

Examen Final

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Ejercicio 2 – Car Rental

Consigna

Una de las empresas líderes en alquileres de automóviles solicita una serie de dashboards y reportes para poder basar sus decisiones en datos. Entre los indicadores mencionados se encuentran total de alquileres, segmentación por tipo de combustible, lugar, marca y modelo de automóvil, valoración de cada alquiler, etc.

Como Data Engineer debe crear y automatizar el pipeline para tener como resultado los datos listos para ser visualizados y responder las preguntas de negocio.

Desarrollo

1. Para crear las tablas en el datawarehouse se utiliza el siguiente script ubicado en `/home/hadoop/scripts/create_database_rental.hql`

El script se corrió manualmente por consola antes de realizar la ingesta mediante `hive -f /home/hadoop/scripts/create_database_rental.hql`

```
-- Create the database
create database car_rental_db;
use car_rental_db;

-- create tables with appropriate schema
create table
    car_rental_analytics (
        fuelType string,
        rating int,
        renterTripsTaken int,
        reviewCount int,
        city string,
        state_name string,
        owner_id int,
        rate_daily int,
        make string,
        model string,
        year int,
        -- Geo_Point string,
        -- Geo_Shape string,
        year_georef string,
        Code_State int,
        Name_State string,
        Area_Code string,
        Type string,
        state_abbreviation string,
        GNIS_Code int
    )
row format delimited
fields terminated by ',';
```

Chequeamos que el esquema de las tablas sea el correcto, con sus nombres y tipos.

```
hive> describe formatted car_rental_analytics;
OK
# col_name          data_type          comment

fueltype            string
rating              int
rentertripstaken    int
reviewcount         int
city                string
state_name          string
owner_id            int
rate_daily          int
make                string
model               string
year                int
year_georef         int
code_state          int
name_state          string
area_code           string
type                string
state_abbreviation  string
gnis_code           int

# Detailed Table Information
Database:            car_rental_db
Owner:               hadoop
CreateTime:          Tue Jun 25 02:21:26 ART 2024
LastAccessTime:      UNKNOWN
Retention:           0
Location:             hdfs://172.17.0.2:9000/user/hive/warehouse/car_rental_db.db/
Table Type:          MANAGED_TABLE
Table Parameters:
    numFiles          1
    numRows           0
    rawDataSize        0
    totalSize         483018
    transient_lastDdlTime 1719292904

# Storage Information
SerDe Library:        org.apache.hadoop.hive.serde2.lazy.LazySimpleSerDe
InputFormat:          org.apache.hadoop.mapred.TextInputFormat
OutputFormat:         org.apache.hadoop.hive ql.io.HiveIgnoreKeyTextOutputFormat
Compressed:           No
Num Buckets:          -1
Bucket Columns:       []
Sort Columns:         []
Storage Desc Params:
    field.delim        ,
    serialization.format ,
Time taken: 0.108 seconds, Fetched: 47 row(s)
```

2. Para realizar la ingesta se utiliza el siguiente script ubicado en `/home/hadoop/scripts/ingest_rental.sh`

```
rm -f /home/hadoop/landing/*.  
  
wget -O /home/hadoop/landing/CarRentalData.csv  
"https://edvaibucket.blob.core.windows.net/data-engineer-edvai/CarRental  
Data.csv?sp=r&st=2023-11-06T12:52:39Z&se=2025-11-06T20:52:39Z&sv=2022-  
11-02&sr=c&sig=J4Ddi2c7Ep230hQLPisbYaerlH472iigPwc1%2FkG80EM%3D"  
  
wget -O /home/hadoop/landing/georef.csv  
"https://dataengineerpublic.blob.core.windows.net/data-engineer/georef-  
united-states-of-america-state.csv"  
  
/home/hadoop/hadoop/bin/hdfs dfs -rm /ingest/*.  
  
/home/hadoop/hadoop/bin/hdfs dfs -put /home/hadoop/landing/*. /ingest
```

3. Para orquestar el realizar las transformaciones en PySpark se utiliza el siguiente script ubicado en `/home/hadoop/scripts/transform_rental.py`

```
from pyspark.context import SparkContext  
from pyspark.sql.session import SparkSession  
sc = SparkContext('local')  
spark = SparkSession(sc)  
from pyspark.sql.functions import *  
from pyspark.sql import HiveContext  
hc = HiveContext(sc)  
  
### Inicio del Script ###  
# Leemos los csv desde HDFS y cargamos en dataframes  
df_rental = spark.read.option("header", "true").option("sep",  
",").csv("hdfs://172.17.0.2:9000/ingest/CarRentalData.csv")  
df_georef = spark.read.option("header", "true").option("sep",  
";").csv("hdfs://172.17.0.2:9000/ingest/georef.csv")  
  
# Dropeamos columnas que no utilizaremos  
df_rental = df_rental.drop('location.country', 'location.latitude',  
'location.longitude', 'vehicle.type')  
df_georef = df_georef.drop('State FIPS Code')  
  
# Normalizamos nombres de columnas  
# Importante mantener linea vacia debajo de cada bucle for  
for column in df_rental.columns:  
    df_rental = df_rental.withColumnRenamed(column,  
column.replace('.', '_'))
```

```

df_rental =
df_rental.withColumnRenamed("location_city","city").withColumnRenamed("location_state","state_name")

for column in df_rental.columns[-3:]:
    df_rental = df_rental.withColumnRenamed(column, column.split('_')[-1])

for column in df_georef.columns:
    df_georef = df_georef.withColumnRenamed(column,
column.replace('.', '_').replace(' ', '_'))

for column in df_georef.columns:
    df_georef =
df_georef.withColumnRenamed(column, '_'.join(column.split('_')[-2:]))

df_georef = df_georef.withColumnRenamed("Year", "year_georef")

# Redondeamos los float de 'rating' y castear a int
# Vuelvo a cargar sql.functions porque levanta error
from pyspark.sql.functions import *
df_rental = df_rental.filter(column("rating").isNotNull())

# Eliminamos 'Texas'
df_rental = df_rental.filter("state_name ≠ 'TX'")

# Joineamos las tablas en 'state_name'
df_rental = df_rental.join( df_georef,
df_rental.state_name==df_georef.state_abbreviation, 'left')

# Casteamos las variables no-string
cast_cols = ['renterTripsTaken', 'reviewCount', 'owner_id', 'rate_daily',
'year', 'year_georef', 'Code_State', 'GNIS_Code']
for column in cast_cols:
    df_rental = df_rental.withColumn( column, col(column).cast('int'))

# Redondeamos los float de 'rating' y castear a int
df_rental = df_rental.withColumn( 'rating',
round('rating').cast('integer') )

# Mayusculas por minusculas en 'fuelType'
df_rental = df_rental.withColumn('fuelType', lower('fuelType') )

# Dropeo columna 'Geo_Point' y 'Geo_Shape'
# --AVERIGUAR COMO TRATAR CON LISTAS GEO-SHAPE
# EL OBJETO DE LISTAS SE CARGA INCORRECTAMENTE EN HIVE--
df_rental = df_rental.drop('Geo_Point', 'Geo_Shape')

# Creamos vistas con la data filtrada
df_rental.createOrReplaceTempView("df_rental_vista")

# Insertamos DFs filtrados en tablas de Hive
hc.sql("insert into car_rental_db.car_rental_analytics select * from
df_rental_vista;")

### Fin del Script ###

```

4. Para orquestar el proceso de ETL mediante Airflow se utilizaron dos scripts de manera que un proceso padre realice la ingesta y llame al proceso hijo para realizar las transformaciones.

Los scripts que implementan los DAGs son:

/home/hadoop/airflow/dags/dag_rental_parent_TriggerDagRun.py

/home/hadoop/airflow/dags/dag_rental_child_TriggerDagRun.py

```
from airflow import DAG
from airflow.operators.bash import BashOperator
from airflow.operators.empty import EmptyOperator
from datetime import datetime, timedelta
from airflow.operators.trigger_dagrun import TriggerDagRunOperator

args = {
    'owner' : 'airflow',
}

with DAG(
    dag_id="rental_parent",
    default_args=args,
    schedule_interval='@once',
    start_date=datetime(2020, 1, 1),
    dagrun_timeout=timedelta(minutes=60),
    tags=['ingest', 'parent'],
) as parent_dag:

    begin = EmptyOperator(
        task_id='begin_processing',
    )

    ingest = BashOperator(
        task_id='ingest_rental',
        # Don't drop the space at the end of the command or Jinja
        # will fail
        bash_command='/usr/bin/sh
/home/hadoop/scripts/ingest_rental.sh ',
    )

    transform_child_task = TriggerDagRunOperator(
        task_id="transform_child_task",
        trigger_dag_id='rental_child',
        wait_for_completion=False,
        reset_dag_run=True,
    )

    end = EmptyOperator(
        task_id='end_processing',
    )

    begin >> ingest >> transform_child_task >> end
```

```

from airflow import DAG
from airflow.operators.bash import BashOperator
from airflow.operators.empty import EmptyOperator
from datetime import datetime, timedelta

args = {
    'owner' : 'airflow',
}

with DAG(
    dag_id="rental_child",
    default_args=args,
    schedule_interval=None,
    start_date=datetime(2020, 1, 1),
    dagrun_timeout=timedelta(minutes=60),
    tags=['transform', 'child'],
) as child_dag:

    begin = EmptyOperator(
        task_id='begin_processing',
    )

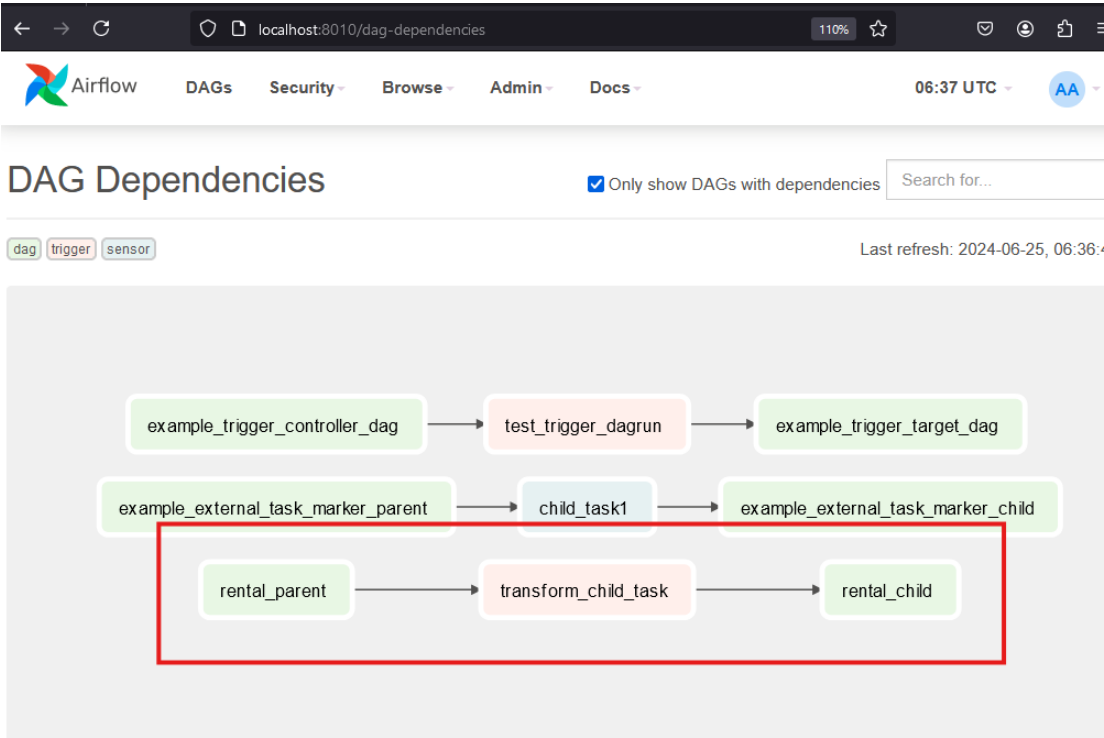
    transform_rental = BashOperator(
        task_id='transform_rental',
        # Don't drop the space at the end of the command or Jinja
will fail
        bash_command='ssh hadoop@172.17.0.2
/home/hadoop/spark/bin/spark-submit --files
/home/hadoop/hive/conf/hive-site.xml
/home/hadoop/scripts/transform_rental.py ',
    )

    end = EmptyOperator(
        task_id='end_processing',
    )

    begin >> transform_rental >> end

```

En la siguiente captura de pantalla se observa la dependencia entre DAGs generada por Airflow:



Esta captura de pantalla muestra los DAGs orquestados por Airflow:

DAGs

All 38Active 2Paused 36

Filter DAGs by tag

Search DAGs

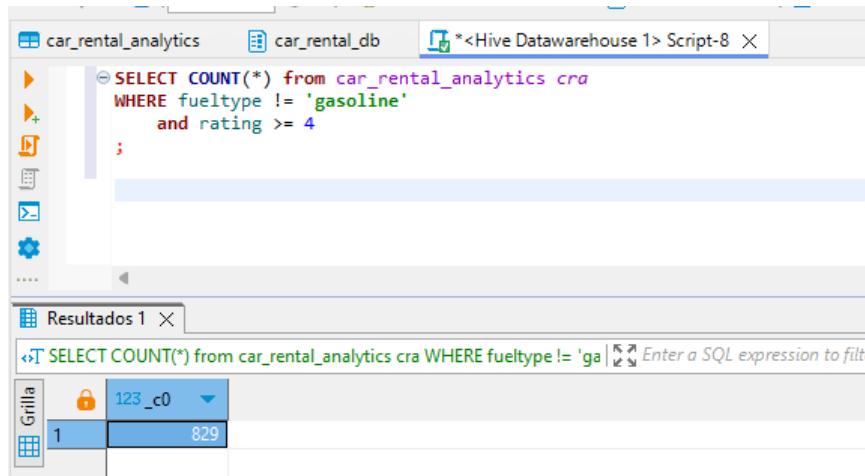
DAG	Owner	Runs	Schedule	Last Run	Next Run	Recent Tasks
<div><div>rental_child</div><div><div>child</div><div>transform</div></div></div>	airflow	<div><div>3</div></div>	None	2024-06-25, 04:17:35		<div><div>3</div></div>
<div><div>rental_parent</div><div><div>ingest</div><div>parent</div></div></div>	airflow	<div><div>1</div></div>	@once	2020-01-01, 00:00:00		<div><div>4</div></div>

« < 1 > »

Showing 1-2 of 2 DAGs

5. Querys

- a. 829 autos alquilados ecológicos con rating de al menos 4.



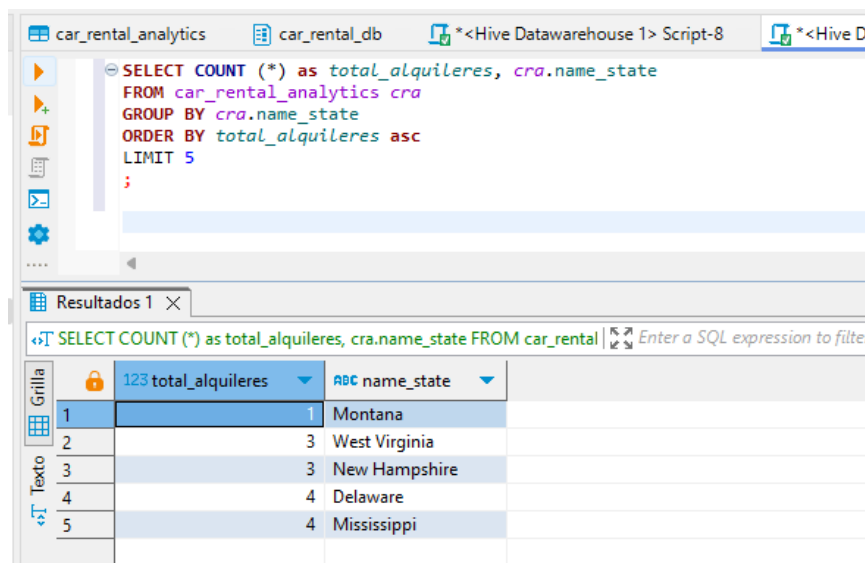
```
SELECT COUNT(*) from car_rental_analytics cra
WHERE fueltype != 'gasoline'
and rating >= 4
;
```

Resultados 1

SELECT COUNT(*) from car_rental_analytics cra WHERE fueltype != 'ga

Grilla	123_c0
1	829

- b. Los 5 estados con menor cantidad de alquileres



```
SELECT COUNT (*) as total_alquileres, cra.name_state
FROM car_rental_analytics cra
GROUP BY cra.name_state
ORDER BY total_alquileres asc
LIMIT 5
;
```

Resultados 1

SELECT COUNT (*) as total_alquileres, cra.name_state FROM car_rental

Grilla	123 total_alquileres	ABC name_state
1	1	Montana
2	3	West Virginia
3	3	New Hampshire
4	4	Delaware
5	4	Mississippi

c. Los 10 modelos (junto con su marca) de autos más rentados

d.

6. _

7. Una arquitectura equivalente podría implementarse en GCP utilizando:

- _ para la Ingesta de csv.
- Google Cloud Storage (GCS Bucket) como datawarehouse.
- DataProc para las Transformaciones.
- BigQuery para las consultas.
- Dataflow para la orquestación del pipeline.
- Looker para las visualizaciones.