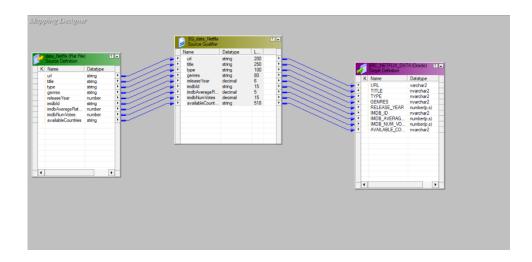
# Project Documentation: Netflix Data Processing Pipeline

# **Project Overview**

This project implements an ETL (Extract, Transform, Load) pipeline for cleaning and managing Netflix data using Informatica PowerCenter, Oracle SQL, and Python. The process loads raw data from a CSV file, applies necessary transformations, and stores the cleaned data in a target table while supporting SCD Type 1 updates for incremental changes.

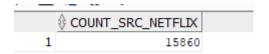
# 1. Data Extraction and Loading (Informatica PowerCenter)

- **Source Data**: The raw dataset was downloaded from Kaggle in CSV format, containing information on Netflix movies and shows.
- **Data Transfer**: Using Informatica PowerCenter, we successfully loaded all 18,860 records from the CSV file into a staging table, src\_netflix\_data, in an Oracle database.



• Sql query to check count of records in the src\_netflix\_data after loading:

SELECT COUNT(\*) as count src netflix FROM src netflix data



## 2. Data Transformation (Python)

After loading the raw data into the staging table, a Python script was executed to perform the following transformations:

- Null Removal: Removed any rows with null values in critical columns.
- **Data Type Conversion**: Converted data types to align with Oracle table specifications
- String Formatting: Standardized string fields for consistent casing and formatting.
- **Data Ordering**: Ordered the dataset based on specified columns to prepare it for analysis and reporting.
- Resulting Data:
  - Post-transformation, the dataset was reduced to 15,134 records, ready for loading into the target table.

#### 3. Data Loading to Target Table

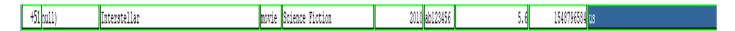
- Target Table: The cleaned data was loaded into the tgt\_netflix\_cleaned\_data table in Oracle, ensuring it was ready for insert and update operations.
- Success Check: The data transfer to tgt\_netflix\_cleaned\_data was successful, with all records appearing as expected.
- Sql query to show count of records in the tgt netflix cleaned data after executing python code:

select count(\*) as count tgt netflix from tgt netflix cleaned data



# 4. Testing the ETL Pipeline

- Insert Test:
  - o A test row was inserted into the src netflix data table with the following values:



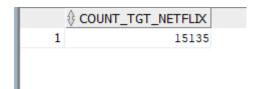
• **Result**: After running the ETL code, the new row appeared in tgt\_netflix\_cleaned\_data, confirming the insert functionality.

```
select * from tgt_netflix_cleaned_data where title =
'Interstellar' and genres = 'Science Fiction'
```

```
    $\text{\text{URL}}$ URL
    $\text{\text{TITLE}}$ TYPE
    $\text{\text{GENRES}}$ GENRES
    $\text{\text{RELEASE_YEAR}}$ MOB_ID
    $\text{\text{IMDB_AVERAGE_RATING}}$ MOB_NUM_VOTES
    $\text{\text{AVAILABLE_COUNTRIES}}$

    1 Unknown Interstellar Movie
    Science Fiction
    2010 ab123456
    5.6
    1548796584 US
```

select count(\*) as count\_tgt\_netflix from tgt\_netflix\_cleaned\_data,
the number of records in the target table increased by one cause of inserted row



# • Update Test:

O We updated the test row in src\_netflix\_data to verify if updates were reflected in the target table:

```
UPDATE src_netflix_data
SET release_year = 2014,
    imdb_average_rating = 9.8
WHERE imdb id = 'ab123456';
```

o **Result**: Running the code successfully updated tgt\_netflix\_cleaned\_data, confirming that SCD Type 1 functionality for updates is working as intended.

select \* from tgt\_netflix\_cleaned\_data where title = 'Interstellar' and genres = 'Science Fiction'



#### 5. Conclusion

This ETL process demonstrates a successful, reliable pipeline for loading, transforming, and maintaining Netflix data. It supports dynamic updates (SCD Type 1) in the target table, making it well-suited for data warehousing and business intelligence applications.

### **GitHub Repository**

For further details and code, visit the GitHub repository: [https://github.com/SherifElshafeyy/Netflix-Data-Pipline]