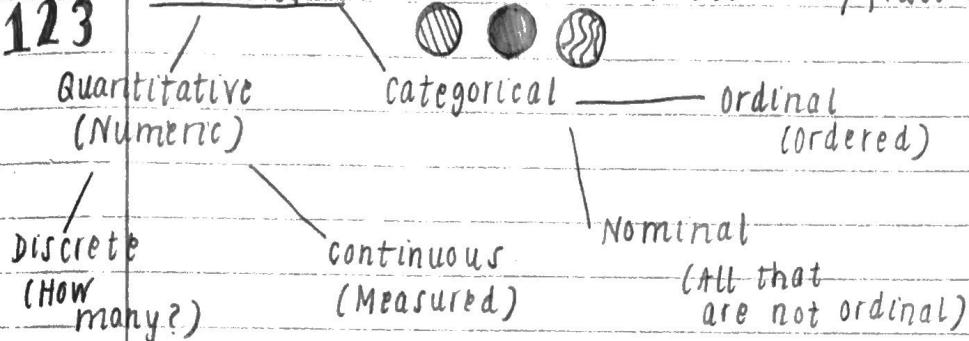


Midterm 1 Review

123

Data Types



* Whole numbers → discrete

* Decimals/fractions → continuous

problem Types

1. Descriptive
2. Predictive
3. Causative

* The JAMA article (discussion 1) looks at how gun mortality changes over time BUT DOES NOT ASK:

- How will mortality be in the future?
- What causes these mortalities?

* Can we take these insurance charges and ages, then predict charges for someone new who gives us their age?

In order to deal with data on the computer we use

R

ggplot2: data visualization
dplyr: data manipulation

dplyr

dataset %>% function()

ggplot2

```
ggplot(dataset, aes( ))  
+ geom_something()
```

mutate(new_col = $\overbrace{\text{old_col}}$)

This is an example! Many more ways to mutate, like log()!

* Arguments go into aes() vary depending on which geom you are using

MORE summarize()
filter()
select()
arrange()
group_by()
rename()

All the plots

x/y/fill

ggplot(your_data, aes()) +
geom_something()



Plot

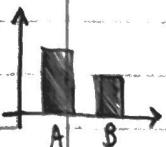
Variable

code

Bar plot

1. Categorical variable
on the x

x=categorical_col
geom_bar()



2. Categorical x
Numeric y
(for bar heights)

x=categorical_col
y=numeric_col

geom_bar(stat =
"identity")



3. 2 categorical variables

x=cat_1_col
fill=cat_2_col

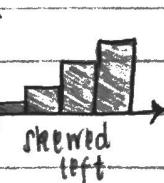
geom_bar(position="dodge")

Histogram

Numeric
Quantitative on x
that can be grouped
into intervals
(bins)

x=numeric_col
geom_histogram()

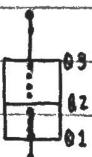
binwidth=?



Mean > Median

Mean < Median

Boxplot



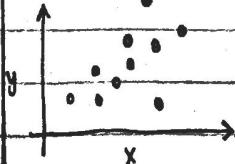
Numeric on y

y=numeric_col
geom_boxplot()

Scatterplot

Numeric on both
x and y

x=num_col_1
y=num_col_2
geom_point()



The IQR Visual: Boxplot

calculations

$$1QR = Q3 - Q1 = 3$$

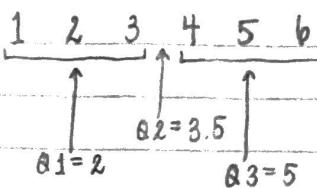
$$\text{Up Whisker} = Q3 + 1.5(IQR)$$

$$= 5 + 4.5 = 9$$

Low Whisker = Q1 - 1.5(IQR)

$$= 2 - 4.5 = -2.5$$

Example 1

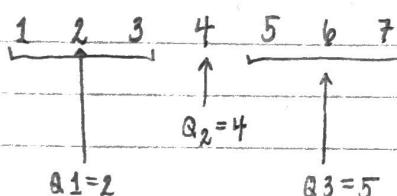


$$\text{Upper Whisker} = Q3 + 1.5(IQR)$$

$$\text{Lower Whisker} = Q1 - 1.5(IQR)$$

$$= 1 - 4.5 = -3$$

Example 2



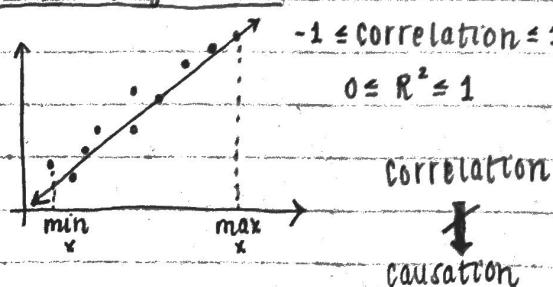
* Outliers are data points outside the range [-2.5, 9]

* TO OUTLIERS...

ROBUST!
Median

NOT!
Mean

Linear Regression



```
your_model <- lm(y ~ x, data = your_data)
```

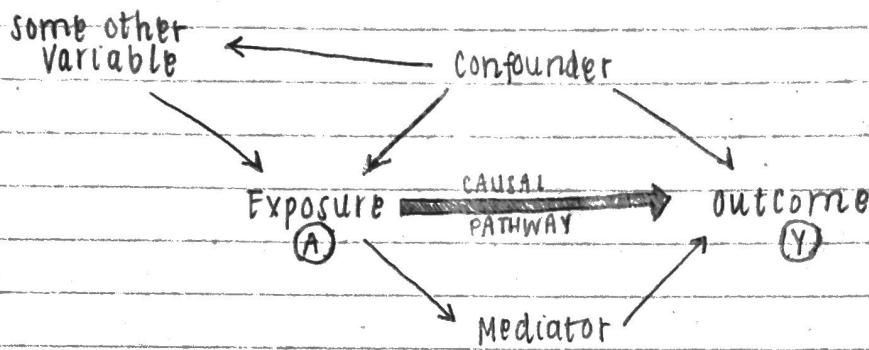
{ glance(your_model) ?
tidy(your_model) }

From the output, you can write

- * Know the interpretations of slope / intercept
(See review slides!)

- * To predict, set x equal to a value and evaluate $y = mx + b$

Causal Graphs



Backdoor Path

A path that does not start at \textcircled{A} but goes through \textcircled{A} and ends at \textcircled{Y}