample size?
  - <http://members.aol.com/johnp71/javastat.html>
  - Sensitivity is an issue
  - # of factors influences sample size
- Recruitment Issues
  - Population selection
  - How do we assign subjects to treatment categories?
- Confounds
  - Experience
  - Self-selection
  - Control techniques



# Other Subject Considerations

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- What is the most efficient way to use human subjects?
  - Between subjects
  - Within subjects
    - Repeated measures
    - Increases power but...
    - Confounds – practice & fatigue
    - Counterbalance
  - Mixed subjects
    - Pre-test/post-test
    - Tests over time



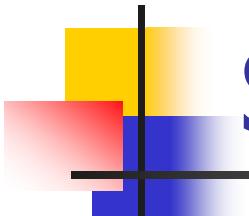
# Pre/post Test Considerations

Between  
Subjects

	Pre-Test	Post-Test
Intervention A		
Intervention B		

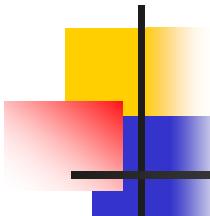
Within Subjects

- Ideally pre-test scores will be equivalent
- You want to see a difference between the experimental and control group.



# Statistical Tests (cont.)

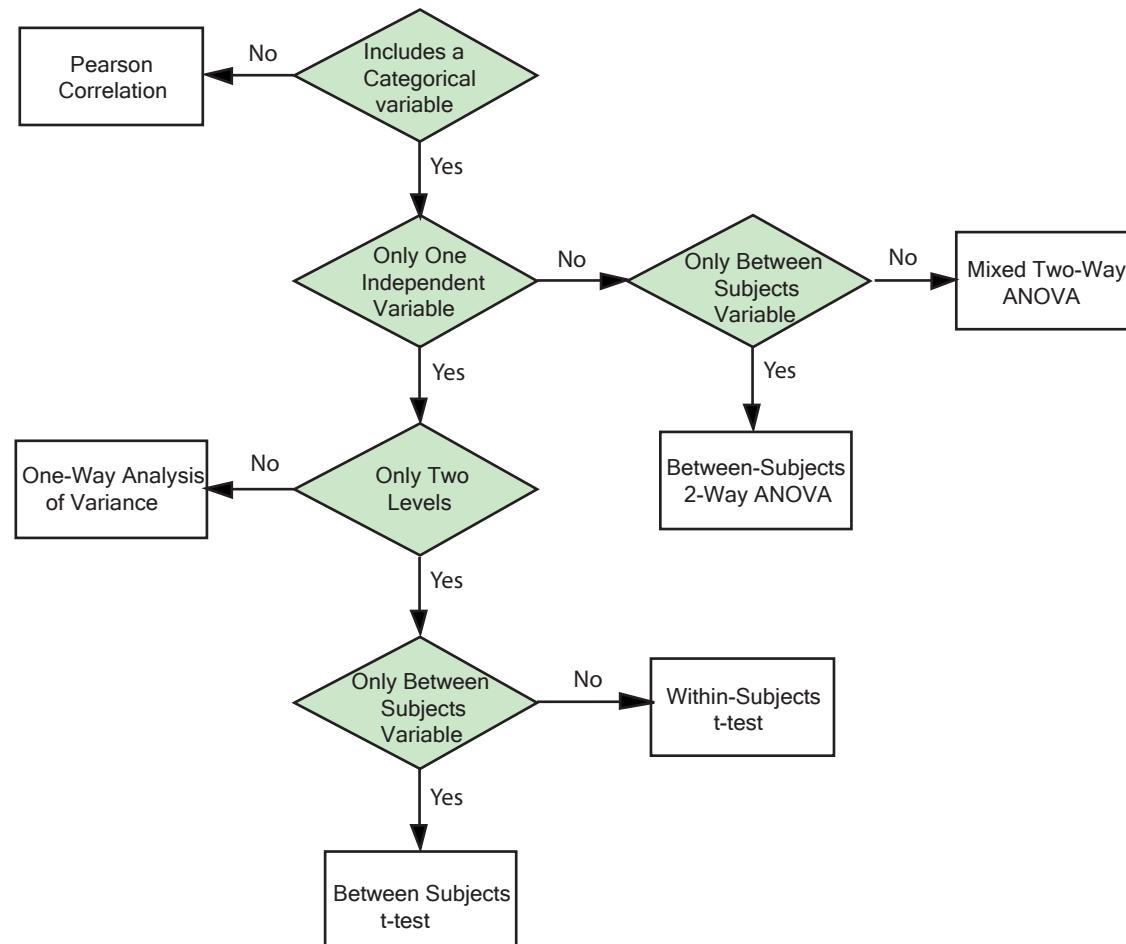
- Analysis of variances (ANOVA)
  - Testing the differences between two or more independent means (or groups) on one dependent measure (either a single or multiple independent variables).
  - One way vs. factorial
  - F test – ratio of variances
  - MANOVA

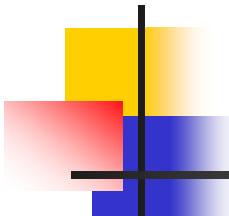


# Other DOE considerations:

- Full Factorial
- Blocking
  - More homogenous grouping
    - Coffee of the day v. another kind
    - Starbuck's at the Marriott vs. Galleria
- Pairing
  - Increases precision by eliminating the variation between experimental units
  - Randomization still possible
- Many others...

# What test to use?

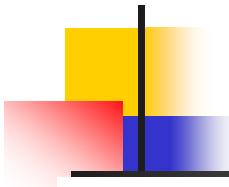




# Example Experiment

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- Are web-based case studies better than print versions.
  - How can we test this?
- This question was tested with 2 classes with 2 different professors.
  - What are the independent & dependent variables?
  - Was it within/between/mixed?
  - What statistical test should we use?



# Results

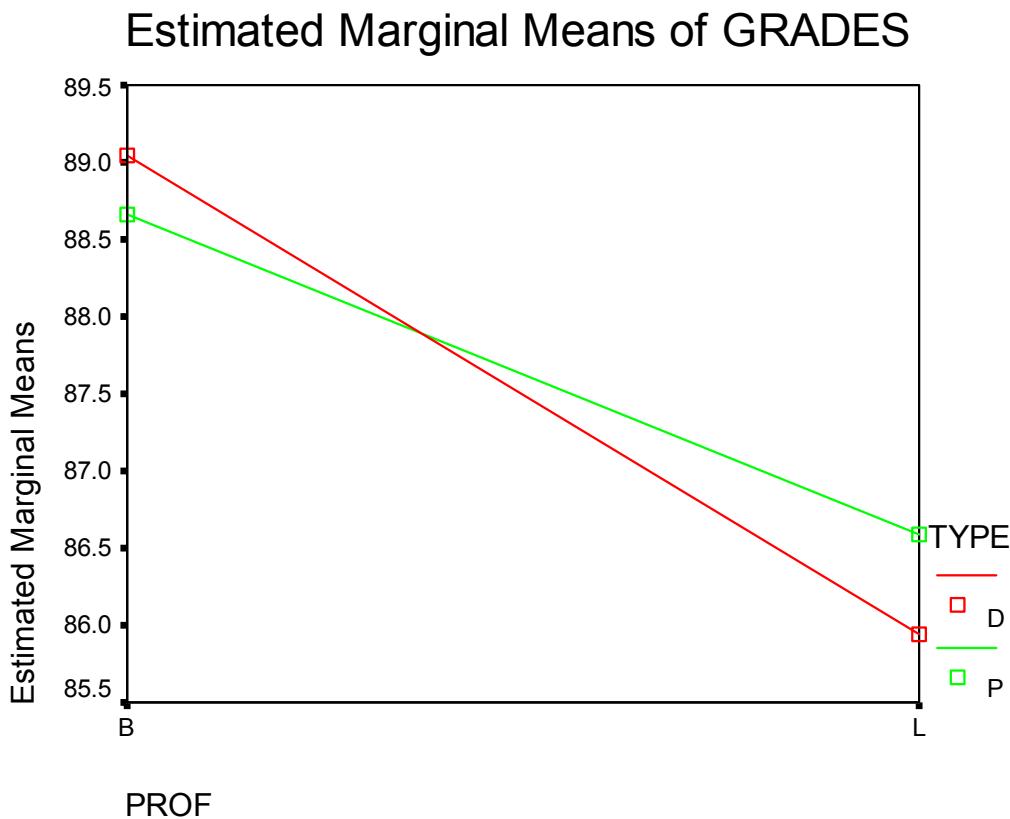
## Tests of Between-Subjects Effects

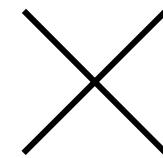
Dependent Variable: GRADES

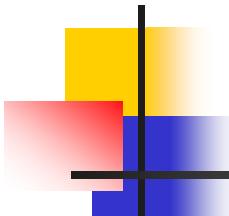
Source	Type III Sum of Squares	df	Mean Square	F	Sig.
Corrected Model	173.681 <sup>a</sup>	4	43.420	.986	.420
Intercept	190832.489	1	190832.489	4333.757	.000
PROF	157.697	1	157.697	3.581	.062
TYPE	26.217	2	13.109	.298	.743
PROF * TYPE	11.840	1	11.840	.269	.605
Error	3654.818	83	44.034		
Total	673001.300	88			
Corrected Total	3828.499	87			

a. R Squared = .045 (Adjusted R Squared = -.001)

# Interactions

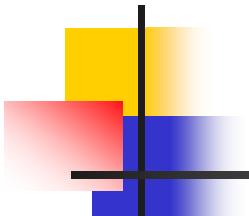


- Interaction effect: the response of one variable depends on effect of another variable
- No interaction – parallel lines
- Significant interaction: 
- Which professor would you rather have?



# Non-Parametric Tests

- Use when you have no good information about an underlying distribution
- Parametric tests:
  - Parametric form - parameters either assumed to be known or estimated from the data
    - The mean and variance of a normal distribution
  - Null hypothesis can be stated in terms of parameters and the test statistic follows a known distribution.
- Non-parametric tests are still hypothesis tests, but they look at the overall distribution instead of a single parameter
- Particularly useful for small samples



# All data is not normal....

## Parametric

- Correlation & Association
  - Pearson
- T-tests
  - Independent & dependent
- ANOVA
  - Factorial
  - Repeated measures
  - MANOVA
- Linear regression

## Non-parametric

- Association
  - Spearman
- Chi-Square
  - Contingency tables
- Kruskal-Wallis test
- Sign-test
- Friedman ANOVA
- Logistic regression