

datadot-e2e-data-pipeline

**An end-to-end data engineering project using Azure Data Factory,
Databricks, Synapse, and Power BI.**

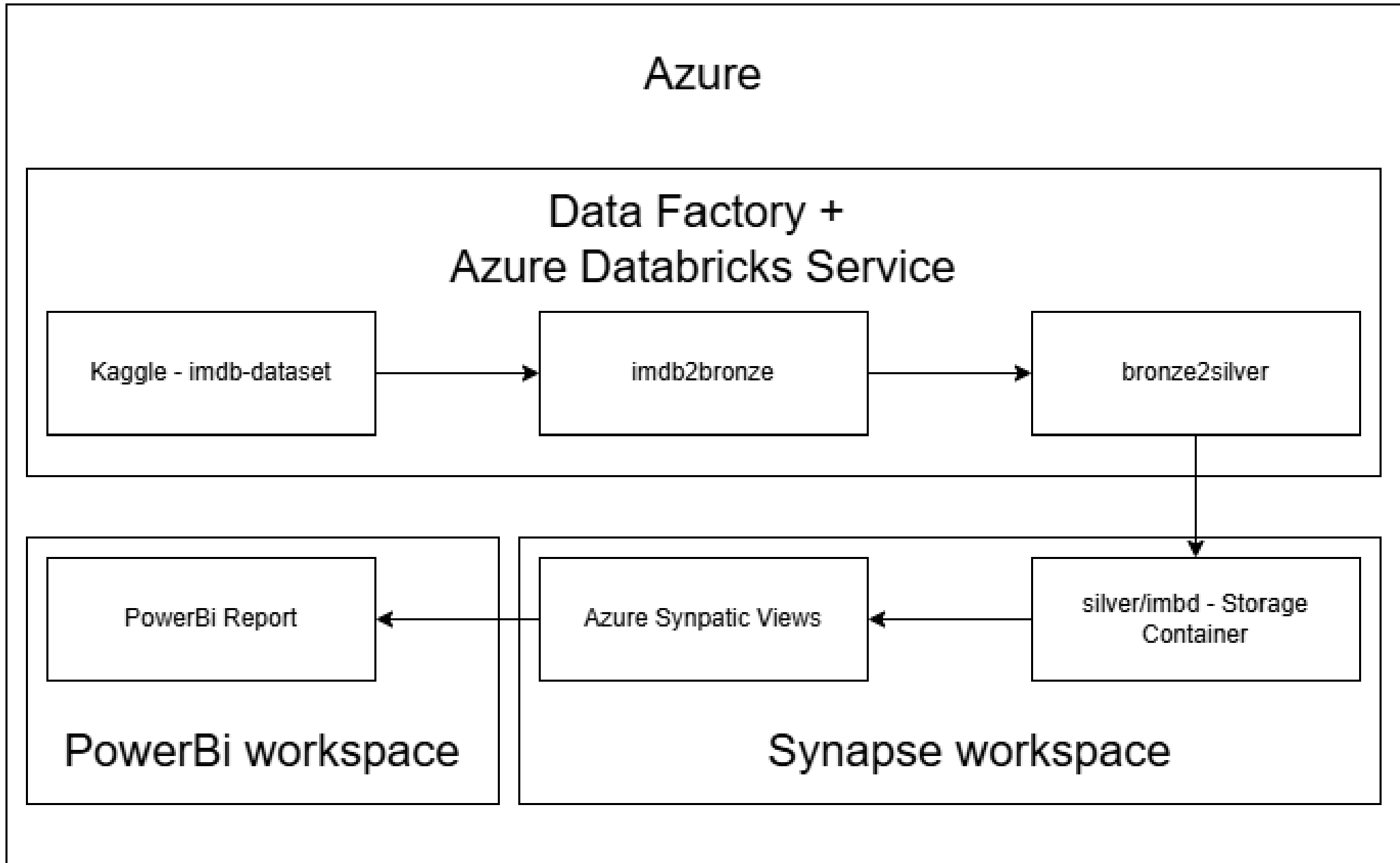
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Architecture

The data pipeline consists of the following components:

- **Azure Data Factory** (Data Ingestion)
- **Azure Databricks** (Data Transformation)
- **Azure Synapse Analytics** (Data Analytics)
- **Power BI** (Data Visualization)



Data Source

- [imdb-dataset \(Kaggle\)](#).

This dataset contains over 8GB of data stored in multiple CSV files, with movie-related attributes such as title, genre, rating, runtime, and number of votes.

Problem Statement

Movies are watched by millions worldwide, and analyzing their data can reveal trends and insights about the industry:

- **Goal:**

- To analyze and provide insights into the distribution of movies by genre, rating, runtime, and region.
- To visualize the results in a Power BI dashboard for deeper understanding.

Challenges

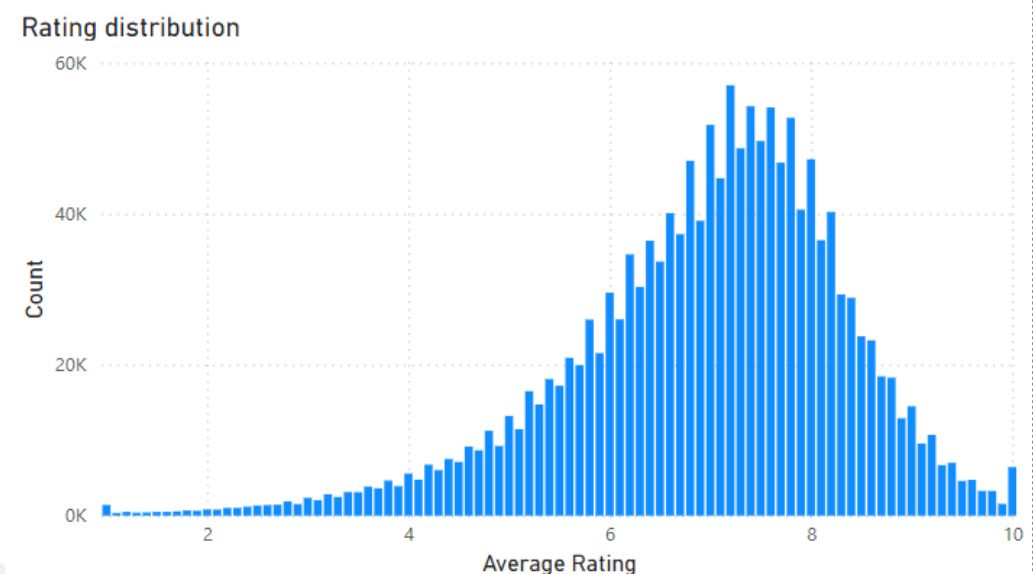
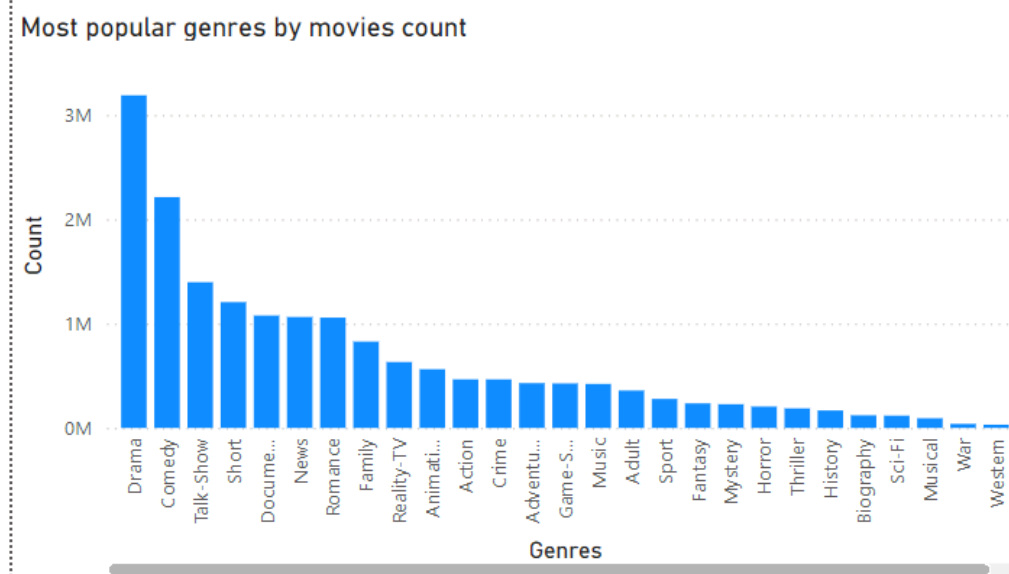
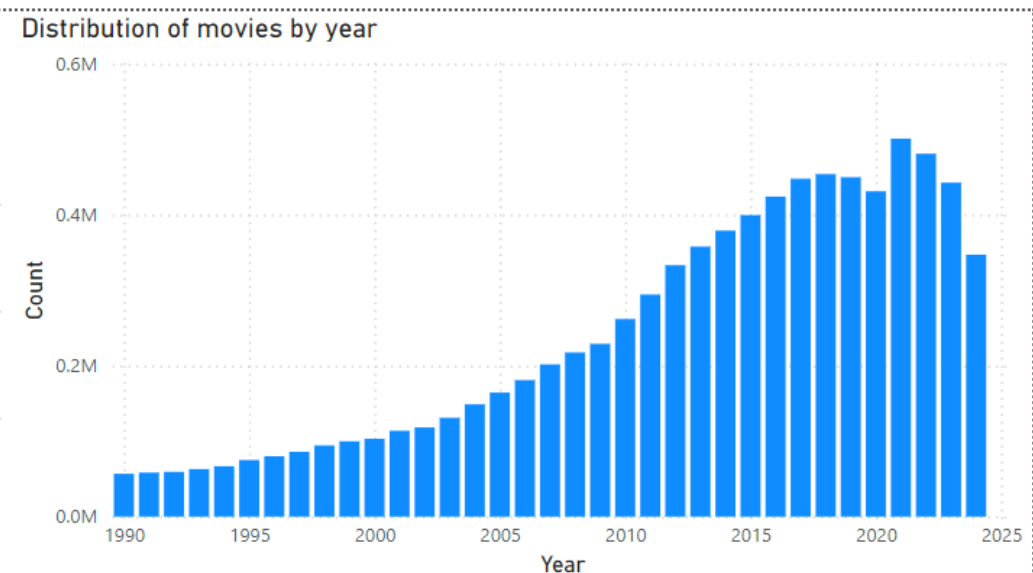
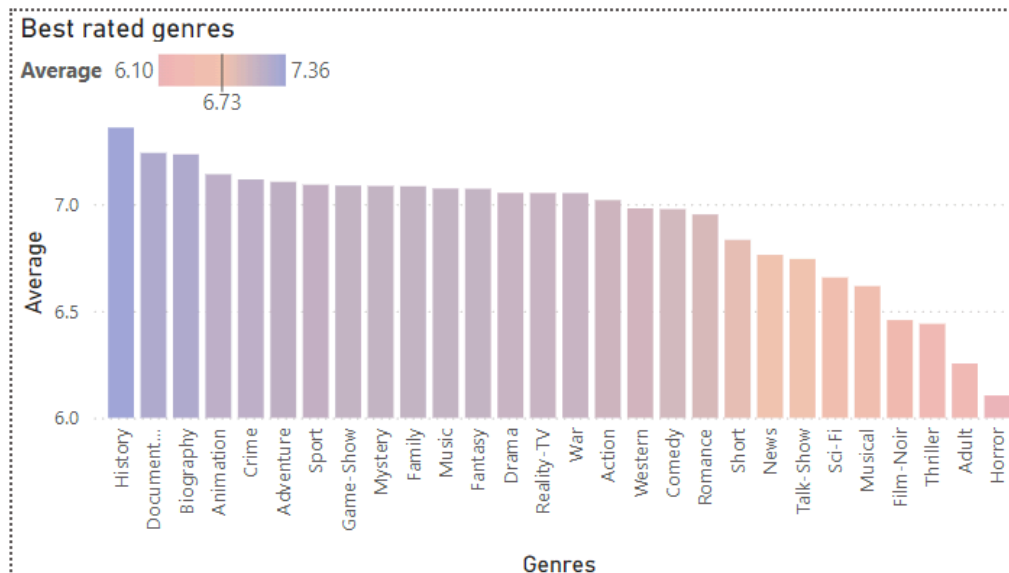
- Large dataset (~8GB).
- Need for automated data transformation (cleaning, aggregation, statistics).
- Leveraging Azure Data Factory, Azure Databricks, Azure Synapse Analytics, and Power BI for an end-to-end pipeline.

Results

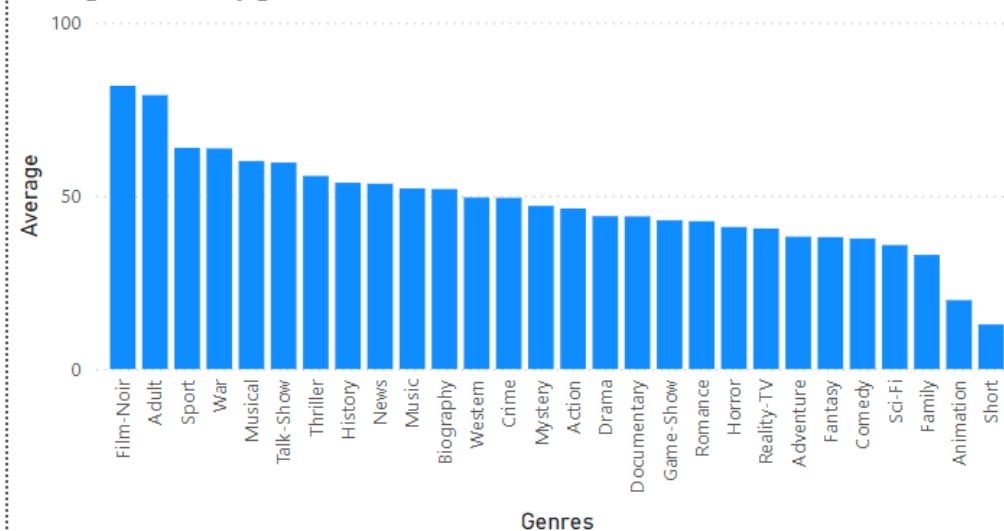
Analysis Aspect	Details
Best rated genres	History, Documentary, Biography, Animation
Distribution of movies (year)	Peak in 2021 (501k movies)
Most popular genres	Drama, Comedy, Talk-Show
Rating distribution	Gaussian distribution with mean ~ 7.4
Runtime average by genre	Film-Noir (82 mins), Adult (79 mins), Sport (63 mins)

Results

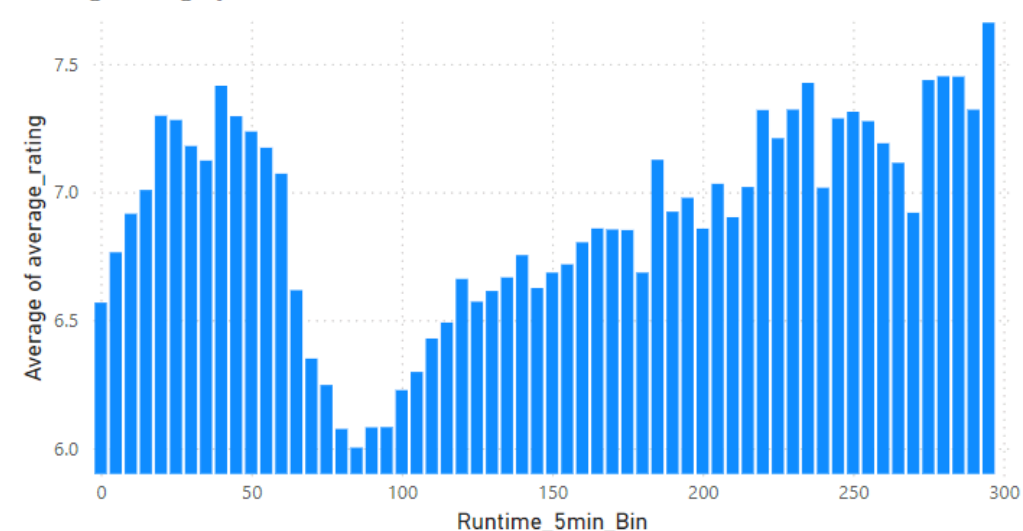
Analysis Aspect	Details
Average rating by runtime	Peak at 40 mins (7.41), dips to ~6 at 85 mins, then gradually rises to ~7.45 at 285 mins
Distribution by runtime	Peaks at 20 mins (450k movies), 30 mins (374k), 60 mins (273k)
Average votes by runtime	Bell-shaped curve centered ~90–200 mins, peak at 170 mins (13k votes)
Count of movies by region	US (946k), GB (118k), IN (334k), CA (194k), FR (185k)



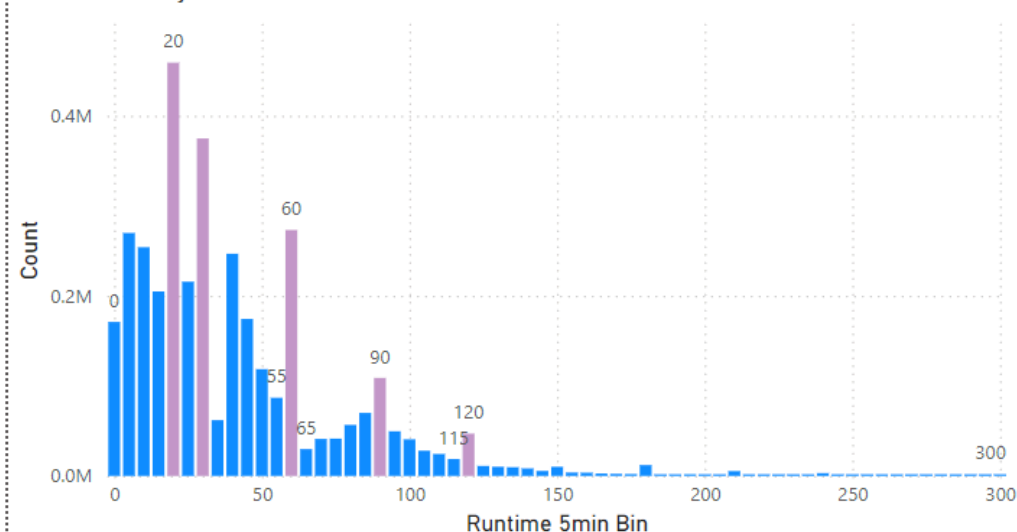
Average runtime by genres



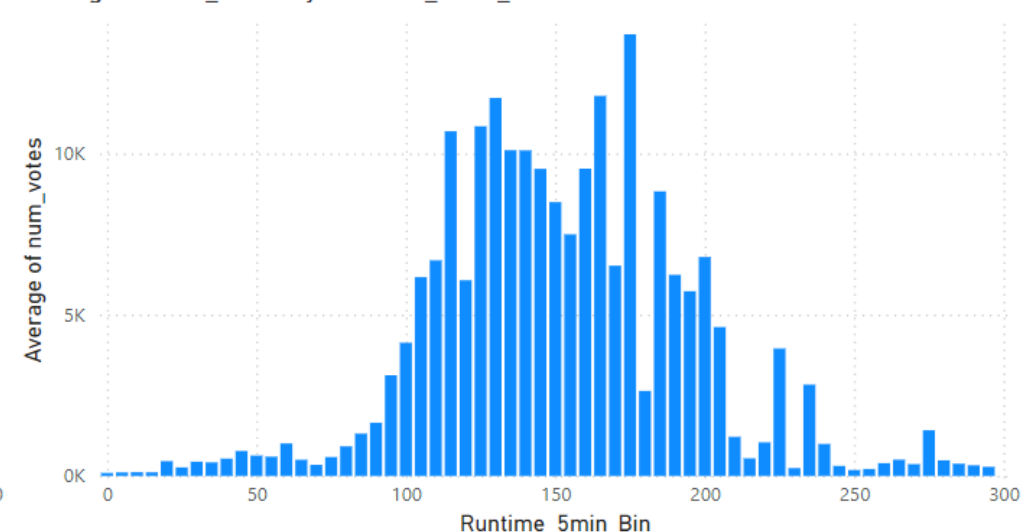
Average rating by runtime

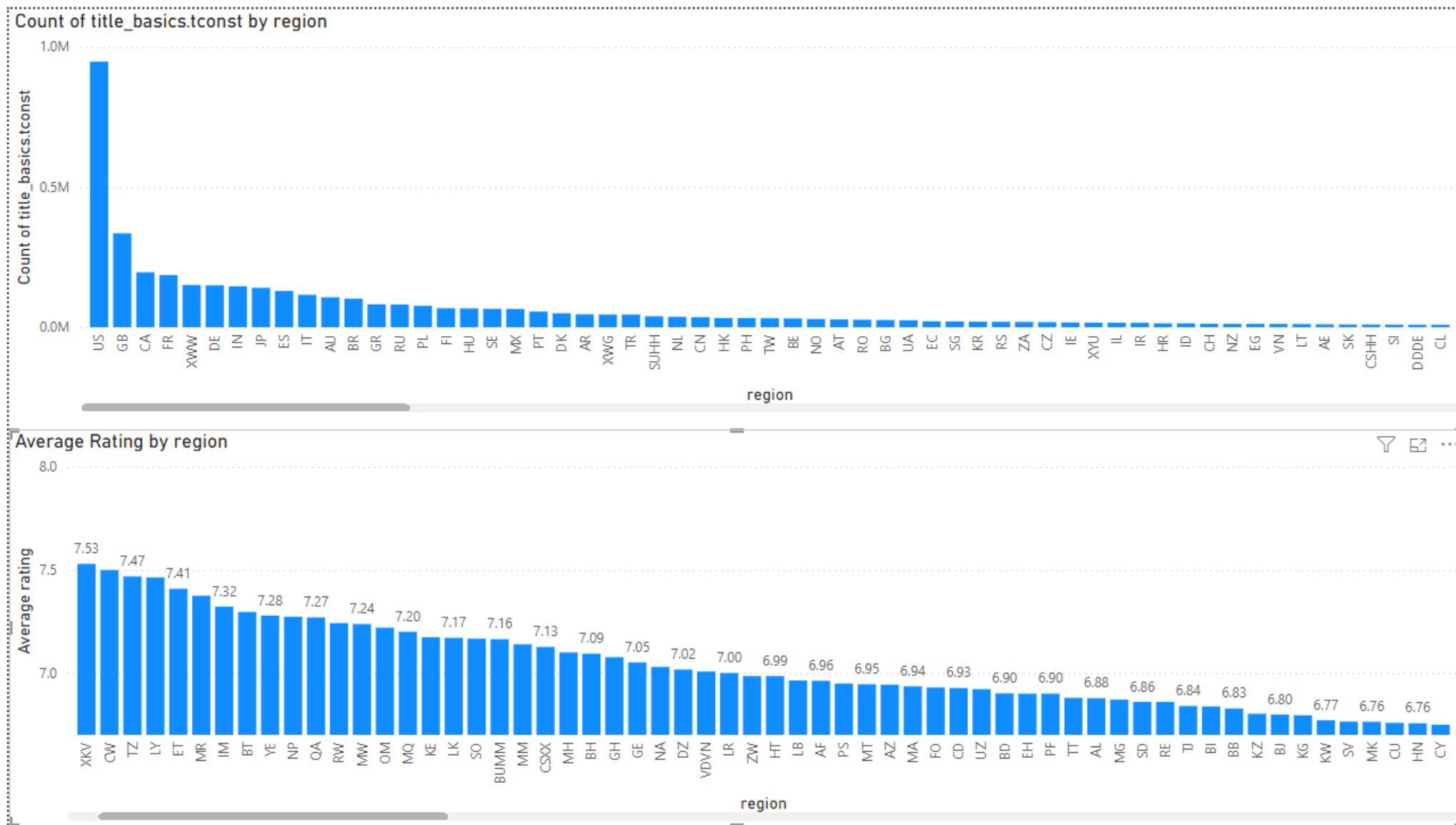


Distribution by Runtime



Average of num_votes by Runtime_5min_Bin





What We Have Learned

- **Building an E2E pipeline in Azure:** Data Factory for ingestion, Databricks for transformation, Synapse for analytics, Power BI for visualization.
- **Data transformation complexity:** Multiple steps (raw → bronze → silver → gold).

What We Have Learned

- **Visualization approach:** Selecting relevant metrics and understanding data distribution is crucial.
- **Performance considerations:** Large dataset (~8GB), multiple CSVs, and parallel processing.
- **Potential data enrichment:** Other datasets (e.g., economic or social factors) could add more context to movie data.

Articles and Resources

- [Modelling Data Pipelines](#)
- [Azure End-to-End Data Engineering Project \(Medium\)](#)
- [Incorporating Deep Learning Model Development With an End-to-End Data Pipeline](#)
- [Azure End-To-End Data Engineering Project for Beginners \(YouTube\)](#)