Data Engineer Take-Home Challenge: ETL Pipeline Development

Objective:

Design and implement an ETL pipeline that extracts data from multiple CSV files, applies transformations based on business rules, and loads the data into a PostgreSQL database. The pipeline should also include robust error handling and logging.

You should create a private GitHub repository and share it with the GitHub usernames (RahmatYousufi and andyphamberkeley) upon completion. There is a 72-hour timeline from the time you receive this assignment. We estimate it will take 4-6 hours to complete.

Challenge Overview

You will be provided with multiple CSV files containing sales and customer data. Your task is to build a dynamic ETL pipeline that processes this data and handles various potential issues along the way.

Sample Data

sales_data.csv

(order_id	customer_i d	product	quantity	price	order_date
•	1	1001	Widget A	2	19.99	2024-01-15 10:00:00
2	2	1002	Widget B	1	29.99	2024-01-16 11:00:00
(3	1003	Widget A	1	19.99	2024-01-16 12:00:00

4	1004	Widget C	0	39.99	2024-01-17 14:00:00
5	1005	Widget B	3	29.99	invalid_date

customer_data.csv

customer_i d	customer_nam e	email	signup_date
1001	John Doe	john.doe@example.com	2023-06-10 09:00:00
1002	Jane Smith	jane.smith@example.com	invalid_date
1003	Alice Johnson	alice.j@example.com	2022-12-01 14:00:00
1004	Bob Brown	bob.brown@example.co m	2023-01-17 10:00:00
1005	Charlie White	charlie.w@example.com	2023-11-15 11:00:00

You can create these files for the challenge.

Requirements

Extraction

1. File Reading:

- Read multiple CSV files (sales_data.csv and customer_data.csv) and load them into suitable data structures (e.g., Pandas DataFrames).
- o Dynamically detect and handle missing files.

2. Validation:

 Validate that the required columns exist in each file. Log errors if columns are missing.

Transformation

1. Data Cleaning:

- Convert order_date and signup_date to a standard format (YYYY-MM-DD).
 Log errors for invalid dates and skip affected records.
- Remove any rows with missing or invalid customer_id or order_id.

2. Data Enrichment:

- Calculate the total value for each order (total_value = quantity * price).
- Join the sales data with customer data on customer_id to enrich the sales table with customer_name and email.

3. Business Rules:

- Exclude orders with a quantity of 0 or less.
- Mark orders with total values exceeding \$1000 as "High-Value Orders" in a new column order_type.
- Add a column customer_tenure representing the number of days since the customer's signup date.

4. Data Aggregation:

 Create a summary table with total sales (SUM(total_value)) per product and order counts (COUNT(order_id)).

Loading

1. Database Schema:

- Create the following tables in PostgreSQL:
 - sales: Contains all transformed sales data with fields: order_id, customer_id, product, quantity, price, order_date, total_value, customer_name, email, order_type, customer_tenure.
 - sales_summary: Aggregates product-level metrics with fields: product, total_sales, order_count.

2. Constraints:

 Enforce primary and foreign key constraints (e.g., customer_id in sales references customer_data).

3. Data Ingestion:

 Batch load data into the database. Implement transaction management to ensure that partial failures do not corrupt data.

4. Logging:

 Log the number of records processed, records loaded, and any errors encountered.

Advanced Requirements

1. Performance Optimization:

- o Optimize the pipeline for handling large files (e.g., processing data in chunks).
- Use indexes in the PostgreSQL database for faster queries.

2. Scalability:

 Design the ETL to handle additional files and tables in the future by parameterizing the pipeline.

3. Configuration Management:

 Store configurable parameters (e.g., file paths, database connection strings) in a config.json file or environment variables.

Deliverables

1. Source Code:

 Include clear and well-organized Python scripts or notebooks for the ETL pipeline.

2. Database Script:

SQL script to create the required PostgreSQL tables with constraints.

3. README:

- Explain the approach, any challenges faced, and how to:
 - Set up the database.
 - Run the ETL pipeline.

4. Sample CSV Files:

Provide sales_data.csv and customer_data.csv.

Evaluation Criteria

1. Code Quality:

Modularity, readability, and adherence to best practices.

2. Error Handling:

Robust handling of edge cases and invalid data.

3. Performance:

o Ability to handle large datasets efficiently.

4. Scalability:

 $\circ\quad$ Flexibility of the pipeline to adapt to future requirements.

5. Documentation:

o Completeness and clarity of instructions.

Deliverables

- 1. Code: Your implementation and database migration files.
- 2. **Documentation**: A README file with setup instructions.
- 3. **Git Workflow**: Ensure your commits reflect logical progress, and commit as you implement.