

# SQL ASSIGNMENT

BY: ANJANA JOSHY (22022447)

## TOPIC: MOBILE SALE DASHBOARD

This assignment is about a mobile sale dashboard that displays customer and mobile details through tables: Customer, Mobile, Mobile sale and stock. I created a database using Python and, used Faker to make the data seem real. I had used some data types, here names is nominal, ratings is ordinal, production date is interval, and prices is ratio.

```
[3] !pip install Faker
import pandas as pd
import numpy as np
from faker import Faker
Fake = Faker()

[2] #setting 1000 rows
n = 1000
#customer : nominal
customer_id = np.random.choice(np.arange(100,99999),size = n, replace = False)
#CUSTOMER NAME : nominal
customer_names = [Fake.name() for i in range(n)]
#Phone number : nominal
country_codes = np.array([f"+{Fake.random_int(1, 99)}" for _ in range(1000)])
phone_number = np.array([f"{code} {Fake.random_number(10)}" for code in country_codes])

# Create DataFrame
df_customer = pd.DataFrame({
    'CustomerID': customer_id,
    'CustomerName': customer_names,
    'Phone Number': phone_number
})

df_customer.set_index('CustomerID', inplace=True)
#SORT WITH CUSTOMER ID
df_customer.sort_index(inplace = True)
#saving as csv file
df_customer.to_csv('customer.csv')
```

This code creates 1000 rows of fake information for each customer, including a unique CustomerID, Customer Name, and Phone Number. Following dataframe sorting by CustomerID, the resultant file is saved as a CSV file called “customer.csv”. The output of the dataframe is shown below.

	CustomerName	Phone Number
CustomerID		
101	Laura Hall	+49 6091584074
155	Edward Russell	+34 2669265862
433	Donna Brown	+50 843041239
438	Casey Harvey	+99 3434543663
553	Brandon Sullivan	+33 6780378570
...	...	...
99225	Yvonne Turner	+84 9165802264
99337	Matthew Payne	+18 8586076830
99623	Katie Johnson	+67 5744120708
99874	Jose Hood	+60 3237927609
99884	Matthew Rivers	+21 3442356204

1000 rows × 2 columns

```
[ ] #mobile id : nominal
mobile_id = np.random.choice(np.arange(100,9999),size = n,replace = False)
# List of specific mobile models :nominal
mobile_models = ['Galaxy S21', 'Galaxy Note 20', 'Galaxy A52', 'Galaxy Z Fold 3', 'Galaxy M32', 'Galaxy XCover Pro', 'Galaxy Z Flip 3', 'Galaxy A32', 'Galaxy S20', 'Galaxy A71']
# Repeat the models to create 1000 rows
repeated_models = np.random.choice(mobile_models, size=1000)

#colors : nominal
colors = ['Bora Purple', 'Phantom White', 'Green', 'Pink Gold', 'Phantom Black']
# Repeat the models to create 1000 rows
phone_color= np.random.choice(colors, size=1000)

#production date : interval
production_year = np.random.randint(2000,2023,n)
production_month = np.random.randint(1,13,n)
production_day = np.random.randint(1,29,n)
production_date = [f'{production_year[i]}-{str(production_month[i]).zfill(2)}-{str(production_day[i]).zfill(2)}' for i in range(n)]
# storage : ordinal
#selecting storages
storage = ['64', '128', '256', '512' ]
# Repeat the models to create 1000 rows
storage= np.random.choice(storage, size=1000)

#setting dataframe for storage
df_mobile = pd.DataFrame({
    'storage':storage
})

random_addition = np.random.uniform(1, 2, n)
#price : ratio
price = df_mobile['storage'].astype(int) * 3 + random_addition
```

```
▶ #dataframe for table mobile
df_mobile = pd.DataFrame({
    'mobileID': mobile_id,
    'models': repeated_models,
    'color': phone_color,
    'production date': production_date,
    'storage':storage,
    'price': price
})

df_mobile['price'] = df_mobile['price'].round(2)
df_mobile.set_index('mobileID', inplace=True)

#saving as csv file
df_mobile.to_csv('mobile.csv')
```

This code generates a dataset of mobile, which contains the columns models, colors, production dates, storage capacities, and unique mobile IDs. Each mobile is assigned a model, color, and ID at random. The date of production is generated at random using the format “YYYY-MM-DD”. Random choices are made regarding the storage options (64GB, 128GB, 256GB, and 512GB), and prices are calculated based on storage with some random variations. After that, the data is saved as a CSV file called “mobile.csv” and the result is shown below.

	models	color	production date	storage	price
mobileID					
9893	Galaxy A71	Pink Gold	2015-04-21	64	193.81
7748	Galaxy A52	Phantom Black	2013-08-17	512	1537.56
2884	Galaxy XCover Pro	Phantom White	2004-02-20	512	1537.10
8519	Galaxy A52	Phantom Black	2012-02-03	256	769.63
2460	Galaxy M32	Green	2021-12-28	512	1537.45
...	...	...	...	...	...
6630	Galaxy Z Flip 3	Bora Purple	2017-10-27	512	1537.25
820	Galaxy S21	Green	2000-11-26	512	1537.83
1704	Galaxy Note 20	Bora Purple	2007-09-20	256	769.48
5414	Galaxy S20	Bora Purple	2011-09-15	512	1537.02
9327	Galaxy M32	Phantom Black	2005-04-05	256	769.69

1000 rows × 5 columns

```

✓ [13] #SETTING Stock ID: nominal
stockid = np.random.choice(np.arange(100,9999),size = n, replace = False)
#quantity available: ratio
quantity_available = np.random.randint(0,999,size = n)
#dataframe for table stock
df_stock = pd.DataFrame({
    'stock id': stockid,
    'quantity available':quantity_available,
    'models': repeated_models,
})
df_stock.set_index('stock id', inplace=True)
#saving as csv file
df_stock.to_csv('stock.csv')
df_stock

```

	quantity available	models
stock id		
4807	218	Galaxy A71
6010	62	Galaxy A52
1631	915	Galaxy XCover Pro
9049	841	Galaxy A52
4925	2	Galaxy M32
...	...	...
7024	337	Galaxy Z Flip 3
9280	658	Galaxy S21

This code generates a dataset that represents stock items. To ensure uniqueness, each stock item is given a unique stock ID, which is generated at random. To represent the amount of each item in stock, random numbers between 0 and 999 are entered into the ‘Quantity Available’ column. Following, the data is placed into a dataframe called “df\_stock”, with the stock ID acting as the index. ‘stock.csv’ is the name of a CSV file that contains this dataframe and the output is shown above.

```

[14] #order id: nominal
order_id = np.random.choice(np.arange(100,9999),size = n,replace = False)

#sales channel:nominal
sales_channels = np.random.choice(['Offline', 'Online'], size=n)
# Generate random ratings:ordinal
rating = np.random.randint(2, 11, size=n).astype(float)
# Set some values to NaN
n_null = 150
random_indices = np.random.choice(rating.size, n_null, replace=True).astype(int)
rating[random_indices] = np.nan

#setting dataframe for table Mobilesale
df_Mobilesale = pd.DataFrame({
    'mobileID': mobile_id,
    'order id': order_id,
    'sales channels': sales_channels,
    'rating': rating,
    'customer id': customer_id
})
# Replace NaN with a string representation
df_Mobilesale['rating'] = df_Mobilesale['rating'].fillna('NaN')

df_Mobilesale['CompoundKey'] = df_Mobilesale['order id'].astype(str) + '_' + df_Mobilesale['mobileID'].astype(str)
df_Mobilesale.set_index('order id', inplace=True)
df_Mobilesale.sort_index(inplace = True)

#saving as csv file
df_Mobilesale.to_csv('mobilesale.csv')
df_Mobilesale

```

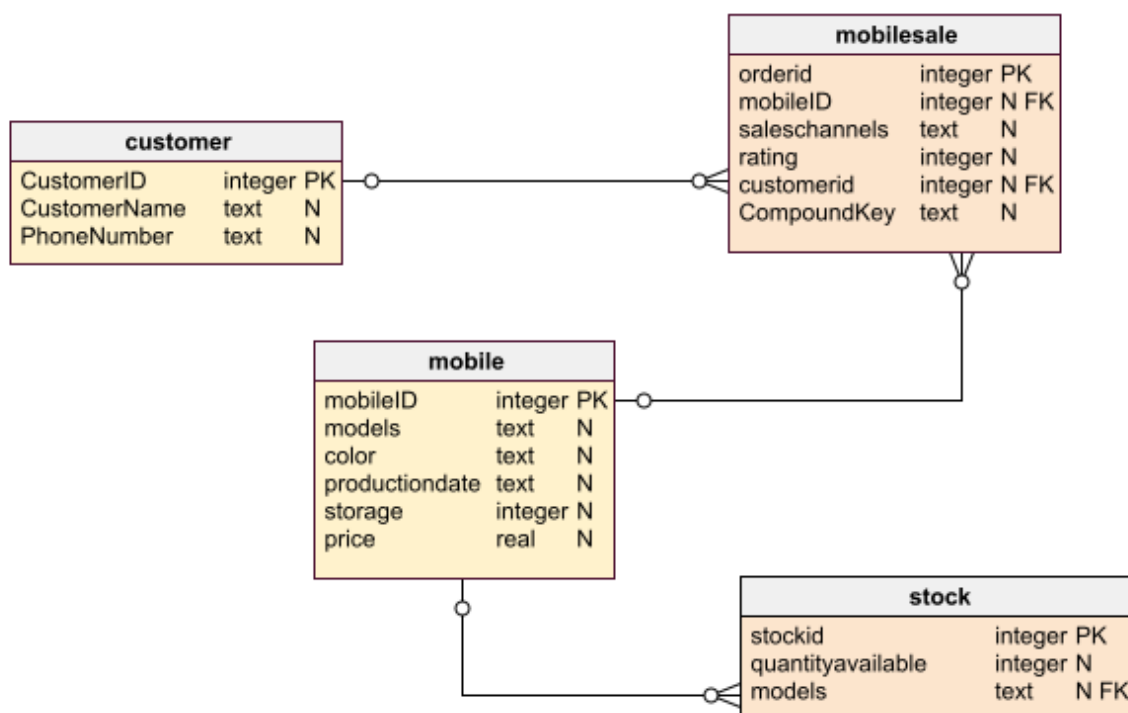
This code creates a dataset for sales on mobile devices and gives each transaction a distinct order ID. The sales channels are assigned at random and indicate if the sale was made “Offline” or “online”. In order to

imitate missing data, some ratings are purposefully set to NaN, in an ordinal scale ranging from 2 to 10. The data that is produced is arranged into a dataframe called “df\_Mobilescale”, which also has Customer IDs and Mobile IDs. A compound key that combines Order IDs and Mobile IDs is created to demonstrate the relationship between order and mobile. The dataframe is saved as a CSV file called “mobile.csv” and the output is given below.

	mobileID	sales channels	rating	customer id	CompoundKey
order id					
103	8649	Online	2.0	30794	103_8649
112	4923	Offline	3.0	76513	112_4923
114	800	Offline	2.0	93068	114_800
117	6512	Offline	4.0	47760	117_6512
119	2917	Offline	NaN	8592	119_2917
...	...	...	...	...	...
9930	217	Offline	4.0	97748	9930_217
9962	5814	Online	4.0	29285	9962_5814
9974	6372	Online	NaN	58664	9974_6372
9976	9296	Offline	3.0	40612	9976_9296
9985	9384	Offline	5.0	21696	9985_9384

1000 rows x 5 columns

The schema establishes the relationship between tables and their structure, ensuring a methodical approach to information management. The customer, mobile, mobilesale, and stock tables in the database schema provide a structured framework for organizing and evaluating mobile sales data. The relation connecting the four tables is displayed in the database schema provided below. Examine the relationship between the customer and mobile sale tables. In mobile sale, the customer id serves as a foreign key that displays the specifics of the mobile sale. Furthermore, models serve as a foreign key in the stock table, allowing us to access the models' stock details.



```

1 SELECT models, rating
2 FROM mobile
3 INNER JOIN mobilesale
4 ON mobile.mobileID = mobilesale.mobileID
5 WHERE rating > 7 AND productiondate BETWEEN 2020 AND 2022
6 ORDER BY rating DESC
7 LIMIT 4;
8

```

	models	rating
1	Galaxy Note 20	10
2	Galaxy XCover Pro	10
3	Galaxy A52	10
4	Galaxy S21	8

The query above selects the models which have ratings above 7 and years between 2020 and 2022.

```

1 SELECT customer.CustomerName, customer.customerid, customer.PhoneNumber, mobilesale.saleschannels
2 FROM customer
3 JOIN mobilesale
4 ON customer.customerid = mobilesale.customerid
5 WHERE saleschannels = 'Online'
6 ORDER BY customer.customerid
7 LIMIT 3

```

	CustomerName	CustomerID	PhoneNumber	saleschannels
1	Mrs. Pamela Wright	27990	+40 5560828985	Online
2	Kyle Brown	61416	+1 5912530705	Online
3	Meredith Stewart	85489	+97 1178740457	Online

The query above selects all the customer details who are purchased online.

```

1 SELECT models, SUM(price) AS total_revenue
2 FROM mobile
3 GROUP BY models;

```

	models	total_revenue
1	Galaxy A32	85221.44
2	Galaxy A52	79849.19
3	Galaxy A71	70793.53
4	Galaxy M32	67929.22
5	Galaxy Note 20	69462.63

The query above calculates the total sum of prices of each model and represents as total revenue.

```

1 SELECT *
2 FROM mobilesale
3 WHERE mobilesale.CompoundKey = "136_1782"

```

	orderid	mobileID	saleschannels	rating	customerid	CompoundKey
1	136	1782	Online	4	22497	136_1782

The query above is designed to retrieve all the information related to the sale identified by the compound key"136\_1782" from the mobilesale table.