# Lecture 26 – Final Review DSC 10, Winter 2024

### **Announcements**

- Lab 7 is due tomorrow at 11:59PM.
- The Final Exam is this Saturday 3/16 from 7-10PM.
  - All sections will take the exam in Catalyst 0125. Seating assignments will be released Friday.
- Collaborative study session on Friday 3/15 from 5-8PM in Solis 104.
- If at least 75% of the class fills out both <u>SETs</u> and the DSC 10-specific <u>End-of-Quarter Survey</u>, then the entire class will have 1% of extra credit added to their overall grade. We value your feedback!

## Agenda

- We'll work through selected problems from the Fall 2023 Final Exam.
- We won't write any code, since you can't run code during the exam. Instead, we'll try to think like the computer ourselves.
- These annotated slides will be posted after lecture is over, as will the solutions to the entire exam.
- Try the problems with us!



Access the exam <a href="here">here</a>. Make sure to read the data info sheet at the top before starting.

Aaron wants to explore the discrepancy in fraud rates between "discover" transactions and "mastercard" transactions. To do so, he creates the DataFrame ds\_mc, which only contains the rows in txn corresponding to "mastercard" or "discover" transactions.

After he creates ds\_mc, Aaron groups ds\_mc on the "card" column using two different aggregation methods. The relevant columns in the resulting DataFrames are shown below.

Aaron decides to perform a test of the following pair of hypotheses:

- Null Hypothesis: The proportion of fraudulent "mastercard" transactions is the same as the proportion of fraudulent "discover" transactions.
- Alternative Hypothesis: The proportion of fraudulent "mastercard" transactions is less than the proportion of fraudulent "discover" transactions.

As his test statistic, Aaron chooses the difference in proportion of transactions that are fraudulent, in the order "mastercard" minus "discover".

ds\_mc.groupby("card").sum()

ds\_mc.groupby("card").count()

is fraud

is\_fraud

#### Problem 6.1

What type of statistical test is Aaron performing?

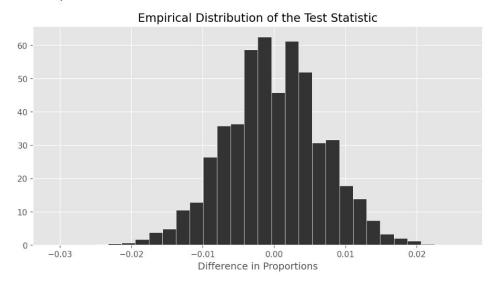
- Standard hypothesis test
- Permutation test

card		card		
discover	160	discover	2000	
mastercard	4000	mastercard	40000	

#### Problem 6.2

What is the value of the observed statistic? Give your answer either as an exact decimal or simplified fraction.

The empirical distribution of Aaron's chosen test statistic is shown below.



#### Problem 6.3

Which of the following is closest to the p-value of Aaron's test?

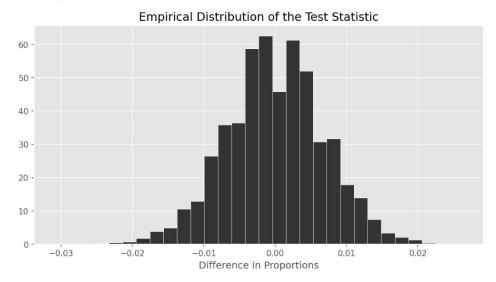
- 0.001
- 0.37
- 0.63
- 0.94
- 0.999

#### Problem 6.4

What is the conclusion of Aaron's test?

- O The proportion of fraudulent "mastercard" transactions is less than the proportion of fraudulent "discover" transactions.
- O The proportion of fraudulent "mastercard" transactions is **greater than** the proportion of fraudulent "discover" transactions.
- The test results are inconclusive.
- None of the above.

The empirical distribution of Aaron's chosen test statistic is shown below.



#### Problem 6.5

Which of the following is closest to the p-value of Aaron's new test?

- 0.001
- 0.06
- 0.37
- 0.63
- 0.94
- 0.999

Aaron now decides to test a slightly different pair of hypotheses.

- Null Hypothesis: The proportion of fraudulent "mastercard" transactions is the same as the proportion of fraudulent "discover" transactions.
- Alternative Hypothesis: The proportion of fraudulent "mastercard" transactions is greater than the proportion of fraudulent "discover" transactions.

He uses the same test statistic as before.

The DataFrame ten\_txns, displayed in its entirety below, contains a simple random sample of 10 rows from txn.

	is_fraud	amount	method	card	lifetime	browser
transaction_id						
3169166	True	100.00	credit	visa	532601.00	chrome 63.0
3093921	False	100.00	debit	mastercard	173276.00	mobile safari 10.0
3137058	False	100.00	debit	visa	120000.00	chrome 63.0
3063164	False	100.00	debit	visa	141342.00	mobile safari 11.0
3051461	False	75.00	credit	visa	153871.00	ie 11.0 for desktop
3171154	False	25.00	debit	visa	182654.00	mobile safari generic
3222420	False	25.00	credit	visa	50199.00	safari generic
3226397	False	22.96	debit	mastercard	122352.00	safari generic
3073572	False	9.33	credit	mastercard	66703.00	mobile safari 11.0
3253371	False	5.00	debit	visa	3007.00	chrome 64.0

#### Problem 9.1

Suppose we randomly select one transaction from ten\_txns. What is the probability that the selected transaction is made with a "card" of "mastercard" or a "method" of "debit"?

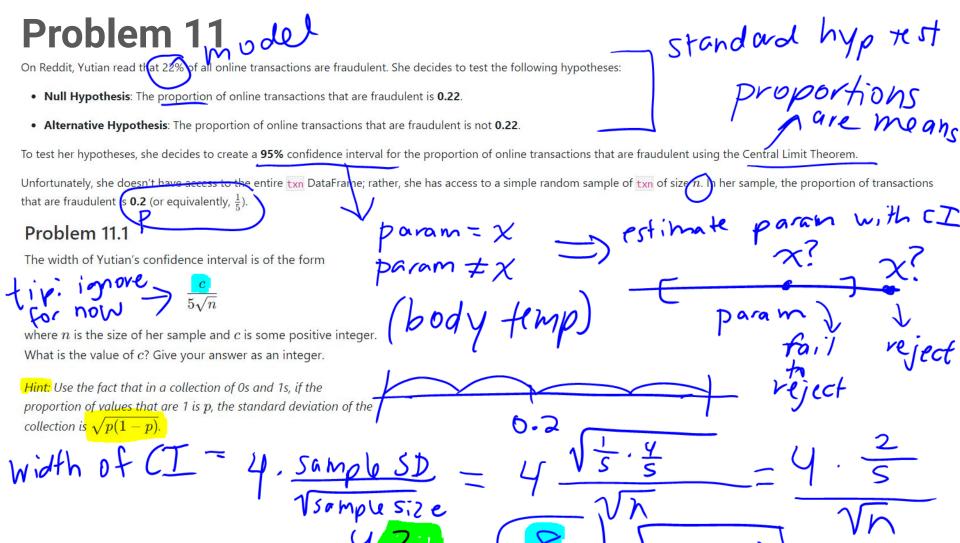
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3063164	False	100.00	debit	visa	141342.00	mobile safari 11.0
3051461	False	75.00	credit	visa	153871.00	ie 11.0 for desktop
3171154	False	25.00	debit	visa	182654.00	mobile safari generic
3222420	False	25.00	credit	visa	50199.00	safari generic
3226397	False	22.96	debit	mastercard	122352.00	safari generic
3073572	False	9.33	credit	mastercard	66703.00	mobile safari 11.0
3253371	False	5.00	debit	visa	3007.00	chrome 64.0

#### Problem 9.2

Suppose we randomly select two transactions from "ten\_txns", without replacement, and learn that neither of the selected transactions is for an amount of 100 dollars. Given this information, what is the probability that:

- the first transaction is made with a "card" of "visa" and a "method" of "debit", and
- the second transaction is made with a "card" of "visa" and a "method" of "credit"?



On Reddit, Yutian read that 22% of all online transactions are fraudulent. She decides to test the following hypotheses:

- Null Hypothesis: The proportion of online transactions that are fraudulent is 0.22.
- Alternative Hypothesis: The proportion of online transactions that are fraudulent is not 0.2

To test her hypotheses, she decides to create a 95% confidence interval for the proportion of online transactions that are fraudulent using the Central Limit Theorem.

Unfortunately, she doesn't have access to the entire txn DataFrame; rather, she has access to a simple random sample of txn of size n. In her sample, the proportion of transactions that are fraudulent is **0.2** (or equivalently,  $\frac{1}{5}$ ).

#### Problem 11.2

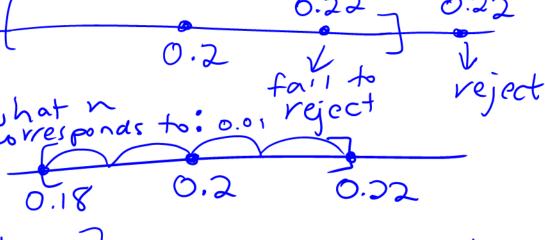
There is a positive integer J such that:

- ullet If n < J, Yutian will fail to reject her null hypothesis at the 0.05 significance level.
- If n > J, Yutian will reject her null hypothesis at the **0.05** significance level.

What is the value of J? Give your answer as an integer.

$$6.01 = \frac{2}{5\sqrt{h}} \sqrt{h} = \frac{2}{0.05}$$

$$\sqrt{n} = 2.20 = 40$$



On Reddit, Keenan also read that 22% of all online transactions are fraudulent. He decides to test the following hypotheses at the **0.16 significance level**:

- Null Hypothesis: The proportion of online transactions that are fraudulent is 0.22.
- Alternative Hypothesis: The proportion of online transactions that are fraudulent is not 0.22.

Keenan has access to a simple random sample of txn of size **500**. In his sample, the proportion of transactions that are fraudulent is **0.23**.

Below is an incomplete implementation of the function <u>reject\_null</u>, which creates a bootstrap-based confidence interval and returns **True** if the conclusion of Keenan's test is to **reject** the null hypothesis, and **False** if the conclusion is to **fail to reject** the null hypothesis, all at the **0.16** significance level.

```
def reject_null():
    fraud_counts = np.array([])
    for i in np.arange(10000):
        fraud_count = np.random.multinomial(500, __(a)__)[0]
        fraud_counts = np.append(fraud_counts, fraud_count)

L = np.percentile(fraud_counts, __(b)__)
R = np.percentile(fraud_counts, __(c)__)

if __(d)__ < L or __(d)__ > R:
        # Return True if we REJECT the null.
        return True

else:
    # Return False if we FAIL to reject the null.
    return False
```

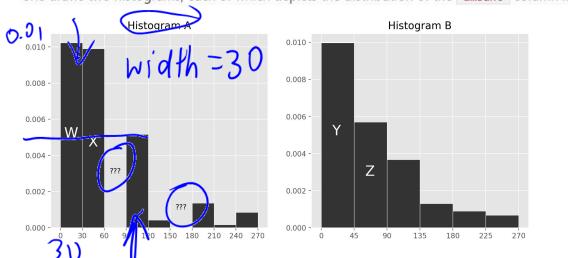
Fill in the blanks so that reject\_null works as intended.

Hint: Your answer to (d) should be an integer greater than 50.

## Problem 13, pp - 120 trans a chons

Ashley doesn't have access to the entire txn DataFrame; instead, she has access to a simple random sample of 400 rows of txn.

She draws two histograms, each of which depicts the distribution of the "amount" column in her sample, using different bins.



- (vactions
area = In/+/

Unfortunately, DataHub is being finicky and so two of the bars in Histogram A are deleted after it is created.

#### Problem 13.1

 $\circ$  [180, 210)

In Histogram A, which of the following bins contains approximately 60 transactions? (A Y e A = ) YO

$$0.30,60)$$
 area = Width \* height  $0.15 = 30 * height$ 

$$0.15 = 30 * height$$

$$0.15 = 30 * height$$

#### Problem 13.2

Let w, x, y, and z be the heights of bars W, X, Y, and Z, respectively. For instance, y is about 0.01.

Which of the following expressions gives the height of the bar corresponding to the [60, 90) bin in Histogram A?

$$\bigcirc (y+z)-(w+x)$$

$$\bigcirc (w+x)-(y+z)$$

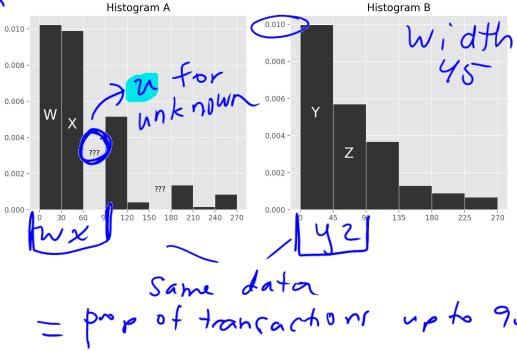
$$\odot rac{3}{2}(y+z)-(w+x)$$

$$\circ (y+z) - rac{3}{2}(w+x)$$

$$\bigcirc \ 3(y+z)-2(w+x)$$

$$\bigcirc 2(y+z)-3(w+x)$$

O None of the above.



$$w \cdot 30 + \chi \cdot 30 + \mu \cdot 30 = y \cdot 45 + \chi \cdot 45$$
  
Algebra:  $30u = 45(y+z) - 30(w+x)$   
 $u = \frac{3}{2}(y+z) - 1(w+x)$ 

As mentioned in the previous problem, Ashley has sample of 400 rows of txn. Coincidentally, in Ashley's sample of 400 transactions, the mean and standard deviation of the "amount" column both come out to 70 dollars.

amour 50=70

#### Problem 14.1

Fill in the blank:

What goes in the blank? Give your answer as an integer. Both blanks are filled in with

athermonuntier.
$$(-\frac{1}{2^2}) + du$$

$$(-\frac{1}{2^2}) + \frac{1}{16} = \frac{15}{16} + \frac{1}{16}$$

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$$Z SDS$$

was  $A = \frac{15}{16}$ 

of data within  $A = \frac{15}{16}$ 
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of the (amount" column both come out to 70 dollars.

Now, we're given that the mean and standard deviation of the "lifetime" column in Ashley's sample are both equal to c do hars. We're also given that the correlation between transaction amounts and lifetime spending in Ashley's sample is  $-\frac{1}{4}$ .

#### Problem 14.3

The predicted lifetime spending, in **dollars**, of a card with a transaction amount of 280 dollars is of the form  $f \cdot c$ , where f is a fraction. What is the value of f? Give your answer as a simplified fraction.

Mun  $\chi = 70$ 

$$SP \times = 70$$

muan y = c

$$y = -\frac{1}{4}$$

prediction when 
$$x$$
 is  $280$ 

$$m = r \cdot SD \cdot y = -\frac{1}{4} \cdot c = -\frac{1}{280} \cdot c$$

$$y = mean y - m. mean x$$

$$y = C + \frac{70}{280} \cdot C \cdot 70 = C + \frac{70}{280} \cdot C \cdot 70 = C + \frac{70}{4} \cdot C = \frac{70}{4} \cdot C \cdot 70 = C + \frac{70}{4} \cdot C = \frac{70}{4} \cdot C \cdot 70 = C + \frac{70}{4} \cdot C = \frac{70}{4} \cdot C$$

$$J = m\chi + b = \frac{1}{280} \cdot c \cdot 280 + \frac{5}{4}$$

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$$\chi = 70$$

regression line in standard your = r. X (su)

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Coincidentally, in Ashley's sample of 400 transactions, the mean and standard deviation of the "amount" column both come out to 70 dollars.

#### Problem 14.4

Suppose the intercept of the regression line, when both transaction amounts and lifetime spending are measured in **dollars**, is 40. What is the value of c? Give your answer as an integer.

production when 
$$x$$
 is  $280$ 
 $m = r \cdot 5Dy = -\frac{1}{4} \cdot c = -\frac{1}{280} \cdot c$ 
 $b = mean y - m. m. an x$ 
 $b = c + \frac{1}{280} \cdot c \cdot 70 = c + \frac{70}{4} \cdot c = 40$ 
 $= c + \frac{1}{4}c = 40$ 

$$\frac{S}{4} = \frac{5}{400}$$
 $C = \frac{400}{8}$ 
 $C = 8 * 4$ 
 $C = 32$ 

#### Problem 1.1

Nate's favorite number is 5. He calls a number "lucky" if it's greater than 500 or if it contains a 5 anywhere in its representation. For example, 1000.04 and 5.23 are both lucky numbers.

Complete the implementation of the function <a href="mailto:check\_lucky">check\_lucky</a>, which takes in a number as a float and returns <a href="mailto:true">true</a> if it is lucky and <a href="mailto:False">False</a> otherwise. Then, add a column named <a href="mailto:"">"is\_lucky"</a> to <a href="mailto:txn">txn</a> that contains <a href="mailto:true">True</a> for lucky transaction amounts and <a href="mailto:False">False</a> for all other transaction amounts, and save the resulting <a href="mailto:DataFrame">DataFrame</a> to the variable <a href="mailto:lucky">lucky</a>.

```
def check_lucky(x):
    return __(a)__
luck = txn.assign(is_lucky = __(b)__)
```

- 1. What goes in blank (a)?
- 2. What goes in blank (b)?

#### Problem 1.2

Fill in the blanks below so that <a href="lucky\_prop">lucky\_prop</a> evaluates to the proportion of fraudulent "visa" card transactions whose transaction amounts are lucky.

```
visa_fraud = __(a)__
lucky_prop = visa_fraud.__(b)_.mean()
```

- 1. What goes in blank (a)?
- 2. What goes in blank (b)?

#### Problem 1.3

Fill in the blanks below so that <a href="lucky\_prop">lucky\_prop</a> is one value in the Series <a href="many\_props">many\_props</a>.

```
many_props = luck.groupby(__(a)__).mean().get(__(b)__)
```

- 1. What goes in blank (a)?
- 2. What goes in blank (b)?

Consider the DataFrame combo, defined below.

```
combo = txn.groupby(["is_fraud", "method", "card"]).mean()
```

#### Problem 2.1

What is the maximum possible value of combo.shape[0]? Give your answer as an integer.

As a senior suffering from senioritis, Weiyue has plenty of time on his hands. 1,000 times, he repeats the following process, creating 1,000 confidence intervals:

- 1. Collect a simple random sample of 100 rows from txn.
- 2. Resample from his sample 10,000 times, computing the mean transaction amount in each resample.
- 3. Create a 95% confidence interval by taking the middle 95% of resample means.

He then computes the width of each confidence interval by subtracting its left endpoint from its right endpoint; e.g. if [2,5] is a confidence interval, its width is 3. This gives him 1,000 widths. Let M be the mean of these 1,000 widths.

#### Problem 10.1

Select the true statement below.

- About 950 of Weiyue's intervals will contain the mean transaction amount of all transactions ever.
- O About 950 of Weiyue's intervals will contain the mean transaction amount of all transactions in txn.
- About 950 of Weiyue's intervals will contain the mean transaction amount of all transactions in the first random sample of 100 rows of txn Weiyue took.
- $^{\circ}$  About 950 of Weiyue's intervals will contain M.