SQL Airflow pipelines

We will start by describing a simple DAG that will perform a series of simple steps in a Postgres table (create the table, load some data, count them and, finally, drop or delete the table randomly). The implementation of all DAGs are placed on the dags directory of the git repository and, in this particular case, the sql comands and even the csv data will be placed in a separate sql subdirectory:

```
# You may need to change the cd command in order to be in the right directory
# Execute this cell
cd ../dags/
ls -1 sql*
If everything went well, you will see an output similar to this:
# Do not execute this cell. Just for information
-rw-r--r-- 1 Angel wheel 3372 Mar 20 15:09 sql_airflow_dag.py
sql:
total 312
-rw-r--r-0 1 Angel
                     wheel
                            137845 Mar 13 11:06 motogp.csv
-rw-r--r-0 1 Angel
                               181 Mar 13 11:06 motogp_create_table.sql
                     wheel
-rw-r--r-- 1 Angel
                                21 Mar 13 11:06 motogp_delete_table.sql
                     wheel
                                20 Mar 13 11:06 motogp_drop_table.sql
-rw-r--r- 1 Angel
                     wheel
-rw-r--r- 1 Angel
                               469 Mar 13 11:06 motogp_load_table.py
                     wheel
-rw-r--r-0 1 Angel
                     wheel
                                30 Mar 13 11:06 motogp_select_table.sql
```

More interesting than the contents of the *.sql files (which are simple sql statements) is the file sql_airflow_dag.py. Even if you are not a python programmer or have no Airflow skills, it is advisable to review the stucture of the code to understand what and how the DAG will do:

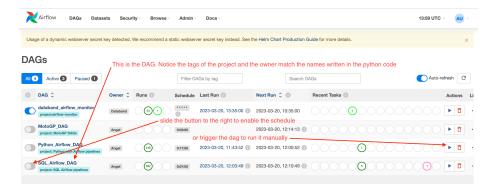
```
and-workshop > dags > 🏺 sql_airflow_dag.py > .
                                                                                                                                                                                                                                                                                    header comments
# File: sql_airflow_dag.py
# Simple DAG for the Databand hands-on workshop
                                                                                                                                                                                                                                                                                    and imports
# These are mandatory imports
from _future__ import annotations
from airflow import DAs
from airflow.operators.bash import BashOperator
from airflow.operators.python import BranchPythonOperator
from airflow.operators.postgres_operator
from dirtlow.operators.postgres_operator
from dirtlow.operators.operator.import PostgresOperator
from datetime import datetime.timedelta
from random import random
                                                                                                                                                                                                                                                                                   auxiliary function
   The header of the DAG with all its tasks.

Notice the argument postgres_conn_id:

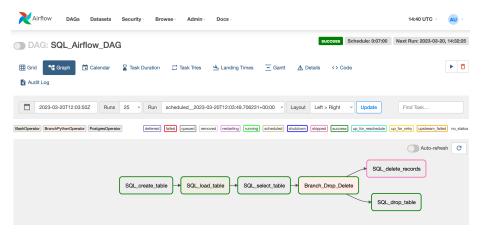
it must match a connection name in Airflow connection menu
                                                                                                                                                                                                                                                                                     DAG header
        h DAG(
dag.id="SQL_Airflow_DAG",
start_date=datetise(2023, 1, 1),
schedule_interval=timedelta(minutes=7),
catchup=false,
default_args = ('owner': 'Angel'),
tags=[
"project: SQL Airflow pipelines "
        SQL_create_table = PostgresOperator (
  task_id="SQL_create_table",
  postgres_conn_id="postgres_motogp",
  sql="sql/motogp_create_table.sql"
                                                                                                                                                                                                                                                                                        Task #1: create
         # AIRTION WALL Spawn a snell and run the python script as indicated 
SOLload table = BashOperator ( 
    task_id="SOL_load_table", 
    bash_command="python3 /opt/airflow/dags/sql/motogp_load_table.py"
                                                                                                                                                                                                                                                                                        Task #2: load
           Still because of the operation automatic
SQL_select_table = PostgresOperator (
task_id="SQL_select_table",
postgres_conn_id="postgres_motogp",
sql="sql/motogp_select_table.sql"
                                                                                                                                                                                                                                                                                        Task #3: select
         # the decision is made = randomity
Branch_Drop_Delete = BranchPythonOperator(
    task_id="Branch_Drop_Delete",
    python_callable=delete_or_drop
                                                                                                                                                                                                                                                                                        Task #4: branch
          # Delete from table...

$QL_delete_records = PostgresOperator (
    task_id="SQL_delete_records",
    postgres_conn_id="postgres_motogp",
    sql="sql/motogp_delete_table.sql"
                                                                                                                                                                                                                                                                                        Task #5: delete
         SQL_drop_table = PostgresOperator (
    task_id="SQL_drop_table",
    postgres_conn_id="postgres_motogp",
    sql="sql/motogp_drop_table.sql"
                                                                                                                                                                                                                                                                                        Task #6: drop
                                                                                                                                                                                                                                                                                        Execution sequence
     These are the task dependencies written with Airflow syntax
L_create_table >> SQL_load_table >> SQL_select_table >> Branch_Drop_Delete
anch_Drop_Delete >> SQL_delete_records
anch_Drop_Delete >> SQL_drop_table
```

As we copied this file to the Airflow containers in the previous section of this workshop, the DAG will be visible on the Airflow console.



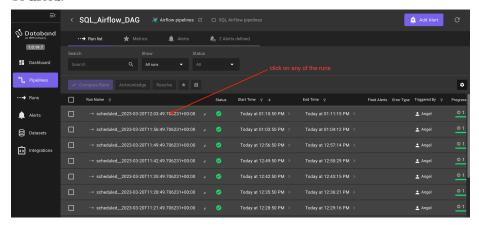
Follow the instructions of the picture above to run the DAG and click on the name of the DAG to see a graphical representation (note that the graph tab is highlighted)



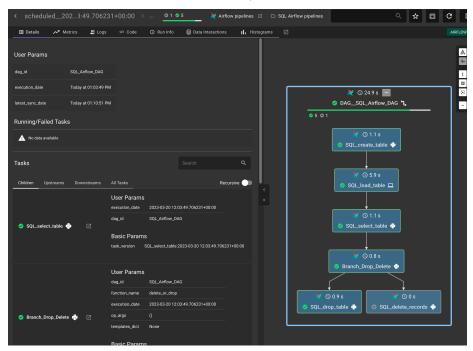
If you open the Databand main interface and navigate to the Pipelines menu on the left, you will see a list of all the pipelines, including the one we are focusing now, labeled as SQL_Airflow_DAG



If you click on the name of the pipeline, all the executions of this pipeline will be listed:



To see the deatils of each run, click on anyone of them:



It is important to remark that this DAG has no sign of Databand at all, i.e. we didn't write special line in the code and nothing implies that it will be monitored by Databand. Actually, it will be scheduled and run by Airflow, which will capture the execution data as any other DAG. The execution data will be pulled by Databand to display it as a pipeline.

The information collected by Databand will include the elapsed runtimes of each

task and its return codes. This is a basic start that will be enhanced in the next chapters where we will see more valuable information.

Next Section: Python pipelines

Previous Section: Preparation

Return to main