Task 1: Movie Ratings Data Ingestion

CSV data representing Movie ratings

UserID	MovieID	Rating	Timestamp	UserID
U001	M001	4	2024-05-01 14:30:00	U001
U002	M002	5	2024-05-01 16:00:00	U002
U003	M001	3	2024-05-02 10:15:00	U003
U001	M003	2	2024-05-02 13:45:00	U001
U004	M002	4	2024-05-03 18:30:00	U004

dbutils.fs.cp("file:/Workspace/Shared/movie_ratings.csv","dbfs:/FileStore/movie_ratings.csv")

```
-- Ingest this CSV data into a Delta table in Databricks.
from pyspark.sql import SparkSession
from pyspark.sql.functions import col, to timestamp
from pyspark.sql.types import StructType, StructField, StringType, IntegerType, TimestampType
import logging
spark = SparkSession.builder \
  .appName("Movie Ratings Ingestion") \
  .getOrCreate()
schema = StructType([
  StructField("UserID", StringType(), True),
  StructField("MovieID", StringType(), True),
  StructField("Rating", IntegerType(), True),
  StructField("Timestamp", StringType(), True)
])
logging.basicConfig(level=logging.INFO, format='%(asctime)s - %(levelname)s - %(message)s')
logger = logging.getLogger()
```

```
try:
  movie ratings df = spark.read.csv("dbfs:/FileStore/movie ratings.csv", schema=schema,
header=True)
  movie ratings df = movie ratings df.withColumn("Timestamp", to timestamp(col("Timestamp"),
                                                                     "yyyy-MM-dd HH:mm:ss"))
  cleaned df = movie ratings df.dropna()
  cleaned df.write.format("delta").mode("overwrite").save("/delta/movie ratings")
- - Ensure proper error handling for missing or inconsistent data, and log errors accordingly.
  logger.info("Movie ratings data ingested successfully.")
except Exception as e:
  logger.error(f"Error during data ingestion: {e}")
    Task 2: Data Cleaning
    - - Ensure that the Rating column contains values between 1 and 5.
   -- Remove any duplicate entries (same UserID and MovieID).
    cleaned ratings df = movie ratings df \
      .filter((col("Rating") \geq 1) & (col("Rating") \leq 5))
      .dropDuplicates(["UserID", "MovieID"])
   - - Save the cleaned data to a new Delta table.
    cleaned ratings df.write.format("delta").mode("overwrite").save("/delta/cleaned movie ratings")
    Task 3: Movie Rating Analysis
    - - Calculate the average rating for each movie.
    avg ratings df = cleaned ratings df.groupBy("MovieID")\
               .agg({"Rating": "avg"})\
               .withColumnRenamed("avg(Rating)", "AverageRating")
```

-- Identify the movies with the highest and lowest average ratings.

```
highest_rated = avg_ratings_df.orderBy(col("AverageRating").desc()).limit(1)
lowest_rated = avg_ratings_df.orderBy(col("AverageRating").asc()).limit(1)
```

- - Save the analysis results to a Delta table.

```
avg_ratings_df.write.format("delta").mode("overwrite")\
.save("/delta/movie rating analysis")
```

Task 4: Time Travel and Delta Lake History

- - Perform an update to the movie ratings data

```
cleaned\_ratings\_df = cleaned\_ratings\_df \\ .withColumn("Rating", when(col("MovieID") == "M001", 5) \\ .otherwise(col("Rating"))) \\ cleaned\_ratings\_df.write.format("delta").mode("overwrite") \\ .save("/delta/cleaned_movie_ratings") \\ \\
```

- - Roll back to a previous version of the Delta table

- - Use DESCRIBE HISTORY to view the history of changes to the Delta table.

```
spark.sql("DESCRIBE HISTORY delta.' /delta/cleaned movie ratings'").show()
```

Task 5: Optimize Delta Table

- - Implement Z-ordering on the MovieID column to improve query performance.
- - Use the OPTIMIZE command to compact the data and improve performance.

```
spark.sql("OPTIMIZE delta.'/delta/cleaned movie ratings' ZORDER BY (MovieID)")
```

- - Use VACUUM to clean up older versions of the table.

```
spark.sql("VACUUM delta.'/path/to/delta/cleaned_movie_ratings' RETAIN 0 HOURS")
```