

# Data Engineering untuk Insight E-Commerce

Muhammad Robi Munawir





# Muhammad Robi Munawir

#### **Education**

Telkom University

#### Working

English one - IT Staff

#### **Overview Project**

Data Engineering untuk Insight E-Commerce

Proyek ini bertujuan untuk membangun pipeline data engineering terautomasi yang menghasilkan insight mendalam dari platform e-commerce menggunakan dataset Olist.





# Project Background





# **Deskripsi Umum Proyek**

Proyek ini membangun pipeline data engineering untuk mengolah dan menganalisis data dari platform E-Commerce Olist menggunakan pendekatan batch processing. Tujuan utamanya adalah menyediakan insight bisnis yang relevan melalui tiga analisis utama:

- 1. Segmentasi Pelanggan (RFM Analysis)
- 2. Analisis SLA Pengiriman dan Kinerja Logistik
- 3. Analisis Review dan Kepuasan Pelanggan

Data mentah dari Olist diekstrak, dibersihkan, ditransformasikan menggunakan Python dan Apache Airflow, disimpan ke dalam **PostgreSQL**, dan divisualisasikan menggunakan **Grafana**. Semua komponen dijalankan menggunakan **Docker** untuk memastikan portabilitas dan reprodusibilitas lingkungan kerja.



# **©** Tujuan Utama Proyek

Mengotomatiskan proses ETL menggunakan Apache Airflow dan Python untuk tiga fokus analisis utama:

- RFM (Recency, Frequency, Monetary) untuk segmentasi pelanggan.
- SLA & Kinerja pengiriman dari data pesanan.
- Review dan sentimen pelanggan berdasarkan rating dan komentar.

Menyimpan hasil analisis ke dalam PostgreSQL untuk keperluan analisis lanjutan dan integrasi ke Grafana.

Menyediakan dashboard interaktif menggunakan Grafana untuk memberikan insight kepada stakeholder bisnis secara real-time.



# Mengapa Proyek Ini Penting

Dalam industri E-Commerce, kemampuan untuk memahami pelanggan, menjaga kualitas pengiriman, dan menanggapi kepuasan pelanggan sangat penting untuk meningkatkan loyalitas dan efisiensi operasional.

Proyek ini memberikan kerangka kerja teknis untuk:

- Mengotomatisasi alur pengolahan data
- Menghasilkan insight yang dapat ditindaklanjuti oleh tim bisnis
- Meningkatkan kecepatan dan akurasi pengambilan keputusan berbasis data



# **Siapa yang Diuntungkan**

#### Tim Bisnis dan Pemasaran

 Untuk memahami segmentasi pelanggan dan menyusun strategi personalisasi.

#### Tim Operasional/Logistik

 Untuk memantau performa pengiriman dan SI A.

#### Manajemen

 Untuk melihat kinerja keseluruhan melalui dashboard dan membuat keputusan strategis.

# Data Engineer & Analyst

 Sebagai contoh pipeline terstruktur untuk proyek serupa di industri lain.





# Problem Statement





# Masalah yang Ingin Diselesaikan

Platform e-commerce seperti Olist menghasilkan data dalam jumlah besar dan bervariasi, namun sering kali data tersebut:

- Tidak langsung siap dianalisis karena formatnya tidak konsisten.
- Tidak memiliki pipeline otomatis untuk ekstraksi, transformasi, dan pemuatan (ETL).
- Sulit digunakan oleh tim bisnis karena tidak tersedia dalam bentuk dashboard atau insight yang mudah dipahami.

Masalah utama yang dihadapi meliputi:

- Tidak adanya segmentasi pelanggan yang membantu dalam kampanye pemasaran yang ditargetkan.
- Kurangnya monitoring terhadap performa pengiriman barang (SLA), sehingga sulit mengevaluasi vendor logistik.
- Tidak adanya sistem otomatis untuk mengolah ulasan pelanggan menjadi metrik kepuasan yang dapat ditindaklanjuti.





# Metrik Kesuksesan Proyek

| Tujuan  | Metrik Kesuksesan   |
|---|---|
| Data berhasil ditarik,<br>dibersihkan, dan disimpan               | Semua tabel target (rfm_cleaned, sla_analysis, review_sentiment) berhasil terbentuk di PostgreSQL |
| Proses ETL berjalan otomatis harian                               | DAG Airflow berjalan sukses tanpa error, minimal 1x/hari  |
| Data engineer dapat<br>menjalankan ulang pipeline<br>dengan mudah | Docker container up & DAG dapat dijalankan<br>ulang tanpa konfigurasi tambahan                    |





# Data Platform Understanding





# 

Proyek ini menggunakan dataset publik dari Olist, yang mencerminkan data ecommerce dari Brazil. Dataset ini mencakup:

| Nama Dataset                     | Deskripsi Singkat                      |
|----------------------------------|--|
| olist_orders_dataset.csv         | Informasi transaksi pesanan            |
| olist_order_items_dataset.csv    | Rincian produk di setiap pesanan       |
| olist_order_reviews_dataset.csv  | Review dan rating pelanggan            |
| olist_order_payments_dataset.csv | Informasi pembayaran                   |
| olist_customers_dataset.csv      | Data pelanggan                         |
| olist_products_dataset.csv       | Informasi produk                       |
| olist_sellers_dataset.csv        | Data penjual                           |
| olist_geolocation_dataset.csv    | Koordinat lokasi pelanggan dan penjual |





# Metode Pengambilan

File .csv disimpan dalam folder data



Proses extract dilakukan menggunakan Python dalam task Airflow (extract.py) yang membaca data dengan pandas.



Data kemudian melalui proses cleaning dan transformasi untuk keperluan analisis





# Apache Airflow digunakan untuk menjadwalkan dan mengatur pipeline ETL dalam bentuk Directed Acyclic Graph (DAG).

ETL terbagi dalam tiga kategori utama:

- rfm\_dag.py: Analisis RFM untuk segmentasi pelanggan.
- logistic dag.py: Mengukur SLA dan kinerja pengiriman.
- review\_dag.py: Mengolah ulasan menjadi metrik kepuasan pelanggan.



# **Penyimpanan**

Setelah transformasi, data dimuat ke PostgreSQL yang bertindak sebagai data warehouse.



Koneksi ke
PostgreSQL dikelola
oleh environment
.env dan Docker
Compose.





# 📊 Visualisasi Menggunakan Grafana

Grafana digunakan sebagai alat visualisasi utama.

Data dari PostgreSQL di-query menggunakan data source PostgreSQL di Grafana.

#### Dashboard Grafana mencakup:

- Segmentasi pelanggan berdasarkan skor RFM.
- Performa pengiriman (SLA actual vs estimated).
- Rangkuman rating dan sentimen dari review pelanggan.





# Data Understanding

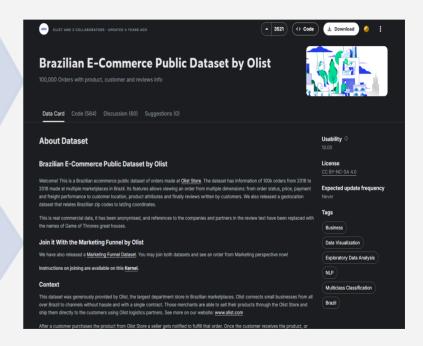




### **Sumber Data**

Olist, sebuah perusahaan e-commerce di Brasil. Dataset ini tersedia secara publik di Kaggle Olist Dataset dan berisi data transaksi.

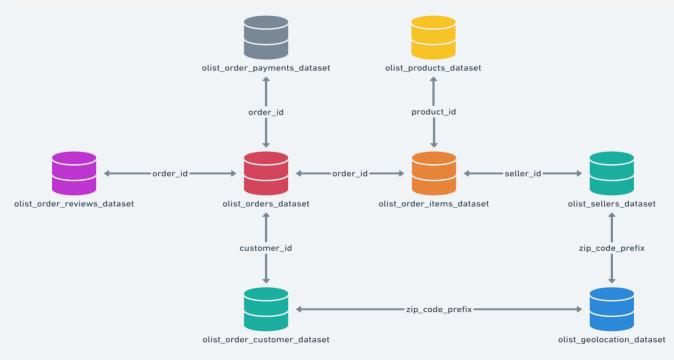
Dataset dikumpulkan dari system operasional Olist dan mencerminkan berbagai aspek ekosistem ecommerce, mulai dari order, pelanggan, penjual, produk, pengiriman, hingga review.







## **Data Schema**





# **Kualitas Data**



Masalah Umum yang Ditemukan:

#### **Nilai Null:**

- Kolom seperti review comment title, review comment message mengandung banyak nilai kosong.
- Kolom product category name juga memiliki null karena tidak semua produk dikategorikan.

#### **Duplikasi:**

Data pelanggan dan order tidak memiliki duplikasi berarti karena sudah memiliki primary key (customer id, order id), namun proses validasi tetap dilakukan untuk mencegah anomali.

#### **Inkonsistensi:**

Format tanggal dan waktu perlu disesuaikan menjadi format standar sebelum dianalisis (order purchase timestamp, review creation date, dll).





# Transformation & Consideration





### **✓** Kenapa bukan Streaming?

Streaming seperti Apache Kafka tidak digunakan karena tidak ada data real-time masuk terus-menerus. Dataset bersifat statis dari file CSV Kaggle.

Proyek ini menggunakan pendekatan batch processing yang berfokus pada batch processing. Berikut penjelasan komponennya:

#### **Apache Airflow (ETL)**

- Menangani penjadwalan dan task ETL harian.
- DAG digunakan untuk mengeksekusi task secara terstruktur, misalnya ekstraksi, pembersihan, transformasi, dan load ke database.





# **Batch Processing**

#### **Python (Transformasi Data)**

Digunakan untuk melakukan:

- Transformasi RFM (Recency, Frequency, Monetary)
- Analisis SLA Pengiriman
- Analisis Kepuasan Pelanggan (Review)

Sumber data: file CSV.

Transformasi dilakukan dalam skrip Python di dalam folder etl\_scripts/.

#### **PostgreSQL** (Data Warehouse)

- Tempat penyimpanan data hasil ekstraksi dan transformasi.
- Struktur tabel terdiri dari tabel raw dan tabel hasil analisis.





# **Batch Processing**

#### Grafana (Visualisasi)

Terhubung ke PostgreSQL.

Visualisasi metrik hasil analisis, misalnya:

- Segmentasi RFM
- Rata-rata waktu pengiriman per kategori produk
- Produk dengan review terburuk/terbaik

#### **Docker (Containerization)**

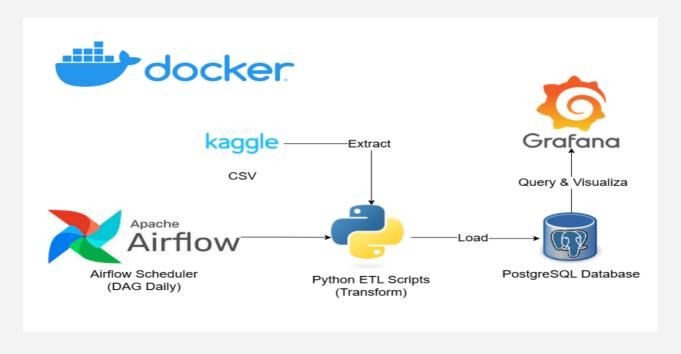
- Seluruh aplikasi dibangun dan dijalankan dalam container Docker.
- Komponen seperti Airflow, PostgreSQL, dan Grafana diorkestrasi dengan docker-compose.yml.





# **II** Diagram Arsitektur Transformasi Data

Berikut diagram arsitektur transformasi data:









# **Tabel hasil dari Extract**



| 2                          | <b>©</b>                   | <b>©</b>                    | <b>&amp;</b>                |  |
|----------------------------|----------------------------|-----------------------------|-----------------------------|--|
| > public                   | lic                        |                             |                             |  |
|                            |                            | raw_order_items             | === raw_orders              |  |
| customer_id                | geolocation_zip_code_prefi | order_id                    | order_id                    |  |
| customer_unique_id         | x                          | order_item_id               | customer_id                 |  |
| customer_zip_code_prefix   | geolocation_lat            | product_id                  | order_status                |  |
| customer_city              | geolocation_Ing            | seller_id                   | order_purchase_timestam     |  |
| customer_state             | geolocation_city           | shipping_limit_date         | р                           |  |
|                            | geolocation_state          | price                       | order_approved_at           |  |
|                            |                            | freight_value               | order_delivered_carrier_dat |  |
| public                     |                            |                             | e order_delivered_customer_ |  |
| raw_review                 | raw_sellers                |                             |                             |  |
| review_id                  | seller_id                  | ◆ public                    | order_estimated_delivery_d  |  |
| order_id                   | seller_zip_code_prefix     | == raw_products             | ate                         |  |
| review_score               | seller_city                | product_id                  |                             |  |
| review_comment_title       | seller_state               | product_category_name       |                             |  |
| review_comment_messag<br>e |                            | product_name_lenght         |                             |  |
| review_creation_date       |                            | product_description_lengh t |                             |  |
| review_answer_timestamp    |                            | product_photos_qty          |                             |  |
|                            |                            | product_weight_g            |                             |  |
|                            |                            | product_length_cm           |                             |  |
|                            |                            | product_height_cm           |                             |  |
|                            |                            | product_width_cm            |                             |  |
|                            |                            |                             |                             |  |



# **Tabel hasil dari Transform**



| <b>8</b>                     | <b>2</b>                | <b>15</b>                 |
|------------------------------|-------------------------|---------------------------|
| public                       | ◆ public                | public                    |
| mesult_rfm                   | result_review           | == result_logistics       |
| customer_unique_id           | review_id               | order_id                  |
| recency                      | order_id                | customer_id               |
| frequency                    | review_score            | seller_id                 |
| monetary                     | review_comment_title    | order_delivered_customer_ |
| r_score                      | review_comment_messag   | date                      |
| f_score                      | e                       | order_estimated_delivery_ |
| m_score                      | review_creation_date    | ate                       |
| rfm_score                    | review_answer_timestamp | delivery_delay            |
| segment                      | customer_id             | logistics_status          |
| a segment                    | order_status            | product_id                |
| 25                           | customer_unique_id      | product_category_name     |
| public                       | review_category         | customer_city             |
| result_logistics_top_product | delay                   | customer_state            |
| product_category_name        |                         | seller_city               |
| late_count                   |                         | seller_state              |
|                              |                         |                           |

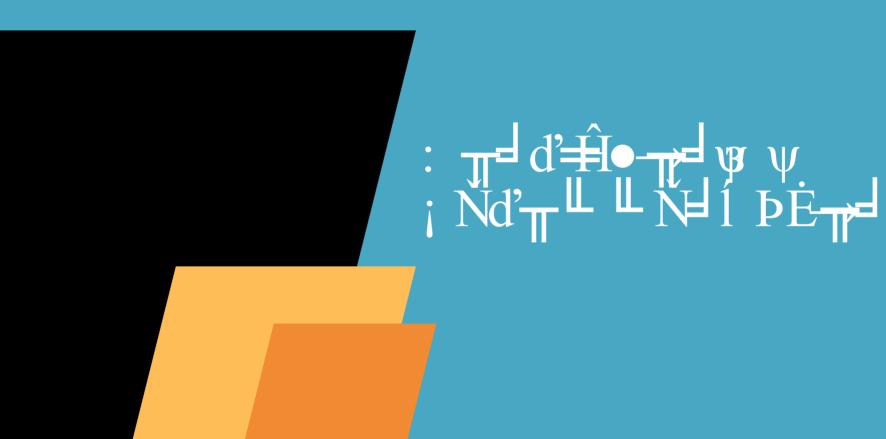


# **Example 1** Kategori Segmen RFM



| Segment            | Kriteria Umum                     | Penjelasan Singkat   |
|--------------------|-----------------------------------|--|
| Champions          | $R \ge 4$ , $F \ge 4$ , $M \ge 4$ | Pembeli paling aktif dan<br>bernilai tinggi.                   |
| Loyal Customers    | F = 5                             | Sangat sering membeli,<br>pelanggan setia.                     |
| Big Spenders       | M = 5                             | Pengeluaran tinggi, walau frekuensi/recency biasa.             |
| Recent Customers   | R = 5, F ≤ 2                      | Baru membeli, perlu dibina<br>loyalitasnya.                    |
| Potential Loyalist | R≥3, F≥3                          | Sering & cukup baru,<br>berpotensi menjadi pelanggan<br>setia. |
| At Risk            | $R \le 2, F \ge 3, M \ge 3$       | Pernah aktif tapi mulai pasif.                                 |
| Can't Lose Them    | R = 1, F ≥ 4, M ≥ 4               | Bernilai tinggi di masa lalu,<br>tapi sudah tidak aktif.       |
| Hibernating        | $R \le 2, F \le 2, M \le 2$       | Tidak aktif, pembelian jarang<br>dan nilai rendah.             |
| Lost               | R = 1, F = 1, M = 1               | Tidak pernah kembali,<br>kontribusi rendah.                    |
| Others             | Selain di atas                    | Segmen umum.   |

# **Q**ibimbing







Proyek ini berhasil membangun sebuah data pipeline end-to-end yang mampu:

- 1. Menjalankan ETL harian secara otomatis menggunakan Apache Airflow dalam container Docker.
- 2. Mengelola dan menyimpan data secara terstruktur di dalam PostgreSQL sebagai data warehouse.
- 3. Melakukan transformasi data yang relevan, seperti:
  - o Segmentasi pelanggan dengan analisis RFM
  - o Evaluasi SLA pengiriman dan performa logistik
  - Penilaian review dan kepuasan pelanggan
- 4. Menyajikan data secara visual dan real-time menggunakan Grafana, memungkinkan stakeholder memahami insight secara cepat dan interaktif.





### Keterbatasan Platform

Meskipun proyek ini fungsional dan efektif untuk skala dataset Olist, terdapat beberapa keterbatasan yang perlu diperhatikan:

#### 1. Skalabilitas & Performa:

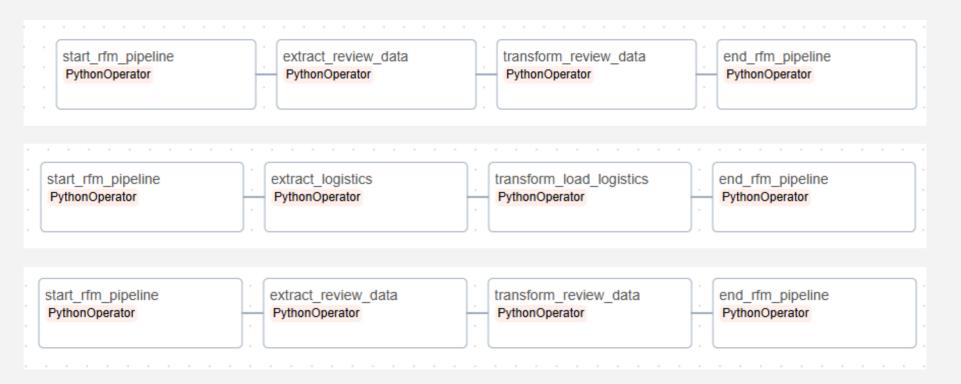
- **PostgreSQL** cukup baik untuk data sedang, tapi kurang optimal jika volume data sangat besar (big data).
- Transformasi dilakukan dengan Python script, bukan engine seperti Apache Spark ini membatasi performa pada volume data besar.

#### 2. Pengolahan Data Real-Time:

- Pipeline ini berbasis **batch processing harian**, sehingga tidak cocok untuk kebutuhan **real- time streaming** seperti deteksi fraud atau live monitoring.
- Tidak ada dukungan untuk platform seperti **Kafka** atau **Spark Streaming**...









# Visualization RFM



| Pelanggan Champion (RFM score tertinggi) |         |           |          |         |         |         |           |           |
|--|---------|-----------|----------|---------|---------|---------|-----------|-----------|
| customer_unique_id                       | recency | frequency | monetary | r_score | f_score | m_score | rfm_score | segment   |
| 012a218df8995d3ec3bt                     | 73      | 2         | 1430     | 5       | 5       | 5       | 555       | Champions |
| 013ef03e0f3f408dd9bf                     | 41      | 2         | 226      | 5       | 5       | 5       | 555       | Champions |
| 0178b244a5c281fb2ade                     | 33      | 2         | 209      | 5       | 5       | 5       | 555       | Champions |
| 01f2285f85a1c603eb7e                     | 91      | 2         | 428      | 5       | 5       | 5       | 555       | Champions |
| 032b3a42598667caf2e                      | 36      | 2         | 388      | 5       | 5       | 5       | 555       | Champions |
| 0341bbd5c969923a0f8                      | 92      | 2         | 1828     | 5       | 5       | 5       | 555       | Champions |



| Top 10 pelanggan dengan nilai pembelian tertinggi (Monetary) |           |          |  |  |  |
|--|-----------|----------|--|--|--|
| customer_unique_id   | frequency | monetary |  |  |  |
| 0a0a92112bd4c708ca5fde585afaa8                               |           | 13440    |  |  |  |
| da122df9eeddfedc1dc1f5349a1a690                              | 2         | 7388     |  |  |  |
| 763c8b1c9c68a0229c42c9fc6f662                                |           | 7160     |  |  |  |
| dc4802a71eae9be1dd28f5d788ceb                                | 1         | 6735     |  |  |  |
| 459bef486812aa25204be022145ca                                |           | 6729     |  |  |  |
| ff4159b92c40ebe40454e3e6a7c35                                | 1         | 6499     |  |  |  |

| Pelanggan Baru (recency rendah, frequency rendah) |         |           |  |  |  |
|---|---------|-----------|--|--|--|
| customer_unique_id                                | recency | frequency |  |  |  |
| 000ec5bff359e1c0ad76a81a45cb59                    | 9       |           |  |  |  |
| 0015752e079902b12cd00b9b7596                      | 27      |           |  |  |  |
| 001928b561575b2821c92254a2327                     |         |           |  |  |  |
| 002471155ecd08d208d1376720e2a                     | 21      |           |  |  |  |
| 00391594070baf8603f5fdc1d0c6a4                    | 24      | 1         |  |  |  |

| Pelanggan Dormant (recency tinggi, frequency rendah) |         |           |              |  |  |
|--|---------|-----------|--------------|--|--|
| customer_unique_id                                   | recency | frequency | segment      |  |  |
| 830d5b7aaa3b6f1e9ad637                               | 714     |           | At Risk      |  |  |
| 8d3a54507421dbd2ce0a1c                               | 695     |           | At Risk      |  |  |
| b8b8726af116a5cfb35b03                               | 695     |           | Others       |  |  |
| cb1bc069e25d9c59773c8                                | 695     |           | Big Spenders |  |  |
| 88bc832f104b6dd684a17c                               | 695     | 1         | At Risk      |  |  |



# **✓** Visualization Logistics



| Pelanggan Champion (RFM score tertinggi) |         |           |          |         |         |         |           |           |
|--|---------|-----------|----------|---------|---------|---------|-----------|-----------|
| customer_unique_id                       | recency | frequency | monetary | r_score | f_score | m_score | rfm_score | segment   |
| 012a218df8995d3ec3bt                     | 73      | 2         | 1430     | 5       | 5       | 5       | 555       | Champions |
| 013ef03e0f3f408dd9bf                     | 41      | 2         | 226      | 5       | 5       | 5       | 555       | Champions |
| 0178b244a5c281fb2ade                     | 33      | 2         | 209      | 5       | 5       | 5       | 555       | Champions |
| 01f2285f85a1c603eb7e                     | 91      | 2         | 428      | 5       | 5       | 5       | 555       | Champions |
| 032b3a42598667caf2e                      | 36      | 2         | 388      | 5       | 5       | 5       | 555       | Champions |
| 0341bbd5c969923a0f8                      | 92      | 2         | 1828     | 5       | 5       | 5       | 555       | Champions |



| Top 10 pelanggan dengan nilai pembelian tertinggi (Monetary) |           |          |  |  |  |
|--|-----------|----------|--|--|--|
| customer_unique_id   | frequency | monetary |  |  |  |
| 0a0a92112bd4c708ca5fde585afaa8                               | 1         | 13440    |  |  |  |
| da122df9eeddfedc1dc1f5349a1a690                              | 2         | 7388     |  |  |  |
| 763c8b1c9c68a0229c42c9fc6f662                                | 1         | 7160     |  |  |  |
| dc4802a71eae9be1dd28f5d788ceb                                | 1         | 6735     |  |  |  |
| 459bef486812aa25204be022145ca                                | 1         | 6729     |  |  |  |
| ff4159b92c40ebe40454e3e6a7c35                                | 1         | 6499     |  |  |  |

| Pelanggan Baru (recency rendah, frequency rendah) |         |           |  |  |  |
|---|---------|-----------|--|--|--|
| customer_unique_id                                | recency | frequency |  |  |  |
| 000ec5bff359e1c0ad76a81a45cb5                     | 9       |           |  |  |  |
| 0015752e079902b12cd00b9b7596                      | 27      |           |  |  |  |
| 001928b561575b2821c92254a2327                     | 5       |           |  |  |  |
| 002471155ecd08d208d1376720e2a                     | 21      |           |  |  |  |
| 00391594070baf8603f5fdc1d0c6a                     | 24      | 1         |  |  |  |

| Pelanggan Dormant (recency tinggi, frequency rendah) |         |           |              |  |  |  |
|--|---------|-----------|--------------|--|--|--|
| customer_unique_id                                   | recency | frequency | segment      |  |  |  |
| 830d5b7aaa3b6f1e9ad637                               | 714     |           | At Risk      |  |  |  |
| 8d3a54507421dbd2ce0a1c                               | 695     |           | At Risk      |  |  |  |
| b8b8726af116a5cfb35b03                               | 695     |           | Others       |  |  |  |
| cb1bc069e25d9c59773c8                                | 695     |           | Big Spenders |  |  |  |
| 88bc832f104b6dd684a17c                               | 695     |           | At Risk      |  |  |  |



# **Visualization Review ✓**



| Distribusi skor review Distribusi review berdasarkan status pesanan |        | Jumlah review per kategori (positive, neutral, negative) |                       |                  |   | Ī            |            |          |         |     |     |     |        |          |
|---|--------|--|-----------------------|------------------|---|--------------|------------|----------|---------|-----|-----|-----|--------|----------|
| review_score  | jumlah | logistics_status   |                       | avg_review_      | score   |              | _          |          |         |     |     |     |        | — jumlah |
| 1   | 1539   | late   |                       |                  | 2.18  | negative     |            |          |         |     |     |     |        |          |
| 2   | 421    | on time  |                       |                  | 3.90  |              |            |          |         |     |     |     |        |          |
| 3   | 716    | Review negatif yang dika                                 | aitkan dengan keterla | mbatan pengirima | n   |              |            |          |         |     |     |     |        |          |
| 4   | 1422   | review_score ↑   | logistics_status      | jı               | umlah   | positive     |            |          |         |     |     |     |        |          |
| 5   | 5402   |  | on_time               |                  | 1711  |              | _          |          |         |     |     |     |        |          |
|   |        |  | late                  |                  | 397   |              |            |          |         |     |     |     |        |          |
|   |        | 2  | on_time               |                  | 502   | neutral      |            |          |         |     |     |     |        |          |
|   |        | 2  | late                  |                  | 62  | 0            | 1 K        | 2 K      | зк      | 4 K | 5 K | 6 K | 7 K    |          |
| Jumlah review positif per produk kategori Top 5                     |        | Top 5 r  | eview negatif y       | ang paling s     | ering mun                                     | cul (judul d | lan pesan) |          |         |     |     |     |        |          |
| product_category_name   |        | jumlah_review_positif revie                              |                       | review           | _comment_title                                | 9            | review_    | comment_ | message |     |     |     | jumlah |          |
| beleza_saude  |        | 788 Falta  |                       | Faltam           | Faltam peças Recebemos o produto faltando per |              | do peça    |          |         | 2   |     |     |        |          |

| product_category_name | jumlah_review_positif |
|-----------------------|-----------------------|
| beleza_saude          | 788                   |
| cama_mesa_banho       | 727                   |
| utilidades_domesticas | 586                   |
| relogios_presentes    | 582                   |
| esporte_lazer         | 521                   |

| Pelanggan dengan review paling banyak dan | rata-rata skor review |                  |
|---|-----------------------|------------------|
| customer_unique_id                        | jumlah_review ↓       | avg_review_score |
| 12c9171ee18219cbf18fcfbc33e0c06           | 4                     | 5                |
| ae4395be33e633d0c87010d59847              | 4                     | 2.50             |
| 1772c0fbee58b154da028487de51b7            | 4                     | 5                |
| 6a9e15d6fa8ce1cabf193c21aa577f6           | 3                     | 5                |

| review_comment_title  | review_comment_message              | jumlah |
|-----------------------|-------------------------------------|--------|
| Faltam peças          | Recebemos o produto faltando peça   | 1      |
| É muito confusa.      | Comprei duas válvulas solenóides i  | 2      |
| Avarias na prateleira | Recebi o produto dentro do prazo p  | .2     |
| Insatisfação          | Negativo                            | 2      |
| Sem recomendação      | Está faltando receber um produto: F | 2      |

| Pelanggan paling sering memberikan review negatif |                       |
|---|-----------------------|
| customer_unique_id                                | jumlah_review_negatif |
| 65571628698194dbda526c22f5ed2317                  | 2                     |
| ae4395be33e633d0c87010d59847e5ed                  | 2                     |
| 50915df398391da01c01bd5cf272f61b                  | 2                     |
| 99bf2e03eb1ef8f29fd57ad10fd75e48                  | 2                     |



### Code Overview – ETL RFM



```
extract_rfm.py 2, U ×
                                                                                                                           transform_load_rfm.py 2, U ×
                                                                                                                                                                                                                                    ⊳∨ ಬ Ш …
 etL scripts > rfm > 👶 extract rfm.pv > 😭 extract rfm
                                                                                                                           etl scripts > rfm > a transform load rfm.pv > .
       import pandas as pd
                                                                                                                                  import pandas as pd
      import os
                                                                                                                                  from sqlalchemy import create engine
       import logging
       from sqlalchemy import create engine
                                                                                                                                  import logging
                                                                                                                                  from datetime import datetime
       logging.basicConfig(tevet=logging.INFO)
                                                                                                                                  logging.basicConfig(tevet=logging.INFO)
       def extract rfm():
                                                                                                                                  def segment(row):
                                                                                                                                     r, f, m = row['r score'], row['f score'], row['m score']
               engine = create engine(
                   f"postgresql://{os.getenv('POSTGRES USER')}:{os.getenv('POSTGRES PASSWORD')}@"
                   f"{os.getenv('POSTGRES HOST')}:{os.getenv('POSTGRES PORT')}/{os.getenv('POSTGRES DB')
                                                                                                                                     if r >= 4 and f >= 4 and m >= 4:
                                                                                                                                         return 'Champions'
                                                                                                                                     elif f == 5:
                                                                                                                                          return 'Loyal Customers'
               customers = pd.read csv('/opt/airflow/data/olist customers dataset.csv')
                                                                                                                                      elif m == 5:
               orders = pd.read csv('/opt/airflow/data/olist orders dataset.csv')
                                                                                                                                          return 'Big Spenders'
               order items = pd.read csv('/opt/airflow/data/olist order items dataset.csv')
                                                                                                                                      elif r == 5 and f <= 2:
                                                                                                                                         return 'Recent Customers'
                                                                                                                                     elif r >= 3 and f >= 3:
               customers.replace('NaN', pd.NA, inplace=True)
                                                                                                                                         return 'Potential Loyalist'
               orders.replace('NaN', pd.NA, inplace=True)
                                                                                                                                      elif r \leftarrow 2 and f >= 3 and m >= 3:
               order items.replace('NaN', pd.NA, inplace=True)
                                                                                                                                         return 'At Risk'
                                                                                                                                      elif r == 1 and f >= 4 and m >= 4:
                                                                                                                                         return 'Can't Lose Them'
               customers.dropna(inplace=True)
                                                                                                                                      elif r <= 2 and f <= 2 and m <= 2:
               orders.dropna(inplace=True)
                                                                                                                                         return 'Hibernating'
                                                                                                                                      elif r -- 1 and f -- 1 and m -- 1:
               order items.dropna(inplace=True)
                                                                                                                                         return 'Lost'
               customers.drop duplicates(inplace=True)
               orders.drop duplicates(inplace=True)
                                                                                                                                         return 'Others'
               order items.drop duplicates(inplace-True)
                                                                                                                                  def transform load rfm():
               engine - create engine(
                   f"postgresql://{os.getenv('POSTGRES USER')}:{os.getenv('POSTGRES PASSWORD')}@"
                                                                                                                                          engine - create engine(
                   f"(os.getenv('POSTGRES HOST')):(os.getenv('POSTGRES PORT'))/(os.getenv('POSTGRES DB')
                                                                                                                                              f"postgresql://{os.getenv('POSTGRES USER')}:{os.getenv('POSTGRES PASSWORD')}@"
                                                                                                                                              f"{os.getenv('POSTGRES HOST')}:{os.getenv('POSTGRES PORT')}/{os.getenv('POSTGRES DB')
               customers.to sql("raw customers", engine, if exists="replace", index=False)
               orders.to sql("raw orders", engine, if exists="replace", index=False)
               order items.to sql("raw order items", engine, if exists="replace", index=False)
                                                                                                                                          customers = pd.read sql('SELECT * FROM raw customers', engine)
                                                                                                                                          orders = pd.read sql('SELECT * FROM raw orders', engine)
               logging.info("Ekstraksi dan pembersihan RFM (null, duplikat, 'NaN' string) selesai.")
                                                                                                                                          order items = pd.read sql('SELECT * FROM raw order items', engine)
           except Exception as e:
               logging.error(f"Gagal extract RFM: {e}")
                                                                                                                                          merged = orders.merge(customers, on='customer id', how='inner') \
                                                                                                                                                         .merge(order items, on='order id', how='inner')
       if name == " main ":
          extract rfm()
                                                                                                                                          merged['order purchase timestamp'] = pd.to datetime(merged['order purchase timestamp'])
```



# Code Overview – ETL Logistik



```
extract_logistics.py 2, U ×
                                                                                                                           transform_load_logistics.py 2, U ×
                                                                                                                                                                                                                                     ▷ ~ ৸ □
etl scripts > logistics > 🚭 extract logistics.pv > ..
                                                                                                                            etl scripts > logistics > 🤔 transform load logistics.pv > 😭 transform load logistics
  1 import pandas as pd
                                                                                                                                  import pandas as pd
      import os
                                                                                                                                  import os
                                                                                                                                  from sqlalchemy import create engine
       import logging
      from sqlalchemy import create_engine
                                                                                                                                  import logging
      logging.basicConfig(tevet=logging.INFO)
                                                                                                                                  logging.basicConfig(tevet=logging.INFO)
      def extract logistics():
                                                                                                                                  def transform load logistics():
               engine = create engine(
                                                                                                                                          engine = create engine(
                   f"postgresgl://{os.getenv('POSTGRES USER')}:{os.getenv('POSTGRES PASSWORD')}@"
                                                                                                                                               f"postgresgl://{os.getenv('POSTGRES_USER')}:{os.getenv('POSTGRES_PASSWORD')}@"
                   f"{os.getenv('POSTGRES_HOST')}:{os.getenv('POSTGRES_PORT')}/{os.getenv('POSTGRES_DB')
                                                                                                                                               f"{os.getenv('POSTGRES_HOST')};{os.getenv('POSTGRES_PORT')}/{os.getenv('POSTGRES_DB')
                                                                                                                                          orders - pd.read sql("SELECT * FROM raw orders", engine)
               sellers = pd.read csv('/opt/airflow/data/olist sellers dataset.csv')
                                                                                                                                          order items = pd.read sql("SELECT * FROM raw order items", engine)
               products = pd.read csv('/opt/airflow/data/olist products dataset.csv')
                                                                                                                                          customers = pd.read sql("SELECT * FROM raw customers", engine)
                                                                                                                                          sellers = pd.read sal("SELECT * FROM raw sellers", engine)
               geolocation = pd.read csv('/opt/airflow/data/olist geolocation dataset.csv')
                                                                                                                                          products = pd.read sql("SELECT * FROM raw products", engine)
                                                                                                                                          geolocation = pd.read sql("SELECT * FROM raw geolocation", engine)
               sellers.replace('NaN', pd.NA, inplace=True)
               products.replace('NaN', pd.NA, inplace=True)
               geolocation.replace('NaN', pd.NA, inplace=True)
                                                                                                                                          df = orders.merge(order items, on="order id", how="inner")
                                                                                                                                          df = df.merge(customers, on="customer id", how="inner")
                                                                                                                                          df = df.merge(sellers, on="seller id", how="inner")
               sellers.dropna(inplace=True)
                                                                                                                                          df - df.merge(products, on "product_id", how "left") # product_id bisa NaN
               products.dropna(inplace=True)
               geolocation.dropna(inplace=True)
                                                                                                                                          df.dropna(subset=[
               sellers.drop duplicates(inplace=True)
                                                                                                                                               'order estimated delivery date',
               products.drop duplicates(inplace=True)
                                                                                                                                               'order delivered customer date',
               geolocation.drop duplicates(inplace=True)
                                                                                                                                               'order id'.
                                                                                                                                               'customer id'.
                                                                                                                                               'seller id'
               sellers.to sql("raw sellers", engine, if exists="replace", index=False)
                                                                                                                                           1. inplace=True)
               products.to sql("raw products", engine, if exists="replace", index=False)
                                                                                                                                          df.drop duplicates(inplace=True)
               geolocation.to sql("raw geolocation", engine, if exists="replace", index=False)
               logging.info("Ekstraksi dan pembersihan Logistics (null. duplikat. 'NaN' string) selesai.
                                                                                                                                          df = df[df['order status'] == 'delivered']
           except Exception as e:
               logging.error(f"Gagal extract logistics: {e}")
                                                                                                                                          df['order_estimated_delivery_date'] = pd.to_datetime(df['order_estimated_delivery_date'])
                                                                                                                                          df['order delivered customer date'] = pd.to datetime(df['order delivered customer date'])
      if __name__ -- "__main__":
          extract logistics()
                                                                                                                                          df['delivery delay'] = (df['order delivered customer date'] - df['order estimated delivery
```



### Code Overview – ETL Review



```
extract review.pv 2, U ×
                                                                                                                          transform load review.pv 2. U X
etl scripts > reviews > 👶 extract review.pv > ...
                                                                                                                           etl scripts > reviews > 👶 transform load review.pv > ...
  1 import pandas as pd
                                                                                                                                 import pandas as pd
      import os
                                                                                                                                  import os
                                                                                                                                  from sqlalchemy import create_engine
       from sqlalchemy import create_engine
       import logging
                                                                                                                                  import logging
      logging.basicConfig(tevet=logging.INFO)
                                                                                                                                  logging.basicConfig(tevet=logging.INFO)
      def extract review():
                                                                                                                                  def transform load review():
              engine = create_engine(
                                                                                                                                          engine = create_engine(
                   f"postgresgl://{os.geteny('POSTGRES USER')}:{os.geteny('POSTGRES PASSWORD')}@"
                                                                                                                                              f"postgresql://{os.getenv('POSTGRES USER')}:{os.getenv('POSTGRES PASSWORD')}@"
                   f"{os.getenv('POSTGRES HOST')}:{os.getenv('POSTGRES PORT')}/{os.getenv('POSTGRES DB')
                                                                                                                                              f"{os.getenv('POSTGRES HOST')}:{os.getenv('POSTGRES PORT')}/{os.getenv('POSTGRES DB')
              reviews = pd.read csv('/opt/airflow/data/olist order reviews dataset.csv')
                                                                                                                                          reviews = pd.read sql("SELECT * FROM raw review", engine)
                                                                                                                                          orders = pd.read sql("SELECT order id, customer id, order status FROM raw orders", engine
                                                                                                                                          customers - pd.read sql("SELECT customer id, customer unique id FROM raw customers", engi
              reviews.dropna(inplace=True)
              reviews = reviews[~reviews['review comment message'].astype(str).str.lower().isin(['nan',
              reviews.drop duplicates(inplace=True)
                                                                                                                                          df = reviews.merge(orders, on='order id', how='left')
                                                                                                                                          df = df.merge(customers, on='customer id', how='left')
              reviews.to_sql("raw_review", engine, if_exists="replace", index=False)
                                                                                                                                          def review category(score):
              logging.info("Extract review selesai.")
                                                                                                                                             if score >= 4:
                                                                                                                                                  return 'positive'
           except Exception as e:
                                                                                                                                              elif score == 3:
               logging.error(f"Gagal extract review: {e}")
                                                                                                                                                  return 'neutral'
       if __name__ == "__main__":
                                                                                                                                                  return 'negative'
          extract review()
                                                                                                                                          df['review category'] = df['review score'].apply(review_category)
                                                                                                                                          df = df[df['order status'].isin(['delivered', 'shipped', 'completed'])]
                                                                                                                                          df.to_sql("result_review", engine, if_exists="replace", index=False)
                                                                                                                                          logging.info("Transform review selesai.")
                                                                                                                                      except Exception as e:
                                                                                                                                          logging.error(f"Gagal transform review: {e}")
                                                                                                                                  if __name__ == "__main__":
                                                                                                                                     transform_load_review()
```

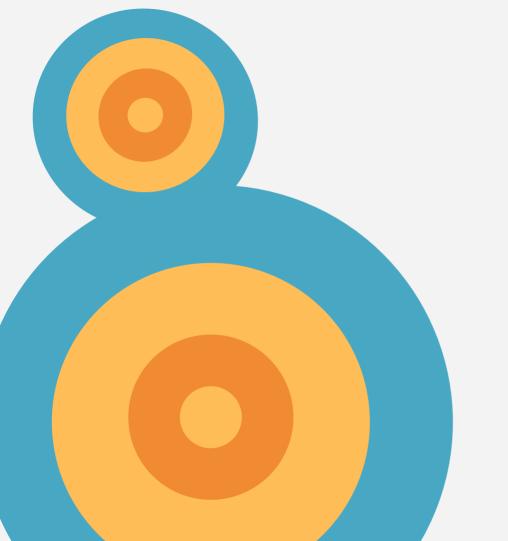


#### Code Overview – DAG



```
logistics_dag.py 2, M ×
                                                                                                                                                                   🕏 rfm_dag.py 2, M 🗙
                                                                                                                                                                                                                                   D ~ 81 M ··
                                                                                 review_dag.py 2, U ×
dags > 👶 logistics dag.pv >
                                                                                  dags > 👶 review dag.pv >
                                                                                                                                                                    dags > 👶 rfm dag.pv >
      from airflow import DAG
                                                                                    1 from airflow import DAG
                                                                                                                                                                      1 from airflow import DAG
       from airflow.operators.python import PythonOperator
                                                                                         from airflow.operators.python import PythonOperator
                                                                                                                                                                          from airflow.operators.python import PythonOperator
       from datetime import datetime, timedelta
                                                                                         from datetime import datetime, timedelta
                                                                                                                                                                          from datetime import datetime, timedelta
       from etl scripts.logistics.extract logistics import extract logistics
       from etl scripts.logistics.transform load logistics import tr
                                                                                         from etl scripts.reviews.extract review import extract review
                                                                                         from etl scripts.reviews.transform load review import transform
                                                                                                                                                                          from etl scripts.rfm.extract rfm import extract rfm
                                                                                                                                                                          from etl scripts.rfm.transform load rfm import transform load
       default args = {
           'owner': 'robi'.
                                                                                        default_args = {
           'depends on past': False.
                                                                                             'owner': 'robi'.
                                                                                                                                                                          default args = {
           'email on failure': True.
                                                                                             'depends on past': False.
                                                                                                                                                                               'owner': 'robi'.
           'email': ['robimunawir1@gmail.com'].
                                                                                             'email on failure': True.
                                                                                                                                                                               'depends_on_past': False,
           'retries': 1.
                                                                                             'email': ['robimunawir1@gmail.com'].
                                                                                                                                                                               'email on failure': True.
           'retry delay': timedelta(minutes=5).
                                                                                             'retries': 1.
                                                                                                                                                                               'email': ['robimunawir1@gmail.com'].
                                                                                             'retry delay': timedelta(minutes=5).
                                                                                                                                                                               'retries': 1.
                                                                                                                                                                               'retry_delay': timedelta(minutes=5),
       with DAG(
           dag id='logistics etl dag',
                                                                                        with DAG(
           default aras=default args.
                                                                                            dag id='review etl dag'.
                                                                                                                                                                          with DAG(
          description='ETL untuk Analisis SLA Pengiriman dan Kinerj
                                                                                            default args=default args,
                                                                                                                                                                              dag id='rfm etl dag',
                                                                                            description='ETL review dan kepuasan pelanggan Olist',
          schedule interval='@daily',
                                                                                                                                                                              default args=default args,
          start date=datetime(2024, 1, 1).
                                                                                            schedule interval='@daily'.
                                                                                                                                                                              description='ETL Analisis Segmentasi RFM'.
                                                                                            start date=datetime(2024, 1, 1),
                                                                                                                                                                              schedule interval='@daily',
           tags=['olist', 'Logistics', 'ETL'],
                                                                                            catchun=False.
                                                                                                                                                                              start date=datetime(2024, 1, 1),
                                                                                             tags=['olist', 'Review', 'ETL'].
                                                                                                                                                                              catchup=False,
       ) as dag:
                                                                                                                                                                              tags=['olist', 'RFM', 'ETL'],
                                                                                                                                                                          ) as dag:
           start task = PythonOperator(
                                                                                            start task = PythonOperator(
               task id='start rfm pipeline',
                                                                                                task id-'start rfm pipeline',
                                                                                                                                                                              start task = PythonOperator(
              python callable=lambda: print("Memulai ETL Logistics.
                                                                                                python callable=lambda: print("Memulai ETL Review..."
                                                                                                                                                                                  task id='start rfm pipeline'.
                                                                                                                                                                                  python callable=lambda: print("Memulai ETL RFM...")
           end task = PythonOperator(
                                                                                            end task = PythonOperator(
               task id='end rfm pipeline'.
                                                                                                task id='end rfm pipeline'.
                                                                                                                                                                              end task = PythonOperator(
              python callable=lambda: print("Selesai ETL Logistics.
                                                                                                python callable=lambda: print("Selesai ETL Riview.")
                                                                                                                                                                                  task id='end rfm pipeline'.
                                                                                                                                                                                  python callable=lambda: print("Selesai ETL RFM.")
           extract task = PythonOperator(
                                                                                            extract task = PythonOperator(
               task id='extract logistics'.
                                                                                                 task id='extract review data'.
                                                                                                                                                                              extract task = PythonOperator(
                                                                                                python callable=extract review
                                                                                                                                                                                  task id='extract rfm'.
                                                                                                                                                                                  python callable-extract rfm
           transform load task = PythonOperator(
                                                                                            transform load task = PythonOperator(
               task id='transform load logistics'.
                                                                                                task_id='transform_review_data',
                                                                                                                                                                              transform_load_task = PythonOperator(
               python_callable=transform_load_logistics
                                                                                                python callable-transform load review
                                                                                                                                                                                  task id='transform load rfm'.
                                                                                                                                                                                  python callable=transform load rfm
           start task >> extract task >> transform load task >> end
                                                                                            start task >> extract task >> transform load task >> end
                                                                                                                                                                              start task >> extract task >> transform load task >> end
```





Terima Kasih.