

RentalSight: An Online Movie Rental Warehouse

Introduction

In the dynamic landscape of online movie rentals, efficient data management is pivotal for success. Introducing Rentalsight, a comprehensive data warehouse crafted to empower online movie rental stores with insightful analytics, seamless inventory management, and strategic rental oversight.

Deliverables

Revenue Trend Analysis:

- Explore revenue trends over specified time periods.
- Visualize revenue patterns based on movie genres, or ratings.
- Identify peak revenue periods and potential areas for revenue growth.

Inventory Management:

- Track inventory levels in real-time.
- Identify slow-moving or obsolete inventory for strategic decision-making.
- Measure the efficiency of inventory management and return policy

Rental Oversight:

- Analyze rental patterns.
- Optimize rental pricing based on demand and historical data.

User-Friendly Interface:

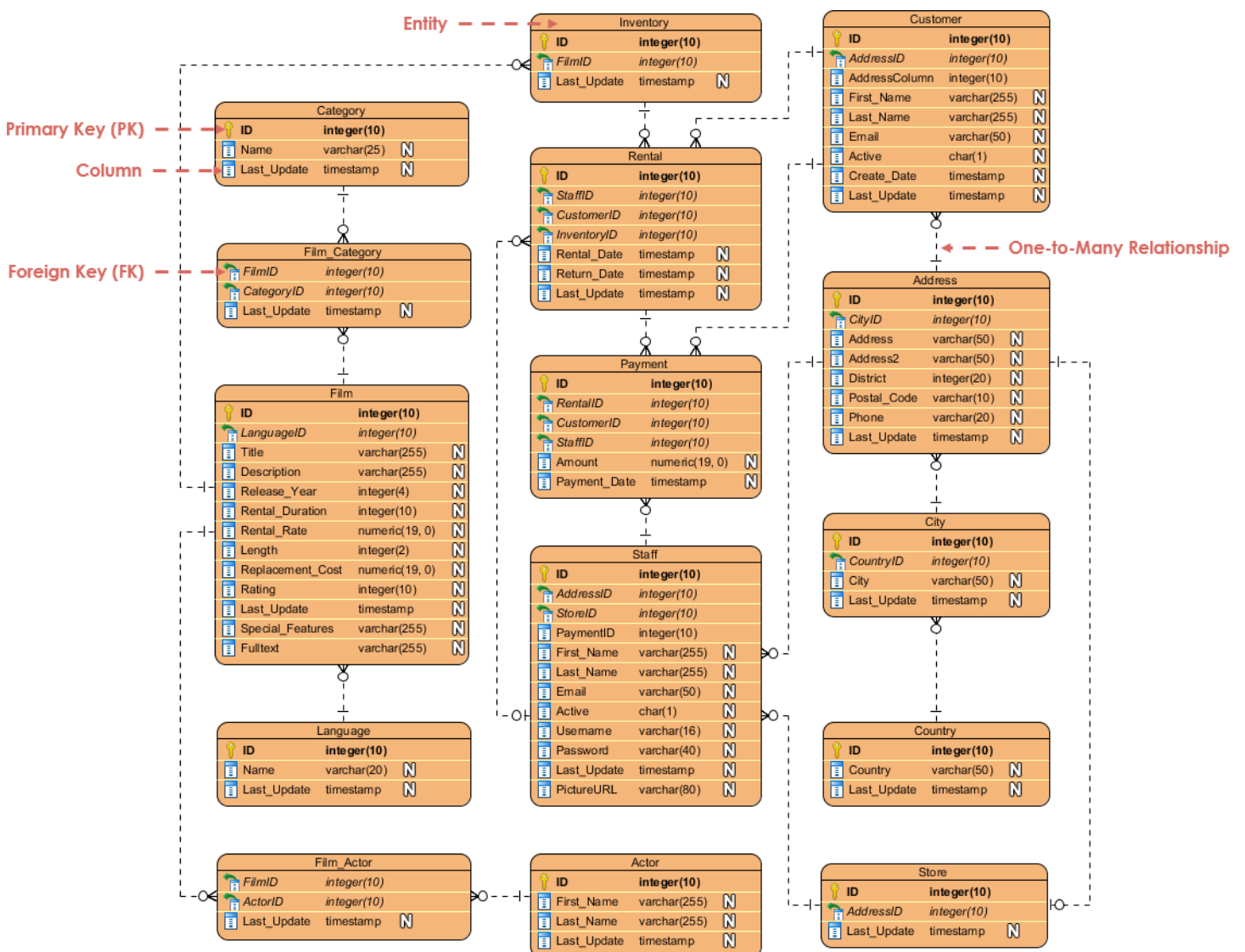
- Intuitive dashboards for easy navigation and quick access to key metrics.
- Customizable reports to tailor insights to specific business requirements.
- Real-time data updates for timely decision-making.

ETL phase

Extraction:

Data was extracted from an online dataset comprised of many individual tables in csv and JSON formats.

Original data can be found in source_data folder. The schema for the original database can be seen here



Transformation:

Data is then transformed using Python and the Pandas library.

The following files were used to individually filter out and transform certain data and save them into csv files located in the main folder:

- [transform_address.py](#): This script reads 3 data files in CSV format as pandas dataframes. It then joins the addresses with the corresponding countries using country_id. “last_update”, “phone”, “address2” are not needed therefore they are discarded from the dataframe. It then joins addresses with cities using city_id and fills in two missing postal codes for “Alberta” and “QLD”. it then saves the output to address.csv
- [transform_film.py](#): reads 4 CSV files, proceeds to drop unneeded columns then joins each film to the corresponding language and then to the correct genre (film_category). It then removes rental_rate column and saves the file.
- [transform_customer.csv](#): reads addresses and customer information. Removes unneeded columns and joins customers to their addresses. formats the strings in “first_name”, “last_name”, and “email”
- [transform_rental.csv](#): reads in rental information, removes unneeded columns, formats rental_date columns, saves day, month, and year information in separate columns. It then establishes primary keys for day_df, month_df, year_df. Calendar library is then imported to encode the month name in the month dataframe.
- [transform_store.py](#): reads in store information and drops many unneeded columns. It then gives a name to the two stores and saves the file.
- The files [create_inventory_snapshot.py](#) and [create_rental_transactions.py](#) are used to create the two fact tables that will be employed in our analysis later on. They use
- the [ETL.ipynb](#) is a jupyter notebook created to execute all transformations, yielding the following csv files found in the main folder: [address.csv](#), [customer.csv](#), [store.csv](#), [film.csv](#), [rental.csv](#), [day.csv](#), [year.csv](#), [month.csv](#), [rental_transactions.csv](#), [inventory_snapshot.csv](#)

Loading:

Data is then loaded onto a snowflake ROLAP database.

This database can be accessed through this link:

<https://vn58684.west-europe.azure.snowflakecomputing.com>

username: slimboyfu

password: Testingtesting1\$

MM mahdi messaai ACCOUNTADMIN

Worksheets

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Search

MOVIE_RENTAL_STORE

INFORMATION_SCHEMA

PUBLIC

Tables

ADDRESS

CUSTOMER

DAY

FILM

INVENTORY_SNAPSHOT

MONTH

RENTAL

RENTAL_TRANSACTIONS

STORE

YEAR

SNOWFLAKE

MOVIE_RENTAL_STORE / PUBLIC

Schema ACCOUNTADMIN 1 day ago

Create

Schema Details

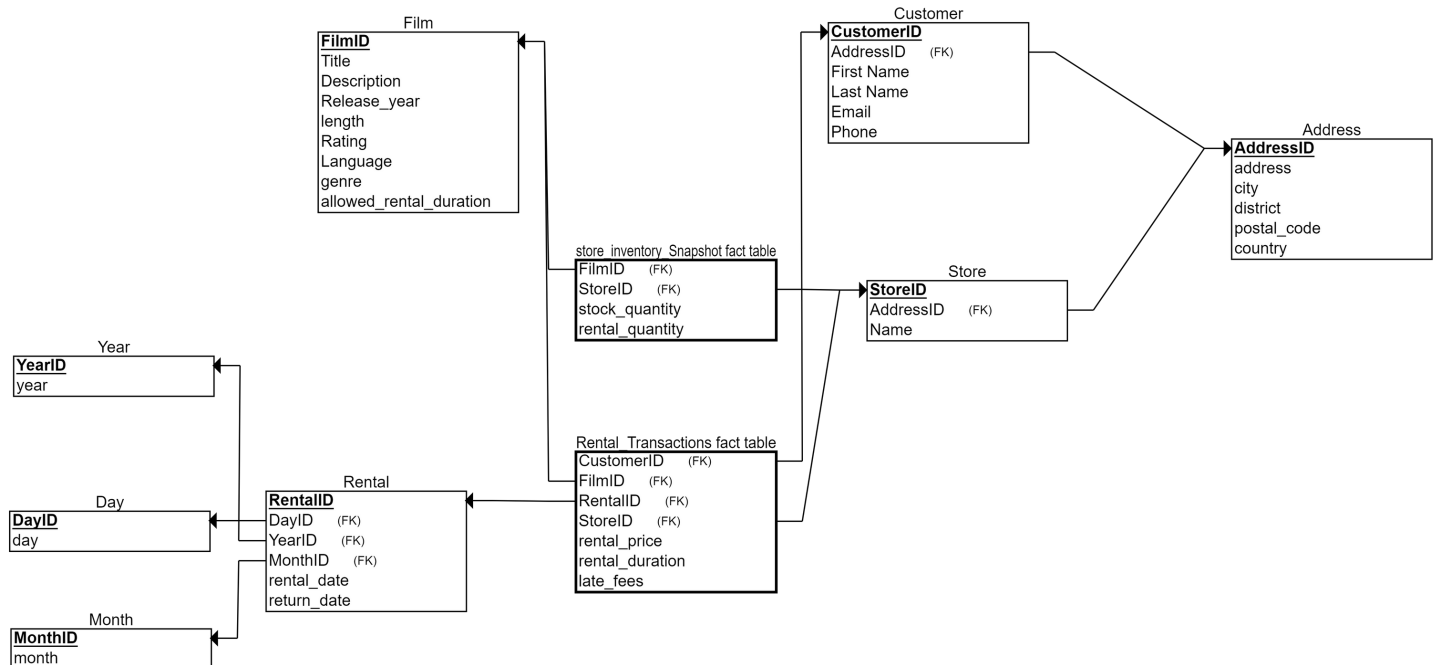
Tables

10 Tables

Search All Tables

NAME	TYPE	OWNER	ROWS	BYTES	CREATED
ADDRESS	Table	ACCOUNTADMIN	603	23.5KB	1 day ago
CUSTOMER	Table	ACCOUNTADMIN	599	24.5KB	1 day ago
DAY	Table	ACCOUNTADMIN	29	1.0KB	1 day ago
FILM	Table	ACCOUNTADMIN	1K	43.5KB	1 day ago
INVENTORY_SNAPSH...	Table	ACCOUNTADMIN	1.5K	5.0KB	1 day ago
MONTH	Table	ACCOUNTADMIN	5	1.5KB	1 day ago
RENTAL	Table	ACCOUNTADMIN	16.0K	256.0KB	1 day ago
RENTAL_TRANSACTIO...	Table	ACCOUNTADMIN	16.0K	100.5KB	1 day ago
STORE	Table	ACCOUNTADMIN	2	1.5KB	1 day ago
YEAR	Table	ACCOUNTADMIN	1	1.0KB	1 day ago

Data Warehousing



Store_inventory_snapshot: used to store historical data on stock and rented quantities by store and movie.

Rental_Transactions: used to store historical data on all transactions occurring in all stores. Its measures are the following:

- **rental_price:** price to rent a certain movie.
- **rental_duration:** actual rental duration of a movie by a customer

- `late_fees`: any incurred late return fees. (By default, 1\$ is charged to the customer for every 1 day delay)

We have also identified the following dimensions:

- Film
- Rental
- Year
- Day
- Month
- Customer
- Store
- Address



Analysis

Analysis was done using Power BI.

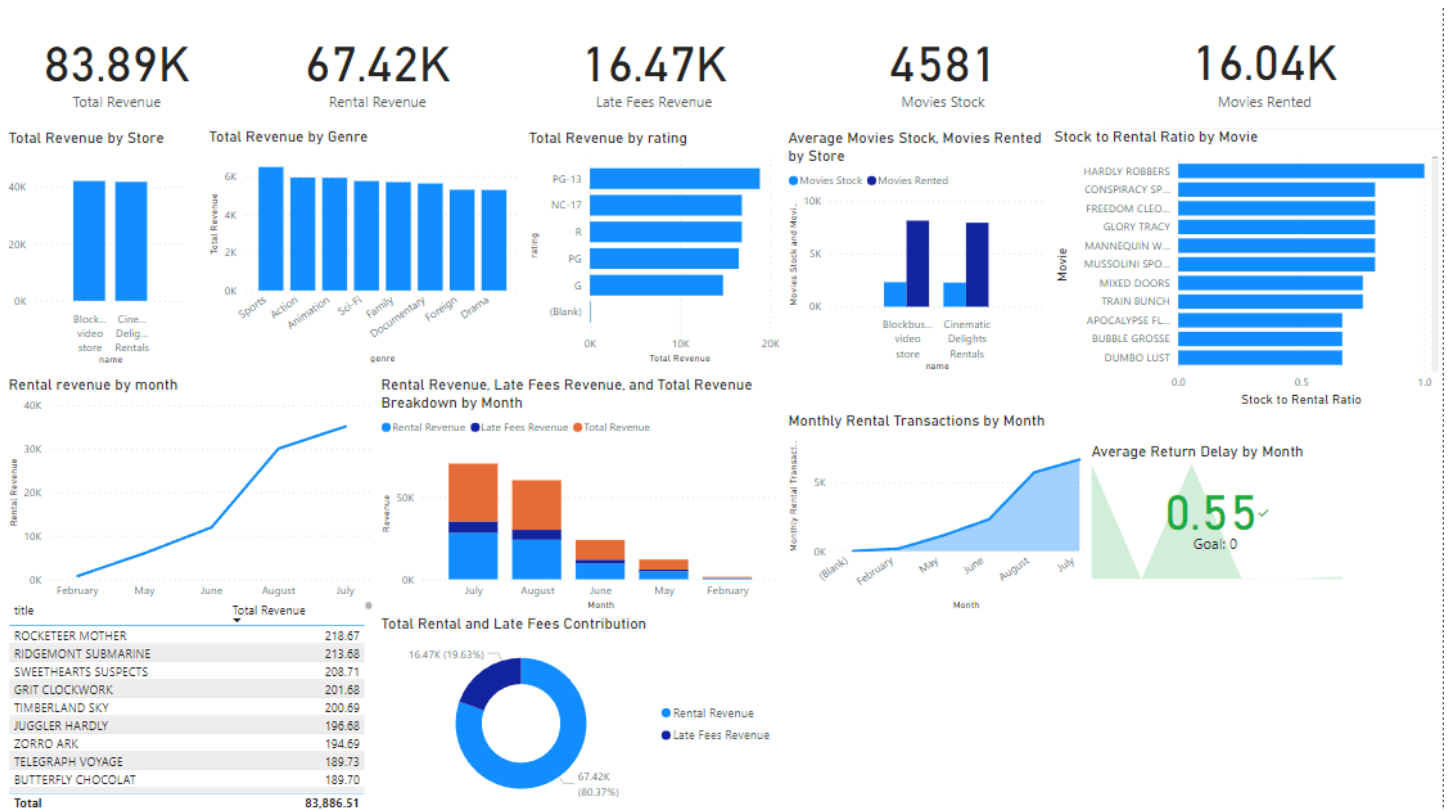
Two PowerBI files are included in the main folder: `snowflake_analysis` and `local_analysis`.

The two are identical. The only difference is `snowflake_analysis` gets its data from snowflake directly. That requires authentication or else it would fail to load any visualizations.

`local_analysis` uses the local files in the folder. This one should be used if you wish not to connect to snowflake.

 <code>local_analysis</code>	1/21/2024 9:33 PM	Microsoft Power B...	2,777 KB
 <code>snowflake_analysis</code>	1/21/2024 2:43 PM	Microsoft Power B...	2,048 KB

Upon loading up `local_analysis.pbix`. You are introduced to the dashboard.



Conclusion

Few things to note about the project:

- Data was not of the highest quality as can be seen from the visualizations. Barely any difference between the two stores. That made for some rather disappointing visualizations.
- Project could have been much more informative had I added more data and probably added a couple more dimensions.
- Working on this alone was a bit tough
- More complex and interesting KPIs were supposed to be done but were cut out due to lack of time/expertise.