

Lesson Plan 4/3

Tuesday, April 3, 2018 12:11 PM



Admin

- Project 2 due on Friday
- HW4 released (due on Tuesday)
- Final project posted
- Grades for Project 1 and HW3 almost done (looking good!)

HDSC:

- Mobility charts from The Upshot: <https://www.nytimes.com/interactive/2018/03/27/upshot/make-your-own-mobility-animation.html?rref=collection%2Fsectioncollection%2Fupshot&action=click&contentCollection=upshot®ion=rank&module=package&version=highlights&contentPlacement=7&pgtype=sectionfront>

Project 2 Questions?

Final Project Description

HW4

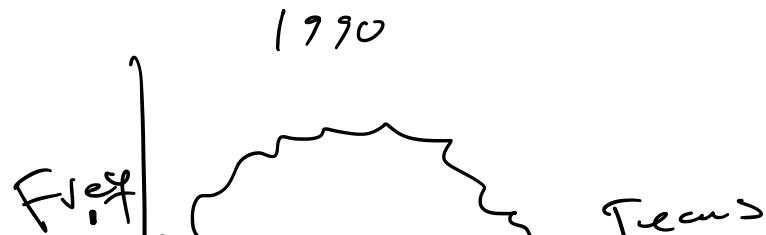
- Discussion
- Example derivation

Bayes Rule

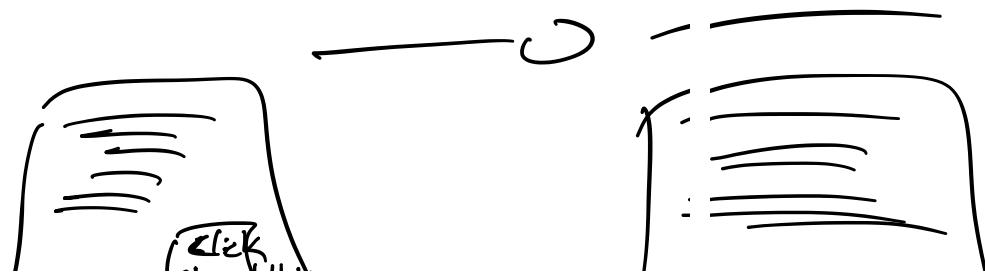
Data Analysis with Geometry

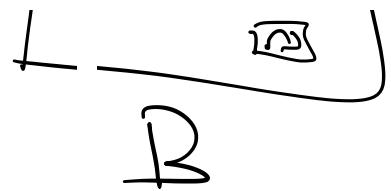
- Preliminaries
- Geometry and distances
- KNN classifier
- Some vector algebra

Proj 2 Problem 0



year	team ID	Payroll	...
		r	





Click rate p_A

Click rate p_B

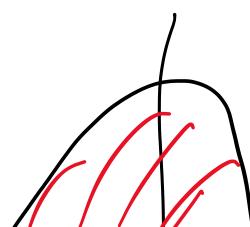
Part 1) Known $p_A = .5$

Part 2) Known $p_A = 0.75$

Part 3)
Part 4) Both p_A and p_B are estimated

Hypothesis: $p_B \leq p_A$

Estimate $S_n = .6$



H_0

if $P(\bar{X} > \hat{P}_B) \leq .05$

then Reject



Approximate as
 $N(0.5, \text{Var}(\bar{X}))$

pnorm(.6, mean = 0.5, sd = ...)

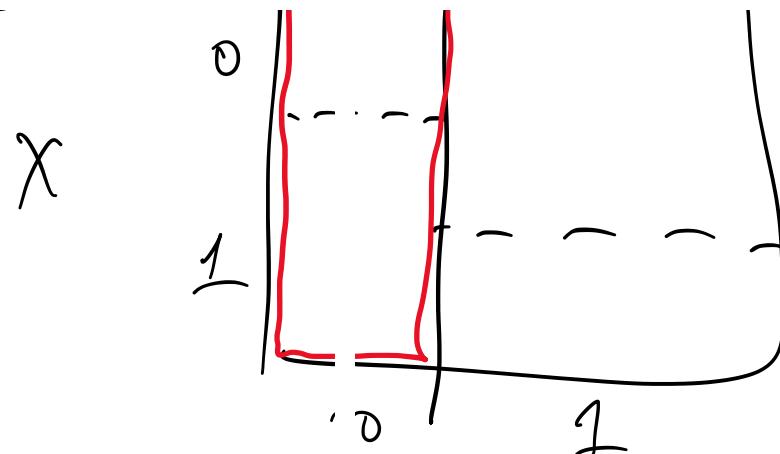
qnorm(-.95, mean = 0.5, sd = ...)

→ 0 →

Joint & Conditional
Probability

1 = 1

$p(x, y)$



a) $p(X=x, Y=y) \geq 0 \quad \forall (x, y) \in D_x \times D_y$

b) $\sum_{(x, y) \in D_x \times D_y} p(X=x, Y=y) = 1$

Conditional

$$P(X=x | Y=o) = \frac{P(X=x, Y=o)}{P(Y=o)}$$

T. dependence

~~sample -~~

$$P(X=x | Y=y) = P(Y=y | X=x) \text{ for all } y \in D_y$$



Easier to think about $P(Y=y | X=x)$
than the opposite $P(X=x | Y=y)$

Sentiment analysis

X: {positive vs. negative}

Y: word frequency

Bayes Rule:

$$\frac{P(X=x | Y=y)}{P(X=x | Y=n)} = \frac{P(Y=y | X=x)}{P(Y=n | X=x)}$$

is very often exchanging conditional probabilities

$$P(X=x | Y=y) = \frac{P(Y=y | X=x)}{P(Y=y)} P(X=x)$$

$$P(X=x | Y=y) = \frac{P(X=x, Y=y)}{P(Y=y)}$$

$$= \frac{P(Y=y | X=x) P(X=x)}{P(Y=y)}$$

$$\therefore P(X=x | Y=y) \approx \underbrace{P(Y=y | X=x)}_{\text{data distribution}} \underbrace{\frac{P(X=x)}{P(X=x)}}_{\text{prior distribution}}$$

$$P(\theta^n | \tau^n)$$

posterior distribution

$$p(Y|y)$$

normalizing
constant

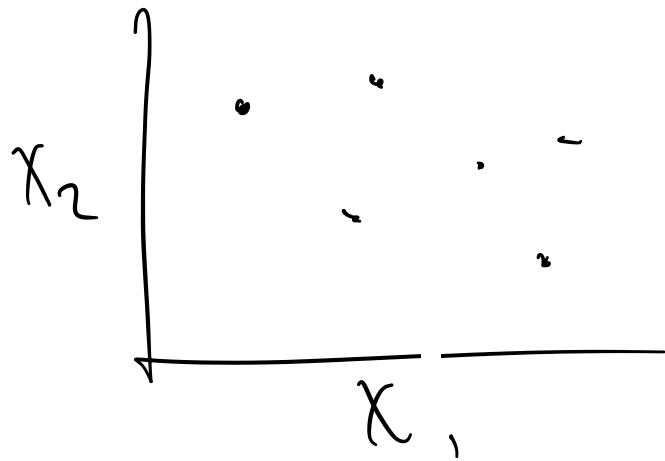
$$\int \dots \int$$

Start thinking about data as
geometric objects

$$P(\text{default} = \text{yes} | \text{student}, \text{balance}, \text{income})$$

outcome: \vec{y}

predictors: x_1, x_2, \dots, x_n



Predictors