

Data Engineering Program Capstone 3 – Outline Property of Rizky Fajar Aditya

## **Table of Contents:**

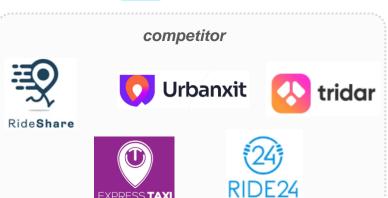
- 1 Background and Study Case
- 2 Project Development and documentation
- 3 Project Results
- 4 Website scraping documentation

# Study Case: GoHailing! company is facing risks that threatens its business sustainability

#### Tight market competition

GoHailing!, a provider of on-demand transportation services, is facing intense competition due to the rise of new players with the same business model, increasing market pressure.

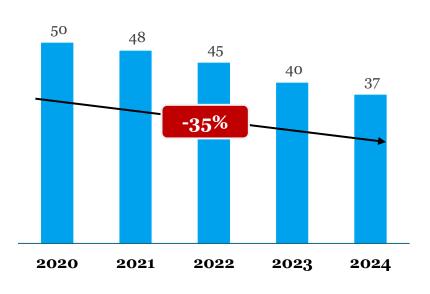




#### Revenue declining sharp

Go-Hailing saw its revenue declining by **35%** over the past few years, significantly threatening its sustainability amid rising costs.

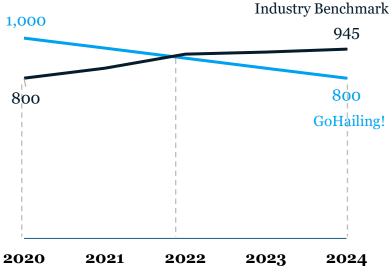
**Go-Hailing revenue**, in IDR Bn 2020 - 2024



#### **Productivity decreases**

GoHailing's business productivity has declined significantly compared to industry benchmarks, signaling increased customer and driver partner churn.

**Productivity competition**, in rides per hour 2020 - 2024



Data Engineering Program

## What kind of data that can help GoHailing in navigating those risks?

As competition intensifies, collecting insights from customers and drivers is crucial for developing an effective contingency plan

### 1 Customer insights

- How many customers that are rarely using the app? How can we stimulate them to use more our app?
- What is the ratio of cancelled rides compared to completed rides?
- How much is the potential revenue loss due to canceled rides?

### 2 Driver insights

- How many drivers that are currently registered within the database?
- What is the ratio of idle driver compared to the active ones in generating revenue?
- ➤ How can we stimulate driver to be more active and gain more revenue?

#### **Contingency Plan**

With all these data available, GoHailing! can formulate contingency plan to mitigate the risks, it could be through:



Driver incentives program



Customer promo program

Currently these data are unavailable. GoHailing! must develop a robust and scalable data pipeline to provide actionable insights for better decision making through Data Engineering

# Three project deliverables for Data Engineer were set to optimize building data pipeline



#### **Design pipeline architecture**

Designing pipeline that's robust and scalable to meet business needs:

- Technology stack selection: script language, data warehouse, transactional database, orchestration tools, etc
- > **Data Processing**: ELT or ETL, etc.
- Dimensional data model: Star Schema, Snowflake, etc
- Data ingestion method: Batch, stream, etc.

# Build automation logic flow

Designing automation logical flow within the data pipeline architecture, for example:

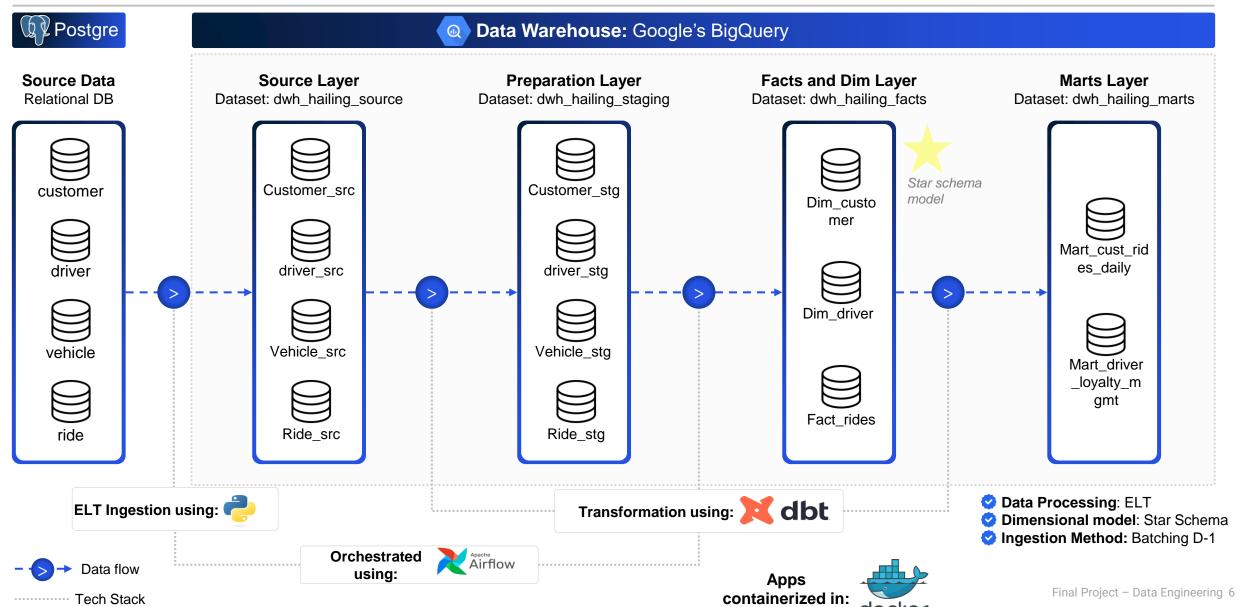
- What happens if error happens when ingesting the data?
- What needs to be done if there's any datatype mismatch between database and data warehouse?
- What would happen if the required table has not existed in data warehouse?

#### **Result effectiveness**

As a data engineer, it's essential to:

- Ensure the data engineering design is well implemented
- Ensure the data generated by the pipeline is accessible and usable for analyst or relevant stakeholders
- Enabling data-driven decision making by providing actionable insights through well-structured and reliable data
- Ensure data privacy is secure

## 1st project deliverable: Design Pipeline Architecture



## 2<sup>nd</sup> project deliverable: Automation logic flow

### DAG\_NAME: DAG1\_load\_to\_postgre.py | OBJECTIVE: Generate data and load to PostgreSQL relational database

Task 1: Ensure table existence in PostgreSQL Task 2: Generate data and load to PostgreSQL/DB Generate data: Check table **Table** Generate data: Start Fetch driver id Customer exist? existence vehicle driver Fetch: Generate data: Driver id ride Look for .sql Cust id .sql file file for table found? creation Discord alert refer to Insert data to DB: .sql file contains DDL Appendix 2 Customer commands - refer to Insert Driver Appendix 1 error? Vehicle Execute .sql ride file Create table Execute Ν in DB error? **Data loaded** End to DB









## 2<sup>nd</sup> project deliverable: Automation logic flow

#### DAG\_NAME: DAG2\_load\_to\_bigquery.py | OBJECTIVE: FETCH data from PostgreSQL and UPSERT to BigQuery

**Task 1:** Ensure table existence in BigQuery Task 2: Extract data and push DF to XCOM Task 3: Pull DF from XCOM and Upsert to BQ Has Pull attempt Check table Fetch data from **JSON** Start increme **ison** from found? existence postgre tables **XCOM** ntal col? Get schema Table N Fetch only **D-1** Fetch all Convert back to from postgres exist? Retrieve JSON (SELECT \*) data DF Has Return DF Return Load data to partition Dataframe (DF) NONE? Merge with the schema temporary field? target table staging table Create table Create table Convert df to DF without with Delete Empty? json partition in BQ partition in staging/tempor BQ ary table Skip Table created in Push json to **BigQuery XCOM** End

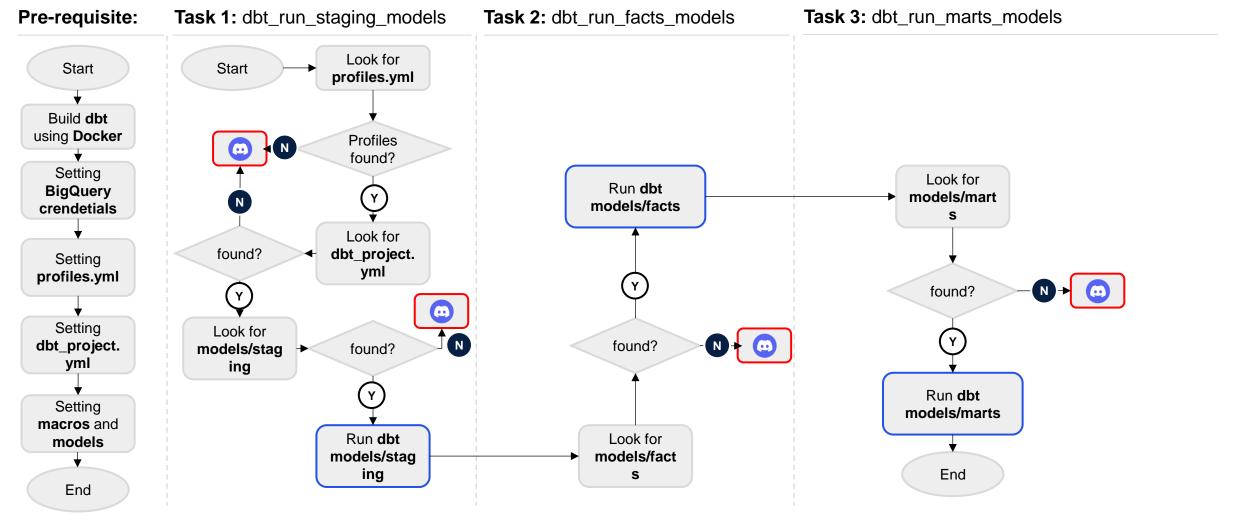






## 2<sup>nd</sup> project deliverable: Automation logic flow

#### DAG\_NAME: DAG3\_dbt\_run\_dwh.py | OBJECTIVE: Data transformation for raw data to business-level data in DWH









## How well did the data pipeline design perform after implementation?

It performs well, ensuring seamless data ingestion into the warehouse and transforming raw data into business-ready insights

#### **Data Generation**

- Successfully generate and load data into PostgreSQL, as well as building relationships between tables – all orchestrated using Airflow
- Send discord alert if either error or retry in logic automation (refer to appendix-2)

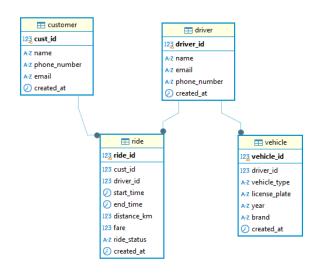


Fig 1. ERD in PostgreSQL

#### **Data Warehouse Modelling**

- Successfully executed ELT process, enabling incremental data ingestion from PostgreSQL to BigQuery using upsert method.
- Transformed data across layers from source to mart, ensuring business-ready insights. Implemented a star schema in facts layer to optimize analytical performance

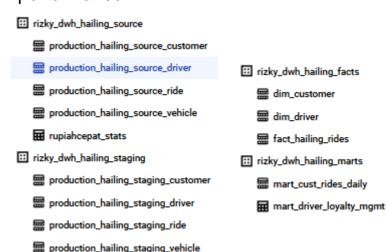


Fig 2. Dataset in BigQuery

#### **Fulfillment of business needs**

➤ The data can answer the questions that are asked in project's background, such as:

**50%** of rides were cancelled.. need further investigations..

**438.2**\$ in potential revenue was lost due to cancellations

**150** drivers were registered within the database...

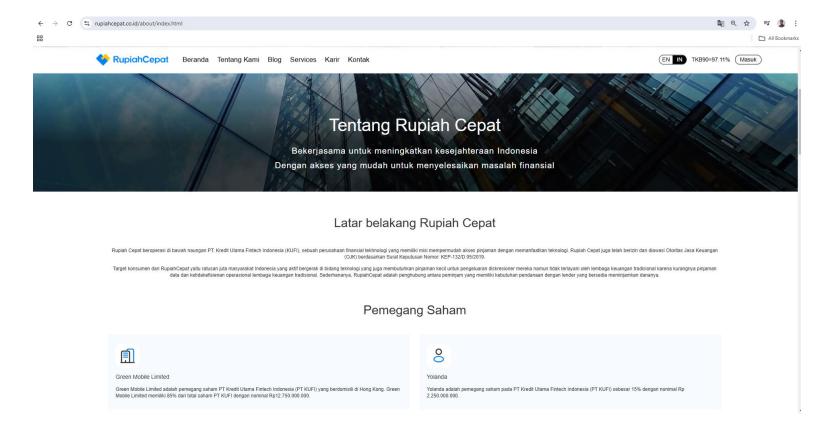
15% of drivers are not generating any revenue.. Need to stimulate for retention...

\*Data as of 12th February 2025

Final Project – Data Engineering10



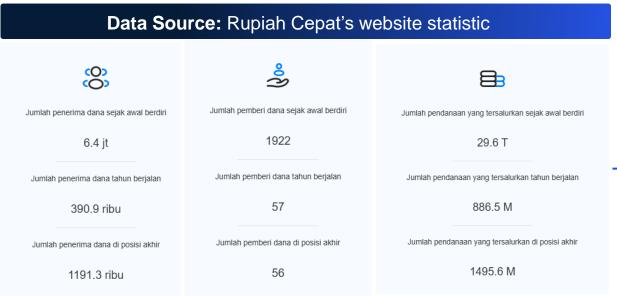
## Study Case: Web Scraping dynamic website of Rupiah Cepat

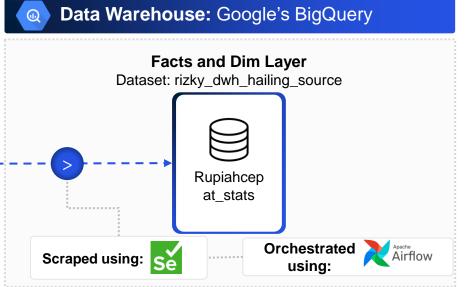


**Project Objective:** Perform web scraping from one of P2P lending website (RupiahCepat) and load to Data Warehouse (BigQuery)

## Website scraping deliverable: Design Pipeline Architecture

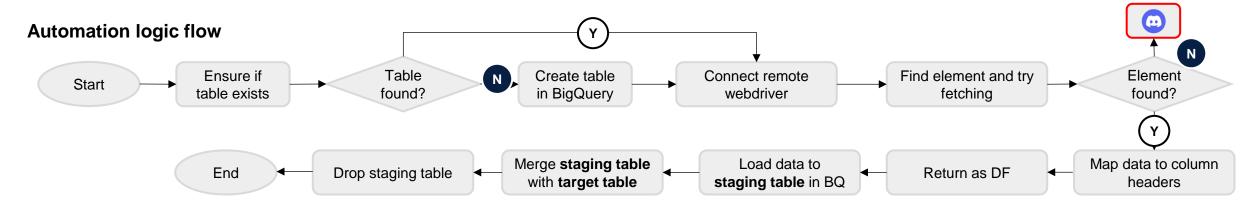






#### Remarks:

- Data Processing: ETL
- Insert new row if there's any changes from the website



## How well did the data pipeline design perform after implementation?

- Data successfully scraped using selenium, and upsert to data warehouse all orchestrated using Airflow
- If there's any changes in website, it will insert a new row of data and set the old data inactive in is active column

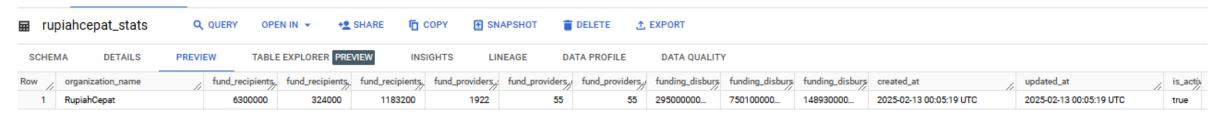


Fig 1. Data Preview in BigQuery

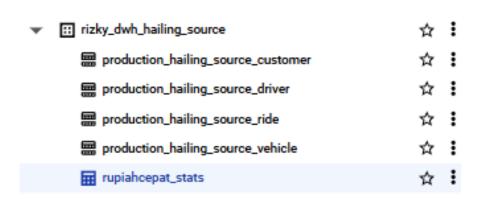


Fig 2. Table preview in BigQuery dataset

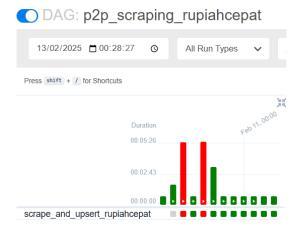
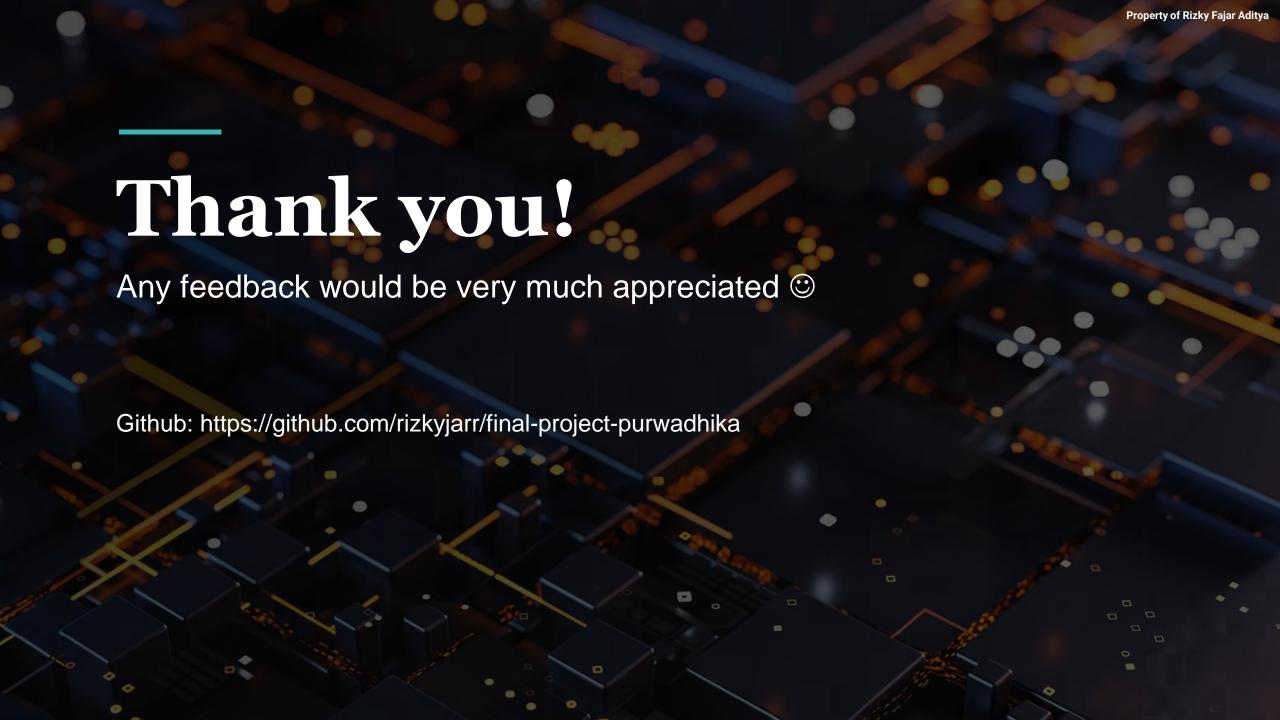


Fig 3. DAG4 Historical Task Logs

\*Data as of 12th February 2025



## Appendix 1 – SQL FILES

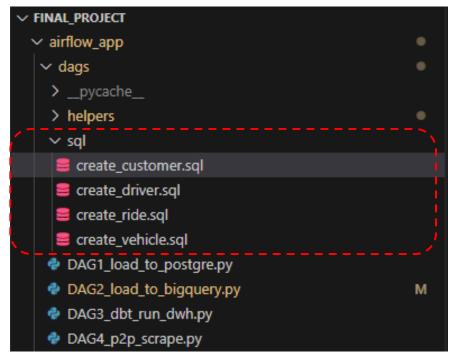
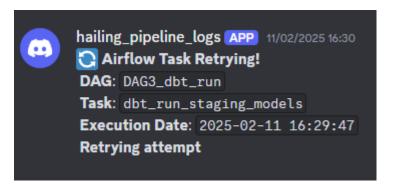


Fig 1. SQL folder location

```
create_ride.sql X
airflow_app > dags > sql > 🛢 create_ride.sql
       CREATE TABLE IF NOT EXISTS ride (
           ride id SERIAL PRIMARY KEY,
           cust id INTEGER NOT NULL,
           driver id INTEGER NOT NULL,
           start time TIMESTAMP NOT NULL,
           end time TIMESTAMP NOT NULL,
           distance km DECIMAL (5,2), -- contoh 999.99 km
           fare DECIMAL(10,2), --contoh 99,999,999.99 RP
           ride_status VARCHAR(20) NOT NULL,
           created at TIMESTAMP NOT NULL,
           FOREIGN KEY (cust_id) REFERENCES customer(cust_id),
 11
           FOREIGN KEY (driver_id) REFERENCES driver(driver_id)
 12
 13
```

Fig 2. Code example

## **Appendix 2 – Discord Webhook Alert**



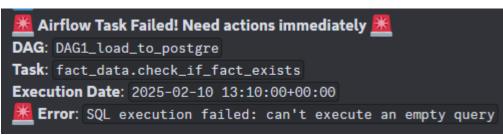


Fig 1. Error and retry alert on Discord

```
DISCORD_WEBHOOK_URL = "https://discord.com/api/webhooks/1338491764238979132/gxkc089HefdilVEvU0JeNmo_Mqsyg3Rg-V_54Rs5k1yzysSA7_e19XvcNts5_6fUXwAU
def send_discord_alert(context, alert_type="failure"):
   dag_id = context.get('dag_run').dag_id
   task id = context.get('task instance').task id
   execution date = context.get('execution date')
   exception = context.get('exception')
   jakarta_tz = pytz.timezone("Asia/Jakarta")
   execution date jakarta = execution date.astimezone(jakarta tz)
   execution date str = execution date jakarta.strftime("%Y-%m-%d %H:%M:%S")
   if alert_type == "failure":
       message = f"≝ **Airflow Task Failed! Need actions immediately** \n" \
                 f" **DAG**: `{dag_id}`\n" \
                 f" **Task**: `{task_id}`\n" \
                 f" **Execution Date**: `{execution date str}`\n" \
                 f" **Error**: `{exception}`
   elif alert_type == "retry":
       message = f" | **Airflow Task Retrying!** \n" \
                 f" **DAG**: `{dag_id}`\n" \
                 f" **Task**: `{task_id}`\n" \
                 f" **Execution Date**: `{execution_date_str}`\n" \
                 f" **Retrying attempt**"
   elif alert type == "success":
       message = f" ✓ **Airflow Task Completed!** \n" \
                 f" **DAG**: `{dag_id}`\n" \
                 f" **Task**: `{task id}`\n" \
                 f" **Execution Date**: `{execution_date_str}`\n" \
   payload = {"content": message}
   headers = {"Content-Type": "application/json"}
   response = requests.post(DISCORD_WEBHOOK_URL, data=json.dumps(payload), headers=headers)
   if response.status_code != 204:
       print(f"Failed to send Discord alert: {response.text}")
```

Fig 2. Discord alert's script

## **Appendix 3 – Data Preview in BigQuery**

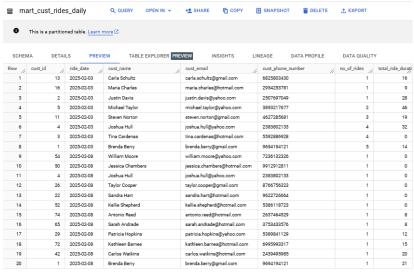


Fig 1. Mart\_cust\_rides\_daily table

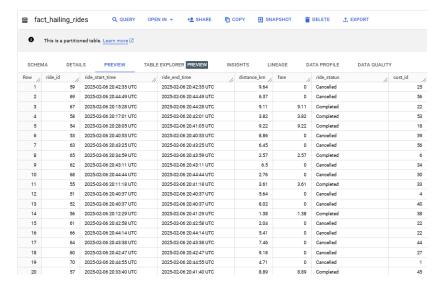


Fig 2. Fact\_hailing\_rides

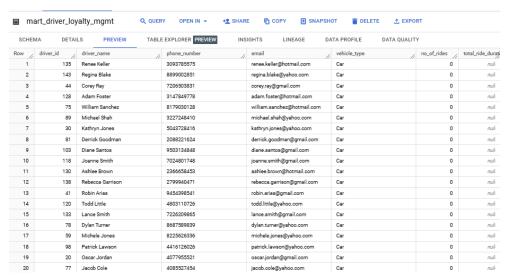


Fig 3. Mart\_driver\_loyalty\_mgmt table

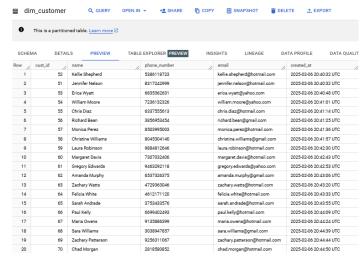


Fig 4. dim\_customer table