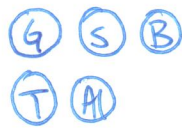


DATA TYPES



NOMINAL.



ORDINAL.

°C / °F.

INTERVAL



RATIO.

"RANK"

← DISCRETE →

← SCALAR →

← CONTINUOUS →

- buckets

- ordering

- regular, meaningful distances b/w each point.
- arbitrary zero.

preference

performance : speed, errors.

|| across people.

DESCRIPTIVE STATISTICS.

CENTRAL TENDENCY.

MEAN.

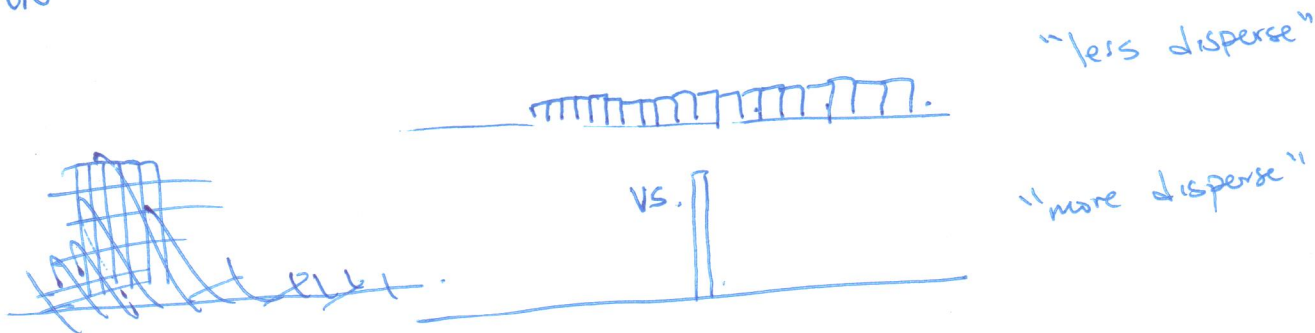
MEDIAN

MODE.

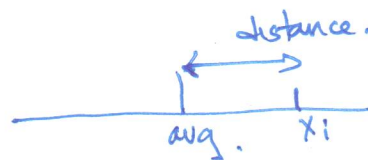


- MEAN: $\frac{\sum x_i}{n}$
- MEDIAN: protection from extreme outliers.
- MODE: useful for nominal data.

DISPERSION.



Common MEASURE " RANGE, VARIANCE



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

← squared values. //

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

← std dev.

~~STANDARD DEVIATION~~

Why Inferential Stats.

2

~~MIKE 6'2"~~

~~MIKE~~

MIKE 5'3"

LARRY 5'5"

"LARRY IS TALLER THAN MIKE" -
- comment about the sample.

NED 5'6"

GAIL 5'9"
HEATHER 6'1"
IVY 5'11"

AVG

$\sim 5'4\frac{1}{2}"$

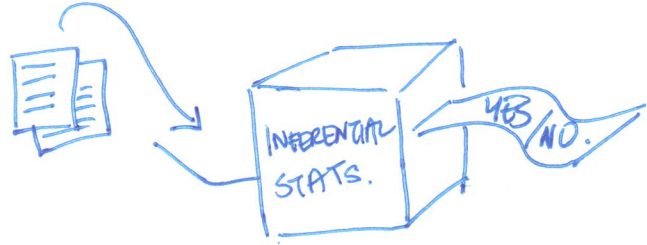
$\sim 5'11"$

"Men are shorter than women"
- comment about the population

- common sense.
- I can manipulate the sample. (sampling bias)
- small group size.
- these are not representative.

IDEAL.

- ① HAVE A HYPOTHESIS.
- ① COLLECT DATA
- ③ INSERT DATA INTO MACHINERY DATA.



- ③ GET YES / NO.

INFERENCE \Rightarrow BEST / INFORMED GUESS.

SIGNIFICANCE

NULL-HYPOTHESIS STATISTICAL TESTING

PROCEDURE

- ① Generate a null hypothesis (H_0)
- ② Generate an alternate hypothesis (H_a)
- ③ Select α value. (Probability of a Type I error)
↳ Reject H_0 when H_0 is true
- ④ Run test, get p-value.
If $p < \alpha$ reject H_0 .
else, cannot reject H_0 .

* Nothing magical going on here. *

KNOWLEDGE TYPE I & TYPE II ERRORS

Honest Ed is accused of theft.

~~Which is worse?
 Ed did not commit the crime but is found guilty.
 Ed committed the crime but is found innocent.~~

Ed

		TRUE	FALSE
VERDICTS. (INNOCENCE)	TRUE	Ed did not commit theft & he is innocent.	TYPE II.
	FALSE	TYPE I	Ed committed theft & is found guilty.

H_0 = Ed is innocent.
 Normally, we say TYPE I errors are worse than TYPE II errors.

TYPE I ERROR = ~~REJECT~~ REJECT H_0 WHEN H_0 IS TRUE
 "GUILTY BILITY"

Ex: New touch screen ATMs vs. RECALL BUTTON ATM.
 H_0 : People make same # of errors w TS as RB
 H_1 : People can use @ same speed TS as RB.
 ⇒ Expensive to change over.

Probability (Type I Error) = $P(\text{rejecting } H_0 \mid H_0 \text{ is true}) = \alpha$. 6/.

$\alpha = 0.05$ or $\alpha = 0.01$, (convention)

STATS TEST.

t-test

ANOVA.

⋮

⇒ p-value.

p-value: likelihood of getting this data if H_0 were true.

NOT
likelihood H_0 is true/false.

If $p < \alpha$, reject H_0 . ⇒ "SIGNIFICANT DIFFERENCE"
else, cannot reject H_0 .

MYTHS OF NULL HYPOTHESIS TESTING

7/

① P is NOT probability of H_0 being true.

It is likelihood of getting this, or more extreme data if H_0 is true.

H_0 true is binary \Rightarrow YES or NO.

② P-VALUE DOESN'T SAY "MORE/LESS" SIGNIFICANT, NOT BORDERLINE SIGNIFICANT.

P-value is what it is

α is set in advance. \Rightarrow no cheating!

③ α IS ARBITRARY.
Nothing magical.

④ SIGNIFICANCE \neq IMPORTANT

\hookrightarrow interpretation is important