EECS 398 W25 Midterm Review

February 23, 2025 • practicaldsc.org • github.com/practicaldsc/wn25 • 📣 See latest announcements <u>here on Ed</u>

Announcements

- The Midterm Exam is on Tuesday, February 25th from 7-9PM.
 - It covers Lectures 1-12, Homeworks 1-6, and Discussions 1-7.
- Midterm Review in lecture tomorrow (going over F24 Final 1-8.2).
- Study Tips
 - Go through lecture notebooks & homeworks to help make cheat sheet (one page, double-sided, handwritten).
 - Do discussion problems.
 - Take F24 Midterm and Problems 1-8.2 of F24 Final (besides SQL question).

Agenda

- We'll be working through <u>https://study.practicaldsc.org/mt-review-sunday/index.html</u>.
- We'll post these annotated slides and the recording after, along with enabling solutions on the study site for this worksheet.

Grouping, Querying, and Merging - Akanksha

The EECS 398 staff are looking into hotels — some in San Diego, for their family to stay at for graduation (and to eat Mexican food), and some elsewhere, for summer trips.

Each row of hotels contains information about a different hotel in San Diego. Specifically, for each hotel, we have:

- "Hotel Name" (str): The name of the hotel. Assume hotel names are unique.
- "Location" (str): The hotel's neighborhood in San Diego.
- "Chain" (str): The chain the hotel is a part of; either "Hilton", "Marriott", "Hyatt", or "Other". A hotel chain is a group of hotels owned or operated by a shared company.
- "Number of Rooms" (int): The number of rooms the hotel has.

The first few rows of hotels are shown below, but hotels has many more rows than are shown.

| | Hotel Name | Location | Chain | Number of Rooms |
|---|---------------------------------------|-----------------|----------|-----------------|
| 0 | Hotel del Coronado | Coronado | Hilton | 680 |
| 1 | Manchester Grand Hyatt | Downtown | Hyatt | 1628 |
| 2 | Hilton San Diego Bayfront | Downtown | Hilton | 1190 |
| 3 | Pendry San Diego | Gaslamp Quarter | Other | 317 |
| 4 | The Westin San Diego Gaslamp Quarter | Gaslamp Quarter | Marriott | 450 |
| 5 | San Diego Marriott La Jolla | La Jolla | Marriott | 462 |
| 6 | La Valencia Hotel | La Jolla | Other | 114 |
| 7 | Coronado Island Marriott Resort & Spa | Coronado | Marriott | 310 |

Now, consider the variable summed, defined below.

```
summed = hotels.groupby("Chain")["Number of Rooms"].sum().idxmax()
```

Problem 1.1

What is type(summed)?

int
str
Series
DataFrame

DataFrameGroupBy

| | Hotel Name | Location | Chain | Number of Rooms |
|---|---------------------------------------|-----------------|----------|-----------------|
| 0 | Hotel del Coronado | Coronado | Hilton | 680 |
| 1 | Manchester Grand Hyatt | Downtown | Hyatt | 1628 |
| 2 | Hilton San Diego Bayfront | Downtown | Hilton | 1190 |
| 3 | Pendry San Diego | Gaslamp Quarter | Other | 317 |
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In one sentence, explain what the value of summed means. Phrase your explanation as if you had to give it to someone who is not a data science major; that is, don't say something like "it is the result of grouping hotels by "Chain", selecting the "Number of Rooms" column, ...", but instead, give the value context.

summed = hotels.groupby("Chain")["Number of Rooms"].sum().idxmax()

| | Hotel Name | Location | Chain | Number of Rooms |
|---|---------------------------------------|-----------------|----------|-----------------|
| 0 | Hotel del Coronado | Coronado | Hilton | 680 |
| 1 | Manchester Grand Hyatt | Downtown | Hyatt | 1628 |
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| 5 | San Diego Marriott La Jolla | La Jolla | Marriott | 462 |
| 6 | La Valencia Hotel | La Jolla | Other | 114 |
| 7 | Coronado Island Marriott Resort & Spa | Coronado | Marriott | 310 |

Consider the variable curious, defined below.

```
curious = frame["Chain"].value_counts().idxmax()
```

Fill in the blank: curious is guaranteed to be equal to summed only if frame has one row for every _____ in San Diego.

hotel

hotel chain

hotel room

neighborhood

| | Hotel Name | Location | Chain | Number of Rooms |
|---|---------------------------------------|-----------------|----------|-----------------|
| 0 | Hotel del Coronado | Coronado | Hilton | 680 |
| 1 | Manchester Grand Hyatt | Downtown | Hyatt | 1628 |
| 2 | Hilton San Diego Bayfront | Downtown | Hilton | 1190 |
| 3 | Pendry San Diego | Gaslamp Quarter | Other | 317 |
| 4 | The Westin San Diego Gaslamp Quarter | Gaslamp Quarter | Marriott | 450 |
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| 6 | La Valencia Hotel | La Jolla | Other | 114 |
| 7 | Coronado Island Marriott Resort & Spa | Coronado | Marriott | 310 |

Fill in the blanks so that popular_areas is an array of the names of the unique neighborhoods that have at least 5 hotels and at least 1000 hotel rooms.

f = lambda df: __(i)__
popular_areas = (hotels
 .groupby(__(ii)__)
 .__(iii)__
 __(iv)__)

1. What goes in blank (i)?

2. What goes in blank (ii)?

○ "Hotel Name"

○ "Location"

○ "Chain"

○ "Number of Rooms"

3. What goes in blank (iii)?

 \bigcirc agg(f)

○ filter(f)

 \bigcirc transform(f)

4. What goes in blank (iv)?

| | Hotel Name | Location | Chain | Number of Rooms |
|---|---------------------------------------|-----------------|----------|-----------------|
| 0 | Hotel del Coronado | Coronado | Hilton | 680 |
| 1 | Manchester Grand Hyatt | Downtown | Hyatt | 1628 |
| 2 | Hilton San Diego Bayfront | Downtown | Hilton | 1190 |
| 3 | Pendry San Diego | Gaslamp Quarter | Other | 317 |
| 4 | The Westin San Diego Gaslamp Quarter | Gaslamp Quarter | Marriott | 450 |
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| 6 | La Valencia Hotel | La Jolla | Other | 114 |
| 7 | Coronado Island Marriott Resort & Spa | Coronado | Marriott | 310 |

Consider the code below.

```
cond1 = hotels["Chain"] == "Marriott"
cond2 = hotels["Location"] == "Coronado"
combined = hotels[cond1].merge(hotels[cond2], on="Hotel Name", how=???)
```

1. If we replace ??? with "inner" in the code above, which of the following will be equal to combined.shape[0]?

```
O min(cond1.sum(), cond2.sum())
```

(cond1 & cond2).sum()

- O cond1.sum() + cond2.sum()
- O cond1.sum() + cond2.sum() (cond1 & cond2).sum()
- O cond1.sum() + (cond1 & cond2).sum()

2. If we replace ??? with "outer" in the code above, which of the following will be equal to combined.shape[0]?

| 0 | <pre>min(cond1.sum(), cond2.sum())</pre> |
|---|--|
| | <pre>(cond1 & cond2).sum()</pre> |
| 0 | <pre>cond1.sum() + cond2.sum()</pre> |
| 0 | <pre>cond1.sum() + cond2.sum() - (cond1 & cond2).sum()</pre> |
| 0 | <pre>cond1.sum() + (cond1 & cond2).sum()</pre> |

| | Hotel Name | Location | Chain | Number of Rooms |
|---|---------------------------------------|-----------------|----------|-----------------|
| 0 | Hotel del Coronado | Coronado | Hilton | 680 |
| 1 | Manchester Grand Hyatt | Downtown | Hyatt | 1628 |
| 2 | Hilton San Diego Bayfront | Downtown | Hilton | 1190 |
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| 4 | The Westin San Diego Gaslamp Quarter | Gaslamp Quarter | Marriott | 450 |
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| 6 | La Valencia Hotel | La Jolla | Other | 114 |
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Random Simulations - Akanksha

Problem 2

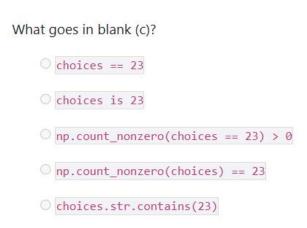
Billina Records, a new record company focused on creating new TikTok audios, has its offices on the 23rd floor of a skyscraper with 75 floors (numbered 1 through 75). The owners of the building promised that 10 different random floors will be selected to be renovated.

Below, fill in the blanks to complete a simulation that will estimate the probability that Billina Records' floor will be renovated.

```
total = 0
repetitions = 10000
for i in np.arange(repetitions):
    choices = np.random.choice(__(a)__, 10, __(b)__)
    if __(c)__:
        total = total + 1
prob_renovate = total / repetitions
```

```
What goes in blank (a)?
    np.arange(1, 75)
    np.arange(10, 75)
    np.arange(0, 76)
    np.arange(1, 76)
```

```
What goes in blank (b)?
```



Merging - Caleb

Suppose the DataFrame today consists of 15 rows — 3 rows for each of 5 different "artist_names". For each artist, it contains the "track_name" for their three most-streamed songs today. For instance, there may be one row for "olivia rodrigo" and "favorite crime", one row for "olivia rodrigo" and "drivers license", and one row for "olivia rodrigo" and "deja vu".

Another DataFrame, genres, is shown below in its entirety.

| | artist_names | genre |
|---|----------------|-------------|
| 0 | harry styles | Рор |
| 1 | olivia rodrigo | Рор |
| 2 | glass animals | Alternative |
| 3 | drake | Hip-Hop/Rap |
| 4 | doja cat | Hip-Hop/Rap |

Problem 3.1

Suppose we perform an **inner** merge between today and genres on "artist_names". If the five "artist_names" in today are the same as the five "artist_names" in genres, what fraction of the rows in the merged DataFrame will contain "Pop" in the "genre" column? Give your answer as a simplified fraction.

| | | artist_names | genre |
|--|---|----------------|-------------|
| | 0 | harry styles | Рор |
| | 1 | olivia rodrigo | Рор |
| | 2 | glass animals | Alternative |
| Suppose the DataFrame today consists of 15 rows — 3 rows for each of 5 different "artist_names". For each artist, it contains the "track_name" for their | 3 | drake | Hip-Hop/Rap |
| three most-streamed songs today. For instance, there may be one row for "olivia rodrigo" and "favorite crime", one row for "olivia rodrigo" and "drivers license", and one row for "olivia rodrigo" and "deja vu". | 4 | doja cat | Hip-Hop/Rap |
| | | | |

genres

Problem 3.2

Suppose we perform an **inner** merge between today and genres on "artist_names". Furthermore, suppose that the only overlapping "artist_names" between today and genres are "drake" and "olivia rodrigo". What fraction of the rows in the merged DataFrame will contain "Pop" in the "genre" column? Give your answer as a simplified fraction.

| | | artist_names | genre |
|--|---|----------------|-------------|
| | 0 | harry styles | Рор |
| | 1 | olivia rodrigo | Рор |
| | 2 | glass animals | Alternative |
| Suppose the DataFrame today consists of 15 rows — 3 rows for each of 5 different "artist_names". For each artist, it contains the "track_name" for their | 3 | drake | Hip-Hop/Rap |
| three most-streamed songs today. For instance, there may be one row for "olivia rodrigo" and "favorite crime", one row for "olivia rodrigo" and "drivers license", and one row for "olivia rodrigo" and "deja vu". | 4 | doja cat | Hip-Hop/Rap |
| | | | |

genres

Problem 3.3

Suppose we perform an **outer** merge between today and genres on "artist_names". Furthermore, suppose that the only overlapping "artist_names" between today and genres are "drake" and "olivia rodrigo". What fraction of the rows in the merged DataFrame will contain "Pop" in the "genre" column? Give your answer as a simplified fraction.

| Suppose the DataFrame today consists of 15 rows — 3 rows for each of 5 different "artist_names". For each artist, it contains the "track_name" for their three most-streamed songs today. For instance, there may be one row for "olivia rodrigo" and "favorite crime", one row for "olivia rodrigo" and | | artist_names | genre |
|--|---|----------------|-------------|
| 2 glass animals Alternative Suppose the DataFrame today consists of 15 rows — 3 rows for each of 5 different "artist_names". For each artist, it contains the "track_name" for their three most-streamed songs today. For instance, there may be one row for "olivia rodrigo" and "favorite crime", one row for "olivia rodrigo" and | 0 | harry styles | Рор |
| Suppose the DataFrame today consists of 15 rows — 3 rows for each of 5 different "artist_names". For each artist, it contains the "track_name" for their three most-streamed songs today. For instance, there may be one row for "olivia rodrigo" and "favorite crime", one row for "olivia rodrigo" and | 1 | olivia rodrigo | Рор |
| three most-streamed songs today. For instance, there may be one row for "olivia rodrigo" and "favorite crime", one row for "olivia rodrigo" and | 2 | glass animals | Alternative |
| | 3 | drake | Hip-Hop/Rap |
| | 4 | doja cat | Hip-Hop/Rap |

genres

Missing Value Imputation - Caleb

The DataFrame random_10 contains the "track_name" and "genre" of 10 randomly-chosen songs in Spotify's Top 200 today, along with their "genre_rank", which is their rank in the Top 200 **among songs in their** "genre". For instance, "the real slim shady" is the 20th-ranked Hip-Hop/Rap song in the Top 200 today. random_10 is shown below in its entirety.

| | track_name | genre rank | genre |
|---|------------------------------|------------|-------------|
| 0 | good looking | 7.0 | Alternative |
| 1 | drowning (feat. kodak black) | NaN | Hip-Hop/Rap |
| 2 | the real slim shady | 20.0 | Hip-Hop/Rap |
| 3 | worldwide steppers | 2.0 | Hip-Hop/Rap |
| 4 | 2055 | 5.0 | Hip-Hop/Rap |
| 5 | drivers license | 9.0 | Рор |
| 6 | cinema | 2.0 | Рор |
| 7 | dos mil 16 | 4.0 | Рор |
| 8 | happier than ever | NaN | Рор |
| 9 | bam bam (feat. ed sheeran) | NaN | Рор |

The "genre_rank" column of random_10 contains missing values. Below, we provide four different imputed "genre_rank" columns, each of which was created using a different imputation technique. On the next page, match each of the four options to the imputation technique that was used in the option.

| Option A | | Op | tion B | Option C | | Option D | |
|------------|-------------|------------|-------------|------------|-------------|------------|-------------|
| genre rank | genre |
| 7.0 | Alternative | 7.0 | Alternative | 7.0 | Alternative | 7.0 | Alternative |
| 5.0 | Hip-Hop/Rap | 7.0 | Hip-Hop/Rap | 2.0 | Hip-Hop/Rap | 9.0 | Hip-Hop/Rap |
| 20.0 | Hip-Hop/Rap | 20.0 | Hip-Hop/Rap | 20.0 | Hip-Hop/Rap | 20.0 | Hip-Hop/Rap |
| 2.0 | Hip-Hop/Rap | 2.0 | Hip-Hop/Rap | 2.0 | Hip-Hop/Rap | 2.0 | Hip-Hop/Rap |
| 5.0 | Hip-Hop/Rap | 5.0 | Hip-Hop/Rap | 5.0 | Hip-Hop/Rap | 5.0 | Hip-Hop/Rap |
| 9.0 | Рор | 9.0 | Рор | 9.0 | Рор | 9.0 | Рор |
| 2.0 | Рор | 2.0 | Рор | 2.0 | Рор | 2.0 | Рор |
| 4.0 | Рор | 4.0 | Рор | 4.0 | Рор | 4.0 | Рор |
| 2.0 | Рор | 7.0 | Рор | 2.0 | Рор | 5.0 | Рор |
| 2.0 | Рор | 7.0 | Рор | 7.0 | Рор | 5.0 | Рор |

In which option was unconditional mean imputation used?

| Opt | Option A | | tion B | Option C | | Option D | |
|------------|-------------|------------|-------------|------------|-------------|------------|-------------|
| genre rank | genre |
| 7.0 | Alternative | 7.0 | Alternative | 7.0 | Alternative | 7.0 | Alternative |
| 5.0 | Hip-Hop/Rap | 7.0 | Hip-Hop/Rap | 2.0 | Hip-Hop/Rap | 9.0 | Hip-Hop/Rap |
| 20.0 | Hip-Hop/Rap | 20.0 | Hip-Hop/Rap | 20.0 | Hip-Hop/Rap | 20.0 | Hip-Hop/Rap |
| 2.0 | Hip-Hop/Rap | 2.0 | Hip-Hop/Rap | 2.0 | Hip-Hop/Rap | 2.0 | Hip-Hop/Rap |
| 5.0 | Hip-Hop/Rap | 5.0 | Hip-Hop/Rap | 5.0 | Hip-Hop/Rap | 5.0 | Hip-Hop/Rap |
| 9.0 | Рор | 9.0 | Рор | 9.0 | Рор | 9.0 | Рор |
| 2.0 | Рор | 2.0 | Рор | 2.0 | Рор | 2.0 | Рор |
| 4.0 | Рор | 4.0 | Рор | 4.0 | Рор | 4.0 | Рор |
| 2.0 | Рор | 7.0 | Рор | 2.0 | Рор | 5.0 | Рор |
| 2.0 | Рор | 7.0 | Рор | 7.0 | Рор | 5.0 | Рор |

| | track_name | genre rank | genre |
|---|------------------------------|------------|-------------|
| 0 | good looking | 7.0 | Alternative |
| 1 | drowning (feat. kodak black) | NaN | Hip-Hop/Rap |
| 2 | the real slim shady | 20.0 | Hip-Hop/Rap |
| 3 | worldwide steppers | 2.0 | Hip-Hop/Rap |
| 4 | 2055 | 5.0 | Hip-Hop/Rap |
| 5 | drivers license | 9.0 | Рор |
| 6 | cinema | 2.0 | Рор |
| 7 | dos mil 16 | 4.0 | Рор |
| 8 | happier than ever | NaN | Рор |
| 9 | bam bam (feat. ed sheeran) | NaN | Рор |

In which option was mean imputation conditional on "genre" used?

| Option A | | Op | otion B | | tion C | Option D | |
|------------|-------------|------------|-------------|------------|-------------|------------|-------------|
| genre rank | genre |
| 7.0 | Alternative | 7.0 | Alternative | 7.0 | Alternative | 7.0 | Alternative |
| 5.0 | Hip-Hop/Rap | 7.0 | Hip-Hop/Rap | 2.0 | Hip-Hop/Rap | 9.0 | Hip-Hop/Rap |
| 20.0 | Hip-Hop/Rap | 20.0 | Hip-Hop/Rap | 20.0 | Hip-Hop/Rap | 20.0 | Hip-Hop/Rap |
| 2.0 | Hip-Hop/Rap | 2.0 | Hip-Hop/Rap | 2.0 | Hip-Hop/Rap | 2.0 | Hip-Hop/Rap |
| 5.0 | Hip-Hop/Rap | 5.0 | Hip-Hop/Rap | 5.0 | Hip-Hop/Rap | 5.0 | Hip-Hop/Rap |
| 9.0 | Рор | 9.0 | Рор | 9.0 | Рор | 9.0 | Рор |
| 2.0 | Рор | 2.0 | Рор | 2.0 | Рор | 2.0 | Рор |
| 4.0 | Рор | 4.0 | Рор | 4.0 | Рор | 4.0 | Рор |
| 2.0 | Рор | 7.0 | Рор | 2.0 | Рор | 5.0 | Рор |
| 2.0 | Рор | 7.0 | Рор | 7.0 | Рор | 5.0 | Рор |

| | track_name | genre rank | genre |
|---|------------------------------|------------|-------------|
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| 4 | 2055 | 5.0 | Hip-Hop/Rap |
| 5 | drivers license | 9.0 | Рор |
| 6 | cinema | 2.0 | Рор |
| 7 | dos mil 16 | 4.0 | Рор |
| 8 | happier than ever | NaN | Рор |
| 9 | bam bam (feat. ed sheeran) | NaN | Рор |

In which option was unconditional probabilistic imputation used?

| Option A | | Op | otion B Opt | | tion C | Option D | |
|------------|-------------|------------|-------------|------------|-------------|------------|-------------|
| genre rank | genre |
| 7.0 | Alternative | 7.0 | Alternative | 7.0 | Alternative | 7.0 | Alternative |
| 5.0 | Hip-Hop/Rap | 7.0 | Hip-Hop/Rap | 2.0 | Hip-Hop/Rap | 9.0 | Hip-Hop/Rap |
| 20.0 | Hip-Hop/Rap | 20.0 | Hip-Hop/Rap | 20.0 | Hip-Hop/Rap | 20.0 | Hip-Hop/Rap |
| 2.0 | Hip-Hop/Rap | 2.0 | Hip-Hop/Rap | 2.0 | Hip-Hop/Rap | 2.0 | Hip-Hop/Rap |
| 5.0 | Hip-Hop/Rap | 5.0 | Hip-Hop/Rap | 5.0 | Hip-Hop/Rap | 5.0 | Hip-Hop/Rap |
| 9.0 | Рор | 9.0 | Рор | 9.0 | Рор | 9.0 | Рор |
| 2.0 | Рор | 2.0 | Рор | 2.0 | Рор | 2.0 | Рор |
| 4.0 | Рор | 4.0 | Рор | 4.0 | Рор | 4.0 | Рор |
| 2.0 | Рор | 7.0 | Рор | 2.0 | Рор | 5.0 | Рор |
| 2.0 | Рор | 7.0 | Рор | 7.0 | Рор | 5.0 | Рор |

| | track_name | genre rank | genre |
|---|------------------------------|------------|-------------|
| 0 | good looking | 7.0 | Alternative |
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| 5 | drivers license | 9.0 | Рор |
| 6 | cinema | 2.0 | Рор |
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In which option was probabilistic imputation conditional on "genre" used?

| Option A | | Option B | | Option C | | Option D | |
|------------|-------------|------------|-------------|------------|-------------|------------|-------------|
| genre rank | genre |
| 7.0 | Alternative | 7.0 | Alternative | 7.0 | Alternative | 7.0 | Alternative |
| 5.0 | Hip-Hop/Rap | 7.0 | Hip-Hop/Rap | 2.0 | Hip-Hop/Rap | 9.0 | Hip-Hop/Rap |
| 20.0 | Hip-Hop/Rap | 20.0 | Hip-Hop/Rap | 20.0 | Hip-Hop/Rap | 20.0 | Hip-Hop/Rap |
| 2.0 | Hip-Hop/Rap | 2.0 | Hip-Hop/Rap | 2.0 | Hip-Hop/Rap | 2.0 | Hip-Hop/Rap |
| 5.0 | Hip-Hop/Rap | 5.0 | Hip-Hop/Rap | 5.0 | Hip-Hop/Rap | 5.0 | Hip-Hop/Rap |
| 9.0 | Рор | 9.0 | Рор | 9.0 | Рор | 9.0 | Рор |
| 2.0 | Рор | 2.0 | Рор | 2.0 | Рор | 2.0 | Рор |
| 4.0 | Рор | 4.0 | Рор | 4.0 | Рор | 4.0 | Pop |
| 2.0 | Рор | 7.0 | Рор | 2.0 | Рор | 5.0 | Рор |
| 2.0 | Pop | 7.0 | Рор | 7.0 | Pop | 5.0 | Рор |

| | track_name | genre rank | genre |
|---|------------------------------|------------|-------------|
| 0 | good looking | 7.0 | Alternative |
| 1 | drowning (feat. kodak black) | NaN | Hip-Hop/Rap |
| 2 | the real slim shady | 20.0 | Hip-Hop/Rap |
| 3 | worldwide steppers | 2.0 | Hip-Hop/Rap |
| 4 | 2055 | 5.0 | Hip-Hop/Rap |
| 5 | drivers license | 9.0 | Рор |
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| 7 | dos mil 16 | 4.0 | Рор |
| 8 | happier than ever | NaN | Рор |
| 9 | bam bam (feat. ed sheeran) | NaN | Рор |

Regular Expressions - Angela

You want to use regular expressions to extract out the number of ounces from the 5 product names below.

| Index | Product Name | Expected Output |
|-------|---|------------------------|
| 0 | Adult Dog Food 18-Count, 3.5 oz Pouches | 3.5 |
| 1 | Gardetto's Snack Mix, 1.75 Ounce | 1.75 |
| 2 | Colgate Whitening Toothpaste, 3 oz Tube | 3 |
| 3 | Adult Dog Food, 13.2 oz. Cans 24 Pack | 13.2 |
| 4 | Keratin Hair Spray 2!6 oz | 6 |

The names are stored in a pandas Series called names. For each snippet below, select the indexes for all the product names that **will not** be matched correctly.

For the snippet below, which indexes correspond to products that will **not** be matched correctly?

```
regex = r'([\d.]+) oz'
names.str.findall(regex)
0
1
2
3
4
```

All names will be matched correctly.

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|-------|---|------------------------|
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| 1 | Gardetto's Snack Mix, 1.75 Ounce | 1.75 |
| 2 | Colgate Whitening Toothpaste, 3 oz Tube | 3 |
| 3 | Adult Dog Food, 13.2 oz. Cans 24 Pack | 13.2 |
| 4 | Keratin Hair Spray 2!6 oz | 6 |

The names are stored in a pandas Series called names. For each snippet below, select the indexes for all the product names that **will not** be matched correctly.

For the snippet below, which indexes correspond to products that will not be matched correctly?

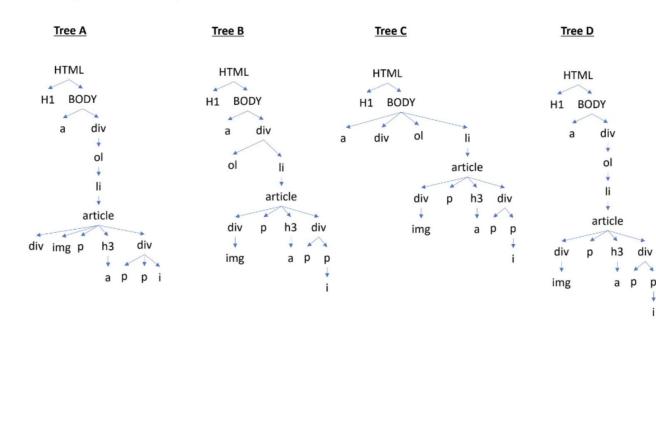
```
regex = r'(\d+?.\d+) oz|Ounce'
names.str.findall(regex)
0
1
```

- 2
- 3
- 4

All names will be matched correctly.

Web Scraping - Abhi

Which is the equivalent Document Object Model (DOM) tree of this HTML file?



<HTML>
<H1>The Book Club</H1>
<BOOY BGCOLOR="FFFFF">
Email us at
support@thebookclub.com.

<div> <article class="product_pod"> <div class="image_container"> </div> <h3>

A Light in the Attic </h3>

<div class="product_price">
 £51.77

 <i class="icon-ok"></i> In stock

</div>

</01>

</div>

.....

</BODY>

</HTML>

Problem 6.2

Rahul wants to extract the <code>finstock availability'</code> status of the book titled 'A Light in the Attic'. Which of the following expressions will evaluate to "In stock"? Assume that Rahul has already parsed the HTML into a BeautifulSoup object stored in the variable named soup.

Code Snippet A

```
soup.find('p',attrs = {'class': 'instock availability'})\
.get('icon-ok').strip()
```

Code Snippet B

```
soup.find('p',attrs = {'class': 'instock availability'}).text.strip()
```

Code Snippet C

```
soup.find('p',attrs = {'class': 'instock availability'}).find('i')\
.text.strip()
```

Code Snippet D

```
soup.find('div', attrs = {'class':'product_price'})\
.find('p',attrs = {'class': 'instock availability'})\
.find('i').text.strip()
```

<HTML>
<H1>The Book Club</H1>
<BODY BGCOLOR="FFFFF">
Email us at
support@thebookclub.com.

<div>

<article class="product_pod"> <div class="image_container">

</div>

<h3>

 A Light in the Attic </h3>

<div class="product_price">
 £51.77

 <i class="icon-ok"></i> In stock

</div>

</article>

145

</div>

</BODY>

</HTML>

Problem 6.3

Rahul also wants to extract the number of stars that the book titled 'A Light in the Attic' received. If you look at the HTML file, you will notice that the book received a star rating of three. Which code snippet will evaluate to "Three"?

Code Snippet A

soup.find('article').get('class').strip()

Code Snippet B

soup.find('p').text.split(' ')

Code Snippet C

soup.find('p').get('class')[1]

None of the above

<HTML>
<H1>The Book Club</H1>
<BODY BGCOLOR="FFFFF">
Email us at
support@thebookclub.com.

<div>

</div>

<h3>

 A Light in the Attic </h3>

<div class="product_price">
 £51.77

 <i class="icon-ok"></i> In stock

</div>

</article>

</div>

(0000)

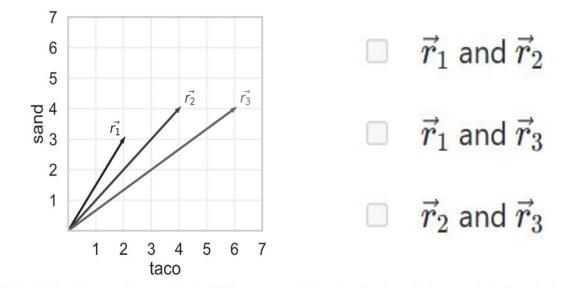
</BODY>

</HTML>

Text as Data - Abhi

Problem 7

Tahseen decides to look at reviews for the same hotel, but he modifies them so that the only terms they contain are "taco" and "sand". The bag-of-words representations of three reviews are shown as vectors below.



Using cosine similarity to measure similarity, which pair of reviews are the most similar? If there are multiple pairs of reviews that are most similar, select them all.

You create a table called gums that only contains the chewing gum purchases of df, then you create a bag-of-words matrix called bow from the name column of gums. The bow matrix is stored as a DataFrame shown below:

| | pur | gum | ••• | paperboard | 80 |
|----|-----|-----|-----|------------|----|
| 0 | 0 | 1 | | 0 | 1 |
| 1 | 0 | 1 | | 1 | 1 |
| | | | | | |
| 38 | 0 | 0 | | 0 | 0 |
| 39 | 0 | 0 | | 0 | 1 |

You also have the following outputs:

| >>> bow_df.s | um(axis=0) | >>> | bow_df.sum(axis=1) | <pre>>>> bow_df.loc[0, 'pur']</pre> |
|--------------|------------|------|--------------------|---|
| pur | 5 | 0 | 21 | 0 |
| gum | 41 | 1 | 22 | |
| sugar | 2 | 2 | 22 | <pre>>>> (bow_df['paperboard'] > 0).sum()</pre> |
| | •• | | | 20 |
| 90 | 4 | 37 | 22 | |
| paperboard | 22 | 38 | 10 | <pre>>>> bow_df['gum'].sum()</pre> |
| 80 | 20 | 39 | 17 | 41 |
| Length: 139 | | Leng | th: 40 | |

For each question below, write your answer as an unsimplified math expression (no need to simplify fractions or logarithms) in the space provided, or write "Need more information" if there is not enough information provided to answer the question.

Problem 8.1

What is the TF-IDF for the word "pur" in document 0?

| | pur | gum | •••• | paperboard | 80 | >>> bow_df.s | um(axis=0) | >>> | bow_df.sum(axis=1) | <pre>>>> bow_df.loc[0, 'pur']</pre> |
|-----|-----|-----|------|------------|----|--------------|------------|------|--------------------|---|
| 0 | 0 | 1 | | 0 | 1 | pur | 5 | 0 | 21 | 0 |
| | • | • | | | | gum | 41 | 1 | 22 | |
| 1 | 0 | 1 | | 1 | 1 | sugar | 2 | 2 | 22 | <pre>>>> (bow_df['paperboard'] > 0).sum()</pre> |
| | | | | | | | •• | | | 20 |
| ••• | | | | | | 90 | 4 | 37 | 22 | |
| 38 | 0 | 0 | | 0 | 0 | paperboard | 22 | 38 | 10 | <pre>>>> bow_df['gum'].sum()</pre> |
| 30 | U | 0 | | 0 | 0 | 80 | 20 | 39 | 17 | 41 |
| 39 | 0 | 0 | | 0 | 1 | Length: 139 | | Leng | th: 40 | |

Problem 8.2

What is the TF-IDF for the word "gum" in document 0?

| | pur | gum | •••• | paperboard | 80 | >>> bow_df.s | um(axis=0) | >>> | bow_df.sum(axis=1) | <pre>>>> bow_df.loc[0, 'pur']</pre> |
|----|-----|-----|------|------------|----|--------------|------------|-------------------|--------------------|---|
| 0 | 0 | 1 | | 0 | 1 | pur | 5 | 0 | 21 | 0 |
| - | • | • | | | • | gum | 41 | 1 | 22 | |
| 1 | 0 | 1 | | 1 | 1 | sugar | 2 | 2 | 22 | <pre>>>> (bow_df['paperboard'] > 0).sum()</pre> |
| | | | | | | | •• | | | 20 |
| | | | | | | 90 | 4 | 37 | 22 | |
| 38 | 0 | 0 | | 0 | 0 | paperboard | 22 | 38 | 10 | <pre>>>> bow_df['gum'].sum()</pre> |
| 50 | U | U | | U | U | 80 | 20 | 39 | 17 | 41 |
| 39 | 0 | 0 | | 0 | 1 | Length: 139 | | Leng ⁻ | th: 40 | |

Problem 8.3

What is the TF-IDF for the word "paperboard" in document 1?

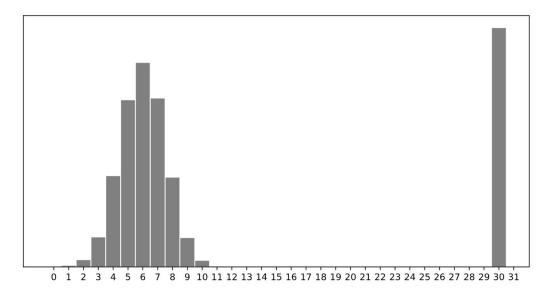
| | pur | gum | •••• | paperboard | 80 | >>> bow_df.s | um(axis=0) | >>> | bow_df.sum(axis=1) | <pre>>>> bow_df.loc[0, 'pur']</pre> |
|----|-----|-----|------|------------|-----|--------------|------------|------|--------------------|--|
| 0 | 0 | 1 | | 0 | 1 | pur | 5 | 0 | 21 | 0 |
| | - | • | | | | gum | 41 | 1 | 22 | |
| 1 | 0 | 1 | | 1 | 1 | sugar | 2 | 2 | 22 | <pre>>>> (bow_df['paperboard'] > 0).sum(</pre> |
| | | | | | | | •• | | | 20 |
| | ••• | | | | ••• | 90 | 4 | 37 | 22 | |
| 38 | 0 | 0 | | 0 | 0 | paperboard | 22 | 38 | 10 | >>> bow_df['gum'].sum() |
| 30 | U | 0 | | U | U | 80 | 20 | 39 | 17 | 41 |
| 39 | 0 | 0 | | 0 | 1 | Length: 139 | | Leng | th: 40 | |

Constant Model - Angela

Which of the following is closest to the constant prediction h^{st} that minimizes:

$$rac{1}{n}\sum_{i=1}^n egin{cases} 0 & y_i=h & \circ 7 \ 1 & y_i
eq h & \circ 11 \ 1 & y_i
eq h & \circ 15 \ \circ 30 \end{array}$$

Consider a dataset of n integers, $y_1, y_2, ..., y_n$, whose histogram is given below:



01

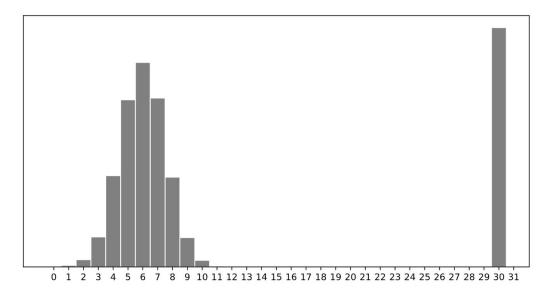
05

06

Which of the following is closest to the constant prediction h^st that minimizes:

$$rac{1}{n}\sum_{i=1}^n |y_i-h|$$

Consider a dataset of n integers, $y_1, y_2, ..., y_n$, whose histogram is given below:



01

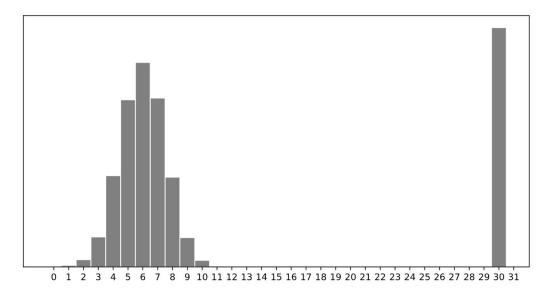
 $\odot 5$

06

Which of the following is closest to the constant prediction h^* that minimizes:

$$rac{1}{n}\sum_{i=1}^n(y_i-h)^2$$
 $\circ 11 \ \circ 15 \ \circ 30$

Consider a dataset of n integers, $y_1, y_2, ..., y_n$, whose histogram is given below:



01 05

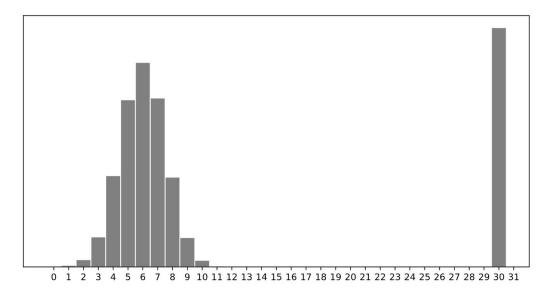
06

07

Which of the following is closest to the constant prediction h^* that minimizes:

$$\lim_{p o\infty}rac{1}{n}\sum_{i=1}^n|y_i-h|^p egin{array}{c} \circ {}^7\ \circ {}^{11}\ \circ {}^{15}\ \circ {}^{30} \end{array}$$

Consider a dataset of n integers, $y_1, y_2, ..., y_n$, whose histogram is given below:



01

○ 5 ○ 6

Regression - Angela

Problem 10

Consider a dataset that consists of y_1, \dots, y_n . In class, we used calculus to minimize mean squared error, $R_{sq}(h) = \frac{1}{n} \sum_{i=1}^n (h - y_i)^2$. In this problem, we want you to apply the same approach to a slightly different loss function defined below:

 $L_{ ext{midterm}}(y,h) = (lpha y - h)^2 + \lambda h$

Problem 10.1

Write down the empiricial risk $R_{
m midterm}(h)$ by using the above loss function.

The mean of dataset is \bar{y} , i.e. $\bar{y} = \frac{1}{n} \sum_{i=1}^{n} y_i$. Find h^* that minimizes $R_{ ext{midterm}}(h)$ using calculus. Your result should be in terms of \bar{y} , α and λ .

$$L_{
m midterm}(y,h) = (lpha y - h)^2 + \lambda h$$