

DS2500
2/7-Fri

Admin

- Hw3 due 9pm
- Exam #1 2/14
- mini projects
- Hw4 graded!
- 2/10 signups
- 2/23 slides due
- 2/24, 2/25 prez

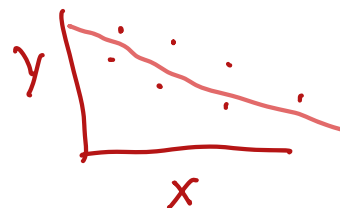
Agenda

1. normalizing data
2. Variance, std dev
3. Python

1. Normalizing Data

→ Tuesday: correlation
linear regression } adm rate, tuition
 $\overline{x}$ (ind) $\overline{y}$ (dep)

- given x , what is y ?
- given 2 new x , what would y be?



conc: adm rate, tuition highly correlated

Another ~~one~~ ~ time series data

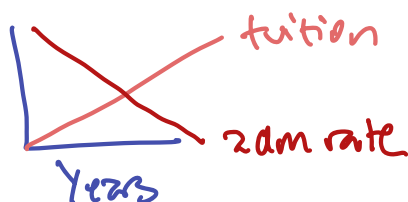
- sequence of observations assoc. with points in time

→ successive, equally spaced

- time series analysis: extracting predictions and filling in gaps

- could end up here if values up and down
- but both are consistently $\uparrow$ or $\downarrow$ so they may also correlate with time

Expectation



time: ind variable
tuition, adm: dep variable

↳ But! $\begin{matrix} \cdot \\ \cdot \\ \cdot \\ \cdot \end{matrix}$ per year: tuition } features
adm rates }

2014 vs. 2015?

2020 vs. other year?

2008 vs. 2024

or

2020 vs. 2024

{euclidean distance}!

Euclidean distance assumes
that: unit change in any
direction is equally significant

(ex) 2014 vs. 2015

Adm	32.22	28.48
tuition	60290	62800

$$\delta(2014, 2015) = \sqrt{(32.22-28.48)^2 + (60290-62800)^2} = 2510$$

2014 vs 2015

Drop in adm rate?

2014 vs. 2016

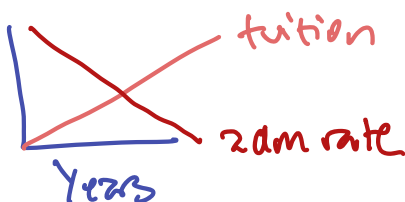
adm rate = 18
tuition = 62000

$$\sqrt{(32.22-18)^2 + (60290-62000)^2}$$

2510.04

2014 vs 2016

Expectation



Rise in tuition?

2014 vs. 2016

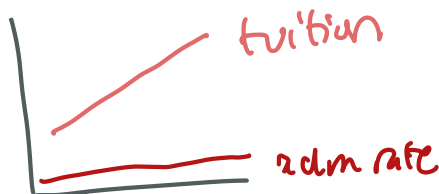
adm rate = 28.48
tuition = 78202

17912

2014 vs. 2016

problem #1: one feature has huge
impact over the other

problem #2: communication



To solve:

tuition: 50,000 to 78,000

adm: 8 to 33

Solution → normalization!

• max = 1 • min = 0

→ smallest val in population: 0
largest val in pop: 1

for every $x_i \dots x_{max}, x_{min}$

everything else scaled

$$x_{scaled} = \frac{x_i - x_{min}}{x_{max} - x_{min}}$$

x_i vs x_{max}
 x_i vs x_{min}

(ex) pop = 1, 2, 3, 4, 5
min 1
max 5

↳ feature scaling

$$\frac{3-1}{5-1} = \frac{2}{4} = \text{cloud with 5 and a smiley face}$$

min/max normalization:
↳ 0 to 1 range

downside: no units
suppress outliers

2. Variance + Std Deviation

↳ How different from the mean?

.....
low var

.....
high var

• Variance avg squared difference from mean

σ^2

used for: comparing 2 datasets
(end up w/ bigger #s)

• Std dev avg difference from mean

σ

used for: one dataset

(keeps units you started with)

$$\sigma^2 = \frac{1}{N-1} \sum (x_i - \bar{x})^2$$

$$\sigma = \sqrt{\sigma^2}$$

(basically a mean, but we divide by $N-1$)

(Bessel's, to correct for not having every data point)

lst2 = [-12, 30, 50, -55, -75, 80]

$\sigma^2(\text{lst2})$ $\bar{x} = 3$

3708

variance: lst 2 varies 2 lot more
than lst 1

$\sigma(\text{lst2}) = 60.89$

↳ on avg, how
far from mean

In Python: `statistics.variance`

lst1 = [-1, 3, 5, -3, 6, 8]

$\sigma^2(\text{lst1})$ $\bar{x} = 3$

18

Standard

$\sigma(\text{lst1}) = 4.24$

↳ on avg, how far from mean

Outlier: > 2 standard deviations from mean