

Unit 1: Introduction to data

3. Introduction to statistical inference

Sta 104 - Summer 2015

Duke University, Department of Statistical Science

May 18, 2015

1. Housekeeping

2. Case study: is yawning contagious?

1. Competing claims
2. Testing via simulation
3. Checking for independence

3. Case study: Tapping on caffeine

- ▶ PS 1 + PA 1 due tonight
- ▶ RA 2 tomorrow

Finish up content from last time...

1. Housekeeping

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Clicker question

Do you think yawning is contagious?

- (a) Yes
- (b) No
- (c) Don't know

An experiment conducted by the MythBusters tested if a person can be subconsciously influenced into yawning if another person near them yawns.



50 people were randomly assigned to two groups:

- ▶ treatment: see someone yawn, $n = 34$
- ▶ control: don't see someone yawn, $n = 16$

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Not Yawn	24	12	36
Total	34	16	50
% Yawners			

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Based on the proportions we calculated, do you think yawning is really contagious, i.e. are seeing someone yawn and yawning dependent?

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- ▶ So we will do just that - well, somewhat - and see what happens
- ▶ Instead of actually conducting the experiment many times, we will *simulate* our results

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2. “There is something going on.”

Seeing someone yawn and yawning are *dependent*, observed difference in proportions of yawners in the treatment and control is not due to chance. → *Alternative hypothesis*



- ▶ H_0 : Defendant is innocent
- ▶ H_A : Defendant is guilty
- ▶ Present the evidence: collect data.
- ▶ Judge the evidence: “Could these data plausibly have happened by chance if the null hypothesis were true?”
- ▶ Make a decision: “How unlikely is unlikely?”

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- ▶ Test results suggest the observed data are
 - reasonably likely to have occurred even if the null is true \rightarrow stick with the null hypothesis (fail to reject H_0)
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 - unlikely to have occurred even if the null is true \rightarrow reject H_0 in favor of H_A
- ▶ Never declare the null hypothesis to be true, because we simply do not know whether it's true or not \rightarrow never “accept the null hypothesis”

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- ▶ Results from the simulations based on the *null hypothesis* look like the data → difference between the proportions of yawners in the treatment and control groups was simply *due to chance* (yawning is not contagious)

... under the assumption of independence, i.e. leaving things up to chance:

- ▶ Results from the simulations based on the *null hypothesis* look like the data → difference between the proportions of yawners in the treatment and control groups was simply *due to chance* (yawning is not contagious)
- ▶ Results from the simulations based on the chance model do not look like the data → difference between the proportions of yawners in the treatment and control groups was not due to chance, but *due to an actual effect of seeding* (yawning is contagious)

- ▶ A regular deck of cards is comprised of 52 cards: 4 aces, 4 of numbers 2-10, 4 jacks, 4 queens, and 4 kings.
- ▶ Take out two aces from the deck of cards and set them aside.
- ▶ The remaining 50 playing cards to represent each participant in the study:
 - 14 face cards (including the 2 aces) represent the people who yawn.
 - 36 non-face cards represent the people who don't yawn.

1. Shuffle the 50 cards at least 7 times to ensure that the cards counted out are from a random process
2. Divide the cards into two decks:
 - deck 1: 16 cards → control
 - deck 2: 34 cards → treatment
3. Count the number of face cards (yawners) in each deck
4. Calculate the difference in proportions of yawners (*treatment - control*), and submit this value using your clicker - **only one submission per team per simulation**
5. Repeat steps (1) - (4) many times

Why shuffle 7 times: http://www.dartmouth.edu/~chance/course/topics/winning_number.html

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Clicker question

Do the simulation results suggest that yawning is contagious, i.e. does seeing someone yawn and yawning appear to be dependent?

(Hint: In the actual data the difference was 0.04, does this appear to be an unusual observation for the chance model?)

(a) Yes

(b) No

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- ▶ In a double-blind experiment a sample of male college students were asked to tap their fingers at a rapid rate.
- ▶ The sample was then divided at random into two groups of 10 students each.
- ▶ Each student drank the equivalent of about two cups of coffee, which included about 200 mg of caffeine for the students in one group but was decaffeinated coffee for the second group.
- ▶ After a two hour period, each student was tested to measure finger tapping rate (taps per minute).

	Taps	Group
1	246	Caffeine
2	248	Caffeine
3	250	Caffeine
4	252	Caffeine
5	248	Caffeine
6	250	Caffeine
...		
16	248	NoCaffeine
17	242	NoCaffeine
18	244	NoCaffeine
19	246	NoCaffeine
20	242	NoCaffeine

Clicker question

What type of plot would be useful to visualize the distributions of tapping rate in the caffeine and no caffeine groups.

- (a) Bar plot
- (b) Mosaic plot
- (c) Pie chart
- (d) Side-by-side box plots
- (e) Single box plot

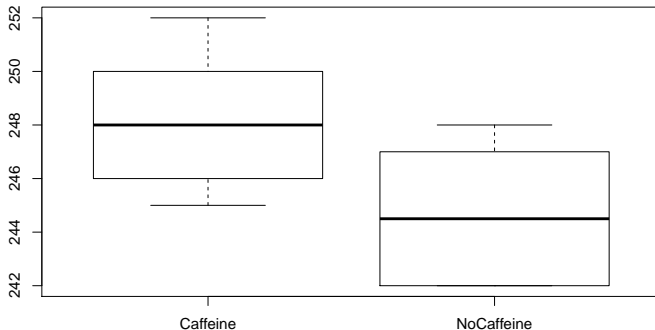
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Compare the distributions of tapping rates in the caffeine and no caffeine groups.

	Caffeine	No Caffeine	Difference
mean	248.3	244.8	3.5
SD	2.21	2.39	-0.18
median	248	245	3
IQR	3.5	4.25	-0.75



Clicker question

We are interested in finding out if caffeine increases tapping rate. Which of the following are the correct set of hypotheses?

(a) $H_0 : \mu_{caff} = \mu_{no\ caff}$

$H_A : \mu_{caff} < \mu_{no\ caff}$

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(c) $H_0 : \bar{x}_{caff} = \bar{x}_{no\ caff}$

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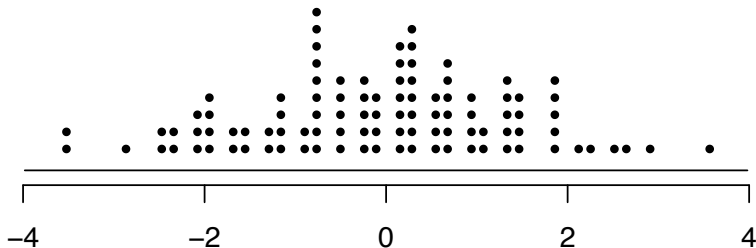
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- ▶ On 20 index cards write the tapping rate of each subject in the study.
- ▶ Shuffle the cards and divide them into two stacks of 10 cards each, label one stack “caffeine” and the other stack “no caffeine”.
- ▶ Calculate the average tapping rates in the two simulated groups, and record the difference on a dot plot.
- ▶ Repeat steps (2) and (3) many times to build a *randomization distribution*.

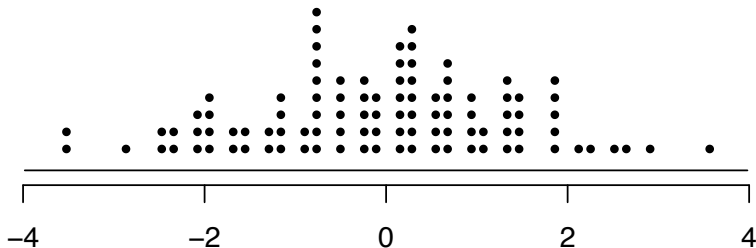
Below is a randomization distribution of 100 simulated differences in means ($\bar{x}_c - \bar{x}_{nc}$). Calculate the p-value for the hypothesis test evaluating whether caffeine increases average tapping rate.

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$$1/100 = 0.01$$

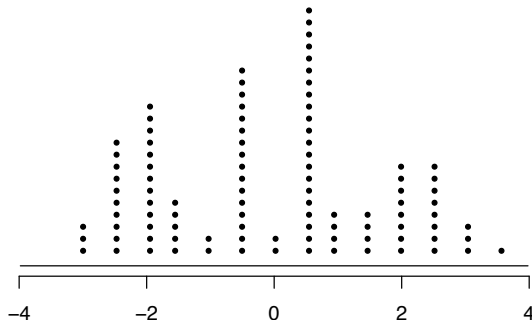
Describe how could we use the same approach to test whether the median tapping rate is higher for the caffeine group?

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Use the same simulation scheme but record the difference between the medians instead of the means, and calculate the p -value as the proportion of simulations where the simulated difference in medians is at least 3.

Below is a randomization distribution of 100 simulated differences in medians ($med_c - med_{nc}$). Do the data provide convincing evidence that caffeine increases median tapping rate?

	Caffeine	No Caffeine	Difference
median	248	245	3



Application exercise: 1.4 Randomization testing

See the course website for instructions.