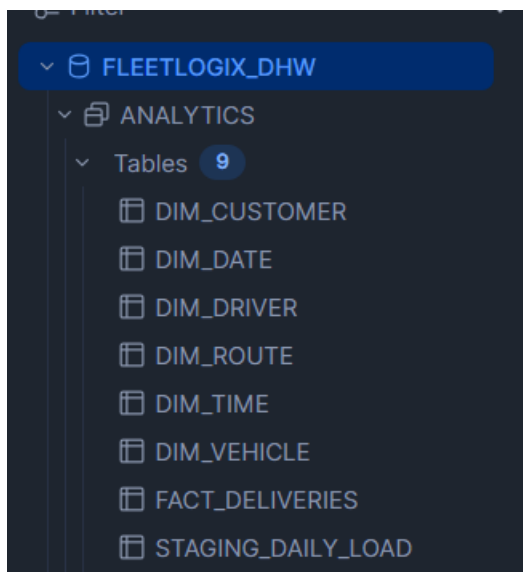


Avance 3: Análisis de Snowflake

Para poder armar el flujo de datos de nuestra base de Postgre a Snowflake lo primero que hay que hacer es crear la estructura de nuestro sistema OLTP:

```
1  --- CREAMOS UN VIRTUAL WAREHOUSE ---
2  USE ROLE ACCOUNTADMIN;
3  CREATE WAREHOUSE IF NOT EXISTS FLEETLOGIX_WH WITH WAREHOUSE_SIZE = 'XSMALL' AUTO_SUSPEND = 60;
4
5  ---CREAMOS LA BASE DE FLEETLOGIX---
6  CREATE DATABASE FleetLogix_dhw;
7  USE DATABASE FleetLogix_dhw;
8
9  ---CREAMOS EL ESQUEMA DE LA BASE DE FLEETLOGIX---
10 CREATE SCHEMA IF NOT EXISTS ANALYTICS;
11 USE SCHEMA ANALYTICS;
12
```



Luego creamos las conexiones entre las herramientas

```
import pandas as pd
import numpy as np
from datetime import datetime
import psycopg2
import snowflake.connector

# PostgreSQL
conn_pg = psycopg2.connect(
    host="localhost",
    database="fleetlogix",
    user="postgres",
    password="Spriest123"
)
cur_pg = conn_pg.cursor()

# Snowflake
conn_sf = snowflake.connector.connect(
    user="ENZ0ZAMBON",
    password='AdiiraelSpriest12345',
    account='GTUNIRG-PU45327',
    warehouse='FLEETLOGIX_WH',
    database='FleetLogix_dhw',
    schema='ANALYTICS'
)
cur_sf = conn_sf.cursor()

[71]

... INFO:snowflake.connector.connection:Snowflake Connector for Python Version: 4.0.0, Python Version: 3.13.0, Platform: Windows-10-10.0.19043-SP0
INFO:snowflake.connector.connection:Connecting to GLOBAL Snowflake domain

def insert_in_batches(cursor, sql, data, batch_size=5000, table_name=""):
    for i in range(0, len(data), batch_size):
        batch = data[i:i+batch_size]
        cursor.executemany(sql, batch)
        conn_sf.commit()
        print(f"[{datetime.now()}] {table_name} - batch {i//batch_size+1} success - {len(batch)} registros")

[72]
```

Posteriormente cargamos la información de las tablas en multiples dataframes:

```
df_deliveries = pd.read_sql("SELECT * FROM deliveries", conn_pg)
df_trip = pd.read_sql("SELECT * FROM trips", conn_pg)
df_vehicle = pd.read_sql("SELECT * FROM vehicles", conn_pg)
df_driver = pd.read_sql("SELECT * FROM drivers", conn_pg)
df_route = pd.read_sql("SELECT * FROM routes", conn_pg)
df_maint = pd.read_sql("SELECT vehicle_id, MAX(maintenance_date) AS last_maintenance_date FROM maintenance GROUP BY vehicle_id", conn_pg)

[73]

... C:\Users\Enzo\AppData\Local\Temp\ipykernel_19296\2084301068.py:1: UserWarning: pandas only supports SQLAlchemy connectable (engine/connection) or d
df_deliveries = pd.read_sql("SELECT * FROM deliveries", conn_pg)
C:\Users\Enzo\AppData\Local\Temp\ipykernel_19296\2084301068.py:2: UserWarning: pandas only supports SQLAlchemy connectable (engine/connection) or d
df_trip = pd.read_sql("SELECT * FROM trips", conn_pg)
C:\Users\Enzo\AppData\Local\Temp\ipykernel_19296\2084301068.py:3: UserWarning: pandas only supports SQLAlchemy connectable (engine/connection) or d
df_vehicle = pd.read_sql("SELECT * FROM vehicles", conn_pg)
C:\Users\Enzo\AppData\Local\Temp\ipykernel_19296\2084301068.py:4: UserWarning: pandas only supports SQLAlchemy connectable (engine/connection) or d
df_driver = pd.read_sql("SELECT * FROM drivers", conn_pg)
C:\Users\Enzo\AppData\Local\Temp\ipykernel_19296\2084301068.py:5: UserWarning: pandas only supports SQLAlchemy connectable (engine/connection) or d
df_route = pd.read_sql("SELECT * FROM routes", conn_pg)
C:\Users\Enzo\AppData\Local\Temp\ipykernel_19296\2084301068.py:6: UserWarning: pandas only supports SQLAlchemy connectable (engine/connection) or d
df_maint = pd.read_sql("SELECT vehicle_id, MAX(maintenance_date) AS last_maintenance_date FROM maintenance GROUP BY vehicle_id", conn_pg)

df_deliveries['scheduled_datetime'] = pd.to_datetime(df_deliveries['scheduled_datetime'])
df_trip['departure_datetime'] = pd.to_datetime(df_trip['departure_datetime'])
df_trip['arrival_datetime'] = pd.to_datetime(df_trip['arrival_datetime'])

[ ]
```

Y poblamos las tablas de Snowflake, procesamiento de datos mediante acorde a la estructura de las tablas previamente creadas

```
# ----- Dimensión VEHICLES -----
# Convertir acquisition_date a datetime
df_vehicle['acquisition_date'] = pd.to_datetime(df_vehicle['acquisition_date'], errors='coerce')

# Traer la última fecha de mantenimiento por vehículo
df_maint['last_maintenance_date'] = pd.to_datetime(df_maint['last_maintenance_date'], errors='coerce')
df_maint_last = df_maint.groupby('vehicle_id', as_index=False)['last_maintenance_date'].max()

# Merge seguro con df_vehicle
if 'last_maintenance_date' in df_vehicle.columns:
    df_vehicle = df_vehicle.drop(columns=['last_maintenance_date'])

df_vehicle = df_vehicle.merge(df_maint_last, on='vehicle_id', how='left')

# Calcular edad en meses
df_vehicle['age_months'] = ((pd.Timestamp.today() - df_vehicle['acquisition_date']).dt.days // 30).fillna(0).astype(int)

# Asegurarse de que last_maintenance_date exista
if 'last_maintenance_date' not in df_vehicle.columns:
    df_vehicle['last_maintenance_date'] = pd.NaT

# Convertir fechas a string para Snowflake
df_vehicle['acquisition_date'] = df_vehicle['acquisition_date'].dt.strftime('%Y-%m-%d %H:%M:%S')
df_vehicle['last_maintenance_date'] = df_vehicle['last_maintenance_date'].dt.strftime('%Y-%m-%d %H:%M:%S')

# Preparar registros para insert
vehicle_records = list(df_vehicle[['vehicle_id', 'license_plate', 'vehicle_type', 'capacity_kg', 'fuel_type',
                                   'acquisition_date', 'age_months', 'status', 'last_maintenance_date']].itertuples(index=False, name=None))

# Insertar en batches
insert_in_batches(cur_sf, """
INSERT INTO dim_vehicle (vehicle_id, license_plate, vehicle_type, capacity_kg, fuel_type, acquisition_date, age_months, status, last_maintenance_date)
VALUES (%s,%s,%s,%s,%s,%s,%s,%s,%s)
""", vehicle_records, batch_size=5000, table_name="dim_vehicle")

[75]
... [2025-10-13 22:38:50.947028] dim_vehicle - batch 1 success - 200 registros
```

```
# ----- Dimensión DRIVERS -----

# Convertir fechas a datetime
df_driver['hire_date'] = pd.to_datetime(df_driver['hire_date'], errors='coerce')
df_driver['license_expiry'] = pd.to_datetime(df_driver['license_expiry'], errors='coerce')

# Calcular experiencia en meses
df_driver['experience_months'] = ((pd.Timestamp.today() - df_driver['hire_date']).dt.days // 30).fillna(0).astype(int)

# Crear full_name
df_driver['full_name'] = df_driver['first_name'].fillna('') + ' ' + df_driver['last_name'].fillna('')

# Contar viajes por conductor
df_trip_counts = df_trip.groupby('driver_id').size().reset_index(name='total_trips')

# Hacer merge
if 'total_trips' in df_driver.columns:
    df_driver = df_driver.drop(columns=['total_trips'])

df_driver = df_driver.merge(df_trip_counts, on='driver_id', how='left')

# Rellenar NaN con 0
df_driver['total_trips'] = df_driver['total_trips'].fillna(0)
df_driver['completed_trips'] = df_driver['total_trips']

# Categoría de desempeño evitando división por 0
df_driver['performance_category'] = np.where(
    df_driver['total_trips'] == 0, 'Bajo', # Si no hizo viajes
    np.where(df_driver['completed_trips']/df_driver['total_trips'] > 0.7, 'Alto',
    np.where(df_driver['completed_trips']/df_driver['total_trips'] > 0.5, 'Medio', 'Bajo'))

# Convertir fechas a string para Snowflake (maneja NaT)
df_driver['hire_date'] = df_driver['hire_date'].dt.strftime('%Y-%m-%d %H:%M:%S')
df_driver['license_expiry'] = df_driver['license_expiry'].dt.strftime('%Y-%m-%d %H:%M:%S')

# Reemplazar NaN por None para Snowflake
df_driver = df_driver.where(pd.notnull(df_driver), None)

# Preparar registros
driver_records = list(df_driver[['driver_id', 'employee_code', 'full_name', 'license_number', 'license_expiry',
                                 'phone', 'hire_date', 'experience_months', 'status', 'performance_category']].itertuples(index=False, name=None))

# Insertar en batches
insert_in_batches(cur_sf, """
INSERT INTO dim_driver (driver_id, employee_code, full_name, license_number, license_expiry, phone, hire_date, experience_months, status, performance_category)
VALUES (%s,%s,%s,%s,%s,%s,%s,%s,%s,%s)
""", driver_records, batch_size=5000, table_name="dim_driver")

[76]
```

```
# ----- Dimensión ROUTES -----
# Llenar valores nulos en numéricos
df_route[['distance_km','estimated_duration_hours','toll_cost']] = df_route[['distance_km','estimated_duration_hours','toll_cost']].fillna(0)

# Asegurar tipos
df_route['distance_km'] = df_route['distance_km'].astype(float)
df_route['estimated_duration_hours'] = df_route['estimated_duration_hours'].astype(float)
df_route['toll_cost'] = df_route['toll_cost'].astype(float)

# Renombrar columnas si es necesario
df_route = df_route.rename(columns={
    'origin': 'origin_city',
    'destination': 'destination_city'
})

# Preparar registros
route_records = list(df_route[['route_id', 'route_code','origin_city','destination_city','distance_km','estimated_duration_hours','toll_cost']].iteruples(index=False, name=None))

# Insertar en batches
insert_in_batches(cur_sf, """
INSERT INTO dim_route (route_id, route_code, origin_city, destination_city, distance_km, estimated_duration_hours, toll_cost)
VALUES (%s,%s,%s,%s,%s,%s,%s)
""", route_records, batch_size=5000, table_name="dim_route")

[77]
```

... [2025-10-13 22:38:55.344954] dim_route - batch 1 success - 50 registros

```
# ----- Dimensión CUSTOMER -----
df_trip = df_trip.merge(df_route[['route_id','destination_city']], on='route_id', how='left')
df_customer = df_deliveries.merge(df_trip[['trip_id','destination_city']], on='trip_id', how='left')
df_customer_group = df_customer.groupby(['customer_name','destination_city']).size().reset_index(name='total_deliveries')

def cat_func(x):
    if x==1: return 'Ocasional'
    elif x==2: return 'Regular'
    else: return 'Habitual'

df_customer_group['customer_category'] = df_customer_group['total_deliveries'].apply(cat_func)
customers_dim = [(i+1, row['customer_name'], row['destination_city'], datetime.today(), row['total_deliveries'], row['customer_category'])
                 for i,row in df_customer_group.iterrows()]

insert_in_batches(cur_sf, """
INSERT INTO dim_customer (customer_key, customer_name, city, first_delivery_date, total_deliveries, customer_category)
VALUES (%s,%s,%s,%s,%s,%s)
""", customers_dim, batch_size=5000, table_name="dim_customer")

[ ]
```

... [2025-10-13 22:39:11.051824] dim_customer - batch 1 success - 5000 registros
[2025-10-13 22:39:13.785869] dim_customer - batch 2 success - 5000 registros
[2025-10-13 22:39:16.333784] dim_customer - batch 3 success - 5000 registros
[2025-10-13 22:39:19.274304] dim_customer - batch 4 success - 5000 registros
[2025-10-13 22:39:22.730543] dim_customer - batch 5 success - 5000 registros
[2025-10-13 22:39:26.134172] dim_customer - batch 6 success - 5000 registros
[2025-10-13 22:39:30.029869] dim_customer - batch 7 success - 5000 registros
[2025-10-13 22:39:33.100676] dim_customer - batch 8 success - 5000 registros
[2025-10-13 22:39:35.737353] dim_customer - batch 9 success - 5000 registros
[2025-10-13 22:39:38.592063] dim_customer - batch 10 success - 5000 registros
[2025-10-13 22:39:41.495406] dim_customer - batch 11 success - 5000 registros
[2025-10-13 22:39:44.085903] dim_customer - batch 12 success - 5000 registros
[2025-10-13 22:39:46.971557] dim_customer - batch 13 success - 5000 registros
[2025-10-13 22:39:49.855782] dim_customer - batch 14 success - 5000 registros
[2025-10-13 22:39:52.794118] dim_customer - batch 15 success - 5000 registros
[2025-10-13 22:39:55.579416] dim_customer - batch 16 success - 5000 registros
[2025-10-13 22:39:58.100893] dim_customer - batch 17 success - 5000 registros
[2025-10-13 22:40:01.778784] dim_customer - batch 18 success - 5000 registros
[2025-10-13 22:40:04.689977] dim_customer - batch 19 success - 5000 registros
[2025-10-13 22:40:07.244356] dim_customer - batch 20 success - 5000 registros
[2025-10-13 22:40:10.068254] dim_customer - batch 21 success - 5000 registros
[2025-10-13 22:40:12.481030] dim_customer - batch 22 success - 5000 registros
[2025-10-13 22:40:15.561460] dim_customer - batch 23 success - 5000 registros
[2025-10-13 22:40:18.576553] dim_customer - batch 24 success - 5000 registros
[2025-10-13 22:40:21.834860] dim_customer - batch 25 success - 5000 registros
[2025-10-13 22:40:22.723200] dim_customer - batch 26 success - 828 registros

171
172
173
174
175

SELECT *

FROM fact_deliveries

LIMIT 10;

Results (just now)

TableChart

10 rows

113ms

	# DELIVERY_KEY	# DATE_KEY	# SCHEDULED_TIME_KEY	# DELIVERED_TIME_KEY	# VEHICLE_KEY	# DRIVER_KEY	# ROUTE_K
1	1	20231216	605	621	97	215	
2	2	20231216	605	621	97	715	
3	3	20231216	648	651	97	215	
4	4	20231216	648	651	97	715	
5	5	20231216	731	737	97	215	
6	6	20231216	731	737	97	715	
7	7	20231216	813	822	97	215	
8	8	20231216	813	822	97	715	
9	9	20231216	856	922	97	215	
10	10	20231216	856	922	97	715	

176
177
178
179

SELECT *

FROM dim_date

LIMIT 10;

Results (just now)

TableChart

10 rows

957ms

	# DATE_KEY	FULL_DATE	# DAY_OF_WEEK	DAY_NAME	# DAY_OF_MONTH	# DAY_OF_YEAR	# WEEK_OF_YEAR	# MON
1	20231024	2023-10-24	2	Tuesday	24	297	43	
2	20230204	2023-02-04	6	Saturday	4	35	5	
3	20241116	2024-11-16	6	Saturday	16	321	46	
4	20230303	2023-03-03	5	Friday	3	62	9	
5	20230325	2023-03-25	6	Saturday	25	84	12	
6	20241023	2024-10-23	3	Wednesday	23	297	43	
7	20230124	2023-01-24	2	Tuesday	24	24	4	
8	20230427	2023-04-27	4	Thursday	27	117	17	
9	20240104	2024-01-04	4	Thursday	4	4	1	
10	20230127	2023-01-27	5	Friday	27	27	4	

```

179
180 SELECT *
181 FROM dim_time
182 LIMIT 10;
183
184 SELECT *

```

Results (just now)

Table Chart 10 rows 679ms

	# TIME_KEY	# HOUR	# MINUTE	# SECOND	TIME_OF_DAY	HOUR_24	HOUR_12	AM_PM	IS_BUSINESS_HOUR
1	0	0	0	0	Madrugada	00:00	12:00 AM	AM	FALSE
2	100	0	1	0	Madrugada	00:01	12:01 AM	AM	FALSE
3	200	0	2	0	Madrugada	00:02	12:02 AM	AM	FALSE
4	300	0	3	0	Madrugada	00:03	12:03 AM	AM	FALSE
5	400	0	4	0	Madrugada	00:04	12:04 AM	AM	FALSE
6	500	0	5	0	Madrugada	00:05	12:05 AM	AM	FALSE
7	600	0	6	0	Madrugada	00:06	12:06 AM	AM	FALSE
8	700	0	7	0	Madrugada	00:07	12:07 AM	AM	FALSE
9	800	0	8	0	Madrugada	00:08	12:08 AM	AM	FALSE
10	900	0	9	0	Madrugada	00:09	12:09 AM	AM	FALSE

```

184 SELECT *
185 FROM dim_route
186 LIMIT 10;
187
188 SELECT *
189 FROM dim_vehicle
190 LIMIT 10;

```

Results (just now)

Table Chart 10 rows 593ms

	# ROUTE_KEY	# ROUTE_ID	ROUTE_CODE	ORIGIN_CITY	DESTINATION_CITY	DISTANCE_KM	ESTIMATED_DURATION
1	1	1	R0001	Buenos Aires	Córdoba	442.99	
2	2	2	R0002	Córdoba	Rosario	226.58	
3	3	3	R0003	Bahía Blanca	La Plata	188.88	
4	4	4	R0004	Buenos Aires	Mar del Plata	569.00	
5	5	5	R0005	Salta	La Plata	373.00	
6	6	6	R0006	La Plata	Rosario	1199.26	
7	7	7	R0007	Salta	La Plata	369.13	
8	8	8	R0008	Mar del Plata	Córdoba	29.56	
9	9	9	R0009	Bahía Blanca	Mar del Plata	1086.90	
10	10	10	R0010	Salta	Rosario	771.21	

188
189
190
191
192

```
SELECT *
FROM dim_vehicle
LIMIT 10;
```

Results (just now)

TableChart

🔍🔗10 rows ⓘ340ms📉📄

	# VEHICLE_KEY	# VEHICLE_ID	🔗 LICENSE_PLATE	🔗 VEHICLE_TYPE	# CAPACITY_KG	🔗 FUEL_TYPE	📅 ACQUISITION_DATE	
1	1	1	AL 0365 FG	Camión grande	18074.00	Nafta	2020-06-19	
2	2	2	6833 RDK	Camión mediano	7143.00	Eléctrico	2019-01-14	
3	3	3	0891 FVJ	Camión grande	18067.00	Eléctrico	2016-03-16	
4	4	4	J 0819 WU	Camión grande	15456.00	Diesel	2019-09-25	
5	5	5	8186 PTG	Camión grande	13791.00	Diesel	2017-12-11	
6	6	6	CO 7874 BM	Camión grande	13628.00	Eléctrico	2016-11-16	
7	7	7	A 1281 LA	Motocicleta	164.00	Eléctrico	2017-08-15	
8	8	8	GI 9632 CX	Van	1653.00	Eléctrico	2019-03-15	
9	9	9	PO 2544 PI	Motocicleta	121.00	Diesel	2018-02-14	
10	10	10	IB 8844 KB	Camión mediano	6837.00	Diesel	2019-03-23	

192
193
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195

```
SELECT *
FROM dim_customer
LIMIT 10;
```

Results (just now)

TableChart

🔍🔗10 rows ⓘ721ms📉📄

	# CUSTOMER_KEY	# CUSTOMER_ID	🔗 CUSTOMER_NAME	🔗 CITY	📅 FIRST_DELIVERY_DATE	# TOTAL_DELIVERIES	
1	1	1	Aarón Abraham Frutos Abad	Mar del Plata	2025-10-14	1	C
2	2	2	Aarón Aguilera Roig	Córdoba	2025-10-14	1	C
3	3	3	Aarón Agullo-Yáñez	Bahía Blanca	2025-10-14	1	C
4	4	4	Aarón Alba Noguera	Mar del Plata	2025-10-14	1	C
5	5	5	Aarón Alcalá Madrid	Bahía Blanca	2025-10-14	1	C
6	6	6	Aarón Alfaro Manzanares	Mar del Plata	2025-10-14	1	C
7	7	7	Aarón Alfaro-Minguez	Buenos Aires	2025-10-14	1	C
8	8	8	Aarón Almagro Calatayud	La Plata	2025-10-14	1	C
9	9	9	Aarón Amigó Dalmau	Buenos Aires	2025-10-14	1	C
10	10	10	Aarón Amor Agustín	Bahía Blanca	2025-10-14	1	C

Y las vistas

247
248
249
250
251
252
253

```
SELECT *
FROM v_sales_deliveries;

SELECT *
FROM v_operations_deliveries;
```

Results (35 minutes ago)

TableChart

🔍🔗24,886 rows ⓘ1.3s📉📄

	📅 FULL_DATE	🔗 CUSTOMER_NAME	🔗 ORIGIN_CITY	🔗 DESTINATION_CITY	# PACKAGE_WEIGHT_KG	🔗 DELIVERY_STATUS	# REVENUE
1	2025-08-01	Oriana del Carmona	Buenos Aires	Córdoba	5150.57	ON_TIME	
2	2025-08-01	Oriana del Carmona	Buenos Aires	Córdoba	5150.57	ON_TIME	
3	2023-03-24	Oriana del Isern	La Plata	Córdoba	4899.77	ON_TIME	
4	2023-03-24	Oriana del Isern	La Plata	Córdoba	4899.77	ON_TIME	
5	2025-09-11	Osvaldo Cañellas Alcara	La Plata	Córdoba	5534.39	ON_TIME	
6	2025-09-11	Osvaldo Cañellas Alcara	La Plata	Córdoba	5534.39	ON_TIME	
7	2024-08-14	Osvaldo Donaire-Saez	Mar del Plata	Córdoba	4072.32	ON_TIME	
8	2024-08-14	Osvaldo Donaire-Saez	Mar del Plata	Córdoba	4072.32	ON_TIME	

250
251
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253

SELECT *
FROM v_operations_deliveries;

Results (just now)

TableChart

4,120 rows2.5s

	FULL_DATE	HORA	LICENSE_PLATE	CONDUCTOR	CUSTOMER_NAME	DELIVERY_TIME_MINUTES	DELAY_MINUT
1	2024-11-23	00:03	SS 8047 JM	Francisca Espinosa	Cecilio del Camacho	24	
2	2024-11-23	00:03	SS 8047 JM	Francisca Espinosa	Cecilio del Camacho	24	
3	2023-05-15	00:00	B 4544 GO	Julio César Planas	Cirino Baños	16	
4	2023-05-15	00:00	B 4544 GO	Julio César Planas	Cirino Baños	16	
5	2024-07-23	00:16	3562 DGS	Carlota Lloret	Miriam Amaya Cortina	23	
6	2024-07-23	00:16	3562 DGS	Carlota Lloret	Miriam Amaya Cortina	23	
7	2024-05-24	00:07	0402 HSX	Amando Velázquez	Sonia Godoy Tejada	14	
8	2024-05-24	00:07	0402 HSX	Amando Velázquez	Sonia Godoy Tejada	14	

Ademas ejecutamos el script para la automatización de la ingesta diaria