Northwind Database

I. Introduction

1. About the Northwind database

The Northwind database is a database developed by Microsoft for guiding their database products over many decades. It contains information about the sales activities of a fictional company named "Northwind Traders," specializing in importing and exporting specialty foods from around the world. Northwind serves as an ideal schema for small business ERP tutorials, encompassing details about customers, orders, inventory, purchasing, suppliers, shipping, employees, and accounting.

2. About Business Requirements

This report dives into SQL-based Northwind analytics, addressing sales business questions through SQL queries. We transformed data from OLTP to OLAP, identified Sales facts and dimensions to easily analyze Northwind's sales results.

 Sales analysis: Provides comprehensive reporting on sales figures to understand which products are being sold to customers, which products sell the most, where they are sold, and which products sell the least. The goal is to get a comprehensive view of the business

These analytics are intended to provide actionable insights and enhance the decision-making process for Northwind in terms of product sales.

3. Business question

The business process helps us understand what business questions we are asking. Here we focus on the what information is available; i.e., the attributes and relations between them. In Northwinds Trading Company database, we might want to try and answer some of these questions during the business modeling process:

- Total revenue, total profit, total sales, total orders, total employees, total customers
- Which country has the highest revenue?
- Which product category has the highest sales?
- Which company has the highest total revenue?
- Which product has the highest sales?
- How does revenue change over the months of the year?
- Which products are sold the most, where they are sold?
- Which products have the most Revenue?
- Which employees have the most total sales?
- Classify customers based on order frequency
- Which customer bought the most products?

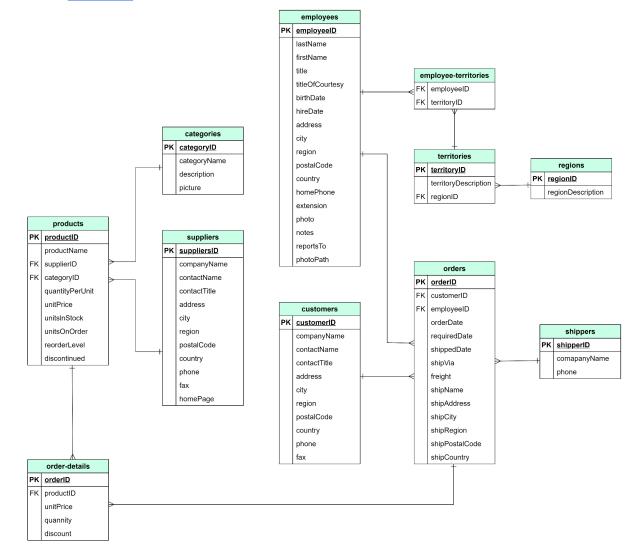
II. Database design

1. Design an entity-relationship diagram

Identifying required tables from ERD

- Customers Individuals or entities purchasing goods from Northwind.
- Employees Individuals working for Northwind.
- Employees territories an entity linking employees with territories, indicating which territories are managed by specific employees.
- Categories Categories of products that Northwind offers for sale.
- Orders Transactions where orders are placed between customers and Northwind.
- Order Details Detailed information about the items in orders placed by customers.
- Products Products currently available for customers to purchase from Northwind.
- Regions Entity that describes a geographic region.
- Shippers Entities responsible for shipping orders from Northwind to customers.
- Suppliers Entities supplying necessary items to Northwind.
- Territories Another entity related to geographical territories, possibly used in conjunction with Employees Territories.

LINK: TAI ĐÂY



2. Build a database based on the ERD

Employees CREATE TABLE "Employees" "EmployeeID" "int" IDENTITY (1, 1) NOT NULL , cmployeeID Int IDENTITY (1, 1) NOT "LastName" nvarchar (20) NOT NULL , "FirstName" nvarchar (10) NOT NULL , "Title" nvarchar (30) NULL , "TitleOfCourtesy" nvarchar (25) NULL , "BirthDate" "datetime" NULL , "HireDate" "datetime" NULL , "Address" nvarchar (60) NULL, "City" nvarchar (15) NULL , "Region" nvarchar (15) NULL , "PostalCode" nvarchar (10) NULL , "Country" nvarchar (15) NULL , "HomePhone" nvarchar (24) NULL , "Extension" nvarchar (4) NULL , "Photo" "image" NULL , "Notes" "ntext" NULL , "ReportsTo" "int" NULL , "PhotoPath" nvarchar (255) NULL , CONSTRAINT "PK_Employees" PRIMARY KEY CLUSTERED "EmployeeID" CONSTRAINT "FK_Employees_Employees" FOREIGN KEY "ReportsTo") REFERENCES "dbo"."Employees" ("EmployeeID" CONSTRAINT "CK Birthdate" CHECK (BirthDate < getdate())) CREATE INDEX "LastName" ON "dbo"."Employees"("LastName") CREATE INDEX "PostalCode" ON "dbo"."Employees"("PostalCode") GO

```
Customers
CREATE TABLE "Customers"
      ATE TABLE "Customers" (
"CustomerID" nchar (5) NOT NULL,
"CompanyName" nvarchar (40) NOT NULL,
"ContactName" nvarchar (30) NULL,
"ContactTitle" nvarchar (30) NULL,
"Address" nvarchar (60) NULL,
"City" nvarchar (15) NULL,
"Region" nvarchar (15) NULL,
"PostalCode" nvarchar (10) NULL,
"Country" nvarchar (15) NULL,
       "Country" nvarchar (15) NULL ,
"Phone" nvarchar (24) NULL ,
      "Fax" nvarchar (24) NULL ,
CONSTRAINT "PK_Customers" PRIMARY KEY CLUSTERED
               "CustomerID"
Ġ0
 CREATE INDEX "City" ON "dbo"."Customers"("City")
 CREATE INDEX "CompanyName" ON "dbo"."Customers"("CompanyName")
 CREATE INDEX "PostalCode" ON "dbo"."Customers"("PostalCode")
 CREATE INDEX "Region" ON "dbo"."Customers"("Region")
```

Categories

```
CREATE TABLE "Categories"
        ALE LABLE "Categories" (
"CategorylO" "int" IDENTITY (1, 1) NOT NULL ,
"CategorylAmme" nvarchar (15) NOT NULL ,
"Description" "ntext" NULL ,
"Picture" "image" NULL ,
CONSTRAINT "PK_Categories" PRIMARY KEY CLUSTERED
                 "CategoryID"
  CREATE INDEX "CategoryName" ON "dbo"."Categories"("CategoryName")
```

Shippers

```
CREATE TABLE "Shippers" (
     "ShipperID" "int" IDENTITY (1, 1) NOT NULL ,
"CompanyName" nvarchar (40) NOT NULL ,
     "Phone" nvarchar (24) NULL ,
CONSTRAINT "PK_Shippers" PRIMARY KEY CLUSTERED
            "ShipperID"
GO
```

Suppliers

```
CREATE TABLE "Suppliers" (
"SupplierID" "int" IDENTITY (1, 1) NOT NULL,
"CompanyName" nvarchar (40) NOT NULL,
"ContactName" nvarchar (30) NULL,
"ContactTitle" nvarchar (30) NULL,
"Address" nvarchar (60) NULL,
"City" nvarchar (15) NULL,
"Region" nvarchar (15) NULL,
"PostalCode" nvarchar (10) NULL,
"Country" nvarchar (15) NULL,
"Phone" nvarchar (24) NULL,
"Fax" nvarchar (24) NULL,
"HomePage" "ntext" NULL,
CONSTRAINT "PK_Suppliers" PRIMARY KEY CLUSTERED
                            "SupplierID'
     CREATE INDEX "CompanyName" ON "dbo"."Suppliers"("CompanyName")
     CREATE INDEX "PostalCode" ON "dbo"."Suppliers"("PostalCode")
```

Region

```
CREATE TABLE [dbo].[Region]
    ( [RegionID] [int] NOT NULL ,
    [RegionDescription] [nchar] (50) NOT NULL
) ON [PRIMARY]
GO
```

Orders CREATE TABLE "Orders" ("OrderID" "int" IDENTITY (1, 1) NOT NULL , "CustomerID" nchar (5) NULL "EmployeeID" "int" NULL , "OrderDate" "datetime" NULL "RequiredDate" "datetime" NULL , "ShippedDate" "datetime" NULL , "ShipPlane" nvarchar (60) NULL , "ShipVia" "int" NULL , "Freight" "money" NULL CONSTRAINT "DF_Orders_Freight" DEFAULT (0), "ShipName" nvarchar (40) NULL , "ShipAddress" nvarchar (60) NULL , "ShipCity" nvarchar (15) NULL , "ShipRegion" nvarchar (15) NULL , "ShipPostalCode" nvarchar (10) NULL , "ShipCountry" nvarchar (15) NULL , CONSTRAINT "PK_Orders" PRIMARY KEY CLUSTERED "OrderID" CONSTRAINT "FK_Orders_Customers" FOREIGN KEY) REFERENCES "dbo"."Customers" ("CustomerID" CONSTRAINT "FK_Orders_Employees" FOREIGN KEY "EmployeeID") REFERENCES "dbo"."Employees" ("EmployeeID" CONSTRAINT "FK Orders Shippers" FOREIGN KEY "ShipVia") REFERENCES "dbo"."Shippers" ("ShipperID" CREATE INDEX "CustomerID" ON "dbo"."Orders"("CustomerID") CREATE INDEX "CustomersOrders" ON "dbo"."Orders"("CustomerID") CREATE INDEX "EmployeeID" ON "dbo"."Orders"("EmployeeID") CREATE INDEX "EmployeesOrders" ON "dbo"."Orders"("EmployeeID") CREATE INDEX "OrderDate" ON "dbo"."Orders"("OrderDate") CREATE INDEX "ShippedDate" ON "dbo"."Orders"("ShippedDate") CREATE INDEX "ShippersOrders" ON "dbo"."Orders"("ShipVia")

Products

Order Details

CREATE INDEX "ShipPostalCode" ON "dbo"."Orders"("ShipPostalCode")

GO

```
CREATE TABLE "Order Details" (
"OrderID" int" NOT NULL ,
"ProductID" int" NOT NULL ,
"ProductID" int" NOT NULL ,
"ProductID" int" NOT NULL CONSTRAINT "DF_Order_Details_UnitPrice" DEFAULT (0),
"Quantity" smallint" NOT NULL CONSTRAINT "DF_Order_Details_Quantity" DEFAULT (1),
"Discount" "real" NOT NULL CONSTRAINT "DF_Order_Details_Quantity" DEFAULT (1),
CONSTRAINT "FK_Order_Details" PRIMARY KEY CLUSTERED

(
    "OrderID",
    "ProductID"
),
CONSTRAINT "FK_Order_Details_Orders" FOREIGN KEY
(
    "OrderID"
) REFERENCES "dbo"."Orders" (
    "OrderID"
),
CONSTRAINT "FK_Order_Details_Products" FOREIGN KEY
(
    "ProductID"
),
CONSTRAINT "FK_Order_Details_Products" FOREIGN KEY
(
    "ProductID"
),
CONSTRAINT "CK_Quantity" CHECK (Discount >= 0 and (Discount <= 1)),
CONSTRAINT "CK_Quantity" CHECK (Quantity > 0),
CONSTRAINT "CK_Quantity" CHECK (Quantity >= 0)
)
GO
CREATE INDEX "OrderID" ON "dbo"."Order Details"("OrderID")
GO
CREATE INDEX "ProductID" ON "dbo"."Order Details"("ProductID")
GO
CREATE INDEX "ProductID" ON "dbo"."Order Details"("ProductID")
GO
CREATE INDEX "ProductSorder_Details" ON "dbo"."Order Details"("ProductID")
GO
CREATE INDEX "ProductSorder_Details" ON "dbo"."Order Details"("ProductID")
GO
CREATE INDEX "ProductSorder_Details" ON "dbo"."Order Details"("ProductID")
```

Territories and EmployeeTerritories:

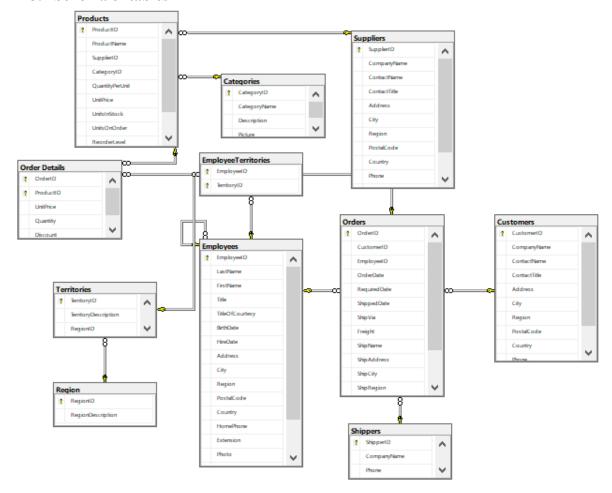
```
CREATE TABLE [dbo].[Territories]
   ([TerritoryID] [nvarchar] (20) NOT NULL ,
   [RegionID] [int] NOT NULL )

ON [PRIMARY]

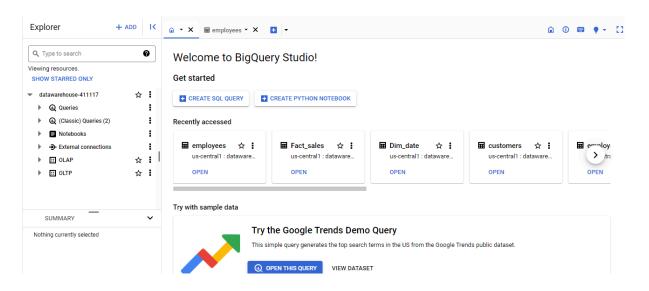
CREATE TABLE [dbo].[EmployeeTerritories]
   ([EmployeeID] [int] NOT NULL,
   [TerritoryID] [nvarchar] (20) NOT NULL

ON [PRIMARY]
```

3. Schema of tables



4. Configure a specific Data Warehouse system (Google Bigquery)



III. Build a data warehouse

1. Choose the Business Process

Business Process: Sales Analysis

In this case, the primary focus is on understanding and improving the sales performance of products. This involves analyzing which products are sold the most, where they are sold, and which products generate the highest revenue.

2. Declare the Grain

The grain is known as the lowest level of the star schema that a business would have. The grain sets the foundation of the entire model development from the order and the inventory business drivers. It determines the level of detailed information that can be made available to the dimensional model. It is quite essential to give this section more time and accurate results because any error on the grain section can cause redesigning of the whole model. The table's grain consists of information on the facts in the data, necessary when building a report for business intelligence. The level at which a row is granular should be atomic, and it holds the most specific details for tables in any data warehouse. Atomic grain refers to the lowest level of grain that is possibly captured by a business on daily and monthly aggregations of individual transactions.

In this case, the grain should be at the transactional level, capturing individual sales transactions, is sales information to understand the number of orders, number of products, number of customers, number of sales staff and the date.

Components of the Grain:

- orderID (Primary Key): A unique identifier for each transaction. This ensures that each record in the data warehouse is distinct and can be easily referenced.
- productID (Foreign Key): A reference to the Product Dimension, linking the sale to the specific product that was sold. This enables analysis at the product level.
- customerID (Foreign Key): A reference to the Customer Dimension, linking the sale to the customer who made the purchase. This facilitates customer-centric analysis, such as identifying the top customers or analyzing sales by customer segments.
- employeeID (Foreign Key): A reference to the Employee Dimension, linking the sale to the employee, analysis of sales performance on an individual employee level.
- categoryID (Foreign Key): A reference to the Category Dimension, linking the sale to the category of product.
- date_key (Foreign Key): A reference to the Date Dimension, indicating the date and time of the sale. This allows for time-based analysis, such as sales trends over days, months, and years.OrderID

3. Identify the Dimensions

The Customers dimension table is a copy of the Customers table and contains an extra attribute: Insertion_timestamp.

The Employees dimension table is a copy of the Employees table and contains an extra attribute: Insertion timestamp.

The Products dimension table is a copy of the Products table and contains an extra attribute: Insertion timestamp.

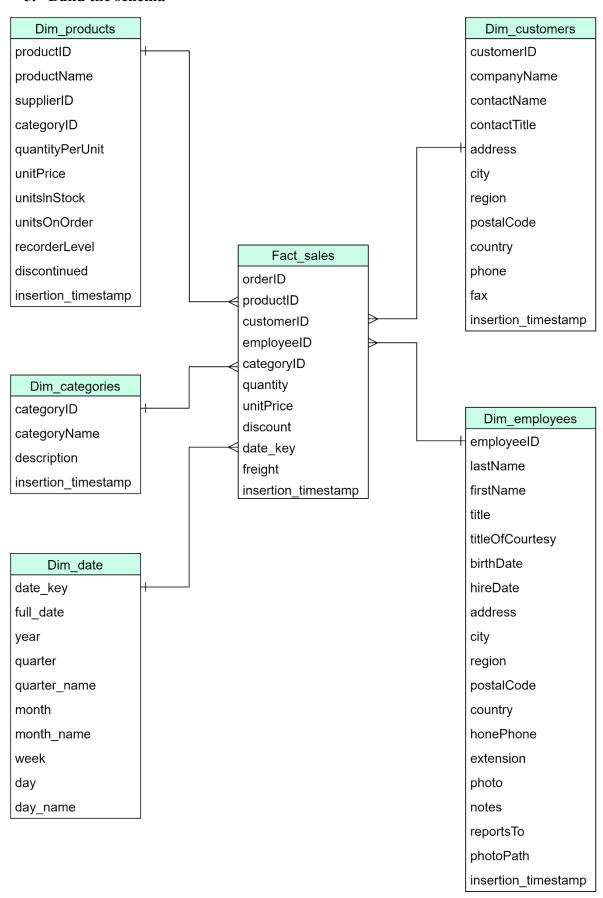
The Categories dimension table is created from the Categories table by cutting out attributes: Picture and contains an extra attribute: Insertion_timestamp.

The Date dimension table includes attributes: Date_key, Full_date, Year, Quarter, Quarter_name, Month, Month_name, Week, Day, Day_name.

4. Identify the Fact

After identifying the dimensions, one must determine the fact tables. Facts are measurable data such as prices per unit, which are in a fact table. The facts are orders and order details. Variables are unit price, quantity, discount, units on stock.

5. Build the schema



IV. Deployment

1. Yml file Docker:

```
x-airflow-common:
 &airflow-common
you can use your extended image.
where you placed the docker-compose.yaml
build` to build the images.
   &airflow-common-env
postgresql+psycopg2://airflow:airflow@postgres/airflow
db+postgresql://airflow:airflow@postgres/airflow
   AIRFLOW CORE FERNET KEY: ''
```

```
https://airflow.apache.org/docs/apache-airflow/stable/administration-an
d-deployment/logging-monitoring/check-health.html#scheduler-health-chec
nfig/gcloud/application default credentials.json
services:
      POSTGRES USER: airflow
      test: ["CMD", "pg isready", "-U", "airflow"]
```

```
redis:
airflow-webserver:
  <<: *airflow-common
 command: webserver
   <<: *airflow-common-depends-on
airflow-scheduler:
  <<: *airflow-common
    <<: *airflow-common-depends-on
```

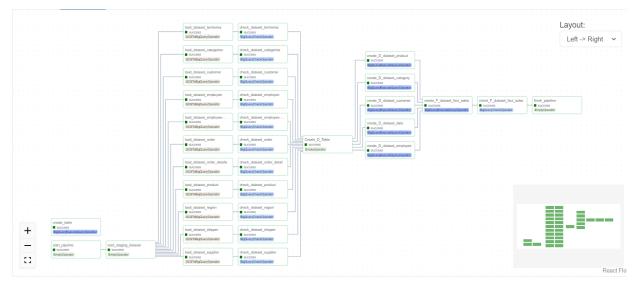
```
airflow-worker:
   <<: *airflow-common
                                                      'celery
airflow.providers.celery.executors.celery_executor.app inspect ping -d
     <<: *airflow-common-env
     <<: *airflow-common-depends-on
   <<: *airflow-common
-hostname "$${HOSTNAME}"']
     <<: *airflow-common-depends-on
```

```
airflow-init:
    <<: *airflow-common
below to set "
be owned by root."
warning with manually created .env file:"
https://airflow.apache.org/docs/apache-airflow/stable/howto/docker-comp
ose/index.html#setting-the-right-airflow-user"
PAGE SIZE) / one meg))
for Docker.\e[0m"
           echo -e "\033[1;33mWARNING!!!: Not enough CPUS available for
Docker.\e[0m"
```

```
"At
                               least 2 CPUs recommended.
                                                                   have
$${cpus available}"
       fi
available for Docker.\e[0m"
iec $$((disk available * 1024 )))"
to run Airflow (see above)!\e[0m"
            echo "Please follow the instructions to increase amount of
resources available:"
https://airflow.apache.org/docs/apache-airflow/stable/howto/docker-comp
ose/index.html#before-you-begin"
     <<: *airflow-common-env
${ AIRFLOW WWW USER USERNAME:-airflow}
${ AIRFLOW WWW USER PASSWORD:-airflow}
```

```
<<: *airflow-common
     <<: *airflow-common-env
     - airflow
docker-compose --profile flower up
up flower.
   <<: *airflow-common
     <<: *airflow-common-depends-on
volumes:
```

2. DAG file:



Building a pipeline transform OLTP to OLAP

```
from datetime import timedelta, datetime
from airflow import DAG
from airflow.utils.dates import days ago
from airflow.operators.dummy operator import DummyOperator
from airflow.providers.google.cloud.transfers.gcs to bigquery
GCSToBigQueryOperator
          airflow.providers.google.cloud.operators.bigquery
BigQueryCheckOperator
             airflow.contrib.operators.bigquery operator
BigQueryOperator
GOOGLE CONN ID = "google cloud default"
PROJECT ID="datawarehouse-411117"
GS PATH = "data/"
BUCKET NAME = 'us-centrall-datawarehouse-ad43d7dc-bucket'
STAGING DATASET = "OLTP"
DATASET = "OLAP"
LOCATION = "us-central1"
default args = {
    'start date': datetime(2024, 1, 1),
    'retry delay': timedelta(minutes=5),
```

```
DAG('northwind', schedule interval=timedelta(days=1),
default args=default args) as dag:
    start pipeline = DummyOperator(
       dag = dag
   load staging dataset = DummyOperator(
       dag = dag
   create table task = BigQueryOperator(
       sql="""
       CREATE OR REPLACE TABLE `datawarehouse-411117.OLTP.customers` (
           customerID STRING,
           address STRING,
   use legacy sql=False,
   dag=dag,
   load dataset customer = GCSToBigQueryOperator(
       bucket=BUCKET NAME,
        source objects=['data/customers.csv'],
destination project dataset table=f'{PROJECT ID}:{STAGING DATASET}.cust
```

```
write disposition='WRITE TRUNCATE',
       source format='CSV',
       field delimiter=',',
       schema fields=[
                  {'name': 'contactTitle', 'type': 'STRING', 'mode':
                    {'name': 'postalCode', 'type': 'STRING', 'mode':
'NULLABLE' },
   load dataset employee = GCSToBigQueryOperator(
       bucket = BUCKET NAME,
       source objects = ['data/employees.csv'],
                                 destination project dataset table
       write disposition='WRITE TRUNCATE',
       field delimiter=',',
   load dataset product = GCSToBigQueryOperator(
       bucket = BUCKET NAME,
       source objects = ['data/products.csv'],
                                destination project dataset table
       write disposition='WRITE TRUNCATE',
```

```
field delimiter=',',
load dataset supplier = GCSToBigQueryOperator(
   bucket = BUCKET NAME,
    source objects = ['data/suppliers.csv'],
                             destination project dataset table
   write disposition='WRITE TRUNCATE',
   field delimiter=',',
load dataset order = GCSToBigQueryOperator(
   source objects = ['data/orders.csv'],
                             destination project dataset table
   write disposition='WRITE TRUNCATE',
   source format = 'csv',
   field delimiter=',',
load dataset order details = GCSToBigQueryOperator(
   source objects = ['data/order-details.csv'],
                             destination project dataset table
   write disposition='WRITE TRUNCATE',
    source format = 'csv',
    field delimiter=',',
load dataset categories = GCSToBigQueryOperator(
   bucket = BUCKET NAME,
   source objects = ['data/categories.csv'],
                             destination project dataset table
   write disposition='WRITE TRUNCATE',
    field delimiter=',',
```

```
load dataset territories = GCSToBigQueryOperator(
   bucket = BUCKET NAME,
    source objects = ['data/territories.csv'],
                             destination project dataset table
   write disposition='WRITE TRUNCATE',
   source format = 'csv',
load dataset employee territories = GCSToBigQueryOperator(
    source objects = ['data/employee-territories.csv'],
                             destination project dataset table
   write disposition='WRITE TRUNCATE',
   source format = 'csv',
    field delimiter=',',
load dataset region = GCSToBigQueryOperator(
   bucket = BUCKET NAME,
    source objects = ['data/regions.csv'],
                             destination project dataset table
   write disposition='WRITE TRUNCATE',
    field delimiter=',',
load dataset shipper = GCSToBigQueryOperator(
   bucket = BUCKET NAME,
    source objects = ['data/shippers.csv'],
                             destination project dataset table
   write disposition='WRITE TRUNCATE',
   source format = 'csv',
    field delimiter=',',
check dataset customer = BigQueryCheckOperator(
```

```
use_legacy_sql=False,
   location = LOCATION,
                                     f'SELECT
                          sql
                                                            FROM
check dataset employee = BigQueryCheckOperator(
   use legacy sql=False,
                          sql = f'SELECT count(*)
check dataset product = BigQueryCheckOperator(
   use legacy sql=False,
                          sql =
                                    f'SELECT
                                                            FROM
check dataset supplier = BigQueryCheckOperator(
   use_legacy sql=False,
                          sql =
                                    f'SELECT
                                                            FROM
check dataset order = BigQueryCheckOperator(
   use_legacy_sql=False,
   location = LOCATION,
                          sql = f'SELECT
                                                            FROM
check dataset order detail = BigQueryCheckOperator(
   use legacy sql=False,
   location = LOCATION,
                              = f'SELECT
                                                            FROM
                          sql
check dataset categories = BigQueryCheckOperator(
```

```
task id = 'check dataset categories',
   use legacy sql=False,
   location = LOCATION,
                          sql = f'SELECT count(*)
check dataset employee territories = BigQueryCheckOperator(
   use legacy sql=False,
                          sql = f'SELECT
check dataset territories = BigQueryCheckOperator(
   use legacy sql=False,
   location = LOCATION,
                          sql = f'SELECT count(*)
                                                           FROM
check dataset region = BigQueryCheckOperator(
   use legacy sql=False,
   location = LOCATION,
                          sql = f'SELECT count(*)
check dataset shipper = BigQueryCheckOperator(
   use legacy sql=False,
                          sql
create_D_Table = DummyOperator(
   dag = dag
create D dataset customer = BigQueryOperator(
   use legacy sql = False,
   location = LOCATION,
```

```
sql = './sql/dim customers.sql'
create D dataset employee = BigQueryOperator(
   use_legacy sql = False,
   location = LOCATION,
    sql = './sql/dim employees.sql'
create D dataset product = BigQueryOperator(
   use legacy sql = False,
   sql = './sql/dim products.sql'
create D dataset category = BigQueryOperator(
   use legacy sql = False,
   location = LOCATION,
   sql = './sql/dim category.sql'
create D dataset date = BigQueryOperator(
   use legacy sql = False,
   location = LOCATION,
   sql = './sql/dim date.sql'
create F dataset fact sales = BigQueryOperator(
   use legacy sql = False,
   sql = './sql/fact sales.sql'
check F dataset fact sales = BigQueryCheckOperator(
   use legacy sql=False,
   location = LOCATION,
                                 = f'SELECT count(*)
                                                                FROM
                           sql
```

```
finish pipeline = DummyOperator(
start pipeline >> load staging dataset
load staging dataset >> [load dataset customer, load dataset employee,
load dataset product, load dataset supplier, load dataset order,
load dataset order details,
load dataset categories, load dataset employee territories,
load dataset territories, load dataset region, load dataset shipper]
load dataset customer >> check dataset customer
load dataset employee >> check dataset employee
load dataset product >> check dataset product
load dataset supplier >> check dataset supplier
load dataset order >> check dataset order
load dataset order details >> check dataset order detail
load dataset categories >> check dataset categories
load dataset employee territories >> check dataset employee territories
load dataset territories >>check dataset territories
load dataset region >>check dataset region
load dataset shipper >> check dataset shipper
[check dataset customer, check dataset employee, check dataset product,
check dataset supplier, check dataset order,
check dataset order detail, check dataset categories,
check dataset employee territories,
check dataset territories,check dataset region,check dataset shipper
>> create D Table
create D Table>> [create D dataset customer, create D dataset employee,
create D dataset_product,create_D dataset_category,
create D dataset date]
[create D dataset customer, create D dataset employee,
create D dataset product,create D dataset category,
create D dataset date] >> create F dataset fact sales
create F dataset fact sales>>check F dataset fact sales>>
finish pipeline
```

- 3. Transform code content to transform Data Warehouse tables:
- dags\sql\dim category.sql:

```
CREATE OR REPLACE TABLE
  `datawarehouse-411117.OLAP.Dim_categories` AS

SELECT
  categoryID,
  categoryName,
  description,
  current_timestamp() as insertion_timestamp,

FROM
  `datawarehouse-411117.OLTP.categories`

QUALIFY ROW_NUMBER() OVER(PARTITION BY categoryID) = 1;
```

• dags\sql\dim customers.sql:

```
CREATE OR REPLACE TABLE

datawarehouse-411117.OLAP.Dim_customers AS

SELECT

customerID,

companyName,

contactName,

contactTitle,

address,

city,

region,

postalCode,

country,

phone,

fax,

current_timestamp() as insertion_timestamp,

FROM

datawarehouse-411117.OLTP.customers

QUALIFY ROW_NUMBER() OVER(PARTITION BY customerID) = 1;
```

• dags\sql\dim date.sql:

```
CREATE OR REPLACE TABLE `datawarehouse-411117.OLAP.Dim_date` AS

SELECT

CAST(FORMAT_DATE('%Y%m%d', date) AS INT64) AS date_key,

date AS full_date,

EXTRACT(YEAR FROM date) AS year,

EXTRACT(QUARTER FROM date) AS quarter,
```

```
CONCAT('Q', CAST(EXTRACT(QUARTER FROM date) AS STRING)) AS

quarter_name,

EXTRACT(MONTH FROM date) AS month,

FORMAT_DATE('%B', date) AS month_name,

EXTRACT(WEEK FROM date) AS week,

EXTRACT(DAY FROM date) AS day,

FORMAT_DATE('%A', date) AS day_name

FROM

UNNEST(GENERATE_DATE_ARRAY(DATE '1996-07-04', DATE '2030-12-31',

INTERVAL 1 DAY)) AS date;
```

• dags\sql\dim employees.sql:

```
CREATE OR REPLACE TABLE
 datawarehouse-411117.OLAP.Dim employees AS
SELECT
 employeeID,
 lastName,
 title,
 titleOfCourtesy,
 birthDate,
 hireDate,
 city,
 region,
 postalCode,
 country,
 homePhone,
 photo,
 notes,
 reportsTo,
 photoPath,
 current timestamp() as insertion timestamp,
FROM
 datawarehouse-411117.OLTP.employees
QUALIFY ROW NUMBER() OVER(PARTITION BY employeeID) = 1;
```

• dags\sql\dim products.sql:

```
CREATE OR REPLACE TABLE
`datawarehouse-411117.OLAP.Dim_products` AS
```

```
SELECT
  productID,
  productName,
  supplierID,
  categoryID,
  quantityPerUnit,
  unitPrice,
  unitsInStock,
  unitsOnOrder,
  reorderLevel,
  discontinued,
  current_timestamp() as insertion_timestamp,

FROM
  `datawarehouse-411117.OLTP.products`
QUALIFY ROW_NUMBER() OVER(PARTITION BY productID) = 1;
```

• dags\sql\fact sales.sql:

```
CREATE OR REPLACE TABLE datawarehouse-411117.OLAP.Fact sales AS
SELECT
 od.orderID,
 od.productID,
 o.customerID,
 o.employeeID,
 p.categoryID,
 od.quantity,
 od.unitPrice,
 od.discount,
 CAST (FORMAT DATE ('%Y%m%d', o.orderDate) AS INT64) AS date key,
 o.freight,
 CURRENT_TIMESTAMP() AS insertion_timestamp
FROM
  `datawarehouse-411117.OLTP.order-details` od
LEFT JOIN `datawarehouse-411117.OLTP.orders` o
ON od.orderID = o.orderID
LEFT JOIN `datawarehouse-411117.OLTP.products` p
ON od.productID= p.productID
WHERE od.orderID IS NOT NULL
QUALIFY ROW_NUMBER() OVER(PARTITION BY o.orderID, od.productID) = 1;
```

V. Visualization

1. Dashboard

Overview

Total Revenue **1,265,793.04**

Total Profit 1,058,486.94

Total Sales 51,317

Total Orders 830

Total Employees

9

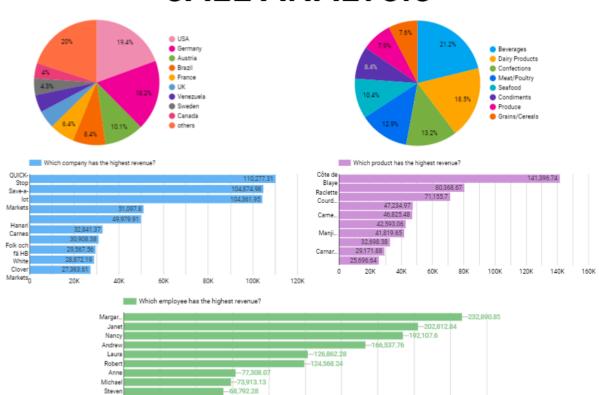
25K

75K

Total Customers

89

SALE ANALYSIS



125K

150K

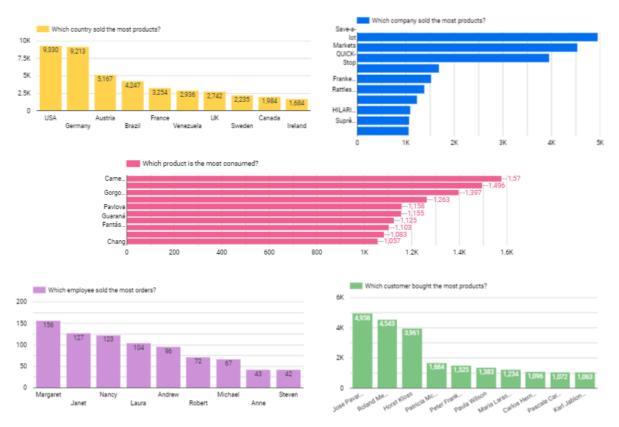
175K

200K

225K

250K

PRODUCT ANALYSIS



- 2. Answer business question
- Total revenue, total profit, total sales, total orders, total employees, total customers

Overview

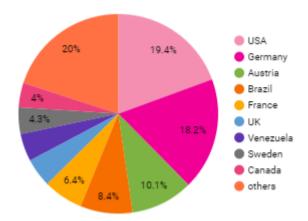
Total Revenue
1,265,793.04

Total Sales
51,317

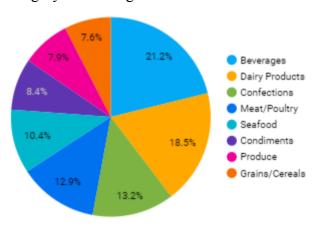
Total Employees
9

Total Customers
89

• Which country has the highest revenue?

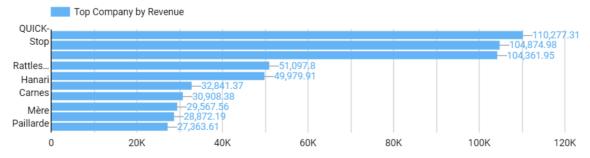


• Which product category has the highest sales?



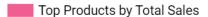
Beverages has the most revenue with 21,2%

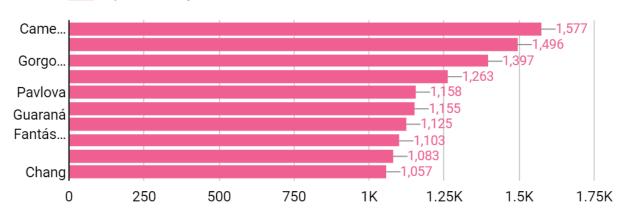
• Which company has the highest total revenue?



QUICK-Stop has the most revenue with \$110,277

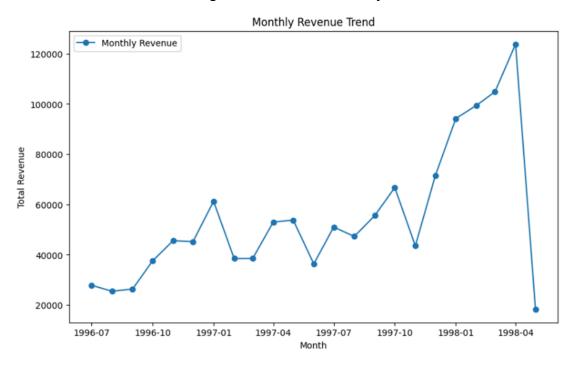
• Which product has the highest sales?



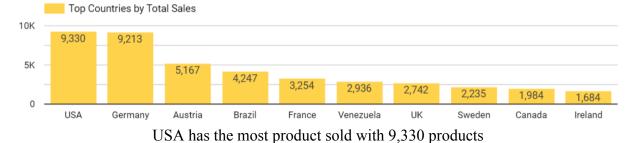


Camembert Pierrot has the most total sales with 1,577 products

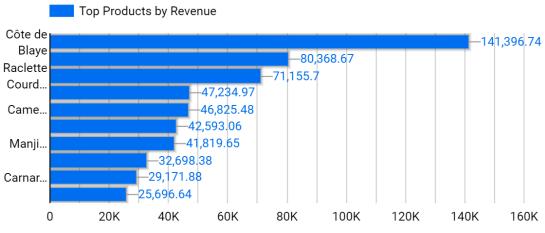
• How does revenue change over the months of the year?



• Which products are sold the most, where they are sold?

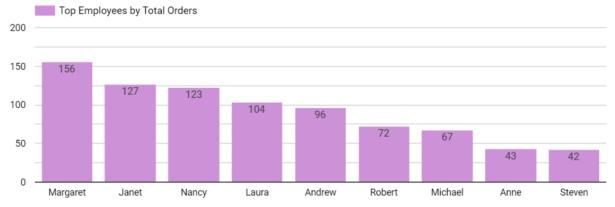


• Which products have the most Revenue?



Côte de Blaye has the most revenue with \$141,396

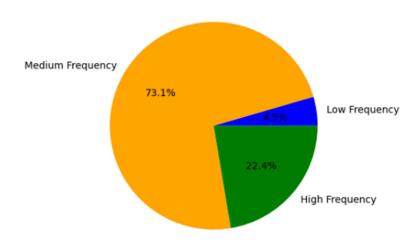
Which employees have the most total sales?



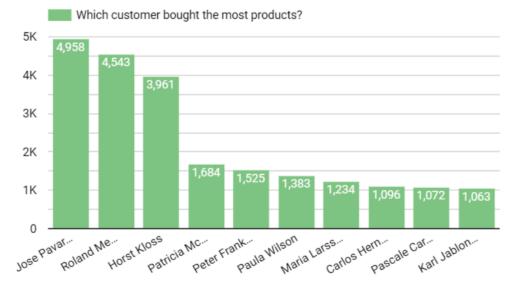
Margaret has the most total sales with 156 ordered

• Classify customers based on order frequency

Percentage of Customers in Each Category



• Which customer bought the most products?



Jose Pavarotti has bought the most products with 4,958

VI. Reference

1. Barnes, G. (n.d.). Implementing a Dimensional Data Warehouse with the SAS@System. [online]

Available at:

https://support.sas.com/resources/papers/proceedings/proceedings/sugi22/DATAWAR <u>E/PAPER129.PDF</u> [Accessed 17 Jan. 2024].

2. Simanjuntak, H., Nainggolan, A., Simatupang, D. and Manurung, D.D.V. (2017). An Automatic Tool to Transform Star Schema Data Warehouse to Physical Data Model. Journal of Telecommunication, Electronic and Computer Engineering (JTEC), [online] 9(2-3), pp.55–59.

Available at: https://jtec.utem.edu.my/jtec/article/view/2273 .

3. Kimball Group. (n.d.). Data Warehouse Lifecycle Toolkit. [online] Available at:

https://www.kimballgroup.com/data-warehouse-business-intelligence-resources/books/data-warehouse-dw-lifecycle-toolkit/ .