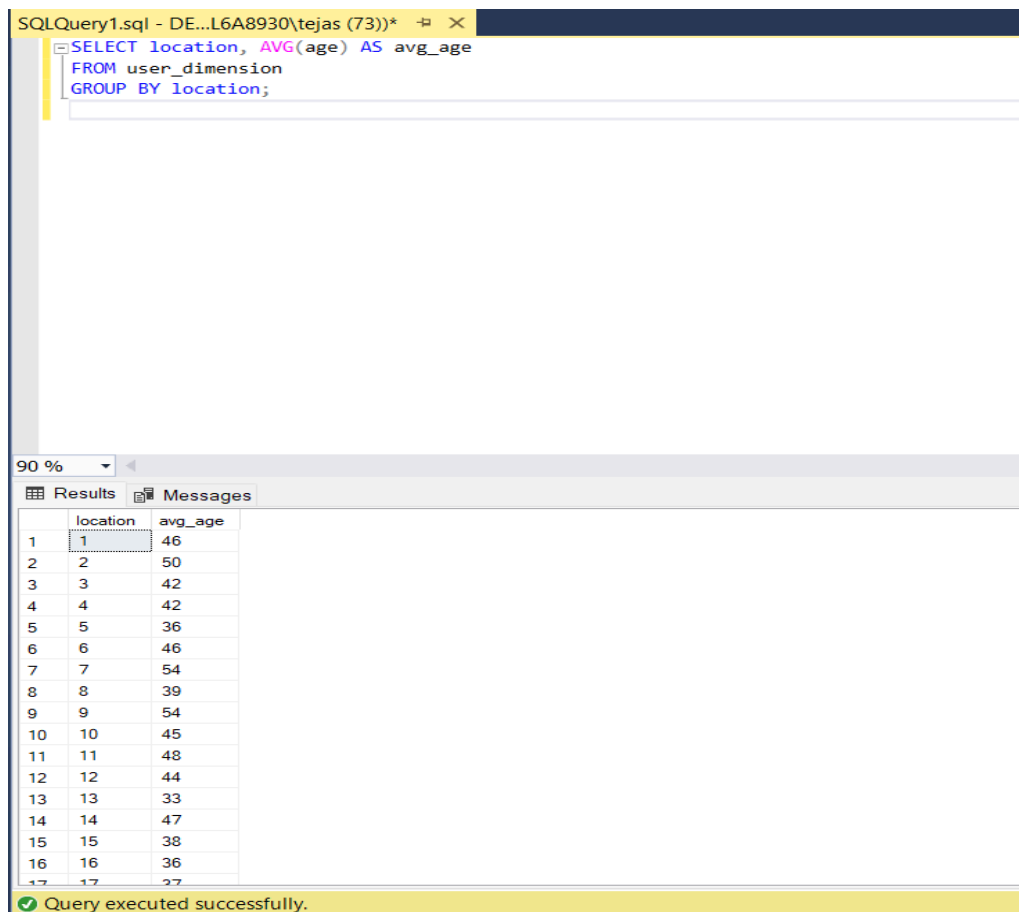


Aim - Perform the OLAP operations on given case study

1. Introduction Online Analytical Processing (OLAP) is a category of data processing that enables the analysis of multidimensional data from multiple perspectives. It is a powerful tool in data warehousing and business intelligence that allows users to perform complex queries and gain insights into the data. OLAP operations are performed on data cubes, which represent data along various dimensions, enabling flexible data analysis.

2. OLAP Operations OLAP provides several operations that allow users to explore and analyze data. The main OLAP operations include:

- **Roll-Up:** This operation aggregates data along a dimension, climbing up a hierarchy of dimensions. For example, data can be aggregated from the individual user level to a location level, summarizing the data by location. Roll-up is useful for generating high-level summaries and identifying trends across different levels of aggregation.



The screenshot shows a SQL query window with the following query:

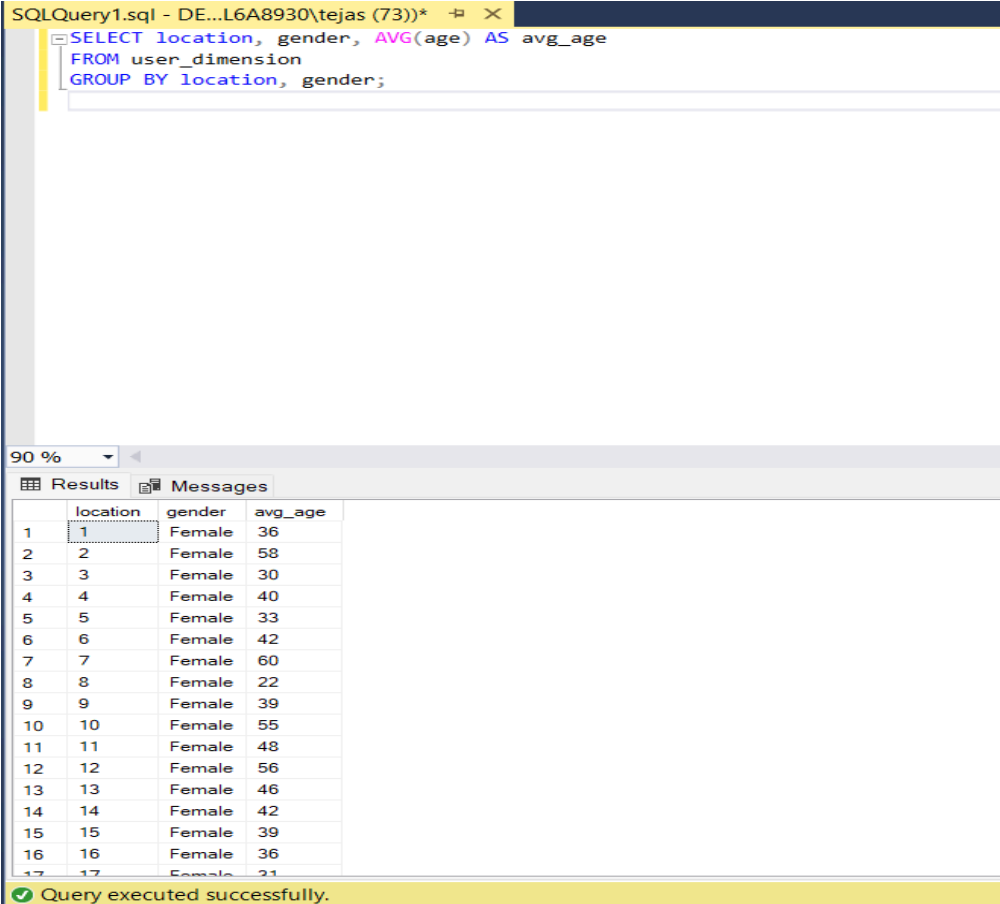
```
SELECT location, AVG(age) AS avg_age
FROM user_dimension
GROUP BY location;
```

The query results are displayed in a table with two columns: 'location' and 'avg_age'. The results are as follows:

	location	avg_age
1	1	46
2	2	50
3	3	42
4	4	42
5	5	36
6	6	46
7	7	54
8	8	39
9	9	54
10	10	45
11	11	48
12	12	44
13	13	33
14	14	47
15	15	38
16	16	36
17	17	27

A status bar at the bottom indicates: "Query executed successfully."

- **Drill-Down:** The drill-down operation is the reverse of roll-up, where data is broken down into more detailed levels. This operation provides finer granularity by moving down a dimension hierarchy. For instance, after aggregating data by location, a drill-down operation can break down the data by age or gender within each location, providing more detailed insights.



SQLQuery1.sql - DE...L6A8930\tejas (73))*

```
SELECT location, gender, AVG(age) AS avg_age
FROM user_dimension
GROUP BY location, gender;
```

90 %

Results Messages

	location	gender	avg_age
1	1	Female	36
2	2	Female	58
3	3	Female	30
4	4	Female	40
5	5	Female	33
6	6	Female	42
7	7	Female	60
8	8	Female	22
9	9	Female	39
10	10	Female	55
11	11	Female	48
12	12	Female	56
13	13	Female	46
14	14	Female	42
15	15	Female	39
16	16	Female	36
17	17	Female	21

Query executed successfully.

- **Slice:** The slice operation allows users to filter the data cube along one dimension, extracting a specific subset of data. For example, slicing the data to show users who are 30 years old provides a focused view on that particular age group, enabling analysis of that specific segment.

SQLQuery1.sql - DE...L6A8930\tejas (73))*

```

SELECT *
FROM user_dimension
WHERE age = 30;

```

90 %

Results Messages

	user_id	username	full_name	location	age	gender
1	86	nicole07	Brett Thompson	23	30	Male
2	215	philipbrooks	Kenneth Ryan	2	30	Other
3	234	shannon28	Wendy Armstrong	41	30	Other
4	260	jeremypitts	Kim Dennis	25	30	Female
5	267	michael43	Dr. Christopher Tucker II	34	30	Female
6	287	bcoleman	Rachel Murillo	50	30	Female
7	299	nmccarthy	Christina Gonzalez	36	30	Male
8	327	james48	William Newman	5	30	Female
9	329	brownjeremy	Juan Allen	45	30	Male
10	349	debra49	Juan Roberts MD	22	30	Other
11	356	zromero	Mr. Nathan Garza	43	30	Male
12	402	simssharon	Mark Brown	33	30	Female
13	435	santoshdakota	Courtney Francis	15	30	Male
14	445	zroberts	Angela Cameron MD	29	30	Male
15	475	danielscott	Luis Herrera	24	30	Other

Query executed successfully.

- Dice:** The dice operation is an extension of the slice operation, applying multiple filters to create a sub-cube. For instance, dicing the data to show male users between the ages of 20 and 40 filters the data by both gender and age, allowing analysis of a targeted group.

SQLQuery1.sql - DE...L6A8930\tejas (73))*

```

SELECT *
FROM user_dimension
WHERE age BETWEEN 20 AND 40
AND gender = 'Male';

```

90 %

Results Messages

	user_id	username	full_name	location	age	gender
1	9	nmiller	Margaret Wilson	30	40	Male
2	11	williammccarty	Amy Reeves	34	22	Male
3	14	gsilva	Mr. Robert Horne	34	37	Male
4	31	lindsey37	Kyle Martinez	25	27	Male
5	33	taylorlkatie	Daniel Hines	16	35	Male
6	50	emueller	Timothy Gomez	10	28	Male
7	52	williamwebb	Christine Mills	36	26	Male
8	58	robertmason	Robert Gonzalez	44	31	Male
9	60	mitchellkatherine	Erica Werner	43	40	Male
10	61	andersonmorgan	Erica Allen	8	39	Male
11	71	carterlee	Jon Lee	15	26	Male
12	77	xdouglas	Michelle Ward	3	31	Male
13	82	swalker	Jack Kim	23	29	Male
14	86	nicole07	Brett Thompson	23	30	Male
15	88	ryan66	Abigail Patel	46	36	Male
16	100	ryandavila	Lydia Wang	5	33	Male
17	102	44	25	Male

Query executed successfully.

- **Pivot (Rotate):** The pivot operation involves rotating the data cube to view it from different perspectives. It is used to reorganize the dimensions of a data cube, providing alternative views of the data. For example, pivoting the data to show the count of users by location and gender helps to compare user distribution across different locations and genders.

SQLQuery1.sql - DE...L6A8930\tejas (73))*

```

SELECT location,
       SUM(CASE WHEN gender = 'Male' THEN 1 ELSE 0 END) AS male_count,
       SUM(CASE WHEN gender = 'Female' THEN 1 ELSE 0 END) AS female_count,
       SUM(CASE WHEN gender = 'Other' THEN 1 ELSE 0 END) AS other_count
FROM user_dimension
GROUP BY location;

```

90 %

Results Messages

	location	male_count	female_count	other_count
1	1	3	1	2
2	2	2	4	4
3	3	3	3	2
4	4	3	2	3
5	5	2	3	5
6	6	2	1	1
7	7	1	2	1
8	8	4	3	4
9	9	5	2	3
10	10	4	4	4
11	11	5	7	1
12	12	4	4	5
13	13	4	1	2
14	14	5	3	3
15	15	5	3	3
16	16	6	4	1
17	17	4	2	2

Query executed successfully.

3. Case Study: User Dimension Analysis In this case study, we have a `user_dimension` table that contains information about users, including their ID, username, full name, location, age, and gender. We will perform OLAP operations on this data to extract insights and analyze user demographics.

- **Roll-Up Example:** We might aggregate the data by location to determine the average age of users in each location. This can help identify which locations have younger or older user populations.
- **Drill-Down Example:** After aggregating by location, we can drill down by gender within each location to see if there are significant differences in age based on gender across locations.
- **Slice Example:** Slicing the data to view users who are exactly 30 years old allows us to focus on a specific age group and understand their distribution across locations and genders.

- **Dice Example:** Dicing the data to view male users between the ages of 20 and 40 helps to analyze the characteristics of this particular demographic segment.
- **Pivot Example:** Pivoting the data to show the count of users by location and gender provides a cross-tabulation that helps compare user distribution across different locations and genders.

4. Conclusion OLAP operations are essential for analyzing multidimensional data in a flexible and dynamic manner. By performing roll-up, drill-down, slice, dice, and pivot operations, users can gain deeper insights into the data and make informed decisions based on the analysis. In the context of the `user_dimension` table, these operations allow us to explore user demographics and uncover patterns and trends that might not be immediately apparent from the raw data.