

AMS 553.430 - Introduction to Statistics

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Fall 2018

Lecture 0 (2018-08-30) **1**

Introduction

Math 553.430 is one of the most important courses that is required/recommended for the engineering-based majors at Johns Hopkins University.

These notes are being live-Texed, through I edot for Typos and add diagrams requiring the *TikZ* package separately. I am using Texpad on Mac OS X.

I would like to thank Zev Chonoles from The University of Chicago and Max Wang from Harvard University for providing me with the inspiration to start live-Texing my notes. They also provided me the starting template for this, which can be found on their personal websites.

Please email any corrections or suggestions to ksriniv4@jhu.edu.

Lecture 0 (2018-08-30)

Introduction to Probability (553.420) Review

Part 1 - Counting

- ① Multiplication rule (Basic Counting Principle)
- ② Combinations/Permutations
 - Sampling with or without replacement. \Rightarrow Inclusion-Exclusion Principle
- ③ Birthday Problem
- ④ Matching Problem (inclusion-exclusion principle)
- ⑤ n balls going into m boxes (all are distinguishable)
- ⑥ Multinomial Coefficients e.g. assign A, B, C, D, to different students \rightarrow anagram problem
- ⑦ Pairing Problem

$$2n \text{ people, paired up } \begin{cases} \text{ordered: } \binom{2n}{2,2,\dots,2} & \text{e.g. different courts for players} \\ \text{unordered: } \frac{\binom{2n}{2,2,\dots,2}}{n!} \end{cases}$$

- ⑧ Partition of integers $\rightarrow \binom{n}{n+k-1}$ where n is the sum of integer and k is the number of partitions

Basics of Probability

Axioms

- ① $0 \leq P(A) \leq 1, \forall A$
- ② $P(\Omega) = 1 \rightarrow$ where Ω is the sample space
- ③ Countable additivity
 - if A_1, \dots, A_n are mutually exclusive, then

$$P\left(\bigcup_{i=1}^{\infty} A_i\right) = P(A_1) + P(A_2) + \dots = \sum_{i=1}^{\infty} P(A_i)$$

$$\Rightarrow P(A) = 1 - P(A^c)$$

$$P(A) = \frac{|A|}{|\Omega|}$$

Conditional Probability

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

Bayes Rule

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{P(B|A)P(A)}{\sum_j P(B|C_j)P(C_j)} \quad \underbrace{\bigcup_j C_j = \Omega}_{\text{partition of } \Omega}$$

Law of Total Probability

$$P(A) = \sum_j P(A|B_j)P(B_j) = \sum P(A \cap B_j) \quad \underbrace{\bigcup_j B_j = \Omega}_{\text{partition of } \Omega}$$

Part 2 - Discrete and Continuous Random Variables

Function	Discrete	Continuous
Probability Function	PMF: $P(X = x)$	PDF: $f_x(x)$
Probability Distribution	$\sum_x P(X = x) = 1$	$\int_x f_x(x)dx = 1$
Expectation	$E[X] = \sum_x xP(X = x)$	$E[X] = \int_x xf(x)dx$
Variance	$Var[X] = E[X^2] - (E[X])^2$	$Var[X] = E[X^2] - (E[X])^2$

Law of the Unconscious Statistician (LOTUS)

$$\text{1-dim} \quad E[g(x)] = \sum_x g(x)P(X = x) \quad \Bigg/ \quad E[g(x)] = \int_x g(x)f(x)dx$$

$$\text{2-dim} \quad E[g(X, Y)] = \sum_y \sum_x g(x, y)P(X = x, Y = y) \quad \Bigg/ \quad E[g(X, Y)] = \int_y \int_x g(x, y)f(x, y)dxdy$$

Special notes on Discrete Distributions

Binomial Distribution

A sum of i.i.d. (identical, independent distribution) Bernoulli(p) R.V.

- Approximation method \Rightarrow if n is large, p very small and $np < 10$.
 - use Poisson (np), otherwise preferably $p \approx \frac{1}{2}$
 - use Normal ($np, np(1 - p)$)
- Mode:
 - if $(n + 1)p$ integer, mode = $(n+1)p$ or $(n+1)p - 1$.
 - if $(n + 1)p \notin \mathbb{Z}$ mode is $\lfloor (n + 1)p \rfloor$
 - **Proof:** consider $\frac{P(X = x)}{P(X = x - 1)}$ going below 1.

Negative Binomial

A sum of i.i.d Geometric(p) R.V.

■ a^{th} head before b^{th} tail

Example 1. A coin has probability p to land on a head, $q = 1 - p$ to land on a tail.
 $P[5^{th} \text{tail occurs before the } 10^{th} \text{ head}]?$