



CAPSTONE PROJECT -1 LOGISTICS & DELIVERY PERFORMANCE ANALYSIS

BD702 - Group 8

(000)





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TABLE OF CONTENT

- DATA SOURCE AND STRUCTURAL OVERVIEW
- 2 DATA EXTRACTION
- PREPROCESSING &TRANSFORMATION
- DATA LOADING AND INTEGRATION
- 5 DATA VISUALIZATION



DATA SOURCE & STRUCTURAL OVERVIEW

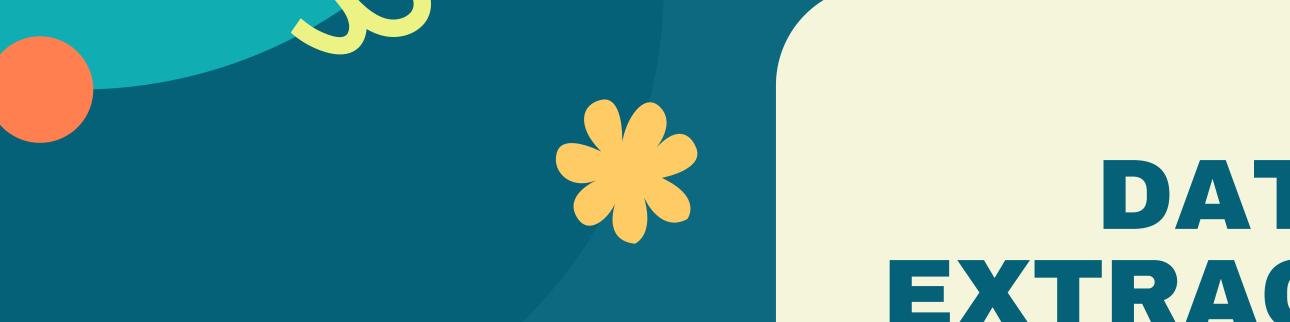
LOGISTICS SUPPLY CHAIN REAL WORLD DATA

Originates from real-world logistics and supply chain operations.

For the purpose of analysis, the dataset has been divided into two main components:

- Logistics Information: Contains details related to shipments, delivery tracking, warehouse handling, and transit times.
- Order Information: Contains details related to individual orders, including order IDs, order_date, category_name, ,department_name, order_item quantity, expected_delivery_date.



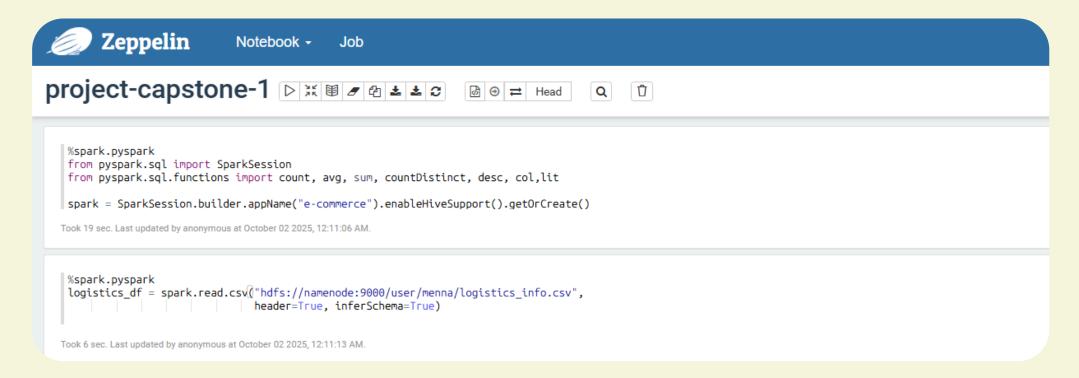


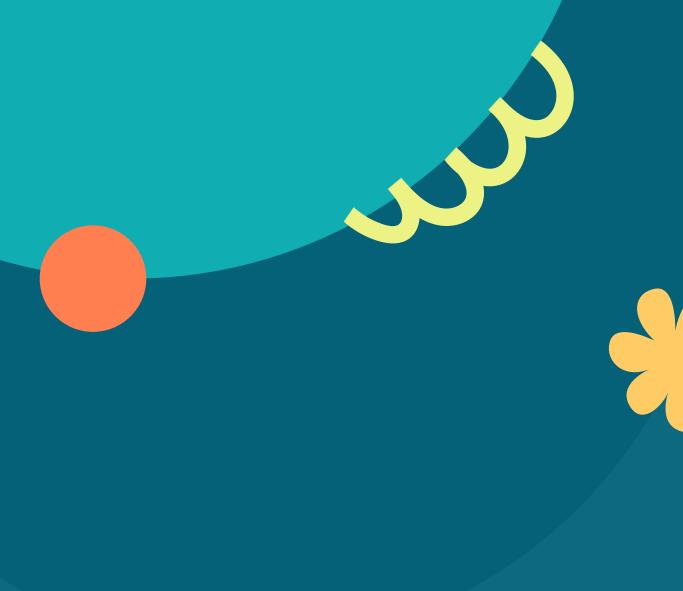
DATA EXTRACTION

1. Uploading Data to HDFS

 hdfs dfs -put /home/menna/projects/Big-Data-Cluster/ logistics_info.csv /user/menna/

2. Reading Data from HDFS in PySpark







DATA TRANSFORMATION



PREPROCESSING

FILTERING OF UNRELTED DATA

ADJUSTING ATTRIBUTES SO THAT EACH COUNTRY HAS A WAREHOUSE ID

ADJUSTING THE ATTRIBUTES
DATA TYPES

TRANSFORMATION

1. JOINING DATASETS

The Logistics Information and Order Information datasets were joined on the order_id column.

2. CALCULATING DELIVERY TIME DELTA

A new column, delivery_time_delta_days, was derived using the Spark datediff function

3. DERIVING KEY PERFORMANCE INDICATOR (KPI) - IS_LATE_DELIVERY

```
# Calculate delivery delta and late delivery
joined_df = joined_df.withColumn(
    "delivery_time_delta_days",
    F.datediff(F.col("actual_delivery_date"), F.col("expected_delivery_date"))
).withColumn(
    "is_late_delivery",
    F.when(F.col("delivery_time_delta_days") > 0, True).otherwise(False)
)
```

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TRANSFORMATION

4. CREATING THE FINAL OUTPUT TABLE: STAGING_LOGISTICS_FACT

We selected only the relevant columns from the joined DataFrame to create a concise table.

The table captures

- Operational metrics (delivery time delta, late delivery flag) a
- Financial/logistics metrics (shipping cost, warehouse_id) for downstream analysis.

```
%spark.pyspark
# Select the columns that exist in joined_df
Staging_Logistics_Fact = joined_df.select(
     "order id",
     "delivery_time_delta_days",
     "is_late_delivery",
     "shipping_cost",
     "warehouse id"
Staging Logistics Fact.show()
order id/delivery time delta days/is late delivery/shipping cost/warehouse id/
|15081.289|
                                                                        29.0
                                             falsel
                                                        84.99157
[56444.684]
                                             false
                                                          181.99
                                                                        25.0
7508.5713
                                             falsel
                                                        93.81015
                                                                         2.0
[56196.926]
                                             false
                                                        99.8906
                                                                        14.0
[5565.5796]
                                             false
                                                       171.07587
                                                                         1.0
32955.824
                                                                         0.0
                                             false
                                                       145.46329
35385.855
                                             falsel
                                                          167.99
                                                                         0.0
  36338.4
                                              truel
                                                          116.99
                                                                         0.0
27692.854
                                             false|
                                                       113.15623
                                                                        21.0
56918.418
                                             falsel
                                                          127.39
                                                                         1.0
67071.39
                                             falsel
                                                      111.575935
                                                                         2.0
58629.56
                                             false
                                                          299.98
                                                                         5.0
15836.077
                                              truel
                                                       290.95166
                                                                         4.0
45317.816
                                             false
                                                           189.0
                                                                         20.0
39779.484
                                             falsel
                                                          123.49
                                                                         0.0
```

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TRANSFORMATION

5. CALCULATING AVERAGE DELIVERY TIME DELTA PER WAREHOUSE

To evaluate the performance of each warehouse, we calculated the average delivery time delta using Spark SQL on the Staging_Logistics_Fact table.

This metric provides insight into how early or late orders are delivered on average for each warehouse.

```
%spark.pyspark
Staging_Logistics_Fact.createOrReplaceTempView("staging_logistics_fact")
avgDeliveryTimeDelta = spark.sql("""
    SELECT
        warehouse id.
        AVG(delivery_time_delta_days) AS avg_delivery_time_delta
    FROM staging_logistics_fact
    GROUP BY warehouse id
avgDeliveryTimeDelta.show()
|warehouse id|avg delivery time delta|
       147.0
                                -1.0
                0.06310679611650485
         8.0
        70.0 | 0.058823529411764705 |
                0.21052631578947367
                0.07115198451113262
                -0.1666666666666666
                 0.09925093632958802
       142.0
       112.0
                 -0.333333333333333333
       124.0
                                1.0
       128.0
                                -0.5
       108.0
                                 0.0
       133.0
        88.0
                  0.11111111111111111
```

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DATA LOADING & INTEGRATION







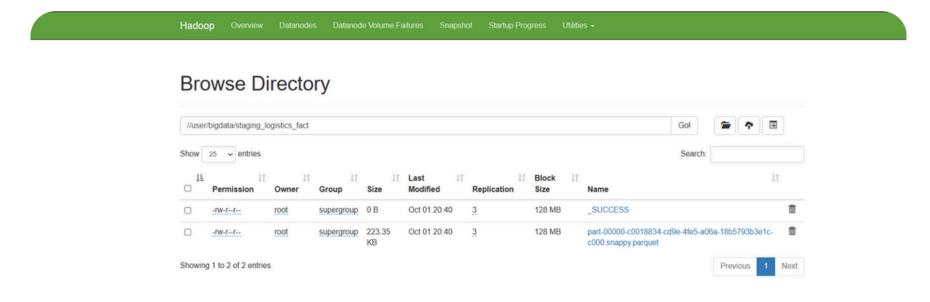
```
%spark.pyspark
from pyspark.sql.functions import col

Staging_Logistics_Fact_fixed = Staging_Logistics_Fact.select(
    col("order_id").cast("int"),
    col("delivery_time_delta_days").cast("int"),
    col("is_late_delivery").cast("boolean"),
    col("shipping_cost").cast("float"),
    col("warehouse_id").cast("int")
)

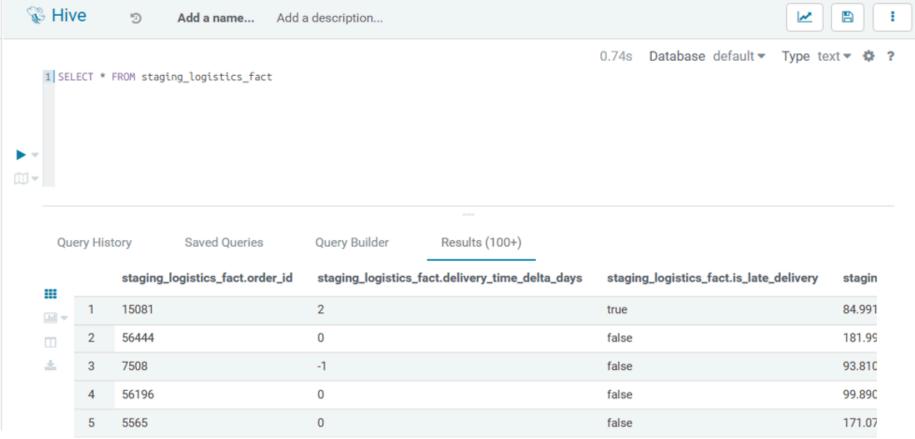
Staging_Logistics_Fact_fixed.write.mode("overwrite").parquet("hdfs://namenode:9000/user/bigdata/staging_logistics_fact")
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```

%spark.pyspark
avgDeliveryTimeDelta.write.mode("overwrite").parquet("hdfs://namenode:9000/user/bigdata/avgDeliveryTimeDelta")

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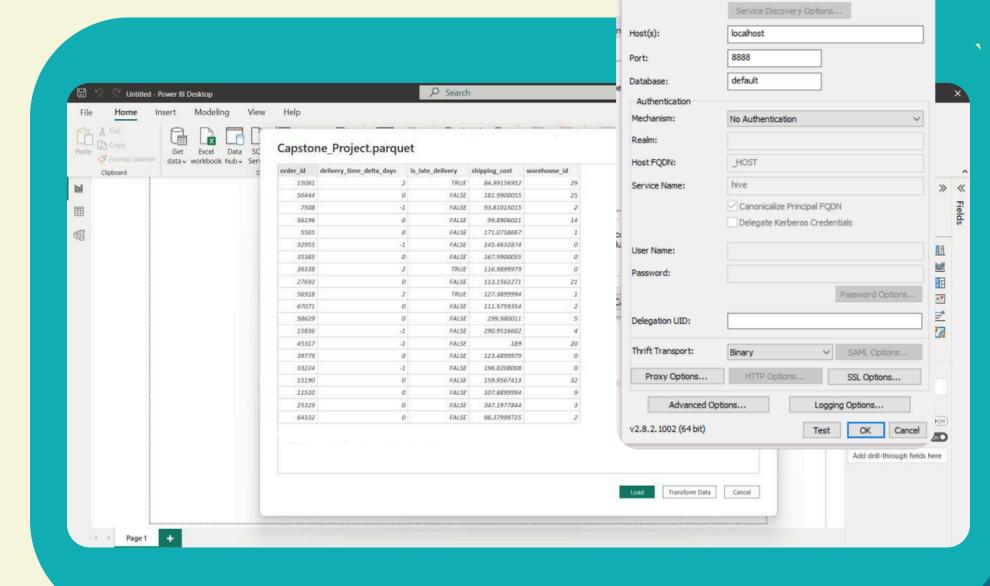
```
projects/Big-Data-Cluster$ docker exec -it hive-server bash
oot@b35050373558:/# beeline -u "jdbc:hive2://localhost:10000/default"
SLF4J: Class path contains multiple SLF4J bindings.
SLF4J: Found binding in [jar:file:/opt/hive/lib/log4j-slf4j-impl-2.6.2.jar!/org/slf4j/impl/StaticLoggerBinder.class]
SLF4J: Found binding in [jar:file:/opt/hadoop-2.7.4/share/hadoop/common/lib/slf4j-log4j12-1.7.10.jar!/org/slf4j/impl/StaticLoggerBinder.class]
SLF4J: See http://www.slf4j.org/codes.html#multiple_bindings for an explanation.
SLF4J: Actual binding is of type [org.apache.logging.slf4j.Log4jLoggerFactory]
Connecting to jdbc:hive2://localhost:10000/default
Connected to: Apache Hive (version 2.3.2)
Driver: Hive JDBC (version 2.3.2)
Transaction isolation: TRANSACTION_REPEATABLE_READ
Beeline version 2.3.2 by Apache Hive
  jdbc:hive2://localhost:10000/default>
  jdbc:hive2://localhost:10000/default> CREATE EXTERNAL TABLE staging_logistics_fact (
 . . . . . . . . . . . . . . . . . . order_id STRING,
  . . . . . . . . . . . . . . . . . . delivery_time_delta_days INT,
  .... is_late_delivery BOOLEAN,
  . . . . . . . . . . . . . . . . . shipping_cost_DOUBLE,
  . . . . . . . . . . . . . . . . warehouse_id STRING
  · · · · · · STORED AS PARQUET
  rows affected (1.449 seconds)
  jdbc:hive2://localhost:10000/default> SHOW TABLES;
       tab name
 staging_logistics_fact |
```



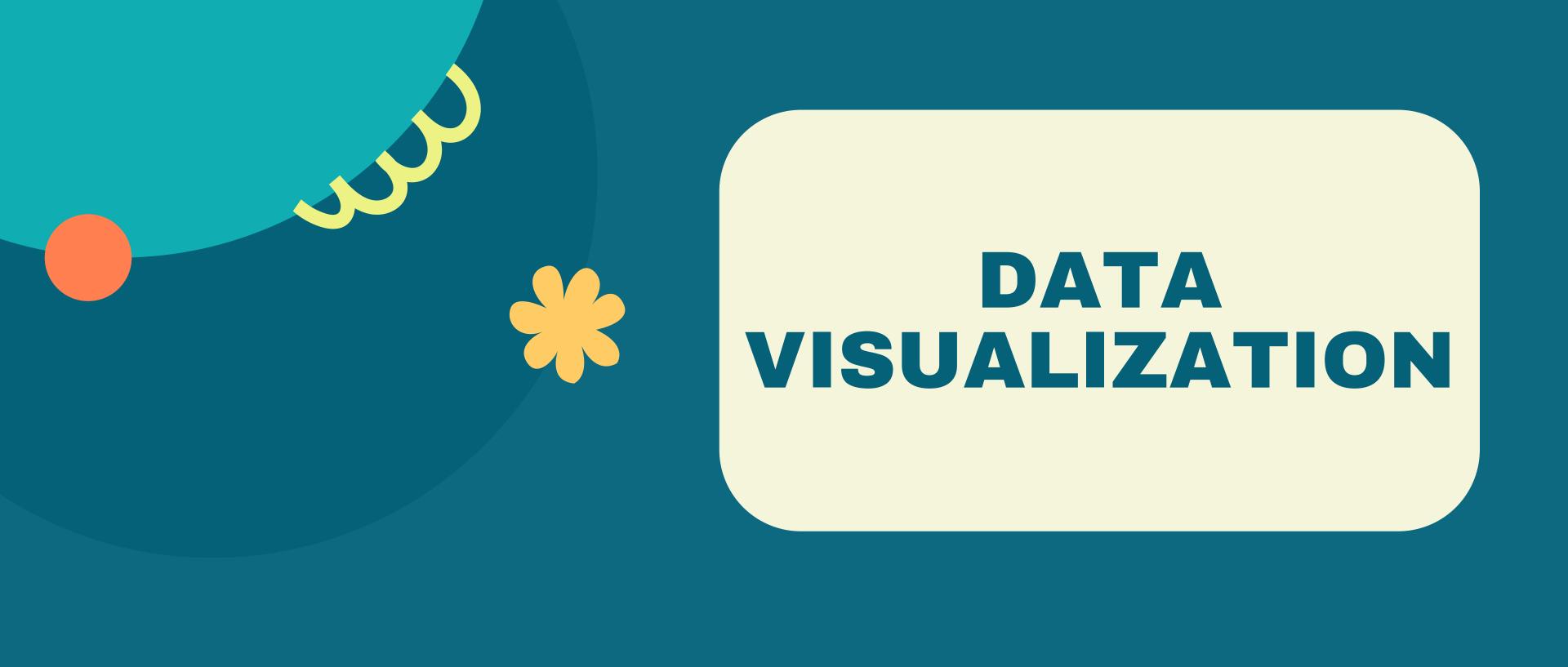


After the previous steps integrated the processed datasets into the Hive ecosystem for analysis, reporting, and visualization.

Data is loaded to powerBi to visualize it using ODBC to connect between Hive server and Power BI



Sample Cloudera Hive DSN





VISUALIZING IN POWER BI



