PROJECT ETL

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Context

His project focuses on the analysis of a health dataset that includes detailed information on 1,879 patients. Each patient is uniquely identified with an ID between 6000 and 7878.

Tools

• Python	• Git hub.
• Postgress.	• Power BI
• Jupyter Notebook.	• Docker
• Dataset (Diabetes_healt).	• AirFlow

Step by step

First we download and configure the Airflow image using docker-compose up.

Airflow containers and networks are created and running correctly.

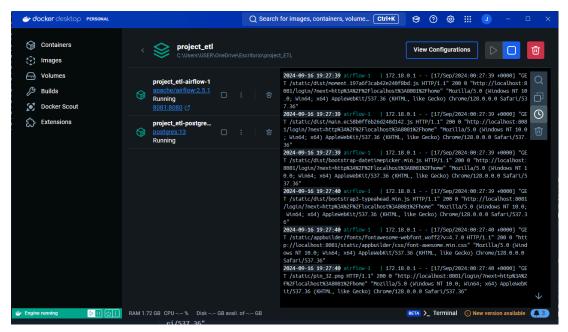
```
✓ Network project_etl_default Created
✓ Container project_etl-airflow-1 Created
Attaching to airflow-1
```

The PostgreSQL and Airflow containers have been started, both are running, and a volume has been created for the PostgreSQL database.

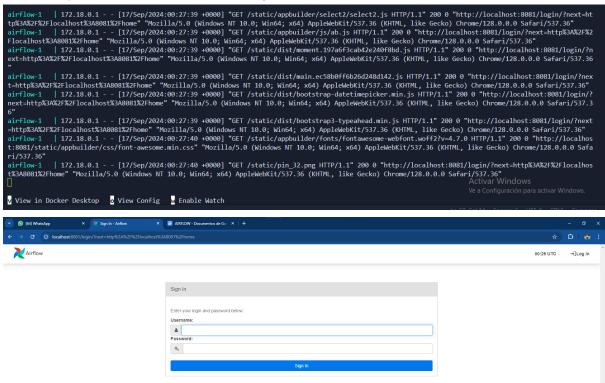
```
C:\Users\USER\OneDrive\Escritorio\project_ETL> docker-compose up --build
me="2024-09-07T14:14:20-05:00" level=warning msg="C:\USers\USER\OneDrive\Escritorio\project_ETL\docker-compose.yaml: the attribute `version` is obsolet
it will be ignored, please remove it to avoid potential confusion"

| Running 15/1
| postgres Pulled | 47.9s |
| Running 4/4
| Network project_etl_default | Created | 0.0s |
| Volume "project_etl_postgres-db-volume" | Created | 0.0s |
| Container project_etl_postgres-1 | Created | 25.1s |
| Container project_etl_airflow-1 | Created | 0.6s |
| Created | 0.6s |
| Container project_etl_airflow-1 | Created | 0.6s |
| Created | 0.6s |
| Container project_etl_airflow-1 | Created | 0.6s |
| Container project_airflow-1 | Creat
```

We have Airflow and PostgreSQL containers running successfully in Docker. I can see Airflow logs with activity on localhost:8080.

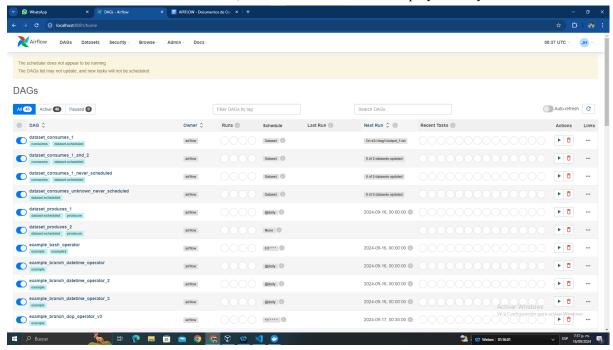


I can see that Airflow is running and it shows several HTTP activity logs in the container. Additionally, I have access to the Airflow login screen at localhost:8081.

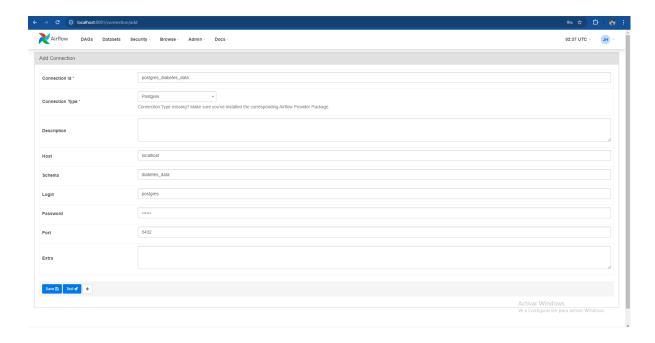


We have created an admin user in Airflow called "admin" with the name Johan Hurtado. The PostgreSQL container is still running, and although we have seen deprecation warnings in the SQLAlchemy configuration, the user was successfully created with the "Admin" role.

We have accessed the Airflow interface, where several DAGs are displayed ready to run.



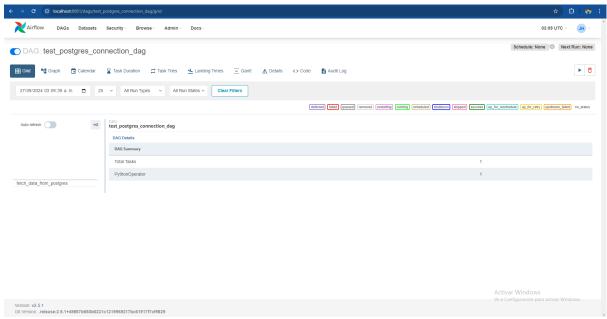
We have set up a connection in Airflow to connect to the PostgreSQL database. The connection type is "Postgres", with the host set to localhost, the schema to clients_data, and the port 5432. The username and password have also been set to allow proper access.



We have created and verified the connection to the diabetes_data database in Airflow, using PostgreSQL. Additionally, we have configured a DAG called test_postgres_connection_dag that performs operations with the connected database, displaying the data extraction task (fetch data from postgres).



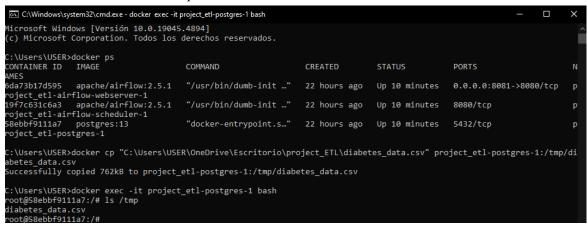
We have configured the test_postgres_connection_dag DAG to interact with the diabetes_data database in Airflow.



We entered our PostgreSQL container in Docker using the docker exec -it project_etl-postgres-1 bash command, which allowed us to access the container console to perform tasks directly in PostgreSQL.

```
PS C:\Users\USER> docker exec -it project_etl-postgres-1 bash root@9ebb5ed9872e:/#
```

Hemos copiado el archivo de dataset diabetes_data.csv desde Windows al contenedor de Docker utilizando el comando docker cp.



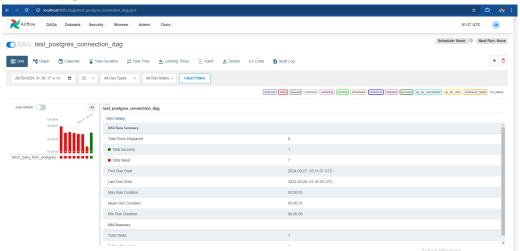
Copied

```
diabetes_data=# COPY diabetes_data FROM '/tmp/diabetes_data.csv' DELIMITER ',' CSV HEADER; COPY 1879
```

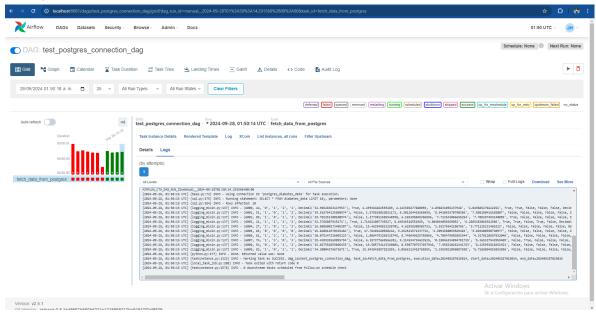
Dag in Airflow



After 8 failed tests, on the ninth run we were able to successfully complete the task in the test_postgres_connection_dag DAG. The error was related to the PostgreSQL connection, as we had not copied the database to the Docker container correctly. After fixing this issue, the DAG runs without any problems.



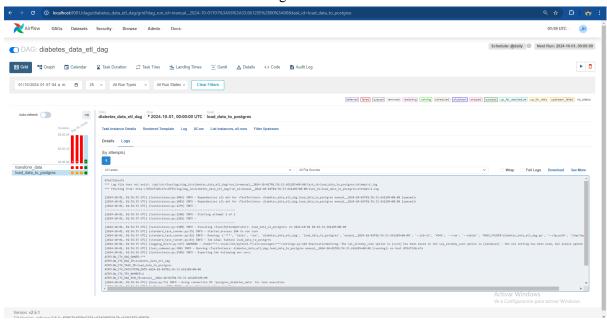
Now we have a problem with the logs, we needed to put the unique key in the docker compose.yml. actually our db in airflow.



CREATION OF THE DIMENSIONAL MODEL

DATA GROUPING AND TRANSFORMATION

After 3 failures the 2 transformation and loading tasks were successful.



We have successfully queried the diabetes_data database tables in PostgreSQL. In the first query, we have obtained the first 5 records from the dim_health_habits table, which shows information about health habits such as smoking, alcohol consumption, physical activity, diet quality, and sleep quality.

```
diabetes_data=# SELECT * FROM dim_health_habits LIMIT 5;
patientid | smoking | alcoholconsumption | physicalactivity |
                                                                 dietauality
                                                                                    sleepquality
                    4.499364662559289 | 2.443385277880059 | 4.898831055237948 | 4.049885278422252
6000
          l true
                    | 1.578919022031171 | 8.301264419669659 | 8.941093370790366 | 7.508150416102007
 6001
            false
                    | 1.1773011585548998 | 6.103395048386896 | 7.722543086655254 | 7.708387493140089
6002
            false
                    1.714621007745527 | 8.64546518551969
                                                            4.804044050369052 | 6.286548306692199
6003
            true
                                        4.62938308903732 | 2.53275642106766 | 9.771125231466325
6004
                    15.4625488312587
            false
(5 rows)
```

In the second query, we have extracted the first 5 records from the dim_demography table, which contains demographic data such as age, gender, ethnicity, socioeconomic status, and education level.

```
diabetes data=# SELECT * FROM dim demography LIMIT 5:
patientid | age | gender | ethnicity | socioeconomicstatus | educationlevel
          44 | 0
                        | 1
                                    | 2
                                                         | 1
6000
6001
          51 | 1
                        0
                                    | 1
                                                         | 2
6002
          89 | 1
                        0
                                    | 1
          | 21 | 1
6003
                        | 1
                                    | 1
                                                        | 2
6004
          27 | 1
                        0
                                    | 1
                                                         | 3
(5 rows)
```

The COPY command in PostgreSQL is used to export the fact_table and dim_health_habits tables to CSV files in the /tmp/ path. The tables are exported in CSV format with the header included.

```
root@9ebb5ed9872e:/# psql -U airflow -d diabetes_data -c "\COPY fact_table TO '/tmp/fact_table.csv' WITH (FORMAT CSV, HEADER)"
COPY 1879
root@9ebb5ed9872e:/# psql -U airflow -d diabetes_data -c "\COPY fact_table TO '/tmp/fact_table.csv' WITH (FORMAT CSV, HEADER)"
COPY 1879
root@9ebb5ed9872e:/# psql -U airflow -d diabetes_data -c "\COPY dim_health_habits TO '/tmp/dim_health_habits.csv' WITH (FORMAT CSV, HEADER)"
COPY 1879
root@9ebb5ed9872e:/#
```

It is verified that the three CSV files (diabetes_data.csv, dim_demography.csv, dim_health_habits.csv) are correctly created in the /tmp/ directory inside the Docker container.

```
root@9ebb5ed9872e:/# ls /tmp
diabetes_data.csv dim_demography.csv dim_health_habits.csv fact_table.csv
root@9ebb5ed9872e:/#
```

Usando el comando docker cp, los archivos CSV se copian desde el contenedor de Docker a una carpeta en el sistema de archivos de Windows, específicamente a la ruta C:\Users\USER\OneDrive\Escritorio\project ETL.

We check folders.

dim_demography	2	30/09/2024 9:44 p. m.	Archivo de valores	30 KB
dim_health_habits	g	30/09/2024 9:42 p. m.	Archivo de valores	153 KB
docker-compose	g	30/09/2024 5:59 p. m.	Archivo de origen	2 KB
▼ fact_table	g	30/09/2024 9:42 p. m.	Archivo de valores	309 KB

we can see that it is fine.

	A
1	patientid, age, gender, ethnicity, socioe conomic status, education level
2	6000,44,0,1,2,1
3	6001,51,1,0,1,2
4	6002,89,1,0,1,3
5	6003,21,1,1,1,2
6	6004,27,1,0,1,3
7	6005,65,0,0,0,0
8	6006,61,1,2,1,3
9	6007,74,1,3,0,3
10	6008,54,0,0,1,2
11	6009,82,1,0,1,1
12	6010,59,1,0,2,2
13	6011,82,1,1,1,3
14	6012,79,1,1,0,2
15	6013,22,0,3,0,2
16	6014,29,0,0,1,2
17	6015,76,0,0,1,3
18	6016,80,0,1,1,2

Dimensional modeling is a design technique used in data warehousing and business intelligence systems. It is organized into two types of tables:

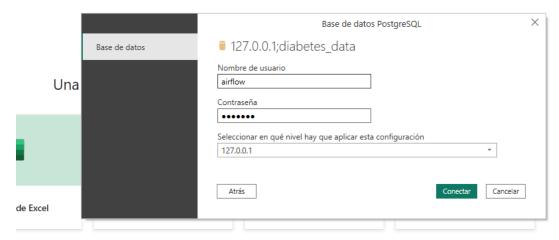
- Fact Table: Contains quantitative data, such as health metrics (bmi, systolicbp, diagnosis) in the fact table table.
- Dimension Tables: Provide context to the facts with descriptive information. In this case, there are the dim_demography (demographic data) and dim_health_habits (health habits) tables.

Reasons for separating the original dataset:

- Efficiency: Improves organization and facilitates quick queries.
- Analysis: Allows data to be analyzed from various perspectives.
- Reduction of redundancy: Minimizes data duplication, improving space and consistency.

BD CONNECTION TO POWER BI





Obtener datos de otro origen \rightarrow

Here are some of the questions we'll be answering in the Power BI dashboard:

- What is the distribution of patients by gender?
- How many patients smoke and how many do not smoke?
- How are patients distributed according to their socioeconomic status?
- How are patient health metrics distributed, including BMI, blood pressure (diastolic and systolic), LDL cholesterol, triglycerides, and quality of life?
- What are the leading causes of death among patients and how many deaths are attributed to each cause?

Finally, we uploaded all the folders to the github repository!