

# Stat 133: Concepts in Computing with Data

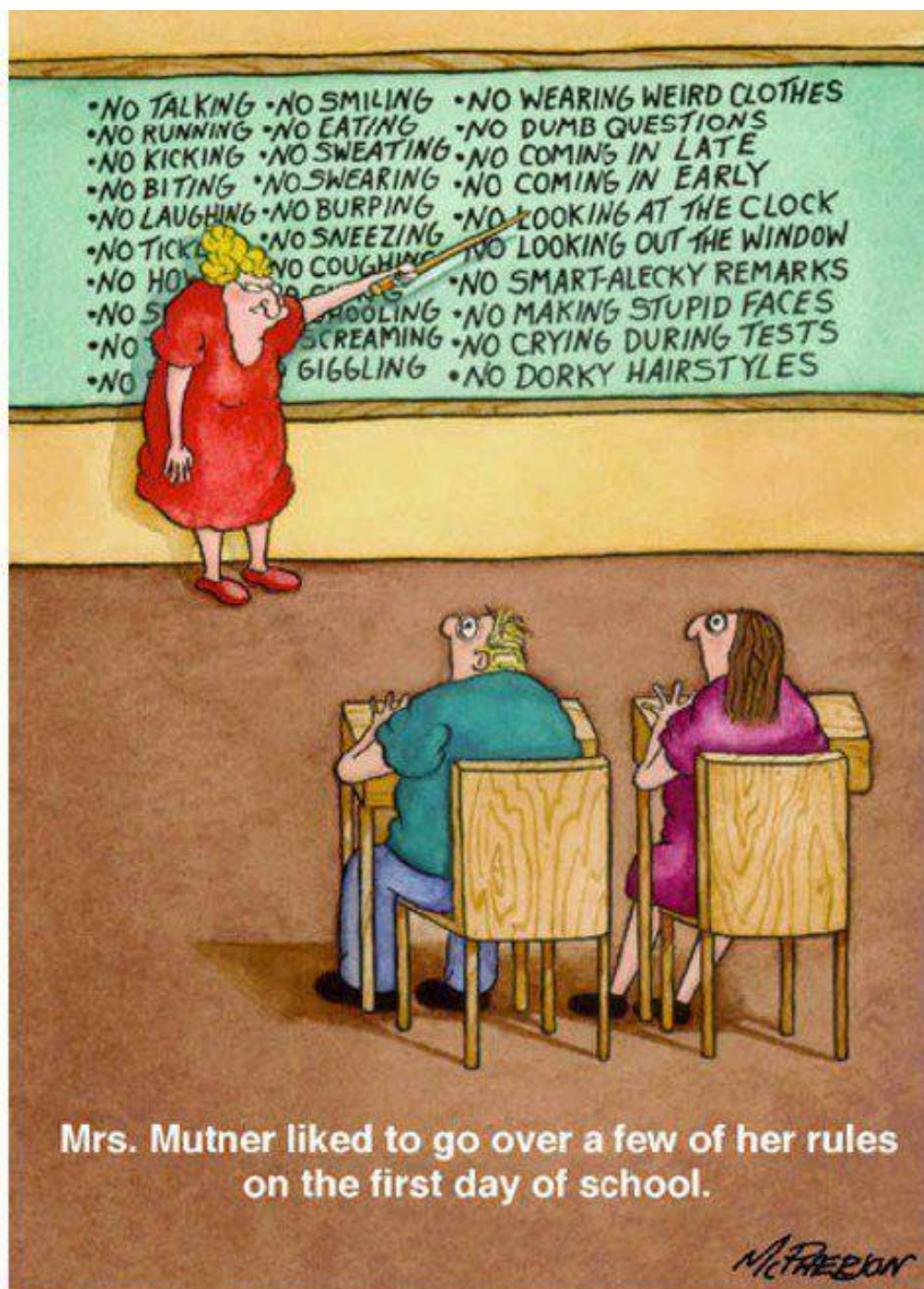
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Stat 133 with Gaston Sanchez

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# Intro survey (google form)





# About Stat 133

# Stat 133

Core Course for Statistics Major

# Stats Major

*Prereqs*

Calculus

Calculus II

Multivariable  
Calculus

Linear  
Algebra

*Core*

**Stat 133  
Computing**

**Stat 134  
Probability**

**Stat 135  
Statistics**

*Elective*

Stat 150  
Stochastic  
Processes

Stat 151A  
Linear  
Modeling

Stat 152  
Sampling  
Surveys

Stat 153  
Times  
Series

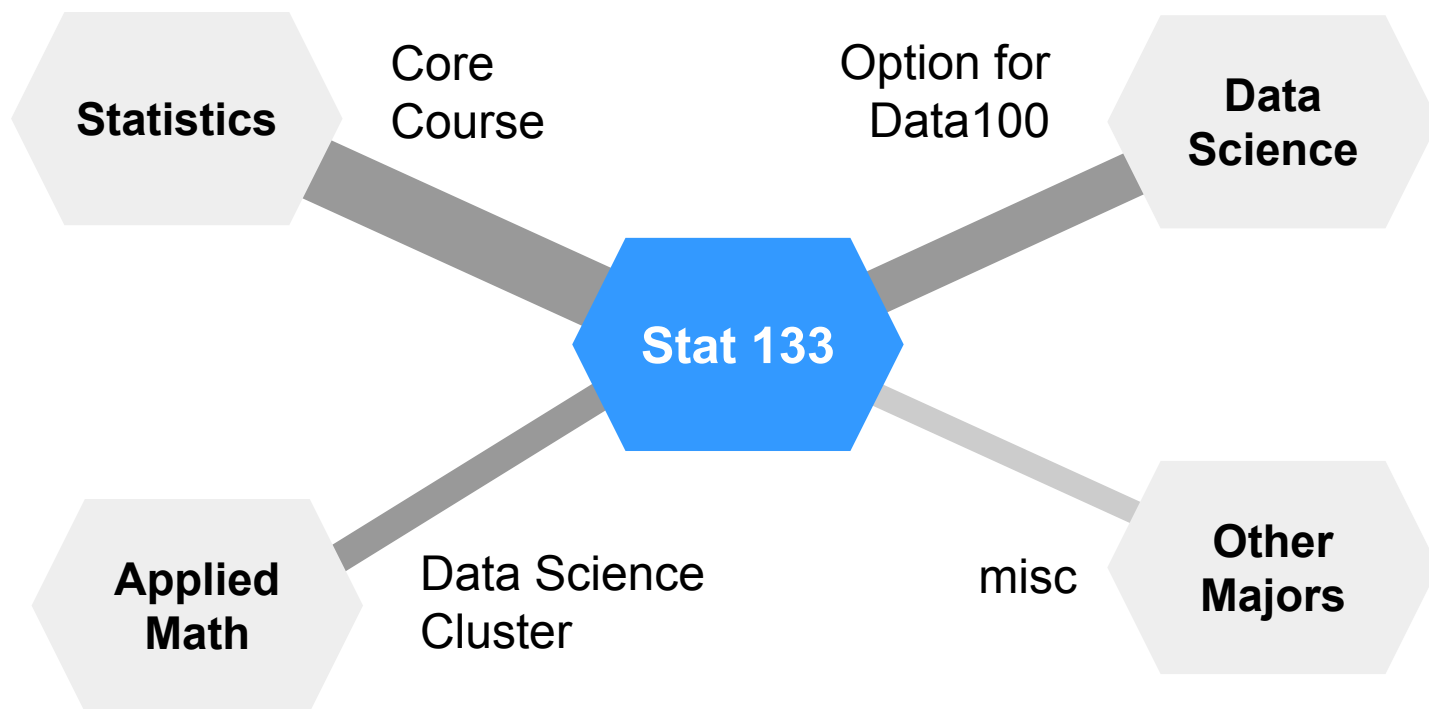
Stat 154  
Predictive  
Modeling

Stat 155  
Game  
Theory

Stat 158  
Design of  
Experiments

Stat 159  
Reproducible  
Research

# Roles for Stat 133



# My Philosophy



# DATA: BY THE NUMBERS



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www.phdcomics.com

<http://www.phdcomics.com/comics/archive.php?comid=462>



## Data Preparation

- Acquisition
- Storage
- Cleaning
- Processing
- Tidying
- Reshaping
- Wrangling



## Analysis

- Exploration
- Description
- Visualization
- Hypothesis Tests
- Inference
- Simulation
- Model Fitting



## Reports

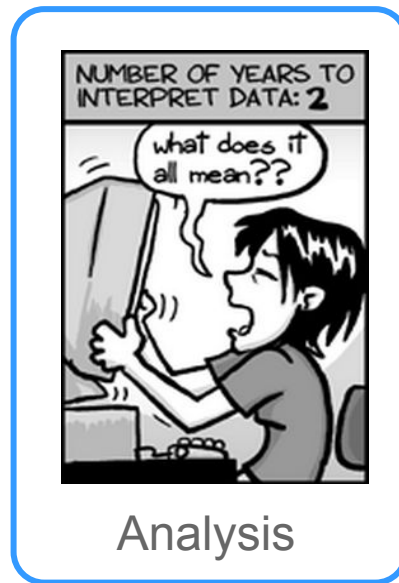
- Document(s)
- Article(s)
- Book(s)
- Poster(s)
- Blog post(s)
- Dissertation
- News



## Communication

- Oral
- Print
- Web
- Audio
- Video
- Multimedia
- Other

# Traditionally ...



Traditionally, this is where most teaching focuses on

However ...

**(ALMOST) NO ONE TEACHES THIS!**



Data



Analysis



Report



Communication

**In practice these are where we spend most of our time**

# Course Content



# Course cornerstones

Data  
Manipulation

Data  
Visualization

Reporting  
Tools

Programming  
Concepts

Data  
Technologies

**R**  
& other tools

# Data Tables

1. Data Tables
2. Selecting and Filtering
3. Reshaping
4. Aggregation & Group by operations
5. Joins and Merges

# Taking Care of Data

1. Storing Tables (files & formats)
2. Data Dictionary (metadata)
3. Data Organization
4. Cleaning
5. Data Tidying

# Data Visualization

1. Visualization basics
2. Colors
3. Design and Aesthetics considerations
4. Efficient displays
5. Good and bad practices

# Programming Concepts

1. Emphasis on **data analysis**
2. Data types and data structures
3. Control flow structures
4. Functions
5. Regular Expressions

# Reporting Tools

1. Markdown syntax
2. LaTeX (mostly equations)
3. Dynamic Documents
4. Shiny Apps
5. Writing reports

## R and other tools

1. R
2. RStudio
3. Command Line (Bash)
4. Unix filters & utilities

# Instruction

In-person instruction

Lecture: more conceptual/theory

Lab: practice



## Website & bCourses

### **Units:** weekly topics

- Slides, readings, cheatsheets, files
- Lab materials
- Assignments
- Submissions

# Grading Structure

8% Lab work (weekly; drop 2 lowest)

35% HW (6 assignments; drop lowest)

27% Apps (3 shiny apps; no drops)

8% Midterm

22% Final exam

# Enrollment

Waitlist

Concurrent-enrollment

# Some Comments

## Remarks

Very hands-on course

Expect to do A LOT OF WORK outside class

Deceptively simple

Very easy to fall behind

## Course Format

**Lecture:** conceptual stuff, demos, case studies, examples, review some code

**Lab:** practical work using R, command line, git

**Homework:** follow the work of labs, plus some challenges

## My Expectations

Don't expect that you'll become a data scientist  
(that takes years of hard work)

Instead: give you solid foundations about data  
analysis

Expose you to different “data technologies”

## Ultimate Goals

Understand different types of data (e.g. files, forms, formats)

Know how to access information stored in different formats

Know how to do data manipulation and processing in R

Be better prepared to crunch data



Becoming a data scientist is  
a (yearslong) **marathon** ...  
not a (one semester) sprint

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