

# CSE 460/560 - DMQL Project Final Report

## I. 1. PROJECT DETAILS

Name of your project, your team, and all team members, everyone's UB id(not the UB number);

**Project Name: Preview before Review Movies**

**Group members:**

**Ryan O'Sullivan - rposulli**

**Sakshi Singhal - ss666**

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## II. 2. PROBLEM STATEMENT [5 POINTS]:

Form a title and problem statement that clearly state the problem and questions you are trying to answer. Why do you need a database instead of an Excel file?

The title of the project is "Preview before Review Movies" and the problem statement is "Lack of prior knowledge about a movie exposes viewers with unexpected and even disappointing entertainment options." The questions that we can answer by using this dataset would be "I am feeling low. Which is the best movie to watch? I don't wanna waste my time like I did for the XYZ movie. It's been a long time since I haven't watched an action movie, let's find something." Because of its scalability, concurrent access, data integrity, security, and capacity for managing data relationships, databases will be a much better option than Excel files. They can handle massive volumes of data, preserve data consistency, and ensure data security because they have advanced querying, backup, and recovery capabilities. Additionally, they would reduce the redundancy that we would otherwise face if we combined several entity sets into a single table. Due to the limitations of Excel files in these areas, databases are a more reliable option for complicated data management requirements.

*A. Discuss the background of the problem leading to your objectives. Why is it a significant problem?*

At the moment, people rely heavily on online reviews and suggestions, which frequently leads to surprising and potentially disappointing movie-watching experiences. Moviegoing consumers want to make sure that they are watching something that is worth their time and money in the theater, and without resources like movie databases, it will be much harder for them to make an informed decision about which movies to attend. This is a significant problem because the absence of resources like ours will worsen the moviegoer's average experience, and it may convince them to visit the theater less often. This will result in less revenue for the theater and less enjoyment from movie fans.

*B. Explain the potential of your project to contribute to your problem domain. Discuss why this contribution is crucial.*

The project will provide moviegoers with a way of attaining information much more efficiently (as opposed to endless scrolling through reviews/suggestions on a variety of sites) and with more quality. Scrolling through random reviews means the user will have to ingest many unbiased and unqualified opinions of many different people who may expect and desire different things in a movie. Our contribution can resolve these issues because all of our data will be in one place (the database) and the data itself is significantly more unbiased because the data consists of reviews and ratings that come from regular users and actual critics, whose opinions are much more valuable than some random person who gave a bad rating because they were upset that they did not get to see a particular scene that they were expecting.

## III. TARGET USER [5 POINTS] :

Who will use your database? Who will administer the database? You are encouraged to give a real-life scenario;

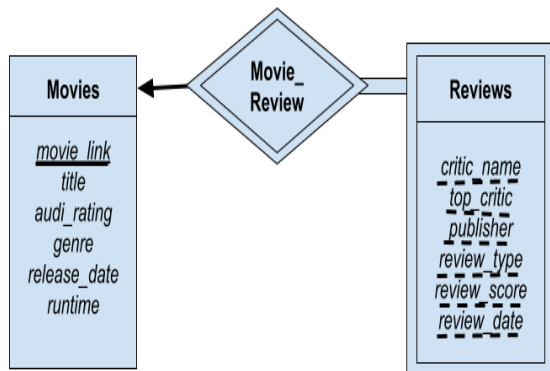
Anyone who wants to get a recommendation on a movie will be using our database. Suppose you are not someone who regularly sees movies, and you buy tickets to the Barbie movie without ever seeing a single review, just because of the branding of the Title. Imagine if the movie completely flopped because it was another corporate cash grab on a recognizable name. Imagine how disappointed you'd feel walking out of the theater. Wouldn't it have been so helpful to see the negative reviews about the movie? That way you could have made a more informed decision, like buying tickets to the Oppenheimer movie which was well-received. The people administering our database would be the popular theater businesses like Regal theaters, or alternatively, this could be presented by streaming services like Netflix or Hulu which have similar programs. They would get increased engagement because viewers would feel comfortable if they knew they were getting their money's worth before they subscribe, or continue their subscription to the service.

## IV. E/R DIAGRAM [10 POINTS]:

Draw an E/R diagram for your database and briefly describe the relationships between different tables. (Do not draw the figure by hand, you may use any tools to design or generate your E/R diagram)

Movies is a strong entity set wherein each tuple corresponds to one movie, and other relevant information about that particular movie. It is identified by the primary key movie\_link which is guaranteed to be unique for each movie on the site. Reviews is a weak entity set which depends on identifying set

Movies via Movie\_Review relationship, because each tuple is a movie review which logically depends on the existence of the movie being reviewed. This is also why all attributes of Reviews must be the partial key. One movie can receive many reviews, but each review can only be for one movie, so there is a one-to-many relationship. Additionally, every review must participate in Movie\_Review because it always has a parent movie, but not every movie must have a review, hence the partial participation arrow.



The only noteworthy FDs are:  $movie\_link \rightarrow movie\_link, title, audi\_rating, genre, release\_date, runtime,$

and  $(movie\_link, critic\_name, top\_critic, publisher, review\_type, review\_score, review\_date) \rightarrow (movie\_link, critic\_name, top\_critic, publisher, review\_type, review\_score, review\_date).$

For the Reviews table, because all attributes are needed to uniquely identify a tuple, there are no nontrivial functional dependencies here. Only the 2nd FD mentioned, the trivial one, holds. Thus, Reviews are in BCNF.

For the Movies table, the only related FD is that  $movie\_link \rightarrow movie\_link, title, audi\_rating, genre, release\_date, runtime.$   $movie\_link$  is the primary key/superkey, and none of the remaining attributes are in the LHS of an FD, so Movies is also in BCNF.

A. 4.1 Define a list of relations and their attributes.

The relation schemas for each relation is as follows: Movies(movie\_link, title, audi\_rating, genre, release\_date, runtime) Reviews(movie\_link, critic\_name, top\_critic, publisher, review\_type, review\_score, review\_date) Indicate the primary key and foreign keys (if any) for each relation. Justify your choice; For Movies, the primary key is movie\_link because each movie is hosted at a different URL on the site, and thus is guaranteed to be a unique identifier for each movie. For Reviews, a weak entity set, there is no single attribute primary key, or even an obvious single-attribute partial key. This is because reviews can have the same critic name, review the same movie, suggest the same score on the same date, etc. Because there are no true identifiers in the weak entity set, we must use ALL attributes as a primary key for reviews. This is common in weak entity sets without an obvious partial key. (movie\_link, critic\_name, top\_critic, publisher, review\_type, review\_score, review\_date)

## V. TASKS (3 AND 4) SHOULD BE COMPLETED [10 POINTS]:

(Task 3) Acquire the large "production" dataset, either by downloading it from a real data source or generating it using a program. Make sure the dataset fits your schema. You might need to write programs/scripts to transform them into a suitable form for loading into a database for real datasets. For program-generated datasets, make sure they contain interesting enough "links" across rows of different tables to show the results of different Advanced SQL queries learned in class.

**Done. See figures 1 and 2 at near the bottom of this report.**

(Task 4) You are required to make sure all of your relations are in Boyce- Codd Normal Form. Provide a list of dependencies for each relation. Decompose them if the tables are not in BCNF. If you decide to keep it in 3NF instead of BCNF, justify the decision for a particular relation. Your report for this milestone should contain a separate section with the details of the transformation from the initial schema to the final schema where the relations are in BCNF

B. 4.2:

Write a detailed description of each attribute (for each table), its purpose, and datatype;

Movies()		
ATTRIBUTE NAME	PURPOSE	DATATYPE
movie_link	URL extension to RT movie's webpage. Unique for each movie.	Varchar(MAX)
title	Title of the movie.	Varchar(MAX)
audi_rating	Audience score of the movie, from 0 to 100.	Integer
genre	Genre of the movie (Science Fiction, Action, Fantasy).	Varchar(MAX)
release_date	Release date of the movie in format yyyy-mm-dd.	Date
runtime	Movie runtime in minutes.	Integer

Reviews()		
ATTRIBUTE NAME	PURPOSE	DATATYPE
movie_link	URL extension to RT movie' webpage. Unique for each movie.	Varchar(MAX)
critic_name	Name of the critic.	Varchar(MAX)
top_critic	Boolean "True" or "False" indicating the critic's authority. (Would be mapped 1 or 0 for BIT)	Varchar(5) or BIT
publisher	Publishing company/organization of the critique.	Varchar(MAX)
review_type	Either "Fresh" or "Rotten", indicates insult or appraisal in the review.	Varchar(MAX)
review_score	Score fraction from 0 to 1. Often represented out of 5 but easily translatable to decimal.	Decimal(4,4)
review_date	Release date of the review in format yyyy-mm-dd.	Date

Movies()	
ATTRIBUTE NAME	DEFAULT VALUE OR NULL
movie_link	No default value. NULL is prohibited because this is a primary key.
title	No default value. Every movie in the set has, and must have, a title.
audi_rating	NULL; 0 cannot be default here because it would bring down average movie ratings unfairly.
genre	NULL; "" is also acceptable but NULL is preferred
release_date	NULL;
runtime	NULL;

Reviews()	
ATTRIBUTE NAME	DEFAULT VALUE OR NULL
movie_link	No default value. NULL is prohibited because this is the primary key of the identifying set.
critic_name	NULL; "" is acceptable but NULL is used in practice.
top_critic	"False". If not specified or new, assume the critic is not a top critic.
publisher	NULL; **In practice this doesn't because all of the critics in the set have publishers
review_type	NULL;
review_score	NULL;
review_date	NULL;

#### C. 4.3

Indicate each attribute's default value (if any) or if the attribute can be set to 'null';

#### D. 4.4

Explain the actions taken on any foreign key when the primary key (that the foreign key refer to) is deleted (e.g., no action, delete cascade, set null, set default)

When a particular movie\_link is deleted from Movies, you would have to remove the entire tuple because that attribute is the primary key of Movies. Thus, the only occurrence of this would be if the entire tuple were deleted from the identifying entity set. Even then, we would take "NO ACTION". Just because a movie is deleted from Movies, doesn't mean that the movie review information is useless. We want to keep the correlated tuples in Reviews for further analysis. Additionally if the related movie\_link value is inserted again later in Movies, we will still have the review data that goes along with it.

**This concludes Phase 2 of the Final Report.**

1 SELECT \* FROM reviews

Results

Messages

	movie_link	critic_name	top_critic	publisher	review_type	review_score	review_date
1	m/8814255	Andrew L. Urban	0	Urban Cinefile	Fresh	NULL	2018-02-06
2	m/8814255	Louise Keller	0	Urban Cinefile	Fresh	NULL	2018-02-06
3	m/8814255	NULL	0	FILMMK (Australia)	Fresh	NULL	2018-02-09
4	m/8814255	Ben McEachen	0	Sunday Mail (Australia)	Fresh	3.5/5	2018-02-09
5	m/8814255	Ethan Alter	1	Hollywood Reporter	Rotten	NULL	2018-02-10
6	m/8814255	David Germain	1	Associated Press	Rotten	NULL	2018-02-10

Fig. 1. Query sample from Reviews() table.

1 SELECT \* FROM movies

ResultsMessages

movie_link	title	audi_rating	genre	release_date	runtime
m/8814255	Percy Jackson & the Olympians: The Lightning Thief	53	Action & Adventure, Comedy, Drama, Science Fiction & Fantasy	2010-02-12	119
m/8778835	Please Give	64	Comedy	2010-06-28	98
m/18	18	53	Comedy, Romance	1979-10-05	122
m/180013-12_angry_men	12 Angry Men (Twelve Angry Men)	97	Classics, Drama	1957-04-13	95
m/1800079-28000_league	28,000 Leagues Under the Sea	74	Action & Adventure, Drama, Kids & Family	1954-01-01	127
m/18000_3c	16,000 B.C.	37	Action & Adventure, Classics, Drama	2000-09-07	109

Fig. 2. Query sample from Movies() table.

## VI. TASK 5: PROBLEMS AND INDEXING CONCEPTS

”Do you specifically run into any problems while handling the larger dataset? Did you try to adopt some indexing concepts to resolve this? Briefly describe the questions you faced and how you solved them.”

Even on our larger dataset, we did not run into any problems handling it using PostgreSQL. However, to demonstrate our understanding, we will discuss some common indexing methods used to handle large datasets. On ordered files like our Movies() table, which is ordered by movie\_link, it would be implemented as a primary index, and NOT a clustered index. This is because records close in the index would also be close in data. Within this primary index, we can choose either a dense index model or a sparse index model. The main difference is that dense indices contain a record for every search-key value in the data file (in our case, an index record for every data record). However, sparse indexes have an advantage because they only include one index key per data block. Because of this and the fact that the data file is only sorted by the search key, we would prefer a sparse index for Movies(). The Reviews() table is a weak entity set and while it is ordered with respect to movie\_link(), it is UNordered with respect to critic\_name, which is a common search key for movie reviews. Because it’s sorted on one of these dimensions, a secondary index would not be necessary. In this case, we would want to use a dense clustered index (variation of a primary index) in order to manage this large dataset. Dense clustered indexes are ideal for this case because the Reviews() table has many possible search keys, including movie\_link, critic\_name and publisher.

## VII. TASK 6: TESTING THE DATABASE ON SQL QUERIES

”Test your database with more than 10 SQL queries. You are supposed to design 1 or 2 queries for each inserting, deleting, and updating operation in your dataset.”

For demonstration, we have performed over ten sample queries on the database, and we have displayed their results

below. Figure 3 inserts a new movie ”Kuch Kuch Hota Hai” into the Movies() table, Figure 4 updates the genre and runtime of the newly inserted tuple, and Figure 5 deletes this movie from the Movies() table entirely. Similarly, Figures 6, 7 and 8 are an insertion, update and deletion of a review for the movie ”Kuch Kuch Hota Hai.” In this case, the update changed the review\_score and publisher name to ’Buffalo News’. In figures Figure 9 and 10, we performed the inner join on the Movies() and Reviews() table.

In figures 11 we first order the movies on their titles from Z-A (descending order). Then in figure 12, we ordered the movies in ascending order by genre.

We performed some similar queries for the Reviews() table. In figure 13, we ordered the reviews by the last name of their critics, from Z to A. Then in reference 14, we reordered the table alphabetically by release date.

Finally, we performed three grouping queries. In figure 15, we count the number of movies with the genre ”Comedy” and find it to be 1263 total movies. Figure 16 is more general and shows us the movie count for every genre. Lastly, in figure 17, we group by movie\_links to get the total number of reviews for each movie.

SQLQuery\_2 - localh...stgres)

SQLQuery\_3 - localh...stgres)

SQLQuery\_1 - localh...stgres)

dbo.movies

dbo.reviews

Run

Cancel

Disconnect

Change

Database: postgres

```
1  --insert into movies
2
3  -- INSERT INTO movies (movie_link, title, audi_rating, genre, release_date, runtime)
4  -- VALUES ('m/Kuch Kuch Hota Hai', 'Kuch Kuch Hota Hai', 49, 'Romance, Comedy & Drama', '2005-04-01', 217);
5
6  select * from movies
7  WHERE title='Kuch Kuch Hota Hai'
8
9
```

Results	Messages					
movie_link	title	audi_rating	genre	release_date	runtime	
1	m/Kuch Kuch Hota Hai	Kuch Kuch Hota Hai	49	Romance, Comedy & Drama	2005-04-01	217

Fig. 3. Insertion of a tuple in the Movies() table.

SQLQuery\_2 - localh...stgres)

SQLQuery\_3 - localh...stgres)

SQLQuery\_1 - localh...stgres)

dbo.movies

dbo.reviews

Run

Cancel

Disconnect

Change

Database: postgres

```
8      --UPDATE into movies
9
10
11     -- UPDATE movies
12     -- SET runtime=90, genre = 'Comedy, In search for love'
13     -- WHERE title = 'Kuch Kuch Hota Hai';
14
15     select * from movies
16     WHERE title='Kuch Kuch Hota Hai'
17
```

Results Messages

movie_link	title	audi_rating	genre	release_date	runtime	
1	m/Kuch Kuch Hota Hai	Kuch Kuch Hota Hai	49	Comedy, In search for love	2005-04-01	90

Fig. 4. Updating in the Movies() table

```

17
18 --Delete from movies
19
20 -- DELETE FROM movies
21 -- WHERE title = 'Kuch Kuch Hota Hai';
22
23 select * from movies
24 WHERE title='Kuch Kuch Hota Hai'
25
Results Messages
movie_link title audi_rating genre release_date runtime

```

Fig. 5. Deletion of a tuple from the Movies() table

```

29
30 --INSERT INTO REVIEWS
31
32 -- INSERT INTO reviews (movie_link, critic_name, top_critic, publisher, review_type, review_score, review_date)
33 -- VALUES ('m/Aladin', 'Ryan', TRUE, 'Times Square', 'Rotten', '3.5', '2021-04-05');
34
35 Select * from reviews
36 WHERE critic_name='Ryan'
37
Results Messages
movie_link critic_name top_critic publisher review_type review_score review_date
1 m/Aladin Ryan True Times Square Rotten 3.5 2021-04-05

```

Fig. 6. Insertion of a tuple in the Reviews() table

```

38
39 --UPDATE IN REVIEWS
40
41 -- UPDATE reviews
42 -- SET review_score=5, publisher = 'Buffalo News'
43 -- WHERE review_date = '2021-04-05';
44
45 Select * from reviews
46 WHERE critic_name='Ryan'
47
Results Messages
movie_link critic_name top_critic publisher review_type review_score review_date
1 m/Aladin Ryan True Buffalo News Rotten 5 2021-04-05

```

Fig. 7. Updating in Reviews() table

```

38
39 --Delete in reviews
40
41 -- DELETE FROM reviews
42 -- WHERE publisher='Buffalo News' and critic_name='Ryan'
43
44 Select * from reviews
45 WHERE critic_name='Ryan'
46
Results Messages
movie_link critic_name top_critic publisher review_type review_score review_date

```

Fig. 8. Deletion of a tuple from the Review() table

```

46
47 --JOIN TWO TABLES
48
49 SELECT *
50 FROM movies
51 INNER JOIN reviews ON movies.movie_link = reviews.movie_link;
52
Results Messages
movie_link title audi_rating genre release_date runtime movie_link
1 m/0814255 Percy Jackson & the Olympians: The Lightning Thief 53 Action & Adventure, Comedy, Drama, Science Fiction & F... 2010-02-12 119 m/0814255
2 m/0814255 Percy Jackson & the Olympians: The Lightning Thief 53 Action & Adventure, Comedy, Drama, Science Fiction & F... 2010-02-12 119 m/0814255
3 m/0814255 Percy Jackson & the Olympians: The Lightning Thief 53 Action & Adventure, Comedy, Drama, Science Fiction & F... 2010-02-12 119 m/0814255
4 m/0814255 Percy Jackson & the Olympians: The Lightning Thief 53 Action & Adventure, Comedy, Drama, Science Fiction & F... 2010-02-12 119 m/0814255
5 m/0814255 Percy Jackson & the Olympians: The Lightning Thief 53 Action & Adventure, Comedy, Drama, Science Fiction & F... 2010-02-12 119 m/0814255
6 m/0814255 Percy Jackson & the Olympians: The Lightning Thief 53 Action & Adventure, Comedy, Drama, Science Fiction & F... 2010-02-12 119 m/0814255
7 m/0814255 Percy Jackson & the Olympians: The Lightning Thief 53 Action & Adventure, Comedy, Drama, Science Fiction & F... 2010-02-12 119 m/0814255
8 m/0814255 Percy Jackson & the Olympians: The Lightning Thief 53 Action & Adventure, Comedy, Drama, Science Fiction & F... 2010-02-12 119 m/0814255
9 m/0814255 Percy Jackson & the Olympians: The Lightning Thief 53 Action & Adventure, Comedy, Drama, Science Fiction & F... 2010-02-12 119 m/0814255
10 m/0814255 Percy Jackson & the Olympians: The Lightning Thief 53 Action & Adventure, Comedy, Drama, Science Fiction & F... 2010-02-12 119 m/0814255

```

Fig. 9. Joining the Movies() and Reviews() tables.

```

53
54 --JOIN TWO TABLES
55
56 SELECT *
57 FROM movies
58 INNER JOIN reviews ON movies.movie_link = reviews.movie_link;
59
Results Messages
release_date runtime movie_link critic_name top_critic publisher review_type review_score review_date
Science Fiction & F... 2010-02-12 119 m/0814255 Andrew L. Urban False Urban Cinefile Fresh NULL 2010-02-06
Science Fiction & F... 2010-02-12 119 m/0814255 Louise Keller False Urban Cinefile Fresh NULL 2010-02-06
Science Fiction & F... 2010-02-12 119 m/0814255 J101010 (Australia) Fresh NULL 2010-02-09
Science Fiction & F... 2010-02-12 119 m/0814255 Ben McEneaney False Sunday Mail (Australia) Fresh 3.5/5 2010-02-09
Science Fiction & F... 2010-02-12 119 m/0814255 Ethan Aker True Hollywood Reporter Rotten NULL 2010-02-10
Science Fiction & F... 2010-02-12 119 m/0814255 David Germain True Associated Press Rotten NULL 2010-02-10
Science Fiction & F... 2010-02-12 119 m/0814255 Nick Schager False Slant Magazine Rotten 4-Jan 2010-02-10
Science Fiction & F... 2010-02-12 119 m/0814255 Bill Goodfellow True Arizona Republic Fresh 3.5/5 2010-02-10
Science Fiction & F... 2010-02-12 119 m/0814255 Jordan Hoffman False USA Fresh n 2010-02-10

```

Fig. 10. Joining the Movies() and Reviews() tables, continued.

```

54
55 --ORDER BY DESC
56
57 SELECT *
58 FROM movies
59 ORDER BY title DESC;
60
Results Messages
movie_link title audi_rating genre
1 m/011014_uhigun_moments_no_tsubasa 011014_uhigun Moments no Tsubasa (Royal Space Forc... 88 Animation, Anime & Manga, Drama, Science Fiction & F...
2 m/011014_uhigun_moments_no_tsubasa 011014_uhigun The Merton Downey Jr. Movie 84 Documentary, Special Interest
3 m/011014_uhigun_moments_no_tsubasa 011014_uhigun (Tie Me Up! Tie Me Down!) 79 Art House & International, Comedy, Drama, Romance
4 m/011014_uhigun_moments_no_tsubasa 011014_uhigun State of the Union 32 Action & Adventure
5 m/011014_uhigun_moments_no_tsubasa 011014_uhigun Return of Xander Xage 37 Action & Adventure, Drama
6 m/011014_uhigun_moments_no_tsubasa 011014_uhigun sleep furiously 67 Documentary, Special Interest
7 m/011014_uhigun_moments_no_tsubasa 011014_uhigun neither 51 Drama, Horror, Mystery & Suspense
8 m/011014_uhigun_moments_no_tsubasa 011014_uhigun IBoy 39 Action & Adventure, Drama, Mystery & Suspense, Sci...
9 m/011014_uhigun_moments_no_tsubasa 011014_uhigun Iboy 58 Drama
10 m/011014_uhigun_moments_no_tsubasa 011014_uhigun Tommy Chong 58 Documentary, Special Interest
11 m/011014_uhigun_moments_no_tsubasa 011014_uhigun Rec 2 68 Drama, Horror, Mystery & Suspense, Special Interest
12 m/011014_uhigun_moments_no_tsubasa 011014_uhigun Rec 2 82 Art House & International, Horror, Special Interest
13 m/011014_uhigun_moments_no_tsubasa 011014_uhigun Rec 4: Apocalypse 34 Art House & International, Drama, Horror
14 m/011014_uhigun_moments_no_tsubasa 011014_uhigun Rec 3 Genesis 27 Art House & International, Drama, Horror, Mystery & ...
15 m/011014_uhigun_moments_no_tsubasa 011014_uhigun Zulu Dawn 62 Action & Adventure, Art House & International, Drama
16 m/011014_uhigun_moments_no_tsubasa 011014_uhigun Zulu 91 Classics, Drama
17 m/011014_uhigun_moments_no_tsubasa 011014_uhigun Zorba the Greek 88 Action & Adventure, Art House & International, Class...

```

Fig. 11. Ordering in Movies() table in descending order.

## VIII. TASK 7: PROBLEMATIC QUERIES AND PERFORMANCE IMPROVEMENTS

"Query execution analysis: identify three problematic queries (show their cost), where the performance can be improved. Provide a detailed execution plan (you may use EXPLAIN in PostgreSQL) on how you plan to improve these queries."

Our first problematic query, shown in figure 18, is a simple query with a glaring inefficiency: We join Reviews() and Tables(), but we only look for attributes from Reviews()! In a database with a small schema, this may be obvious, but in larger schemas it is important to only join tables which would provide you with the desired attributes. This strategy of "only necessary joins" is how we would improve

```

56 -- SELECT *
57 -- FROM movies
58 -- ORDER BY title DESC;
59
60 --ORDER BY ASC
61
62 SELECT *
63 FROM movies
64 order by genre

```

movie_link	title	audi_rating	genre	release_date	runtime
1	m/rush_hour	78	Action	1998-09-18	98
2	m/lethal_weapon_4	64	Action & Adventure	1998-07-10	127
3	m/wolf_totem	58	Action & Adventure	2015-09-11	118
4	m/lethal_weapon_3	61	Action & Adventure	1992-05-15	118
5	m/machete_kills	35	Action & Adventure	2013-10-11	107
6	m/lucy_2014	47	Action & Adventure	2014-07-25	88
7	m/linheart_1998	52	Action & Adventure	1991-01-11	105
8	m/bad_ass_2012	39	Action & Adventure	2012-04-13	105
9	m/london_has_fallen	51	Action & Adventure	2016-03-04	99
10	m/death_wish_v_the_face_of_death	32	Action & Adventure	1993-08-01	95
11	m/death_wish_ii	44	Action & Adventure	1982-02-19	93
12	m/death_wish_4_the_crackdown	48	Action & Adventure	1987-11-06	99
13	m/death_wish_3	54	Action & Adventure	1985-11-01	90
14	m/1862483-bad_boys	78	Action & Adventure	1995-05-19	118

Fig. 12. Ordering in Movies() table in ascending order.

```

65 -- order by desc
66
67 select * from reviews
68 order by critic_name desc

```

movie_link	critic_name	top_critic	publisher	review_type	review_score	review_date
17649	m/black_christmas_2019	Zofia Wjaskna	File Inquiry	Fresh	NULL	2020-05-04
17650	m/live_a_boy	Zofia Wjaskna	File Inquiry	Fresh	NULL	2020-05-04
17651	m/batman_the_jay_ii	Zofia Wjaskna	File Inquiry	Rotten	NULL	2020-05-04
17652	m/phantom_thread	Zoe Rose Smith	Bust Magazine	Fresh	NULL	2018-02-06
17653	m/volody	Zoe Rose Smith	JumpCut Online	Fresh	5-Apr	2019-08-10
17654	m/daniel_jant_real	Zoe Rose Smith	JumpCut Online	Fresh	5-Apr	2020-03-29
17655	m/knives_and_skin	Zoe Rose Smith	JumpCut Online	Rotten	5-Jan	2019-09-14
17656	m/renter_2017	Zoe Rose Smith	Zoho With A Shotgun	Fresh	5-May	2019-09-14
17657	m/the_wild_2017	Zoe Rose Smith	Zoho With A Shotgun	Fresh	4.5/5	2019-09-14
17658	m/the_wretched_2020	Zoe Rose Smith	JumpCut Online	Rotten	5-Feb	2020-05-08

Fig. 13. Ordering in Reviews() table in descending order.

```

72 select * from reviews
73 order by review_date
74
75

```

movie_link	critic_name	top_critic	publisher	review_type	review_score	review_date
1	m/a_sumer_in_genoa	Thomas Caldwell	Cinema Autopsy	Fresh	4.5/5	1000-01-01
2	m/4321	NULL	Empire Magazine	Fresh	5-Apr	1000-01-01
3	m/4321	Jennie Kermode	Eye for Film	Fresh	3.5/5	1000-01-01
4	m/a_sumer_in_genoa	Coco Forsythe	Future Movies UK	Fresh	10-Jul	1000-01-01
5	m/dick_jason_2006	Dave White	Reviews.com	Fresh	A	1000-01-01
6	m/losing_in_emergency_stories_of_doctors_without_bor...	John P. McCarthy	Heartflic Magazine	Fresh	5-Apr	1000-01-01
7	m/was_better_blue	NULL	Entertainment Weekly	Fresh	B-	1000-01-01
8	m/sup_2018	Matthew Turner	VindLondon	Rotten	5-Feb	1000-01-01

Fig. 14. Ordering in Reviews() table in ascending order.

on queries like this in large databases. Otherwise, as shown in the EXPLAIN function, you will incur extra time costs from the join operation.

The second problematic query we present is shown in figure 19. In this case, we join Reviews() and Movies(), only to later apply a condition on the release date of the movie in question. This is problematic because the joined table will have more total tuples than Movies(), and will require a longer scan than if we had just applied the release date condition to the Movies() table BEFORE joining it, and this inefficiency is reflected in the JOIN cost in the figure. In bigger databases it is important to apply conditions before joins when possible, in order to shrink the tables and lower the cost of the join. Applying conditions before joins is another useful data strategy to manage queries like this one.

The screenshot displays the Azure Data Studio application window. The title bar at the top includes the application icon and the text "Azure Data Studio". The menu bar contains "File", "Edit", "View", "Window", and "Help". The main editor area shows a SQL query with line numbers 75 through 82. The query is as follows:

```
75
76 --group by
77
78 SELECT genre, COUNT(*) AS movie_count
79 FROM movies
80 WHERE genre = 'Comedy'
81 GROUP BY genre;
82 |
```

Below the editor, the "Results" tab is active, showing a table with two columns: "genre" and "movie\_count". The table contains one row of data:

	genre	movie_count
1	Comedy	1263

Fig. 15. Performing GroupBy operation to get the total count of Comedy Movies in the Movies() table.

SQLQuery\_2 - localh...stgres) • SQLQuery\_3 - localh...stgres) • SQLQuery\_1 - lc

Run Cancel Disconnect Change Database: postgres

```
75
76 --group by
77
78 SELECT genre, COUNT(*) AS movie_count
79 FROM movies
80 GROUP BY genre;
81
```

Results Messages

	genre	movie_count
1	NULL	19
2	Action & Adventure, Horror	31
3	Action & Adventure, Classics, Cult Movies, Horror, S...	1
4	Animation, Comedy, Kids & Family, Musical & Perform...	1
5	Documentary, Science Fiction & Fantasy, Special Inte...	1
6	Action & Adventure, Comedy, Drama, Kids & Family, Sc...	1
7	Animation, Comedy, Musical & Performing Arts	1
8	Action & Adventure, Art House & International	37

Fig. 16. Performing GroupBy operation in the Movies() table, grouping movies by genre.

Our third and final query, presented in figure 20, has a number of problematic components that increase the total execution time of the process. For one thing, including the phrase "HAVING COUNT > 0" is redundant because if a movie exists at all, it will have a title since we're grouping by movie link. Additionally, we scan on two conditions in a nested select statement instead of just applying them together. This is cumbersome and leads to many of the inefficiencies you see in the explanation table associated with the figure. In this case, the query is problematic because it includes redundant clauses and unnecessary nesting which bloats the runtime of the query. To combat this, we would modify the query to put both WHERE clauses in a single SELECT statement, and

movie_link	movie_link_count
m/accidental_love	35
m/return_of_swamp_thing	18
m/amour_fou	51
m/priceless_2016	10
m/das_finistere_tal	6
m/table_19	117
m/10009929-a_christmas_tale	132
m/the_other_side_of_the_wind	98
m/the_last_mountain	42
m/sleepy_hollow	244
m/mark_felt_the_man_who_brought_down_the_white_house	113
m/7_guardians_of_the_tomb	15
m/september_issue	228

Fig. 17. Performing GroupBy operation using to acquire the total number of movie links for each movie.

remove the HAVING clause entirely. In general you want to simplify overcomplicated queries to prevent running excessive scans, and this is especially true for larger databases where each scan takes much longer.

**This concludes Phase 2 of the Final Report.**

QUERY PLAN
1 Hash Join (cost=642.52..26331.44 rows=1848575 width=46)
2 Hash Cond: ((reviews.movie_link)::text = (movies.movie_link)::text)
3 -> Seq Scan on reviews (cost=0.00..22935.75 rows=1848575 width=65)
4 -> Hash (cost=421.12..421.12 rows=17712 width=19)
5 -> Seq Scan on movies (cost=0.00..421.12 rows=17712 width=19)

Fig. 18. Problem Query 1: It is inefficient to join several tables when you're only searching for values in one of them.

**References :- [Link to Rotten Tomatoes Dataset](#)**

QUERY PLAN
1 Hash Join (cost=489.19..26178.10 rows=112660 width=143)
2 Hash Cond: ((reviews.movie_link)::text = (movies.movie_link)::text)
3 -> Seq Scan on reviews (cost=0.00..22935.75 rows=1848575 width=65)
4 -> Hash (cost=465.40..465.40 rows=1903 width=78)
5 -> Seq Scan on movies (cost=0.00..465.40 rows=1903 width=78)
6 Filter: (release_date > '2020-01-01'::date)

Fig. 19. Problem Query 2: It is inefficient to join before applying a condition; We should apply the condition to lower the amount of tuples before joining.

QUERY PLAN
1 Hash Join (cost=25649.83..26117.46 rows=5515 width=78)
2 Hash Cond: ((movies.movie_link)::text = (movies_1.movie_link)::text)
3 -> Seq Scan on movies (cost=0.00..421.12 rows=17712 width=78)
4 -> Hash (cost=25580.89..25580.89 rows=5515 width=19)
5 -> Finalize HashAggregate (cost=25318.93..25525.74 rows=5515 width=19)
6 Group Key: movies_1.movie_link
7 Filter: (count(movies_1.title) > 0)
8 -> Gather (cost=21679.83..25153.48 rows=33090 width=27)
9 Workers Planned: 2
10 -> Partial HashAggregate (cost=20679.83..20844.48 rows=16545 width=27)
11 Group Key: movies_1.movie_link
12 -> Hash Join (cost=672.21..18638.43 rows=408120 width=37)
13 Hash Cond: ((reviews.movie_link)::text = (movies_1.movie_link)::text)
14 -> Parallel Seq Scan on reviews (cost=0.00..16819.06 rows=436906 width=19)
15 -> Hash (cost=465.40..465.40 rows=16545 width=37)
16 -> Seq Scan on movies movies_1 (cost=0.00..465.40 rows=16545 width=37)
17 Filter: (release_date > '2020-01-01'::date)

Fig. 20. Problem Query 3: There are several inefficiencies here; The conditions are separated and the addition of a COUNT call should be optimized out.