

MATH 205: Statistical methods

Vu Dinh

Departments of Mathematical Sciences
University of Delaware

September 1st, 2021

- Lectures:

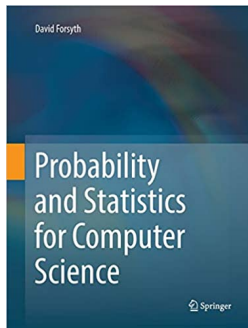
MW 3:35pm-4:50pm, Kirkbride 205

- Labs:

- Section 050L: M 2:30pm - 3:20pm, Ewing 101
- Section 051L: W 2:30pm - 3:20pm, Ewing 101

- Office hours

- TTh 2:00pm - 3:30pm, Ewing 312
- or by appointments



Lectures:

Probability and Statistics for Computer Science.

David Forsyth (2018)

Labs:

simpleR – Using R for Introductory Statistics.

John Verzani (2002)

The safety of our learning environment

We will adhere to the practice of wearing face masks and cleaning your seat and desk area at the beginning of class:

- Must wear a cloth mask that covers your nose and mouth
- Must not eat or drink in class
- Upon entering the classroom, wipe down your seat and desk area

MATH205-Fa... Search MATH205-Fall2021

general 54

+ Add a bookmark

Saturday, August 28th

Vu Dinh 7:21 PM

Important infos about the class:

1. Course webpage: Information about the class, including the syllabus and infos about the midterm, can be found on the course webpage:
<https://vucdinh.github.io/m205f21.html>

Note that all handouts (homework assignments, lectures) will be posted at this webpage. Canvas will only be used to access your grades (and is set up to prepare for the scenarios when the course has to be moved online).

2. The references for this class are:
 - **Lectures:**
Probability and Statistics for Computer Science. David Forsyth (2018)
 - **Labs:**
simpleR – *Using R for Introductory Statistics*. John Verzani (2002) (edited)
Probability and Statistics for Computer Science. David Forsyth (2018)

PDF

Probability and Statistics for Computer Science.pdf
6 MB PDF

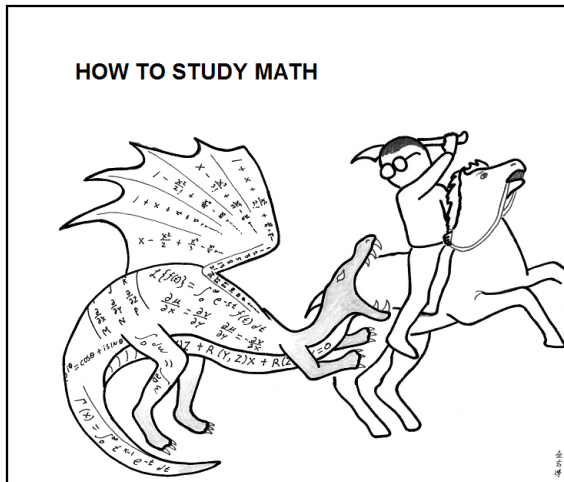
David Forsyth

Other classroom settings

- The lectures will be recorded by UD Capture, accessible through Canvas.
Note that there will be no camera in class, so work on the board wouldn't be seen in the records.
- The lab doesn't have UD Capture. I will provide a Zoom session for each lab.

Evaluation

- Overall scores will be computed as follows:
25% homework, 10% quizzes, 25% midterm, 40% final
- No letter grades will be given for homework, midterm, or final. Your letter grade for the course will be based on your overall score.
- The lowest homework scores and the lowest quiz score will be dropped.
- Letter grades you can achieve according to your overall score.
 - $\geq 90\%$: At least A
 - $\geq 75\%$: At least B
 - $\geq 60\%$: At least C
 - $\geq 50\%$: At least D



Don't just read it; fight it!

— Paul R. Halmos

Homework

- There are 5 homework assignments throughout the semester
- Assignments will be posted on Monday (starting from the third week) and will be due on Wednesday of *the following week, at the beginning of lecture*.
- No late homework will be accepted.
- Your lowest homework scores will be dropped in the calculation of your overall homework grade.

- At the end of some chapter, there will be a short quiz during class.
- The quiz dates will be announced at least one class in advance.
- The lowest quiz score will be dropped.

- There will be an in-class midterm exam during the week of October 25-27. The exam consists of two parts: a written exam during the Oct 27 lecture, and the computational exam during the lab sessions of that week.
- Final exam (written) during the final week.

Open source statistical system R

<http://cran.r-project.org/>

Tentative schedule

Date	Theme/Topic	Labs	Assignments
Sep 1	Syllabus		
Sep 8	Chapter 1: Describing dataset	Section 2: Handling data	
Sep 13 - 15	Chapter 2: Looking at Relationships	Section 3: Univariate data	
Sep 20-22	Chapter 3: Basic Ideas in Probability	Section 4: Bivariate Data	Homework 1 (due 09/22)
Sep 27-29	Chapters 3-4	Section 4: Correlation	
Oct 4-6	Chapter 4: Random variables and expectations	Section 6: Random data	Homework 2 (due 10/06)
Oct 11-13	Chapter 5: Useful distributions	Section 7: The central limit theorem	
Oct 18-20	Chapter 6: Samples and populations	Section 9: Confidence interval estimation	Homework 3 (due 10/20)
Oct 25-27	Review and midterm exam		Midterm: Oct 27 (lecture), Oct 25-27 (labs)
Nov 1-3	Chapter 7: The significance of evidence	Section 10: Hypothesis testing	
Nov 8-10	Goodness of Fit	Section 12: Goodness of Fit	Homework 4 (due 11/10)
Nov 15-17	Linear Regression	Section 13: Linear regression	
Nov 22-24	Thanksgiving break		
Nov 29 - Dec 1	One-Way Analysis of Variance	Section 15: Analysis of variance	Homework 5 (due 12/01)
Dec 6-8	Selected topics + Review		
Exam week			

Chapter 1: Describing dataset

Categorical and continuous data

Statistics deal with the collection, organization, analysis, interpretation and presentation of data:

- Categorical:
 - data that records categories
 - each data item can take a (typically small) set of prescribed values
 - example: students' majors or programs
- Continuous:
 - can receive any value in a particular range
 - example: height or weight or body temperature

Dataset as d -tuples

- A d -tuple is an ordered list of d elements
- We think of a dataset as a collection of d -tuples
- Example:
A dataset has entries for ID, Email, Name, Audit, Units, Program and Plan, Level, Grade, Weight for 55 students
→ $d = 9$, $N = 55$.

...but this class is about $d = 1$ and $d = 2$

- Chapter 1: Looking at 1D data
- Chapter 2: Looking at 2D data
- Confidence interval, hypothesis testing, goodness of fit: analyzing 1D data
- Linear regression: analyzing 2D data

Chapter 1: Describing univariate data

Summarizing univariate data:

- Mean
- Median
- Standard deviation
- Interquartile Range

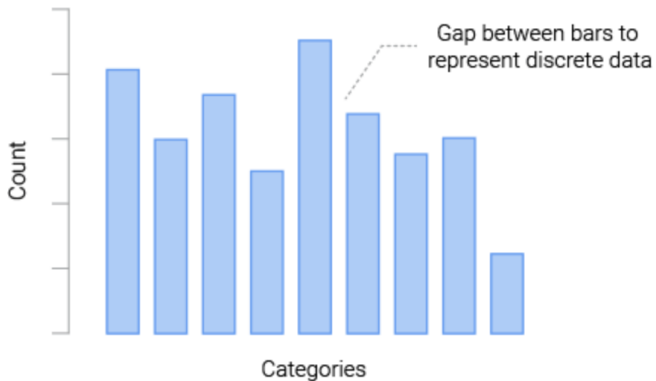
Visualizing univariate data:

- Bar chart
- Pie chart
- Histogram
- Box plot

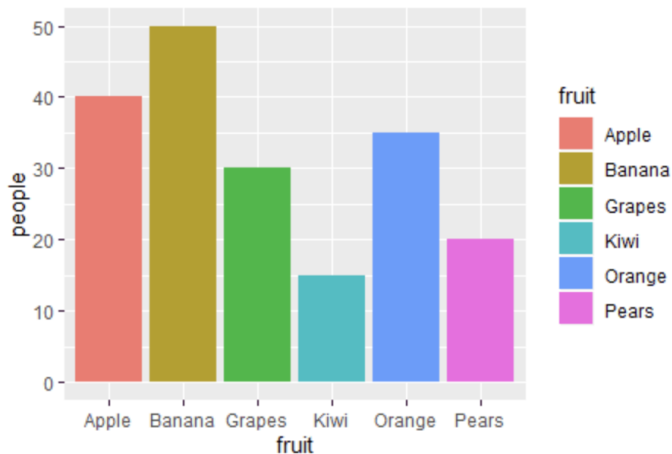
Categorical data: bar charts

- A bar chart is a set of bars, one per category
- the height of each bar is proportional to the number of items in that category
- the height could be given by the frequency, or the proportion

Bar charts



Example: People's favorite fruit in a survey

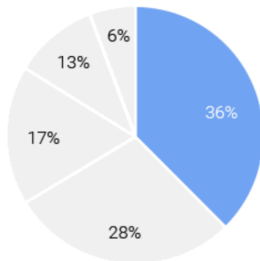


Categorical data: pie charts

- each slice of the pie corresponds to one category
- the area of the slice is proportional to the number of items in that category

Bar charts

A Pie Chart is a special chart that shows relative sizes of data using pie slices.

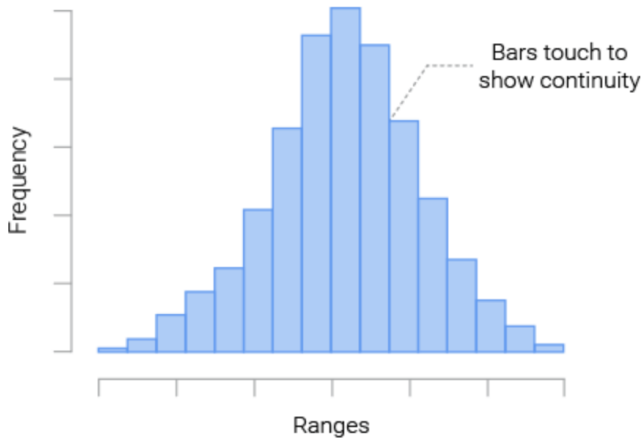


They are good if you are trying to compare parts of a single data series to the whole.

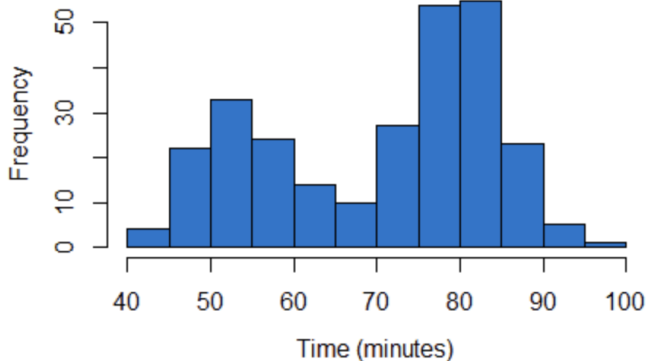
Continuous data: histograms

- a simple generalization of a bar chart
- We divide the range of the data into intervals, which do not need to be equal in length
- We then build a set of boxes, one per interval. Each box sits on its interval on the horizontal axis.
- The area of the box is proportional to the number of elements in the box.

Histograms



Time between eruptions of Old Faithful



Summarizing univariate data

- Mean
- Median
- Standard deviation
- Variance
- Interquartile Range

Definition 1.1 (Mean) Assume we have a dataset $\{x\}$ of N data items, x_1, \dots, x_N . Their mean is

$$\text{mean}(\{x\}) = \frac{1}{N} \sum_{i=1}^{i=N} x_i.$$

Properties of the Mean

Useful Facts 1.1 (Properties of the Mean)

- Scaling data scales the mean: or

$$\text{mean}(\{kx_i\}) = k \text{mean}(\{x_i\}).$$

- Translating data translates the mean: or

$$\text{mean}(\{x_i + c\}) = \text{mean}(\{x_i\}) + c.$$

- The sum of signed differences from the mean is zero: or,

$$\sum_{i=1}^N (x_i - \text{mean}(\{x_i\})) = 0.$$

Definition 1.4 (Median) The median of a set of data points is obtained by sorting the data points, and finding the point halfway along the list. If the list is of even length, it's usual to average the two numbers on either side of the middle. We write

$$\text{median}(\{x\})$$

for the operator that returns the median.

Median is not affected by outliers

Median

The risk of developing iron deficiency is especially high during pregnancy. The problem with detecting such deficiency is that some methods for determining iron status can be affected by the state of pregnancy itself. Consider the following data on transferrin receptor concentration for a sample of women with laboratory evidence of overt iron-deficiency anemia (“Serum Transferrin Receptor for the Detection of Iron Deficiency in Pregnancy,” *Amer. J. Clin. Nutr.*, 1991: 1077–1081):

$$\begin{array}{cccccc} x_1 = 15.2 & x_2 = 9.3 & x_3 = 7.6 & x_4 = 11.9 & x_5 = 10.4 & x_6 = 9.7 \\ x_7 = 20.4 & x_8 = 9.4 & x_9 = 11.5 & x_{10} = 16.2 & x_{11} = 9.4 & x_{12} = 8.3 \end{array}$$

The list of ordered values is

$$7.6 \quad 8.3 \quad 9.3 \quad 9.4 \quad 9.4 \quad 9.7 \quad 10.4 \quad 11.5 \quad 11.9 \quad 15.2 \quad 16.2 \quad 20.4$$

Since $n = 12$ is even, we average the $n/2 =$ sixth- and seventh-ordered values:

$$\text{sample median} = \frac{9.7 + 10.4}{2} = 10.05$$

Measures of variability: deviation from the mean

Definition 1.2 (Standard Deviation) Assume we have a dataset $\{x\}$ of N data items, x_1, \dots, x_N . The standard deviation of this dataset is:

$$\begin{aligned}\text{std}(\{x_i\}) &= \sqrt{\frac{1}{N} \sum_{i=1}^{i=N} (x_i - \text{mean}(\{x\}))^2} \\ &= \sqrt{\text{mean}(\{(x_i - \text{mean}(\{x\}))^2\})}.\end{aligned}$$

Properties of the standard deviation

Useful Facts 1.2 (Properties of Standard Deviation)

- Translating data does not change the standard deviation, i.e. $\text{std}(\{x_i + c\}) = \text{std}(\{x_i\})$.
- Scaling data scales the standard deviation, i.e. $\text{std}(\{kx_i\}) = k\text{std}(\{x_i\})$.
- For any dataset, there can be only a few items that are many standard deviations away from the mean. For N data items, x_i , whose standard deviation is σ , there are at most $\frac{1}{k^2}$ data points lying k or more standard deviations away from the mean.
- For any dataset, there must be at least one data item that is at least one standard deviation away from the mean, that is, $(\text{std}(\{x\}))^2 \leq \max_i (x_i - \text{mean}(\{x\}))^2$.

The standard deviation is often referred to as a scale parameter; it tells you how broadly the data spreads about the mean.

Definition 1.3 (Variance) Assume we have a dataset $\{x\}$ of N data items, x_1, \dots, x_N , where $N > 1$. Their variance is:

$$\begin{aligned}\text{var}(\{x\}) &= \frac{1}{N} \left(\sum_{i=1}^{i=N} (x_i - \text{mean}(\{x\}))^2 \right) \\ &= \text{mean}(\{(x_i - \text{mean}(\{x\}))^2\}).\end{aligned}$$