Data Warehouse Project Using Pentaho and Mondrian

0. Steps:

- 1. Data Description
- 2. Input Data Format
- 3. ETL with Pentaho
- 4. Some Visualisation with Python
- 5. Create Schema with Pentaho Schema Workbench
- 6. Use Mondrian

1. Data Description

→ The data set used is sales-data-sample, it has 9994 observations and it contains information about sales transactions, with each row representing a single transaction.

Columns Description:

- 1. **OrderDate:** The date when the order was placed.
- 2. **Category:** The category of the product.
- 3. **City:** The city where the order was placed.
- 4. **Country:** The country where the order was placed.
- 5. **CustomerName:** The name of the customer who placed the order.
- 6. **Discount:** The discount applied to the order.
- 7. **OrderID:** The unique identifier for the order.
- 8. **PostalCode:** The postal code of the location where the order was placed.
- 9. **ProductName:** The name of the product.
- 10. **Profit:** The profit generated from the sale.
- 11. **Quantity:** The quantity of the product ordered.

- 12. **Region:** The region where the order was placed.
- 13. **Sales:** The total sales amount for the transaction.
- 14. **Segment:** The customer segment.
- 15. **ShipDate:** The date when the order was shipped.
- 16. **ShipMode:** The shipping mode for the order.
- 17. **State:** The state where the order was placed.
- 18. **Sub_Category:** The sub-category of the product.
- 19. **DaystoShipActual:** The actual days taken to ship the order.
- 20. **SalesForecast:** Forecasted sales amount.
- 21. **ShipStatus:** The status of the shipment.
- 22. **DaystoShipScheduled:** The scheduled days to ship the order.
- 23. **OrderProfitable:** Whether the order was profitable.
- 24. **SalesperCustomer:** Sales per customer.
- 25. **ProfitRatio:** The ratio of profit to sales.
- 26. **SalesaboveTarget:** Sales above target.
- 27. **latitude:** Latitude of the location.
- 28. **longitude:** Longitude of the location.

2. Input Data Format

To experience creating ETL using data from multiple type of sources I devided the observations into two seperate csv files and loaded one of the data into a postgresql table called **"input-data_1"** and populated it from one of the csv files using a python script.

 \rightarrow The data is relatively clean.

Data description with python:

Data n	oroperties:							
DF1:	nopercies.							
	discount	postalcode	profit	quantity	profitratio	salesabovetarget	latitude	longi
tude count	4893.000000	4893.000000	4893.000000	4893.000000	4893.000000	0.0	4893.000000	4893.00
0000 mean	0.157893	54074.331290	27.733088	3.793174	11.667300	NaN	37.612121	-93.71
8947 std	0.207810	32024.515397	193.483750	2.225195	46.932372	NaN	4.855674	17.90
2021								
min 5200	0.000000	1040.000000	-3702.000000	1.000000	-275.000000	NaN	25.813000	-123.05
25% 1800	0.000000	22153.000000	2.000000	2.000000	7.000000	NaN	34.011600	-115.17
50% 3600	0.200000	54880.000000	9.000000	3.000000	27.000000	NaN	38.744900	-87.63
75% 9700	0.200000	89115.000000	31.000000	5.000000	36.300000	NaN	40.801100	-77.09
max	0.800000	99301.000000	4630.000000	14.000000	50.000000	NaN	47.835300	-68.79
1800								
DF2:								
	discount	postalcode	profit	quantity	 profitratio	salesabovetarget	latitude	longi
tude count	discount 5101.000000	postalcode 5101.000000	profit 5101.000000	quantity 5101.000000	 profitratio	salesabovetarget	latitude 5101.000000	longi 5101.00
tude count 0000 mean								J
tude count 0000 mean 5328 std	5101.000000	5101.000000	5101.000000	5101.000000	5101.000000	0.0	5101.000000	5101.00
tude count 0000 mean 5328	5101.000000 0.154581	5101.000000 56260.919232 32067.883763	5101.000000 29.533229	5101.000000 3.786120	5101.000000	- 0.0 NaN	5101.000000 37.955183	5101.00
tude count 0000 mean 5328 std 8082	5101.000000 0.154581 0.205148	5101.000000 56260.919232 32067.883763	5101.000000 29.533229 267.606838	5101.000000 3.786120 2.225241	5101.000000 12.391806 46.434286	0.0 NaN NaN	5101.000000 37.955183 4.906674	5101.00 -94.97 18.18
tude count 0000 mean 5328 std 8082 min 9800 25% 2900	5101.000000 0.154581 0.205148 0.000000 0.000000	5101.000000 56260.919232 32067.883763 1752.000000 27604.000000	5101.000000 29.533229 267.606838 -6600.000000 2.000000	5101.000000 3.786120 2.225241 1.000000 2.000000	5101.000000 12.391806 46.434286 -275.000000 7.500000	0.0 NaN NaN NaN	5101.000000 37.955183 4.906674 25.476600 34.066000	5101.00 -94.97 18.18 -123.09 -117.97
tude count 0000 mean 5328 std 8082 min 9800 25% 2900 50% 7100	5101.000000 0.154581 0.205148 0.000000 0.000000 0.200000	5101.000000 56260.919232 32067.883763 1752.000000 27604.000000 60505.000000	5101.000000 29.533229 267.606838 -6600.000000 2.000000 9.000000	5101.000000 3.786120 2.225241 1.000000 2.000000 3.0000000	5101.000000 12.391806 46.434286 -275.000000 7.500000 27.000000	0.0 NaN NaN NaN NaN	5101.000000 37.955183 4.906674 25.476600 34.066000 39.205500	5101.00 -94.97 18.18 -123.09 -117.97 -88.29
tude count 0000 mean 5328 std 8082 min 9800 25% 2900 50%	5101.000000 0.154581 0.205148 0.000000 0.000000	5101.000000 56260.919232 32067.883763 1752.000000 27604.000000	5101.000000 29.533229 267.606838 -6600.000000 2.000000	5101.000000 3.786120 2.225241 1.000000 2.000000	5101.000000 12.391806 46.434286 -275.000000 7.500000	0.0 NaN NaN NaN	5101.000000 37.955183 4.906674 25.476600 34.066000	5101.00 -94.97 18.18 -123.09 -117.97
tude count 0000 mean 5328 std 8082 min 9800 25% 2900 50% 7100 75%	5101.000000 0.154581 0.205148 0.000000 0.000000 0.200000	5101.000000 56260.919232 32067.883763 1752.000000 27604.000000 60505.000000	5101.000000 29.533229 267.606838 -6600.000000 2.000000 9.000000	5101.000000 3.786120 2.225241 1.000000 2.000000 3.0000000	5101.000000 12.391806 46.434286 -275.000000 7.500000 27.000000	0.0 NaN NaN NaN NaN	5101.000000 37.955183 4.906674 25.476600 34.066000 39.205500	5101.00 -94.97 18.18 -123.09 -117.97 -88.29

Number of null	values	in	each	column:
DF1:				
orderdate			0	
category			0	
city			0	
country			0	
customername			0	
discount			0	
orderid			0	
postalcode			0	
productname			0	
profit			0	
quantity			0	
region			0	
sales			0	
segment			0	
shipdate			0	
shipmode			0	
state			0	
sub_category			0	
dauetachinaetus	.1		0	

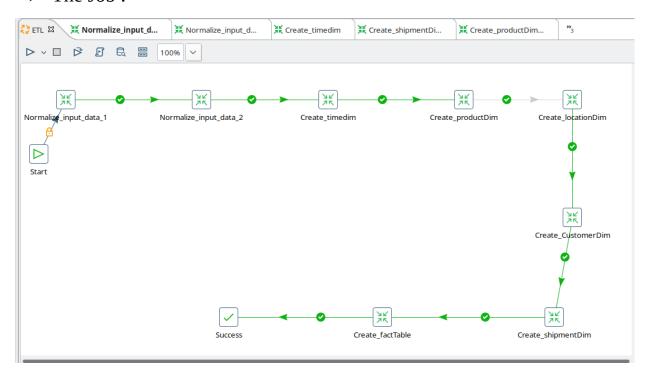
profit	0	
quantity	0	
region	0	
sales	0	
segment	0	
shipdate	0	
shipmode	0	
state	0	
sub_category	0	
daystoshipactual	0	
salesforecast	0	
shipstatus	0	
daystoshipscheduled	0	
orderprofitable	4893	
salespercustomer	0	
profitratio	0	
salesabovetarget	4893	
latitude	0	
longitude	0	
dtype: int64		

			2
DE3.		productname	0
DF2:		profit	0
		quantity	0
orderdate	0	region	0
category	0	sales	0
city	0	segment	0
country	0	shipdate	0
customername	0	shipmode	0
discount	0	state	0
orderid	0	sub_category	0
postalcode	0	daystoshipactual	0
productname	0	salesforecast	0
profit	0	shipstatus	0
quantity	0	daystoshipscheduled	0
region	0	orderprofitable	5101
sales	0	salespercustomer	0
segment	0	profitratio	0
shipdate	ø		
		salesabovetarget	5101
shipmode	0	latitude	0
state	0	longitude	0
sub_category	0	dtype: int64	

→ "orderprofitable" and "salesabovetarget" are almost always null so we can just remove them.

3. ETL with Pentaho

➤ The Job:

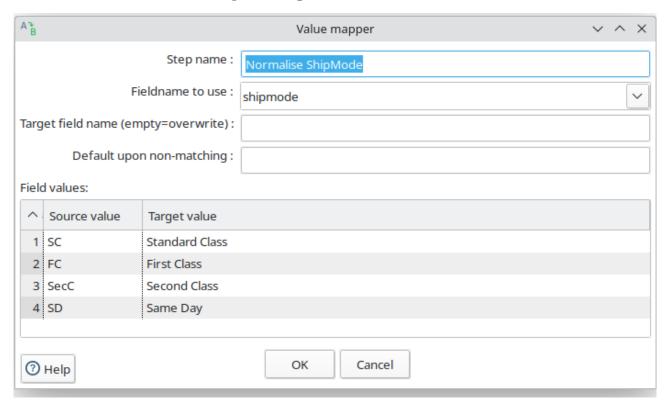


Each Transformation Description :

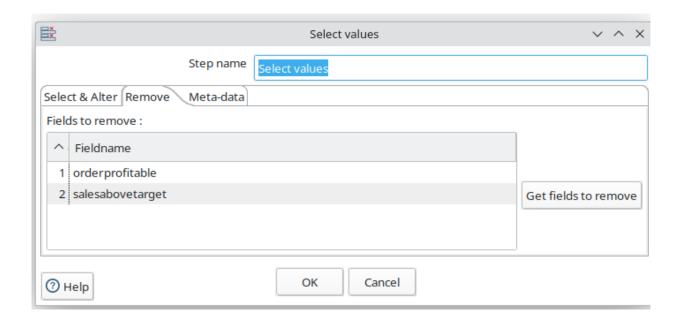
1. Normalize_input_data_1:



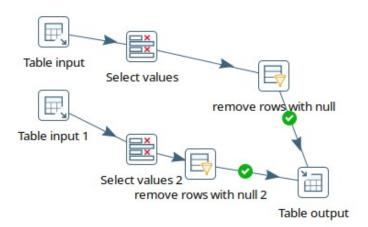
- → The input is the first csv file ('I changed the values of country and shipmode for this file using a shell script to add a step of normalising value in the ETL because the data was clean' ex: sed -i 's/United States/US/g' data_1.csv).
 - So the first step is normalising the country values to match the other input file. Then normalising the shipmode.



 The we would remove the columns "orderprofitable" and "salesabovetarget".



2. Normalize_input_data_2:



x This step is mainly about merging the data from both sources into one table called "input_data"

dwproject=# \d input_data Table "public.input_data"							
Column		Collation	Nullable	Default			
orderdate	text						
category city	text text						
country	text						
customername	text						
discount	double precision						
orderid	text		i				
postalcode	bigint		į	İ			
productname	text	ĺ		ĺ			
profit	bigint			l			
quantity	bigint						
region	text						
sales	bigint						
segment	text		!				
shipdate	text		!				
shipmode	text		!				
state	text		!				
sub_category	text		!				
daystoshipactual salesforecast	bigint						
	bigint text						
shipstatus daystoshipscheduled	bigint						
salespercustomer	double precision		!				
profitratio	double precision						
latitude	double precision						
longitude	double precision		i				
			'	'			

3. Creating fact and dimensions table :

-- Create Dimension Tables

```
CREATE TABLE CustomerDim (
customerid SERIAL PRIMARY KEY,
customername VARCHAR(255),
segment VARCHAR(50),
country VARCHAR(100),
state VARCHAR(100),
city VARCHAR(100),
postalcode VARCHAR(20)
);
```

```
CREATE TABLE ProductDim (
  productid SERIAL PRIMARY KEY,
  productname VARCHAR(255),
  category VARCHAR(100),
  sub_category VARCHAR(100)
);
CREATE TABLE LocationDim (
  locationid SERIAL PRIMARY KEY,
  city VARCHAR(100),
  state VARCHAR(100),
  country VARCHAR(100),
  postalcode VARCHAR(20),
  latitude DECIMAL(9,6),
  longitude DECIMAL(9,6),
  region VARCHAR(100)
);
CREATE TABLE TimeDim (
  timeid SERIAL PRIMARY KEY,
  year INT,
  month INT,
  day INT,
  quarter INT,
  weekday INT
);
CREATE TABLE ShipmentDim (
  shipmentid SERIAL PRIMARY KEY,
  daystoshipscheduled bigint,
  shipstatus text,
  daystoshipactual bigint,
```

```
shipmode text,
  shipdate text
);
→ There are file dimensions: 'customer dimension', 'location dimension',
'product dimension', 'shipment dimension', 'time dimension'.
-- Create Fact Table
CREATE TABLE FactTable (
  FactID SERIAL PRIMARY KEY,
  CustomerID INT,
  ProductID INT,
  LocationID INT,
  TimeID INT,
  ShipmentID INT,
  OrderID TEXT,
  Discount DECIMAL,
  Profit DECIMAL,
  Quantity INT,
  Sales DECIMAL,
  SalesForecast DECIMAL,
  SalesPerCustomer DECIMAL,
  ProfitRatio DECIMAL,
  CONSTRAINT fk_customer FOREIGN KEY (CustomerID) REFERENCES
CustomerDim(CustomerID),
  CONSTRAINT fk_product FOREIGN KEY (ProductID) REFERENCES
ProductDim(ProductID),
  CONSTRAINT fk_location FOREIGN KEY (LocationID) REFERENCES
LocationDim(LocationID),
  CONSTRAINT fk_time FOREIGN KEY (TimeID) REFERENCES timedim(timeid),
  CONSTRAINT fk_shipment FOREIGN KEY (ShipmentID) REFERENCES
ShipmentDim(ShipmentID)
);
```

 \rightarrow The fact table has the foreign keys of all the dimension tables + the other remaining columns.

dwproject=# \d facttable									
Column	Table "public.facttable" Column Type Collation Nullable Default								
Cotumn	Type	Collation	Nullable	Default					
factid	integer		not null	nextval('facttable_factid_seq'::regclass)					
customerid	integer								
productid	integer		i						
locationid	integer		i						
timeid	integer		i						
shipmentid	integer		i						
orderid	text		i	i					
discount	numeric		İ	i e e e e e e e e e e e e e e e e e e e					
profit	numeric	i	İ						
quantity	integer	i	İ						
sales	numeric	ĺ	ĺ	İ					
salesforecast	numeric		İ	İ					
salespercustomer	numeric	ĺ	ĺ	İ					
profitratio	numeric	ĺ	ĺ						
Indexes:									
"facttable_pkey	y" PRIMARY	KEY, btree	(factid)						
Foreign-key constra	Foreign-key constraints:								
"fk_customer"	"fk_customer" FOREIGN KEY (customerid) REFERENCES customerdim(customerid)								
				ES locationdim(locationid)					
				productdim(productid)					
				ES shipmentdim(shipmentid)					
"fk_time" FORE	IGN KEY (t	imeid) REFER	ENCES timed	im(timeid)					

-- Populating the Dimension Tables and Fact Table

→ Using the transformations 'Create_timeDim', 'Create_ProductDim', 'Create_LocationDim', 'Create_ShipmentDim', 'Create_facttable'.

They contain and ExecuteSQLScript each one contain the script to populate a table.

-- Populate Dimension Tables

INSERT INTO public.customerdim (customername, segment, country, state, city, postalcode)
SELECT DISTINCT

```
customername,
segment,
country,
state,
city,
postalcode
```

FROM public.input_data;

```
INSERT INTO public.productdim (productname, category, sub_category)
SELECT DISTINCT
  productname,
  category,
  sub_category
FROM public.input_data;
INSERT INTO public.locationdim (city, state, country, postalcode, latitude, longitude, region)
SELECT DISTINCT
  city,
  state,
  country,
  postalcode,
  latitude,
  longitude,
  region
FROM public.input_data;
INSERT INTO TimeDim (Year, Month, Day, Quarter, Weekday)
SELECT DISTINCT
  DATE_PART('YEAR', CAST(orderdate AS TIMESTAMP)) AS Year,
  DATE_PART('MONTH', CAST(orderdate AS TIMESTAMP)) AS Month,
  DATE_PART('DAY', CAST(orderdate AS TIMESTAMP)) AS Day,
  DATE_PART('QUARTER', CAST(orderdate AS TIMESTAMP)) AS Quarter,
  EXTRACT(ISODOW FROM CAST(orderdate AS TIMESTAMP)) AS Weekday
FROM public.input_data;
INSERT INTO public.shipmentdim (daystoshipscheduled, shipstatus, daystoshipactual, shipmode,
shipdate)
SELECT DISTINCT
  daystoshipscheduled,
  shipstatus,
```

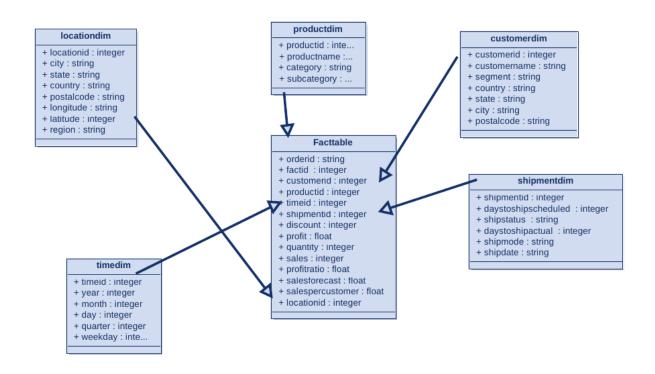
```
daystoshipactual,
  shipmode,
  shipdate
FROM public.input_data;
```

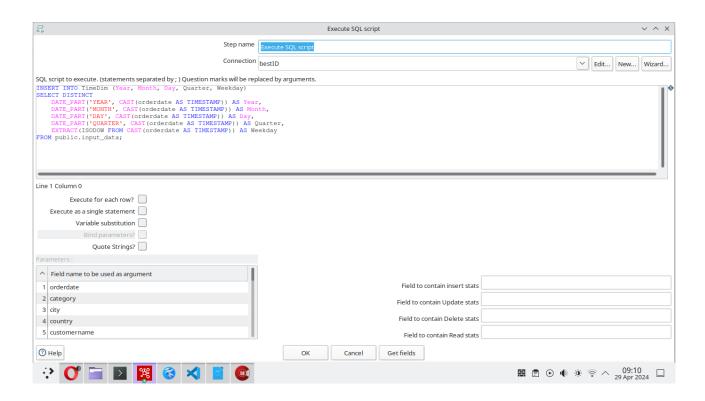
-- Populate Fact Table

INSERT INTO FactTable (CustomerID, ProductID, LocationID, TimeID, ShipmentID, OrderID, Discount, Profit, Quantity, Sales, SalesForecast, SalesPerCustomer, ProfitRatio)

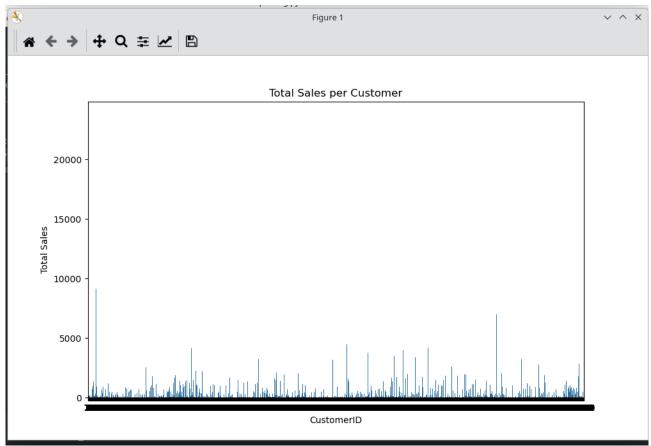
```
SELECT
  cd.CustomerID,
  pd.ProductID,
  ld.LocationID,
  td.TimeID,
  sd.ShipmentID,
  od.OrderID,
  od.discount,
  od.profit,
  od.quantity,
  od.sales,
  od.salesforecast,
  od.salespercustomer,
  od.profitratio
FROM
  public.input_data od
JOIN
  public.customerdim cd
  ON od.customername = cd.customername
  AND od.segment = cd.segment
  AND od.country = cd.country
  AND od.state = cd.state
  AND od.city = cd.city
```

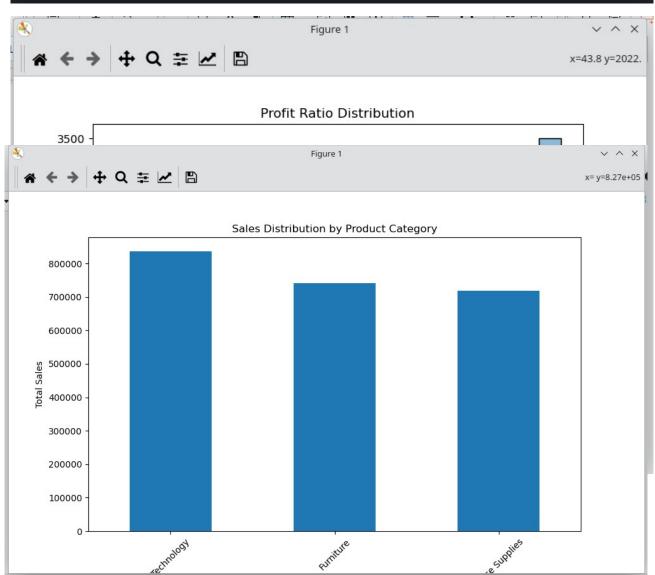
```
AND od.postalcode::text = cd.postalcode
JOIN
  public.productdim pd
  ON od.productname = pd.productname
  AND od.category = pd.category
  AND od.sub_category = pd.sub_category
JOIN public.locationdim ld
  ON od.city = ld.city
  AND od.state = ld.state
  AND od.country = ld.country
  AND od.latitude = ld.latitude
  AND od.longitude = ld.longitude
  AND od.region = ld.region
  AND od.postalcode::text = ld.postalcode
JOIN public.timedim td
  ON DATE_PART('YEAR', CAST(od.orderdate AS TIMESTAMP)) = td.vear
  AND DATE_PART('MONTH', CAST(od.orderdate AS TIMESTAMP)) = td.month
  AND DATE_PART('DAY', CAST(od.orderdate AS TIMESTAMP)) = td.day
  AND DATE_PART('QUARTER', CAST(od.orderdate AS TIMESTAMP)) = td.quarter
  AND EXTRACT(ISODOW FROM CAST(od.orderdate AS TIMESTAMP)) = td.weekday
JOIN public.shipmentdim sd ON od.daystoshipscheduled = sd.daystoshipscheduled
              AND od.daystoshipactual = sd.daystoshipactual
              AND od.shipstatus = sd.shipstatus
              AND od.shipmode = sd.shipmode
              AND DATE(od.shipdate) = sd.shipdate::date;
```

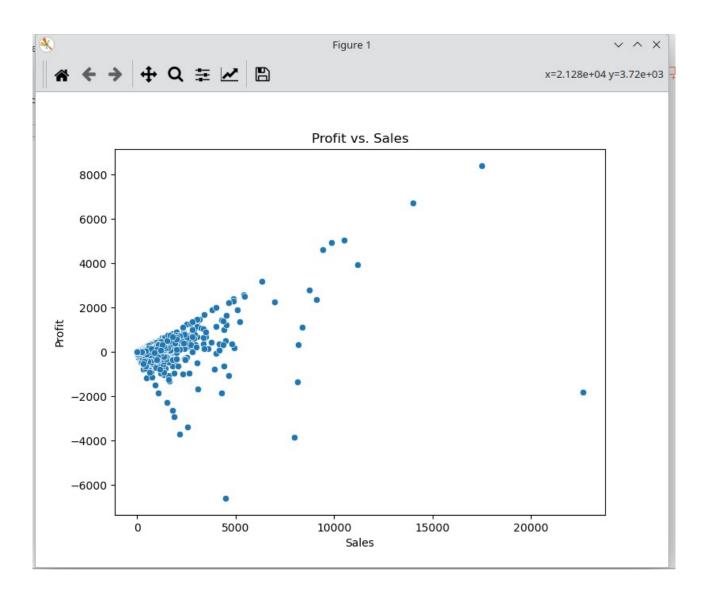




4. Some Visualisation with Python



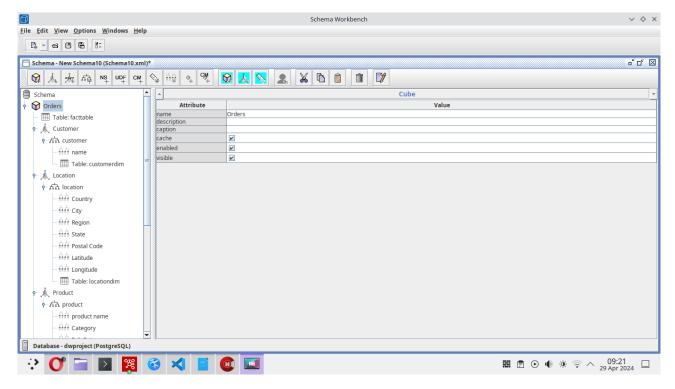


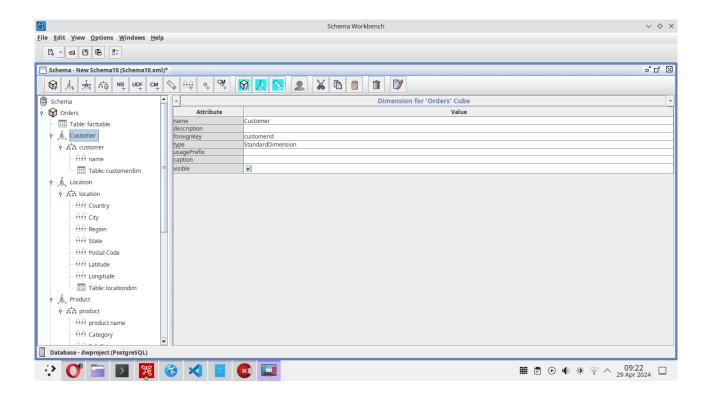


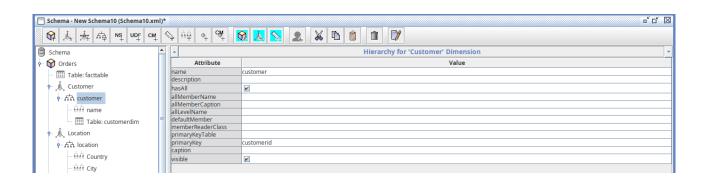
5. Create Schema with Pentaho Schema Workbench

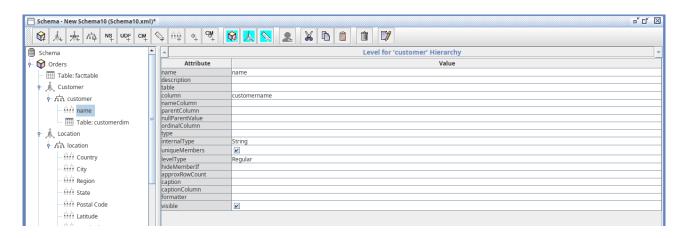
a) Connecting to database:

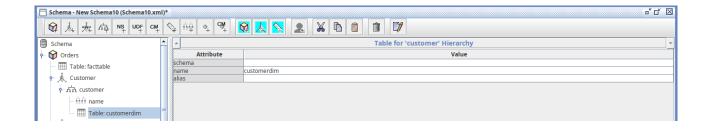
b) Creating Cube and Dimensions:





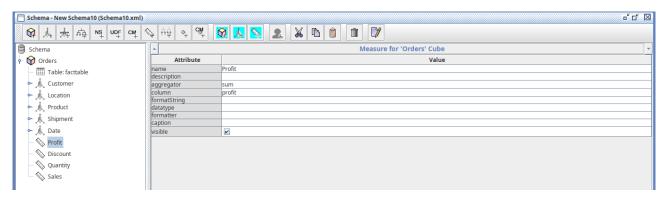




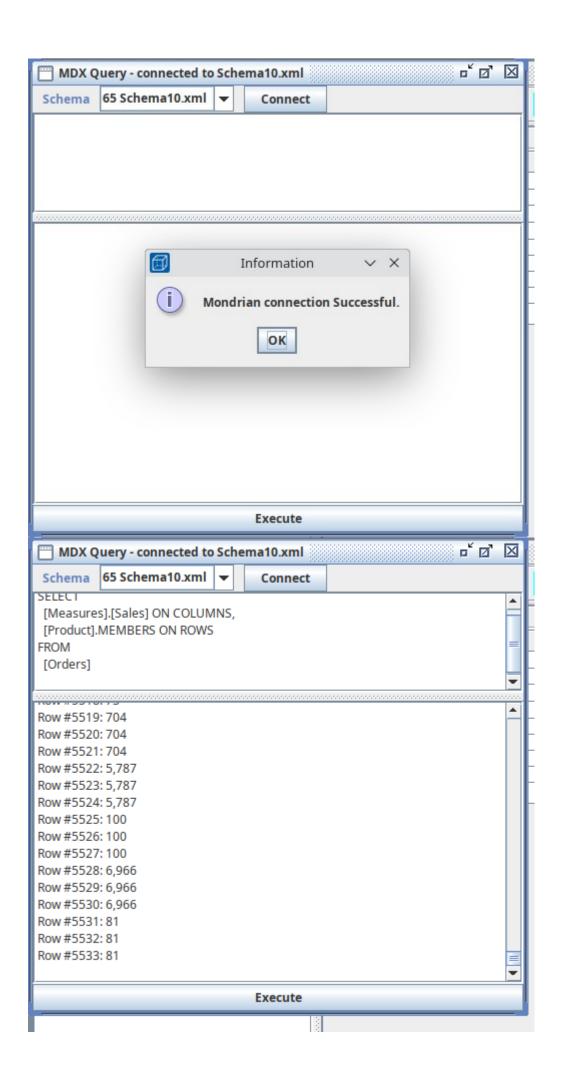


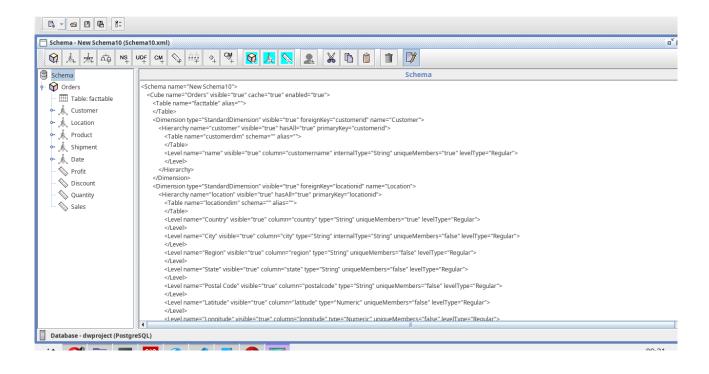
c) Creating Measures:

I created 4 measures: 'Profit', 'Discount', 'Quantity', 'Sales'.



d) Try running MDX query:

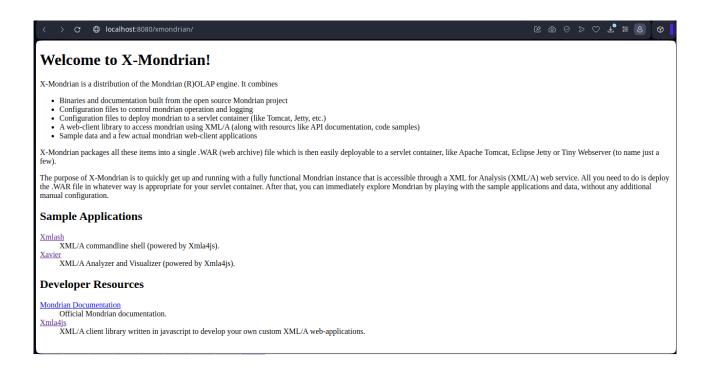




6. Use Mondrian

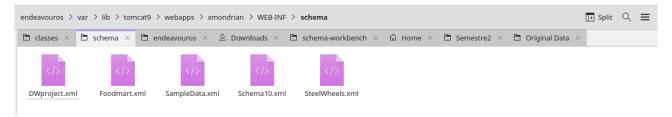
a) Setting up:

To start mondrian you need to download the .war file and put it in the tomcat/webapp directory then you would be able to access mondrian (http://localhost:8080/xmondrian/)



b) Access the schema created:

To be able to acces the schema created you need to add the .xml file to ('/tomcat9/webapps/xmondrian/WEB-INF/schema/').



Change the configuration in the file ('/tomcat9/webapps/xmondrian/WEB-INF/classes/mondrian.properties') to add the configuration for the relational database used.

```
Provider=mondrian;
516 | mondrian.dwproject.jdbcURL=jdbc:postgresql://localhost:5432/dwproject;
mondrian.dwproject.jdbcURs=postgres
mondrian.dwproject.jdbcPassword=12345
mondrian.dwproject.jdbcPassword=12345
mondrian.jdbcDrivers=org.postgresql.Driver
Catalog=file:/home/rania/Documents/CI (University)/CI2/Semestre2/Data Warehouse/Project/project-f/Mondrian/DWproject.xml
```

Configure the data in the

('/tomcat9/webapps/xmondrian/WEB-INF/datasources.xml') file.

c) Result:

To get help about a specific shell command, type HELP <commmand>. Refer to the MDX specification for more information about writing MDX queries.

Check out the tutorial here:

https://github.com/rpbouman/pash/wiki/Pash---The-Pentaho-Analysis-Shell

Hint:

Type SHOW CATALOGS to list the available catalogs.
Then, type USE [<catalog-name>] to select a particular catalog and enter MDX queries.
MDX> SHOW CATALOGS;

CATALOG_NAME	DESCRIPTION	R0LES	DATE_MODIFIED
DWproject	No description available		null

1 rows in set.

Query took: 17 miliseconds; Rendering took: 0 miliseconds. MDX> _

