

Curso Data Engineer: Creando un pipeline de datos

Agenda



- Transformaciones con SCALA
- Transformaciones en cloud (Spark)
- Orquestación
- Ejercitación

Transformaciones SCALA



Para ingresar a Spark Python

pyspark

Para ingresar a Spark Scala

spark-shell

Para salir de Spark Python

exit()

Para salir de Spark Scala

:q



Extract

```
scala> val df = spark.read.option("header", "true").csv("/ingest/yellow_tripdata_2021-01.csv")
df: org.apache.spark.sql.DataFrame = [VendorID: string, tpep_pickup_datetime: string ... 16 more fields]
```



Transform

```
scala> val df_cast = df.select( df("tpep_pickup_datetime").cast("date"), df("passenger_count").cast("int"), df("trip_distance").cast("float") )
df_cast: org.apache.spark.sql.DataFrame = [tpep_pickup_datetime: date, passenger_count: int ... 1 more field]
scala> val df_trans = df_cast.filter( (df_cast("passenger_count") > 1) && (df("trip_distance") > 10) )
df_trans: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [tpep_pickup_datetime: date, passenger_count: int ... 1 more field]
```



Load

scala> df_trans.write.insertInto("tripsdb.trips")

Transformaciones en cloud

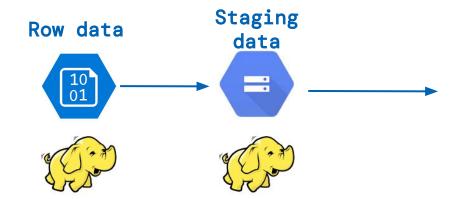


Row data

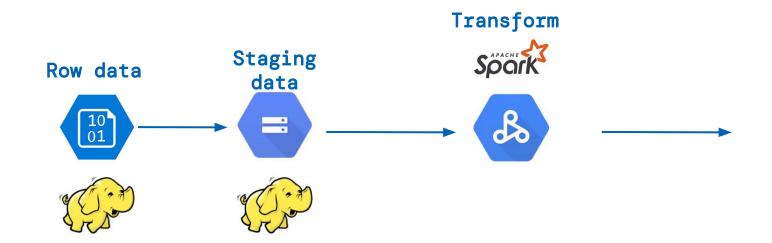




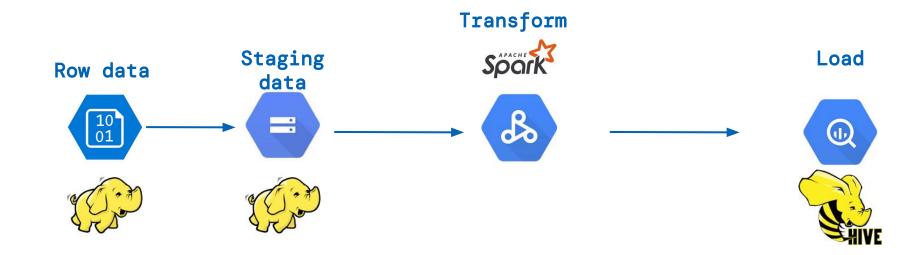




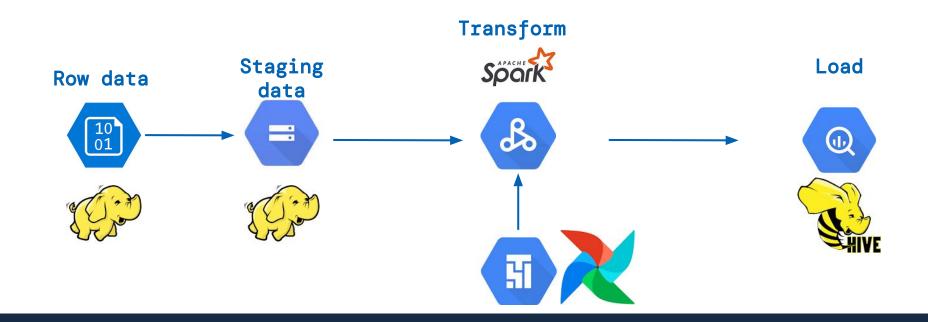




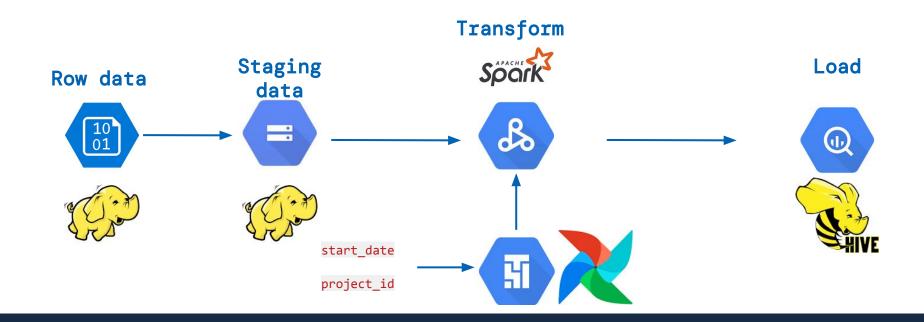








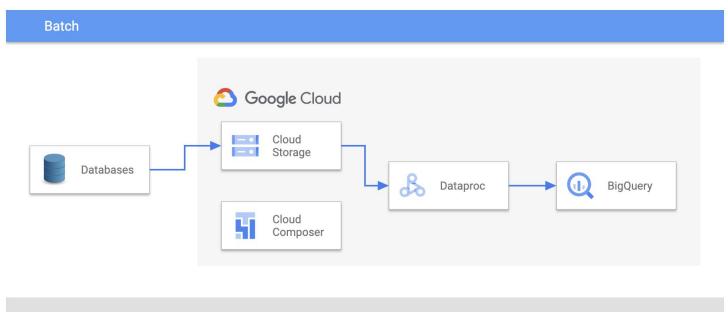




Orquestación

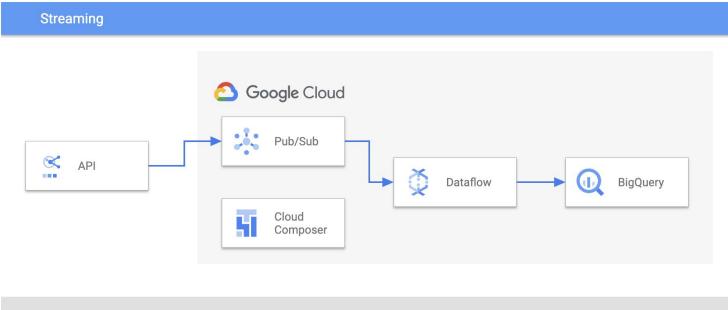
Batch process





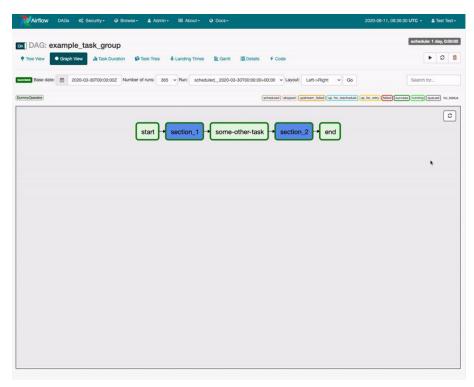
Streaming process





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Apache Airflow





Apache Airflow

Importing Modules

An Airflow pipeline is just a Python script that happens to define an Airflow DAG object. Let's start by importing the libraries we will need.

```
from datetime import datetime, timedelta
from textwrap import dedent

# The DAG object; we'll need this to instantiate a DAG
from airflow import DAG

# Operators; we need this to operate!
from airflow.operators.bash import BashOperator
```



Apache Airflow

Default Arguments

We're about to create a DAG and some tasks, and we have the choice to explicitly pass a set of arguments to each task's constructor (which would become redundant), or (better!) we can define a dictionary of default parameters that we can use when creating tasks.

```
airflow/example_dags/tutorial.py
                                                                                                                    [source]
# These args will get passed on to each operator
# You can override them on a per-task basis during operator initialization
default_args={
    "depends_on_past": False,
    "email": ["airflow@example.com"],
    "email_on_failure": False,
    "email_on_retry": False,
    "retries": 1.
    "retry_delay": timedelta(minutes=5),
    # 'queue': 'bash_queue',
    # 'pool': 'backfill',
    # 'priority_weight': 10,
    # 'end_date': datetime(2016, 1, 1),
    # 'wait_for_downstream': False,
    # 'sla': timedelta(hours=2),
    # 'execution_timeout': timedelta(seconds=300),
    # 'on_failure_callback': some_function,
    # 'on_success_callback': some_other_function,
    # 'on_retry_callback': another_function,
    # 'sla_miss_callback': yet_another_function,
    # 'trigger_rule': 'all_success'
```

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Apache Airflow

Instantiate a DAG

We'll need a DAG object to nest our tasks into. Here we pass a string that defines the default argument dictionary that we just defined and define a schedule of 1 day for the DAG.

```
airflow/example_dags/tutorial.py
                                                                                                                                 source
with DAG(
   "tutorial",
   # These args will get passed on to each operator
   # You can override them on a per-task basis during operator initialization
   default_args={
        "depends_on_past": False,
       "email": ["airflow@example.com"].
       "email on failure": False.
       "email_on_retry": False,
       "retries": 1,
       "retry_delay": timedelta(minutes=5),
       # 'queue': 'bash_queue',
       # 'pool': 'backfill',
       # 'priority_weight': 10,
       # 'end_date': datetime(2016, 1, 1),
       # 'wait_for_downstream': False,
       # 'sla': timedelta(hours=2),
       # 'execution_timeout': timedelta(seconds=300),
       # 'on_failure_callback': some_function,
       # 'on_success_callback': some_other_function,
       # 'on_retry_callback': another_function,
       # 'sla_miss_callback': yet_another_function,
       # 'trigger_rule': 'all_success'
   description="A simple tutorial DAG",
   schedule=timedelta(days=1),
   start_date=datetime(2021, 1, 1),
   catchup=False,
   tags=["example"],
) as dag:
```

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Apache Airflow

Operators

An Operator is conceptually a template for a predefined Task, that you can just define declaratively inside your DAG:

```
with DAG("my-dag") as dag:
    ping = SimpleHttpOperator(endpoint="http://example.com/update/")
    email = EmailOperator(to="admin@example.com", subject="Update complete")
    ping >> email
```

Airflow has a very extensive set of operators available, with some built-in to the core or pre-installed providers. Some popular operators from core include:

- BashOperator executes a bash command
- PythonOperator calls an arbitrary Python function
- EmailOperator sends an email



Apache Airflow

If the operator you need isn't installed with Airflow by default, you can probably find it as part of our huge set of community provider packages. Some popular operators from here include:

- SimpleHttpOperator
- MySqlOperator
- PostgresOperator
- MsSqlOperator
- OracleOperator
- JdbcOperator
- DockerOperator
- HiveOperator
- S3FileTransformOperator
- PrestoToMySqlOperator
- SlackAPIOperator



Apache Airflow

Tasks

To use an operator in a DAG, you have to instantiate it as a task. Tasks determine how to execute your operator's work within the context of a DAG.

In the following example, we instantiate the BashOperator as two separate tasks in order to run two separate bash scripts. The first argument for each instantiation, task_id, acts as a unique identifier for the task.

```
airflow/example_dags/tutorial.py

t1 = BashOperator(
   task_id="print_date",
   bash_command="date",
)

t2 = BashOperator(
   task_id="sleep",
   depends_on_past=False,
   bash_command="sleep 5",
   retries=3,
)
```

Ejercicio

Apache Airflow

Username: airflow Password: airflow



http://<direccion ip>:8010



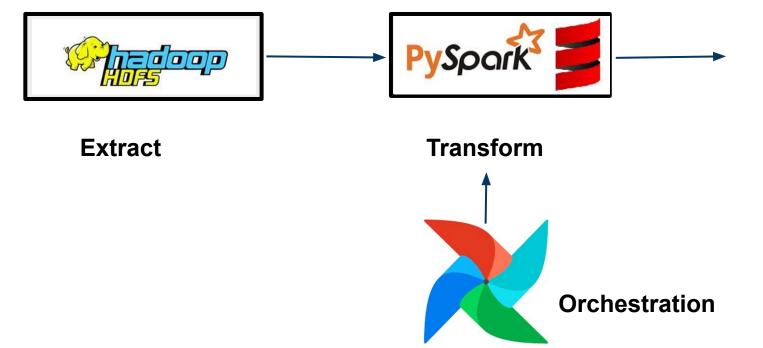
ejercicio

Crear un pipeline a través de un DAG, creando las tareas:

- Ingest
- Transform
- Load

Load

https://dataengineerpublic.blob.core.windows.net/data-engineer/yellow_tripdata_2021-01.csv





Load

Workaround possible issue

```
root@0485e86b7577:/# ping www.google.com
ping: www.google.com:_Temporary failure in name resolution
```

```
root@0485e86b7577:/# wget -P /home/hadoop/landing/ https://data-engineer-edvai.s3.amazonaws.com/yellow_tripdata_2021-01.csv
--2023-04-17 18:52:31-- https://data-engineer-edvai.s3.amazonaws.com/yellow_tripdata_2021-01.csv
Resolving data-engineer-edvai.s3.amazonaws.com (data-engineer-edvai.s3.amazonaws.com)... failed: Tempo rary failure in name resolution.
wget: unable to resolve host address 'data-engineer-edvai.s3.amazonaws.com'
```

Dentro de consola_Hadoop, modificar el archivo /etc/resolv.conf, comentado todos los DNS y dejando solamente el de Google, de la siguiente manera:

nameserver 8.8.8.8 options timeout:2 attempts:5