

## ETL Project Writeup - Group 1

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**Extract:** For this project, we chose to extract data from Rotten Tomatoes, specifically a table of the Top 100 Movies of 2020. With this data, we ran these movies through the OMDB API to pull relevant information. We scraped the Rotten Tomatoes website using Pandas and created a table. We then used the OMDB API to create a more detailed table using the information from Rotten Tomatoes. This new table was exported into a CSV.

**Transform:** In order to clean our data, a variety of things were required. We first cleaned our data from Rotten Tomatoes. We removed parenthesis' from movie titles to give the titles more consistency and then removed the percent sign from ratings to make the column an integer. The rank column was a float due to the numbers being decimals so we converted that column into integers as well.

Next, we transformed the data collected from the OMDB API. We initialized the lists of columns we wanted to add to our cleaned Rotten Tomatoes table. We then created a loop that would parse the responses and append the information to our list of columns. If information was missing from the API, we used a Try/Except to move onto the next movie. Next, columns were renamed and data was exported to a CSV. The final thing we did to our API data was to go through columns we wanted to write queries for in pgAdmin 4. The columns we chose were Genre, Language, Actor, Country, Director, and Movies. From these columns, we created queries to match our ERD diagram and used pgAdmin 4 to extract specific insights we wanted from our data.

**Load:** We also created a database that we could each access using Google Console. We connected this database to pgAdmin and were able to create each table. We connected our Jupyter Notebook to our database and were able to load each data frame into its designated table. We then chose five queries to run on pgAdmin 4. Shown below are the outputs of the queries:

1) Top 10 Movies - we wanted to see the top 10 movies in 2020.

```
28 select
29     r.rank,
30     o.title
31 from
32 movies o
33     join ratings r on o.film_id = r.film_id
34 limit 10
```

|    | rank    | title                          |
|----|---------|--------------------------------|
|    | integer | character varying (100)        |
| 1  | 1       | Portrait of a Lady on Fire     |
| 2  | 2       | 1917                           |
| 3  | 3       | The Invisible Man              |
| 4  | 4       | Never Rarely Sometimes Alwa... |
| 5  | 5       | Hamilton                       |
| 6  | 6       | Da 5 Bloods                    |
| 7  | 7       | Sound of Metal                 |
| 8  | 8       | Nomadland                      |
| 9  | 9       | Miss Juneteenth                |
| 10 | 10      | Onward                         |

2) Average Rating by Genre - we looked for what genre had the highest rating based on the number of reviews. As seen below, Horror movies had the highest rating but the least number of reviews, which was an interesting insight.

```
1 select
2     round(avg(r.rating),2) as avg_rating,
3     round(avg(r.num_reviews),2) as avg_num_reviews,
4     e.genre
5 from
6 movies o
7     join ratings r on o.film_id = r.film_id
8     join film_genres f on o.film_id = f.film_id
9     join genres e on f.genre_id = e.genre_id
10 group by e.genre
11 order by
12     avg_rating desc
13
```

|   | avg_rating | avg_num_reviews | genre                   |
|---|------------|-----------------|-------------------------|
|   | numeric    | numeric         | character varying (100) |
| 1 | 99.00      | 63.00           | Horror                  |
| 2 | 98.17      | 66.96           | Documentary             |
| 3 | 96.18      | 133.03          | Drama                   |
| 4 | 95.53      | 104.24          | Comedy                  |
| 5 | 95.00      | 144.25          | Animation               |
| 6 | 93.00      | 99.00           | Crime                   |
| 7 | 92.67      | 163.67          | Adventure               |
| 8 | 89.67      | 212.33          | Action                  |

- 3) Count of Number of Films per Country - This showed us which country had the most movies shown for the top 100 2020 movies. The USA had the highest count by far.

```
14 select
15     count(i.country),
16     round(avg(r.rating),2) as avg_rating,
17     round(avg(r.num_reviews),2) as avg_num_reviews,
18     country
19 from
20 movies o
21     join film_countries f on o.film_id = f.film_id
22     join ratings r on o.film_id = r.film_id
23     join countries i on f.country_id = i.country_id
24 group by i.country
25 order by
26     count DESC
```

|   | count<br>bigint | avg_rating<br>numeric | avg_num_reviews<br>numeric | country<br>character varying (100) |
|---|-----------------|-----------------------|----------------------------|------------------------------------|
| 1 | 58              | 95.79                 | 120.36                     | USA                                |
| 2 | 12              | 94.67                 | 115.92                     | UK                                 |
| 3 | 3               | 95.00                 | 126.67                     | Australia                          |
| 4 | 3               | 97.33                 | 87.00                      | Ireland                            |
| 5 | 2               | 99.50                 | 71.50                      | South Korea                        |
| 6 | 2               | 94.50                 | 221.00                     | Canada                             |
| 7 | 2               | 98.00                 | 48.00                      | Germany                            |
| 8 | 1               | 97.00                 | 99.00                      | Norway                             |

- 4) Count of films per director with their average rating and number of reviews - This showed us which director had produced the most movies in the top 100 movies of 2020 and what the average rating for those movies was.

```
36 select
37     count(o.title),
38     d.director_name,
39     round(avg(r.rating),2) as avg_rating
40 from
41 movies o
42     join ratings r on o.film_id = r.film_id
43     join film_directors f on o.film_id = f.film_id
44     join directors d on f.director_id = d.director_id
45 group by d.director_name
46 order by
47     count desc
```

|   | count<br>bigint | director_name<br>character varying (100) | avg_rating<br>numeric |
|---|-----------------|--|-----------------------|
| 1 | 2               | Spike Lee                                | 94.50                 |
| 2 | 2               | Hlynur Palmason                          | 87.00                 |
| 3 | 1               | Andrew Ahn                               | 100.00                |
| 4 | 1               | Juliano Dornelles                        | 91.00                 |
| 5 | 1               | Channing Godfrey Peoples                 | 99.00                 |
| 6 | 1               | Jack Bender                              | 100.00                |
| 7 | 1               | Bridget Savage Cole                      | 98.00                 |
| 8 | 1               | Ki-duk Kim                               | 99.00                 |

- 5) Count of number of films by actor - Finally, we wanted to see the number of movies actors were in for the top 100 movies and if there were actors in multiple top movies.

```
49 select
50     count(o.title),
51     d.actor_name
52 from
53 movies o
54     join ratings r on o.film_id = r.film_id
55     join film_actors f on o.film_id = f.film_id
56     join actors d on f.actor_id = d.actor_id
57 group by d.actor_name
58 order by
59     count desc
60
```

| Data Output |        |                          |  | Explain | Messages | Notifications |
|-------------|--------|--------------------------|--|---------|----------|---------------|
|             | count  | actor_name               |  |         |          |               |
|             | bigint | character varying (100)  |  |         |          |               |
| 1           | 2      | Elisabeth Moss           |  |         |          |               |
| 2           | 2      | Lucas Jaye               |  |         |          |               |
| 3           | 1      | Walter Mercado           |  |         |          |               |
| 4           | 1      | Eliza Scanlen            |  |         |          |               |
| 5           | 1      | Folajomi 'FJ' Akinmurele |  |         |          |               |
| 6           | 1      | Julia Garner             |  |         |          |               |
| 7           | 1      | Shane McRae              |  |         |          |               |
| 8           | 1      | Karl Bertil-Nordland     |  |         |          |               |

As a bonus, we made a Flask App on VS Code to show our data as well.