ANALYTICS ENGINEERING ASSIGNMENT

TASK 1

Parquet File Conversion to CSV

Download the <u>parquet</u> file and save it locally.

AWS Free Tier Account Setup

- Create an Amazon Web Services (AWS) account if you don't already have one.
- Use an IAM user account instead of the root user account for security best practices.
- Assign appropriate permissions and roles to the IAM user.
- Log in using the IAM user credentials.

S3 Bucket Creation

- Navigate to the S3 service in the AWS Management Console.
- Create two S3 buckets within the same AWS region.
- Ensure both bucket names follow AWS's global uniqueness and naming conventions.

Lambda Function Creation

- Search for Lambda in the AWS Console and create a new Lambda function.
- Select the **Author from scratch** option to create the function.
- Provide a meaningful name for the Lambda function.

- Choose the runtime environment (e.g., Python 3.13) and architecture (e.g., x86_64).
- Under permissions, select a role with basic Lambda execution permissions.
- · Create the function.

Lambda Function Development

- Develop a script that:
 - Triggers when a parquet file is uploaded to the source bucket.
 - Converts the parquet file to csv.
 - Uploads the resulting csv file to the destination bucket.
- Add a trigger for the Lambda function:
 - Choose S3 as the trigger source.
 - Select the source bucket (e.g., source-bucket-analytics-v1).
 - Set the event type to All object create events.
 - Add a suffix filter for parquet files.
- Specify the destination bucket (e.g., processed-bucket-analytics-v1) and provide its ARN.
- Add necessary Lambda layers, such as **pandas**, to handle data processing.
- Attach an IAM policy to the Lambda execution role granting required permissions to access both S3 buckets.

Testing and Deployment

- Upload a .parquet file to the source bucket (source-bucket-analytics-v1).
- Test the Lambda function to verify it converts the file correctly.
- Deploy the Lambda function.
- Monitor CloudWatch metrics:
 - Confirm Invocations count reflects the test.
 - Verify **Error count** is zero.

- Check Success rate (%) is 100% (green status).
- Review CloudWatch logs for detailed execution information, ensuring no errors occurred.
- Verify that the converted .csv file appears in the destination bucket (processed-bucket-analytics-v1).
- Download the converted file for further use.

TASK 2

TXT File Conversion

- Download the .txt file and save it locally.
- Choose and install an IDE of your preference; in this case, **PyCharm** is used.
- Navigate to the directory containing the .txt file using basic Linux commands such as cd, is, etc.
- Create a new Python script named txt_file_conversion.py and save it in your working directory.
- Open the <u>ltxt</u> file to review its structure and identify the delimiter used (e.g., comma, tab, pipe).
- Install the required Python libraries, primarily pandas and os, if not already installed.
- Import the necessary libraries in your script.
- Read the .txt file using pandas.read_csv(), specifying the correct delimiter.
- Convert the data to .csv format and save it to your desired location.
- Implement error handling using try and except blocks to catch and display any errors during the conversion process.
- Verify that the converted <u>.csv</u> file is created successfully in the specified location.

TASK 3

DBeaver Schema and Table Creation

- Download and install **DBeaver** from https://dbeaver.io/download.
- Ensure PostgreSQL is installed on your machine. If not, download and install
 the latest stable version compatible with your OS
 from https://www.postgresql.org/download/.
- Launch DBeaver, then add a new connection by selecting PostgreSQL.
- Configure the connection using your PostgreSQL instance details: Host, Port, Username, Password, and Database name.
- Test the connection; if successful, you will receive a confirmation message.
- Create a new database using the SQL DDL command:

CREATE DATABASE your_database_name;

- · Create a schema within the database as needed.
- Import CSV files into your database by right-clicking on the target Tables node and selecting Import Data.
- Verify successful ingestion by inspecting the table structure (columns) and running a simple query such as:

SELECT * FROM table_name LIMIT 10;

Data Cleaning

- Perform a quick Exploratory Data Analysis (EDA) to ensure data quality:
 - 1. Verify that the unique identifier column contains no duplicates.

- 2. Identify any NULL or missing values and determine an appropriate strategy to handle them.
- 3. Standardize categorical data to maintain consistency (e.g., unify representations such as "M" and "Male" to a single standard value).
- 4. Confirm that each column's data type is appropriate for its contents.
- 5. Assess overall data quality and integrity.
- 6. Apply necessary changes using SQL DDL commands such as ALTER TABLE to clean and adjust the schema or data as required.

TASK 4

Data Integration via SQL

- Perform SQL joins on tables using common columns to create a unified, consolidated dataset.
- Use the integrated data to answer the required business questions; the corresponding SQL queries are provided in a separate document.
- For question 3, the justification for using **average imputation** to fill null values is that the question specifically focuses on average metrics.
 - Replacing nulls with the average preserves the integrity of the analysis and yields a more accurate and meaningful result compared to using zeros or removing records entirely.

TASK 5

Dashboard Creation

- Export the unified dataset into a BI tool of your choice—Power BI or Amazon
 QuickSight.
- In this case, Power BI was used. If not already installed, download it from:
 https://www.microsoft.com/en-us/power-platform/products/power-bi/downloads

- Develop a comprehensive dashboard or report by applying appropriate visualizations that best communicate the data insights.
- Choose chart types based on the nature of the data (e.g., bar charts for comparisons, line charts for trends, pie/donut charts for proportions, etc.).
- Ensure the dashboard is clean, interactive, and provides a clear narrative for decision-makers.

DBT ASSESMENT

- Install Docker if not already installed: Windows download link
- Install Git if not already installed: macOS download link
- Download or clone the project files to your local machine
- Launch the Docker daemon and ensure it is running
- Open the project folder using Visual Studio Code
- Run docker build -t dbt_project. to build the Docker image
- Run docker run -it -v \$(pwd):/app dbt_project to start the container and mount the project directory
- Run dbt --version to verify dbt is installed and working
- Run dbt debug to confirm the profiles.yml is correctly configured and the database connection works
- Run dbt seed to load the .csv seed files into the database
- Navigate to the /models/dim/ folder and edit dim_patients.sql to define the patient dimension model
- Navigate to the /models/fct/ folder and edit fct_patient_claims.sql to define the fact model for patient claims
- Add tests and documentation in schema.yml:

- Include column- and model-level tests such as not_null, unique, etc.
- Write clear documentation for each model and its columns
- Run dbt test to execute the defined tests
- Run dbt run to build all dbt models
- Zip the entire project folder
- Push the zipped (or unzipped) folder to a Git repository
- Create a README.md file containing:
 - Assumptions made during the project
 - Steps to build and run the dbt models
 - List of tools and technologies used