



ASSIGNMENT

~Data and Cloud Analytics internship~

MUŞAT ANDRA

CUPRINS

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A) Provision an Azure SQL Database

The screenshot shows the Microsoft Azure portal interface. At the top, there's a search bar and navigation icons. Below the header, the 'Azure services' section displays icons for various services like 'Create a resource', 'Azure SQL', 'Resource groups', etc. The 'Resources' section is active, showing a table of recent resources.

Name	Type	Last Viewed
andra-internship-srv	SQL server	a few seconds ago
InternshipResources-rg	Resource group	a minute ago
andra-internship-db (andra-internship-srv/andra-internship-db)	SQL database	a minute ago
andra-covid-reporting-adf	Data factory (V2)	2 weeks ago
andracovidreportingsa	Storage account	3 weeks ago
andracovidreportingsa-d486342a-9c39-493f-a512-f847b1388378	Event Grid System Topic	3 weeks ago
andra-covid-db	SQL database	3 weeks ago
andracovidreportingdl	Storage account	3 weeks ago
covid-reporting-ws	Log Analytics workspace	3 weeks ago
andratesticddemodl	Storage account	3 weeks ago

The screenshot shows the configuration page for the 'andra-internship-srv' SQL server. The left sidebar contains navigation links like 'Overview', 'Activity log', 'Access control (IAM)', etc. The main content area is divided into 'Essentials' and 'Features' sections.

Essentials:

- Resource group (move): InternshipResources-rg
- Status: Available
- Location: East US
- Subscription (move): Azure for Students
- Subscription ID: f9acd0bd-c6a3-4a38-8c3f-aea5e6b53cff
- Tags (edit): Add tags
- Server admin: andra
- Networking: Show networking settings
- Microsoft Entra admin: Not configured
- Server name: andra-internship-srv.database.windows.net

Features (6):

- Microsoft Entra admin:** Allows you to centrally manage identity and access to your Azure SQL databases. NOT CONFIGURED
- Microsoft Defender for SQL:** Vulnerability Assessment and Advanced Threat Protection. NOT CONFIGURED
- Automatic tuning:** Monitors and tunes your database automatically to optimize performance. CONFIGURED
- Auditing:** Track database events and writes them to an audit log in Azure storage. NOT CONFIGURED
- Failover groups:** Automatically manages replication, connectivity and failover for a set of databases. NOT CONFIGURED
- Transparent data encryption:** Encryption at rest for your databases, backups, and logs. SERVICE-MANAGED KEY

Microsoft Azure Search resources, services, and docs (0+)

Home > **andra-internship-db (andra-internship-srv/andra-internship-db)** SQL database

Search Copy Restore Export Set server firewall Delete Connect with... Feedback

Overview Activity log Tags Diagnose and solve problems Query editor (preview) Settings Data management Integrations Power Platform Security Intelligent Performance Monitoring Automation Help

Essentials Resource group (move) : [InternshipResources-rg](#) Status : Online Location : East US Subscription (move) : [Azure for Students](#) Subscription ID : f9acd0bd-c6a3-4a38-8c3f-aea5e6b53cff Tags (edit) : [Add tags](#)

Server name : [andra-internship-srv.database.windows.net](#) Connection strings : [Show database connection strings](#) Pricing tier : General Purpose - Serverless: Gen5, 1 vCore Auto-pause delay : [1 hour](#) Earliest restore point : 2024-05-16 13:31 UTC

[Getting started](#) Monitoring Properties Features Notifications (0) Integrations Tutorials

Start working with your database
Connect to your database and start working with data with a few simple steps. [Learn more](#)

Configure access
Configure network access to your SQL server. [Learn more](#)

Connect to application
Use connection strings to connect to your SQL database from your applications and favorite tools.

Start developing
Work in your database by using tools to add, modify and query data. [Compare tools](#)

Activate Windows
Go to Settings to activate Windows.

Microsoft Azure Search resources, services, and docs (0+)

Home > Azure SQL > andra-internship-db (andra-internship-srv/andra-internship-db)

Azure SQL Academia de Studii Economice din Bucuresti (aser... Create Reservations ...

Filter for any field... Name ↑

andra-covid-db (andra-covid-srv/andra-... andra-covid-srv andra-internship-db (andra-internship-... andra-internship-srv

andra-internship-db (andra-internship-srv/andra-internship-db) | Query editor (preview) SQL database

Search Login New Query Open query Feedback Getting started

Overview Activity log Tags Diagnose and solve problems Query editor (preview) Settings Data management Integrations Power Platform Security Intelligent Performance Monitoring Automation Help

Showing limited object explorer here. For full capability please click here to open Azure Data Studio.

Tables
dbo.flowers
flower_id (PK, int, not null)
flower_name (varchar, not null)
color (varchar, null)
bloom_season (varchar, null)

Views
Stored Procedures

Query 1
Run Cancel query Save query Export data as Show only Editor

```
1 CREATE TABLE flowers (
2   flower_id INT PRIMARY KEY,
3   flower_name VARCHAR(50) NOT NULL,
4   color VARCHAR(25),
5   bloom_season VARCHAR(35)
6 );
```

Results Messages

Query succeeded: Affected rows: 0

Activate Windows
Go to Settings to activate Windows.

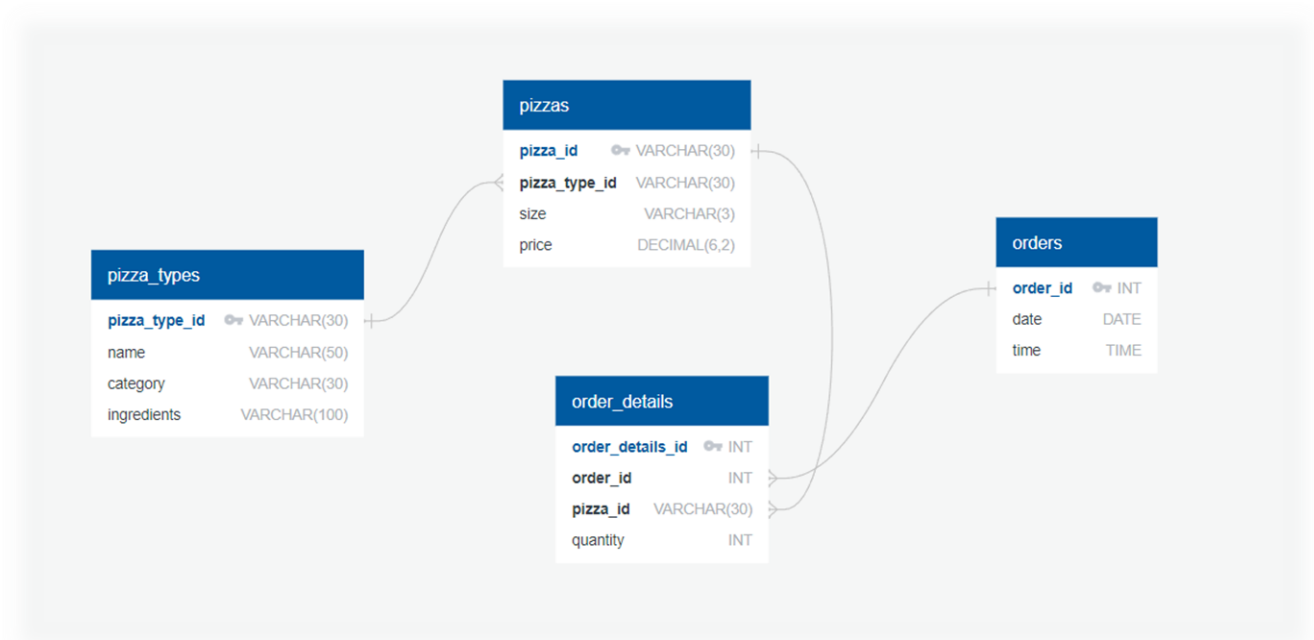
Query succeeded 1.0s

As a student account holder, I took the following steps:

- * Before provisioning any resources, I created a new resource group to organize my resources logically. I named the resource group "InternshipResources-rg" and selected the East US location;
- * Provisioning a single SQL database: From the search bar in Azure, I looked for Azure SQL, clicked on SQL databases, and left the Resource type set to Single database, then selected Create. I selected my student subscription and proceeded to give my database a name, "andra_internship-db". For Server, I selected Create new and filled out the New server form with the following values: Server name: "andra-internship-srv", Location: East US, Authentication method: Use SQL authentication, created a Server admin login and password below. After that, I selected the Workload Environment to be Development and clicked on Configure Database. I selected the Service Tier to be General Purpose and Compute tier to be Serverless, and applied the changes, ensuring the SQL elastic pool is set to No and Backup storage redundancy option is locally redundant storage. Next, I clicked on Networking and set the Connectivity method to Public endpoint. For Firewall rules, I set Add current client IP address to Yes, and left Allow Azure services and resources to access this server set to No. Then, I clicked on security to proceed and kept Enable Microsoft Defender for SQL set to No. Next, I clicked on additional settings and kept Use existing data set to None to create an empty database. Finally, I clicked on Review and Create and then on Create.
- * Waited for Deployment: Azure deployed my SQL Server instance and database. This process took a few minutes. I monitored the deployment progress from the notification area in the upper right area of the portal;
- * Verified Deployment: Once the deployment was complete, I visited the Azure SQL area again to ensure that both the SQL Server instance and the database were provisioned successfully and were in a running state;
- * As an additional task, I went to my newly created database, navigated to the Query editor, and created a simple table (flowers table) using an SQL query;
- * At the end, I took screenshots.

B) In debt analysis of a dataset

1. I used QuickDBD



2. I expect the type of relationship between tables to be:

- * The relationship between "pizzas" and "pizza_types" is a one to many relationship. Each pizza belongs to only one pizza type, but each pizza type can have multiple pizzas associated with it.
- * The relationship between "pizzas" and "order_details" is a one to many relationship. Each pizza can have multiple order details associated with it, but each order detail is related to only one pizza.
- * The relationship between "orders" and "order_details" is a one to many relationship. Each order can have multiple order details associated with it, but each order detail is related to only one order.

3. For this requirement:

- * I split the sheets with tables from the given Excel file and saved them in CSV format. Then, using MS SQL from the online SQL server, I imported the CSV files with tables, and for each column in the tables, I adjusted the data type and added the necessary constraints.
- * The code:

```
SELECT * FROM pizza_types;
```

```
ALTER TABLE pizza_types
ALTER COLUMN pizza_type_id VARCHAR(30) NOT NULL;
ALTER TABLE pizza_types
ALTER COLUMN name VARCHAR(50);
ALTER TABLE pizza_types
ALTER COLUMN category VARCHAR(30);
ALTER TABLE pizza_types
ALTER COLUMN ingredients VARCHAR(100);
ALTER TABLE pizza_types
ADD CONSTRAINT pk_pizza_type_id PRIMARY KEY (pizza_type_id);
```

```
SELECT * FROM pizzas;
```

```
ALTER TABLE pizzas
ALTER COLUMN pizza_id VARCHAR(30) NOT NULL;
ALTER TABLE pizzas
ALTER COLUMN pizza_type_id VARCHAR(30);
ALTER TABLE pizzas
ALTER COLUMN size VARCHAR(3);
ALTER TABLE pizzas
ALTER COLUMN price DECIMAL(6,2);
ALTER TABLE pizzas
ADD CONSTRAINT pk_pizza_id PRIMARY KEY (pizza_id);
ALTER TABLE pizzas
add CONSTRAINT fk_pizza_type_id FOREIGN KEY(pizza_type_id) REFERENCES
pizza_types(pizza_type_id);
```

```
SELECT TOP 100 * FROM orders;
```

```
ALTER TABLE orders
ALTER COLUMN order_id INT NOT NULL;
ALTER TABLE orders
ALTER COLUMN date DATE;
ALTER TABLE orders
ALTER COLUMN time TIME;
alter TABLE orders
```

```
ADD CONSTRAINT pk_order_id PRIMARY KEY (order_id);
```

```
SELECT TOP 100 * FROM order_details;
```

```
ALTER TABLE order_details
```

```
ALTER COLUMN order_details_id INT NOT NULL;
```

```
ALTER TABLE order_details
```

```
ALTER COLUMN order_id INT NOT NULL;
```

```
ALTER TABLE order_details
```

```
ALTER COLUMN pizza_id VARCHAR(30);
```

```
ALTER TABLE order_details
```

```
ALTER COLUMN quantity INT;
```

```
ALTER TABLE order_details
```

```
ADD CONSTRAINT pk_order_details_id PRIMARY KEY (order_details_id);
```

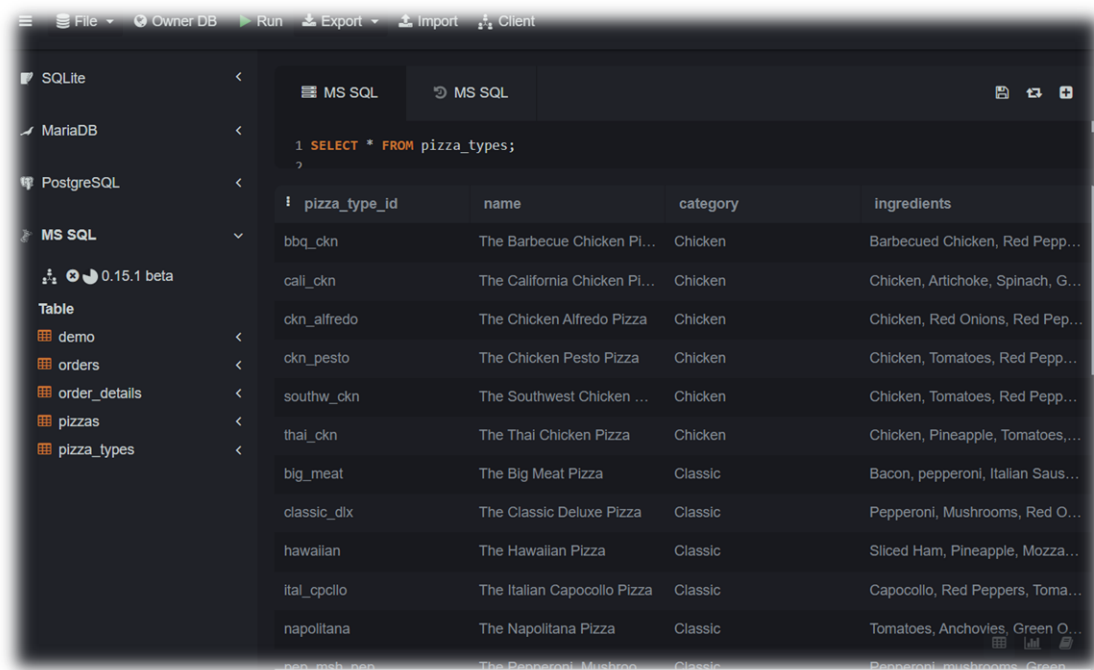
```
ALTER TABLE order_details
```

```
ADD CONSTRAINT fk_order_id FOREIGN KEY(order_id) REFERENCES orders(order_id);
```

```
ALTER TABLE order_details
```

```
ADD CONSTRAINT fk_pizza_id FOREIGN KEY(pizza_id) REFERENCES pizzas(pizza_id);
```

* The results:



pizza_type_id	name	category	ingredients
bbq_chn	The Barbecue Chicken Pi...	Chicken	Barbecued Chicken, Red Pepp...
cali_chn	The California Chicken Pi...	Chicken	Chicken, Artichoke, Spinach, G...
ckn_alfredo	The Chicken Alfredo Pizza	Chicken	Chicken, Red Onions, Red Pep...
ckn_pesto	The Chicken Pesto Pizza	Chicken	Chicken, Tomatoes, Red Pepp...
southw_chn	The Southwest Chicken ...	Chicken	Chicken, Tomatoes, Red Pepp...
thai_chn	The Thai Chicken Pizza	Chicken	Chicken, Pineapple, Tomatoes,...
big_meat	The Big Meat Pizza	Classic	Bacon, pepperoni, Italian Saus...
classic_dlx	The Classic Deluxe Pizza	Classic	Pepperoni, Mushrooms, Red O...
hawaiian	The Hawaiian Pizza	Classic	Sliced Ham, Pineapple, Mozza...
ital_cpollo	The Italian Capocollo Pizza	Classic	Capocollo, Red Peppers, Toma...
napolitana	The Napolitana Pizza	Classic	Tomatoes, Anchovies, Green O...
pepp_msh_nop	The Pepperoni Mushroom	Classic	Pepperoni, mushrooms, Green...

pizza_types table

The screenshot shows a database client interface with a sidebar on the left listing databases (SQLite, MariaDB, PostgreSQL, MS SQL) and tables (demo, orders, order_details, pizzas, pizza_types). The main pane displays the results of a SQL query: `SELECT * FROM pizzas;`. The results are shown in a table with columns: pizza_id, pizza_type_id, size, and price. The data includes various pizza types like bbq_ckn, cali_ckn, and ckn_alfredo in sizes S, M, and L, with prices ranging from 12.75 to 20.75. A 'USERWAY' banner for a compliance checker is visible in the bottom left corner.

pizza_id	pizza_type_id	size	price
bbq_ckn_s	bbq_ckn	S	12.75
bbq_ckn_m	bbq_ckn	M	16.75
bbq_ckn_l	bbq_ckn	L	20.75
cali_ckn_s	cali_ckn	S	12.75
cali_ckn_m	cali_ckn	M	16.75
cali_ckn_l	cali_ckn	L	20.75
ckn_alfredo_s	ckn_alfredo	S	12.75
ckn_alfredo_m	ckn_alfredo	M	16.75
ckn_alfredo_l	ckn_alfredo	L	20.75
ckn_pesto_s	ckn_pesto	S	12.75
ckn_pesto_m	ckn_pesto	M	16.75
ckn_pesto_l	ckn_pesto	L	20.75

pizzas table

The screenshot shows the same database client interface, but with a different SQL query: `SELECT TOP 100 * FROM orders;`. The results are shown in a table with columns: order_id, date, and time. The data lists 12 orders, all dated 1/1/2015, with times ranging from 11:38:36 to 13:04:41. The 'USERWAY' banner is also present in the bottom left corner.

order_id	date	time
1	1/1/2015	11:38:36
2	1/1/2015	11:57:40
3	1/1/2015	12:12:28
4	1/1/2015	12:16:31
5	1/1/2015	12:21:30
6	1/1/2015	12:29:36
7	1/1/2015	12:50:37
8	1/1/2015	12:51:37
9	1/1/2015	12:52:01
10	1/1/2015	13:00:15
11	1/1/2015	13:02:59
12	1/1/2015	13:04:41

orders table

order_details_id	order_id	pizza_id	quantity
1	1	hawaiian_m	1
2	2	classic_dix_m	1
3	2	five_cheese_l	1
4	2	ital_supr_l	1
5	2	mexicana_m	1
6	2	thal_ckn_l	1
7	3	ital_supr_m	1
8	3	prsc_argla_l	1
9	4	ital_supr_m	1
10	5	ital_supr_m	1
11	6	bbq_ckn_s	1

order_details table

4. I used python

* The code:

```
import pandas as pd
```

```
#load the CSV file specifying the encoding(specifying the encoding parameter
#when reading the CSV file helps Pandas interpret the file's contents correctly,
#especially when dealing with characters that fall outside the standard ASCII range.)
df=pd.read_csv(r"D:\internship_diconium\tables\pizza_types.csv", encoding='latin1')
```

```
#initialize a dictionary to count ingredient occurrences
ingredient_counts={ }
```

```
#iterate over each row in the DataFrame and parse the ingredients
```

```
for ingredients_str in df['ingredients']:
    ingredients=ingredients_str.split(", ")
    for ingredient in ingredients:
        if ingredient in ingredient_counts:
            ingredient_counts[ingredient]+=1
        else:
            ingredient_counts[ingredient]=1
```

```
#find the ingredient with the highest occurrence count
most_used_ingredient=max(ingredient_counts, key=ingredient_counts.get)
frequency_of_most_used_ingredient=ingredient_counts[most_used_ingredient]

print("The most used ingredient is:", most_used_ingredient)
print("It appears", frequency_of_most_used_ingredient, "times.")
```

* The result:

The screenshot shows a Python IDE with a file explorer on the left and a code editor on the right. The file explorer shows a project named 'INTERNSHIP_DICONIUM' with several folders and files, including 'src' and 'tables'. The code editor shows a Python script named 'ex4.py' with the following code:

```
src > ex4.py > ...
4 #when reading the CSV file helps Pandas interpret the file's content
5 #especially when dealing with characters that fall outside the standard
6 df=pd.read_csv(r"D:\internship_diconium\tables\pizza_types.csv", encoding='utf-8')
7
8 #initialize a dictionary to count ingredient occurrences
9 ingredient_counts={}
10
11 #iterate over each row in the DataFrame and parse the ingredients
12 for ingredients_str in df['ingredients']:
13     ingredients=ingredients_str.split(", ")
14     for ingredient in ingredients:
15         if ingredient in ingredient_counts:
16             ingredient_counts[ingredient]+=1
17         else:
18             ingredient_counts[ingredient]=1
19
20 #find the ingredient with the highest occurrence count
21 most_used_ingredient=max(ingredient_counts, key=ingredient_counts.get)
22 frequency_of_most_used_ingredient=ingredient_counts[most_used_ingredient]
23
24 print("The most used ingredient is:", most_used_ingredient)
```

The terminal window at the bottom shows the output of the script:

```
PS D:\internship_diconium> & C:/Users/andra/AppData/Local/Programs/Python/Python311/python.exe d:/internship_diconium/src/ex4.py
The most used ingredient is: Garlic
It appears 17 times.
PS D:\internship_diconium>
```

5. I used python and the jupyter notebook extension to display the result more clearly

* The code:

```
import pandas as pd

#load the CSV file into a DataFrame
df=pd.read_csv(r"D:\internship_diconium\tables\pizza_types.csv", encoding='latin1')

#initialize a set to store unique types of cheese
cheese_types=set()

#function to split ingredients by both commas and semicolons
def split_ingredients(ingredients_str):
    return [item.strip() for item in ingredients_str.replace('; ', ',').split(',')]

#iterate over each row in the DataFrame and parse the ingredients
for ingredients_str in df['ingredients']:
    ingredients=split_ingredients(ingredients_str)
    for ingredient in ingredients:
        if "cheese" in ingredient.lower(): #check if the word 'cheese' is in the ingredient
            cheese_types.add(ingredient.strip())

#convert the set of unique cheeses to a sorted list
unique_cheeses=sorted(cheese_types)

# Print the results
print(f"The restaurant uses {len(unique_cheeses)} types of cheese.")
print("The types of cheese used are:")
for cheese in unique_cheeses:
    print(cheese)

#create a DataFrame with an ID column and a Cheese column
cheese_df=pd.DataFrame({'ID': range(1, len(unique_cheeses)+1), 'Cheese': unique_cheeses})

#print the DataFrame with a title
print("List of Cheese Types Used by the Restaurant")
cheese_df
```

* The result:

```
1 import pandas as pd
2
3 #load the CSV file into a Dataframe
4 df=pd.read_csv("D:\internship_diconium\tables\pizza_types
5
6 #initialize a set to store unique types of cheese
7 cheese_types=set()
8
9 #function to split ingredients by both commas and semicolons
10 def split_ingredients(ingredients_str):
11     return [item.strip() for item in ingredients_str.repla
12
13 #iterate over each row in the Dataframe and parse the ingr
14 for ingredients_str in df['ingredients']:
15     ingredients=split_ingredients(ingredients_str)
16     for ingredient in ingredients:
17         if "cheese" in ingredient.lower(): #check if the
18             cheese_types.add(ingredient.strip())
19
20 #convert the set of unique cheeses to a sorted list
21 unique_cheeses=sorted(cheese_types)
22
23 # print the results
24 print(f"The restaurant uses {len(unique_cheeses)} types of
25 print(f"the types of cheese used are:")
26 for cheese in unique_cheeses:
27     print(cheese)
28
29 #create a DataFrame with an ID column and a Cheese column
30 cheese_df=pd.DataFrame({'ID': range(1, len(unique_cheeses)
31
32 #print the Dataframe with a title
33 print("\nList of cheese types used by the restaurant luigi
34 cheese_df
```

Connected to Python 3.11.0

✓ import pandas as pd

... The restaurant uses 15 types of cheese.

The types of cheese used are:

- Asiago Cheese
- Blue Cheese
- Brie Carre Cheese
- Feta Cheese
- Fontina Cheese
- Goat Cheese
- Gorgonzola Piccante Cheese
- Gouda Cheese
- Mozarella Cheese
- Mozarella Cheese
- Parmigiano Reggiano Cheese
- Provolone Cheese
- Ricotta Cheese
- Romano Cheese
- Smoked Gouda Cheese

List of cheese types used by the restaurant Luigi's Pizza

ID	Cheese
0 1	Asiago Cheese
1 2	Blue Cheese
2 3	Brie Carre Cheese
3 4	Feta Cheese

	ID	Cheese
0	1	Asiago Cheese
1	2	Blue Cheese
2	3	Brie Carre Cheese
3	4	Feta Cheese
4	5	Fontina Cheese
5	6	Goat Cheese
6	7	Gorgonzola Piccante Cheese
7	8	Gouda Cheese
8	9	Mozarella Cheese
9	10	Mozarella Cheese
10	11	Parmigiano Reggiano Cheese
11	12	Provolone Cheese
12	13	Ricotta Cheese
13	14	Romano Cheese
14	15	Smoked Gouda Cheese

6. I used python and the jupyter notebook extension to display the result more clearly

- * The code:

```
import pandas as pd
```

```
#load the CSV file into a pandas DataFrame
```

```
pizzas_df=pd.read_csv(r"D:\internship_diconium\tables\pizzas.csv")
```

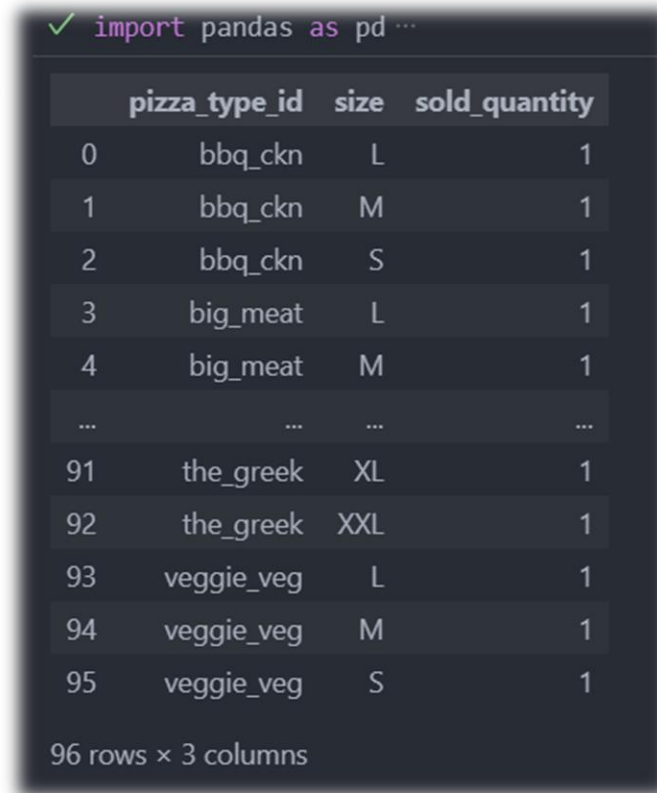
```
#group by pizza_type_id and size, then count the number of pizzas sold for each group
```

```
sales_summary=pizzas_df.groupby(['pizza_type_id', 'size']).size().reset_index(name='sold_quantity')
```

```
#display the sales summary DataFrame as a table
```

```
sales_summary
```

- * The result:



```
✓ import pandas as pd ...
```

	pizza_type_id	size	sold_quantity
0	bbq_ckn	L	1
1	bbq_ckn	M	1
2	bbq_ckn	S	1
3	big_meat	L	1
4	big_meat	M	1
...
91	the_greek	XL	1
92	the_greek	XXL	1
93	veggie_veg	L	1
94	veggie_veg	M	1
95	veggie_veg	S	1

96 rows × 3 columns

7. I used python and the jupyter notebook extension to display the result more clearly

* The code:

```
import pandas as pd
```

```
#load the order details CSV file into a pandas DataFrame
```

```
order_details_df=pd.read_csv(r"D:\internship_diconium\tables\order_details.csv")
```

```
#group by pizza_id and sum the quantities to get the total quantity of each pizza ordered
```

```
pizza_orders=order_details_df.groupby('pizza_id')['quantity'].sum().reset_index()
```

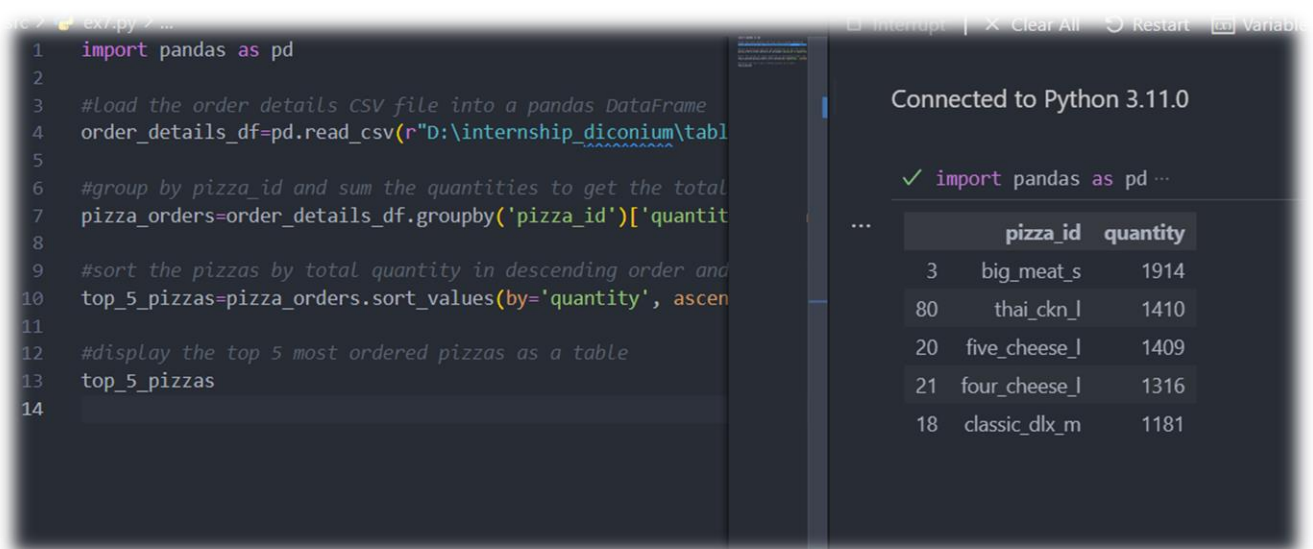
```
#sort the pizzas by total quantity in descending order and select the top 5
```

```
top_5_pizzas=pizza_orders.sort_values(by='quantity', ascending=False).head(5)
```

```
#display the top 5 most ordered pizzas as a table
```

```
top_5_pizzas
```

* The result:



The screenshot shows a Jupyter Notebook interface. On the left, the code editor contains the following Python code:

```
1 import pandas as pd
2
3 #load the order details CSV file into a pandas DataFrame
4 order_details_df=pd.read_csv(r"D:\internship_diconium\tabl
5
6 #group by pizza_id and sum the quantities to get the total
7 pizza_orders=order_details_df.groupby('pizza_id')['quantit
8
9 #sort the pizzas by total quantity in descending order and
10 top_5_pizzas=pizza_orders.sort_values(by='quantity', ascen
11
12 #display the top 5 most ordered pizzas as a table
13 top_5_pizzas
14
```

On the right, the output area shows the result of the code execution. It indicates that the notebook is connected to Python 3.11.0 and shows the output of the last cell, which is a table of the top 5 most ordered pizzas:

	pizza_id	quantity
3	big_meat_s	1914
80	thai_ckn_l	1410
20	five_cheese_l	1409
21	four_cheese_l	1316
18	classic_dlx_m	1181

8. I used python and the jupyter notebook extension to display the result more clearly

* The code:

```
import pandas as pd
from pandasql import sqldf

#reading data from CSV files
order_details_df=pd.read_csv(r"D:\internship_diconium\tables\order_details.csv")
orders_df=pd.read_csv(r"D:\internship_diconium\tables\orders.csv")
pizzas_df=pd.read_csv(r"D:\internship_diconium\tables\pizzas.csv")

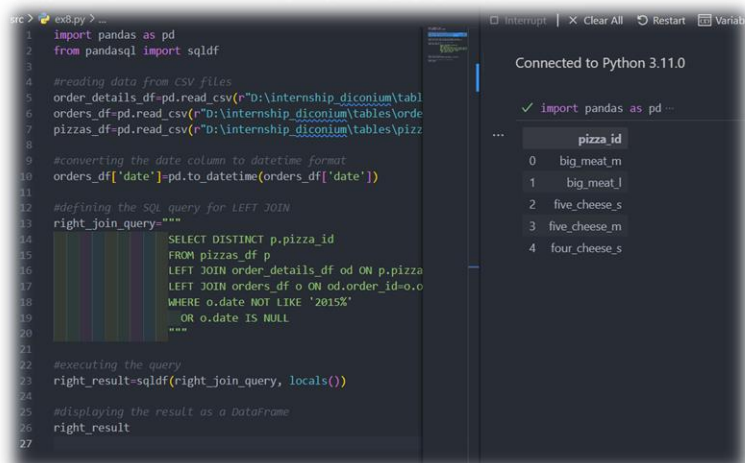
#converting the date column to datetime format
orders_df['date']=pd.to_datetime(orders_df['date'])

#defining the SQL query for LEFT JOIN
right_join_query="""
    SELECT DISTINCT p.pizza_id
    FROM pizzas_df p
    LEFT JOIN order_details_df od ON p.pizza_id=od.pizza_id
    LEFT JOIN orders_df o ON od.order_id=o.order_id
    WHERE o.date NOT LIKE '2015%'
    OR o.date IS NULL
    """

#executing the query
right_result=sqldf(right_join_query, locals())

#displaying the result as a DataFrame
right_result
```

* The result:



```
src> ex8.py - ...
1 import pandas as pd
2 from pandasql import sqldf
3
4 #reading data from CSV files
5 order_details_df=pd.read_csv(r"D:\internship_diconium\tables\order_details.csv")
6 orders_df=pd.read_csv(r"D:\internship_diconium\tables\orders.csv")
7 pizzas_df=pd.read_csv(r"D:\internship_diconium\tables\pizzas.csv")
8
9 #converting the date column to datetime format
10 orders_df['date']=pd.to_datetime(orders_df['date'])
11
12 #defining the SQL query for LEFT JOIN
13 right_join_query="""
14     SELECT DISTINCT p.pizza_id
15     FROM pizzas_df p
16     LEFT JOIN order_details_df od ON p.pizza_id=od.pizza_id
17     LEFT JOIN orders_df o ON od.order_id=o.order_id
18     WHERE o.date NOT LIKE '2015%'
19     OR o.date IS NULL
20     """
21
22 #executing the query
23 right_result=sqldf(right_join_query, locals())
24
25 #displaying the result as a DataFrame
26 right_result
27
```

Connected to Python 3.11.0

	pizza_id
0	big_meat_m
1	big_meat_l
2	five_cheese_s
3	five_cheese_m
4	four_cheese_s

9. I used python and the jupyter notebook extension to display the result more clearly

* The code:

```
import pandas as pd
```

```
#read the pizzas CSV file into a DataFrame
```

```
pizzas_df = pd.read_csv(r"D:\internship_diconium\tables\pizzas.csv")
```

```
#define a function to categorize the sizes
```

```
def categorize_size(size):
```

```
    if 's' in size.lower():
```

```
        return 'Small'
```

```
    elif 'm' in size.lower():
```

```
        return 'Medium'
```

```
    elif 'l' in size.lower():
```

```
        return 'Large'
```

```
    else:
```

```
        return 'Unknown'
```

```
#apply the function to create a new column 'size_category'
```

```
pizzas_df['size_category']=pizzas_df['size'].apply(categorize_size)
```

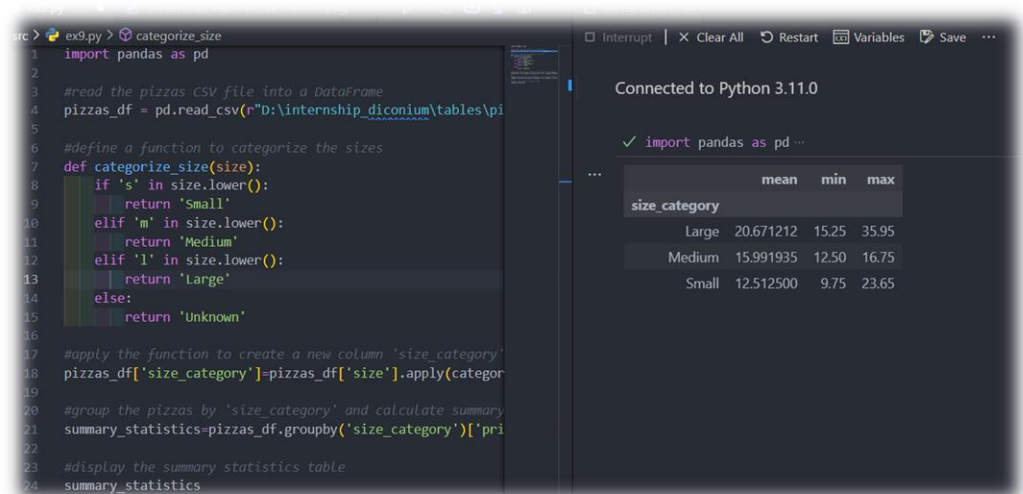
```
#group the pizzas by 'size_category' and calculate summary statistics for 'price'
```

```
summary_statistics=pizzas_df.groupby('size_category')['price'].agg(['mean', 'min', 'max'])
```

```
#display the summary statistics table
```

```
summary_statistics
```

* The result:



```
src > ex9.py > categorize_size
1 import pandas as pd
2
3 #read the pizzas CSV file into a DataFrame
4 pizzas_df = pd.read_csv(r"D:\internship_diconium\tables\pi
5
6 #define a function to categorize the sizes
7 def categorize_size(size):
8     if 's' in size.lower():
9         return 'Small'
10    elif 'm' in size.lower():
11        return 'Medium'
12    elif 'l' in size.lower():
13        return 'Large'
14    else:
15        return 'Unknown'
16
17 #apply the function to create a new column 'size_category'
18 pizzas_df['size_category']=pizzas_df['size'].apply(categor
19
20 #group the pizzas by 'size_category' and calculate summary
21 summary_statistics=pizzas_df.groupby('size_category')['pri
22
23 #display the summary statistics table
24 summary_statistics
```

Connected to Python 3.11.0

✓ import pandas as pd ...

	mean	min	max
size_category			
Large	20.671212	15.25	35.95
Medium	15.991935	12.50	16.75
Small	12.512500	9.75	23.65

10. I used python and the jupyter notebook extension to display the result more clearly

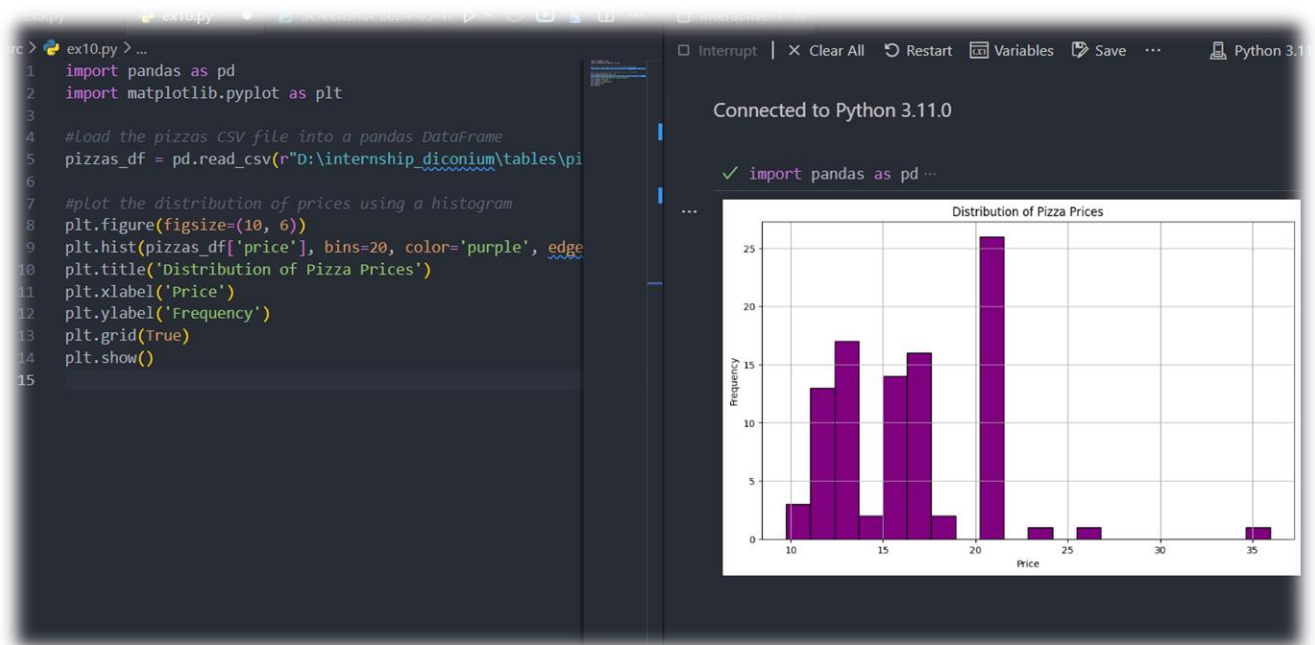
* The code:

```
import pandas as pd
import matplotlib.pyplot as plt

#load the pizzas CSV file into a pandas DataFrame
pizzas_df = pd.read_csv(r"D:\internship_diconium\tables\pizzas.csv")

#plot the distribution of prices using a histogram
plt.figure(figsize=(10, 6))
plt.hist(pizzas_df['price'], bins=20, color='purple', edgecolor='black')
plt.title('Distribution of Pizza Prices')
plt.xlabel('Price')
plt.ylabel('Frequency')
plt.grid(True)
plt.show()
```

* The result:



11. I used python and the jupyter notebook extension to display the result more clearly

- * The code:

```
import pandas as pd
from pandasql import sqldf

# Load the CSV files into pandas DataFrames
pizzas_df=pd.read_csv(r"D:\internship_diconium\tables\pizzas.csv")
orders_df=pd.read_csv(r"D:\internship_diconium\tables\orders.csv")
order_details_df=pd.read_csv(r"D:\internship_diconium\tables\order_details.csv")

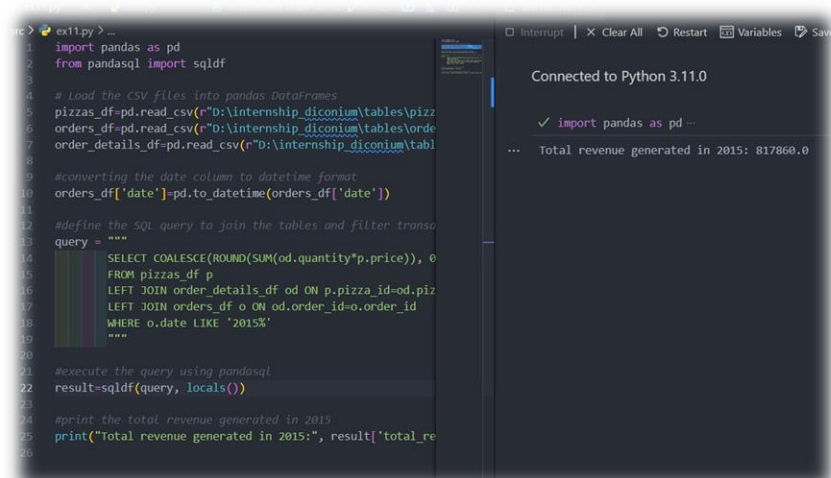
#converting the date column to datetime format
orders_df['date']=pd.to_datetime(orders_df['date'])

#define the SQL query to join the tables and filter transactions in 2015
query = """
SELECT COALESCE(ROUND(SUM(od.quantity*p.price)), 0) AS total_revenue_2015
FROM pizzas_df p
LEFT JOIN order_details_df od ON p.pizza_id=od.pizza_id
LEFT JOIN orders_df o ON od.order_id=o.order_id
WHERE o.date LIKE '2015%'
"""

#execute the query using pandasql
result=sqldf(query, locals())

#print the total revenue generated in 2015
print("Total revenue generated in 2015:", result['total_revenue_2015'].values[0])
```

- * The result:



The screenshot shows a Jupyter Notebook interface with a code cell on the left and a console output on the right. The code cell contains the same Python code as shown in the previous block. The console output on the right shows the execution of the code, including the connection to Python 3.11.0 and the final result: "Total revenue generated in 2015: 817860.0".

12. I used python and the jupyter notebook extension to display the result more clearly

* The code:

```
import pandas as pd
import matplotlib.pyplot as plt
from pandasql import sqldf

#load the CSV files into pandas DataFrames
pizzas_df=pd.read_csv(r"D:\internship_diconium\tables\pizzas.csv")
orders_df=pd.read_csv(r"D:\internship_diconium\tables\orders.csv")
order_details_df=pd.read_csv(r"D:\internship_diconium\tables\order_details.csv")

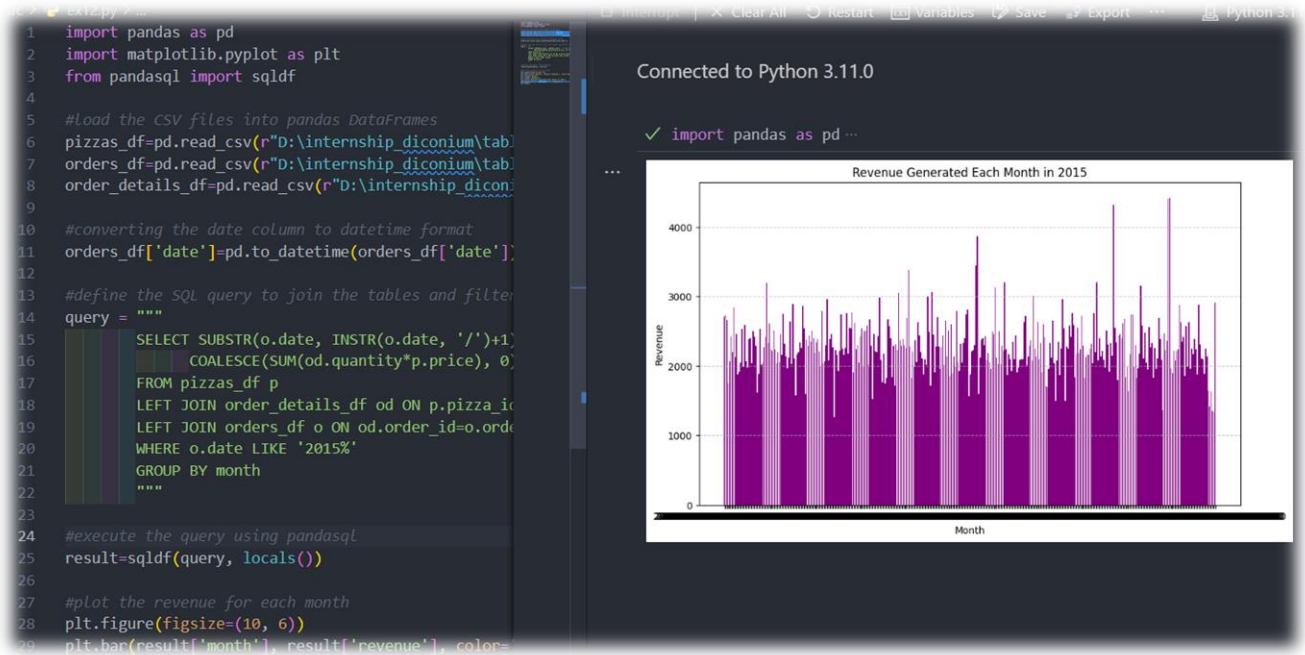
#converting the date column to datetime format
orders_df['date']=pd.to_datetime(orders_df['date'])

#define the SQL query to join the tables and filter transactions in 2015, grouped by month
query = """
    SELECT SUBSTR(o.date, INSTR(o.date, '/')+1) AS month,
           COALESCE(SUM(od.quantity*p.price), 0) AS revenue
    FROM pizzas_df p
    LEFT JOIN order_details_df od ON p.pizza_id=od.pizza_id
    LEFT JOIN orders_df o ON od.order_id=o.order_id
    WHERE o.date LIKE '2015%'
    GROUP BY month
    """

#execute the query using pandasql
result=sqldf(query, locals())

#plot the revenue for each month
plt.figure(figsize=(10, 6))
plt.bar(result['month'], result['revenue'], color='purple')
plt.xlabel('Month')
plt.ylabel('Revenue')
plt.title('Revenue Generated Each Month in 2015')
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.show()
```


- * The result:



- * Conclusion: I observe that revenues seem to vary depending on the month. For example, we can see that revenues appear to be higher during the warm months and lower during the cold months. This variation may be influenced by seasonal factors such as vacations or holidays.

13. I used python and the jupyter notebook extension to display the result more clearly

- * The code:

```

import pandas as pd
import matplotlib.pyplot as plt

#load the orders CSV into a pandas DataFrame
orders_df=pd.read_csv(r"D:\internship_diconium\tables\orders.csv")

#convert the 'time' column to datetime with specified format
orders_df['time']=pd.to_datetime(orders_df['time'], format='%H:%M:%S')

```

```

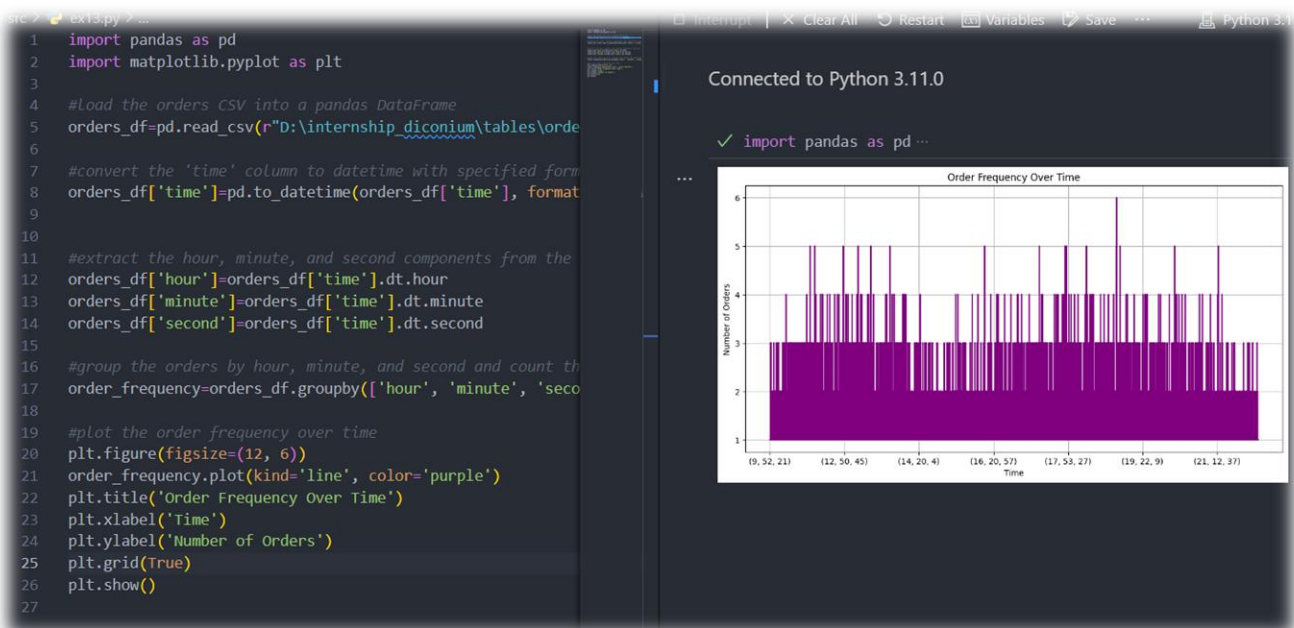
#extract the hour, minute, and second components from the 'time' column
orders_df['hour']=orders_df['time'].dt.hour
orders_df['minute']=orders_df['time'].dt.minute
orders_df['second']=orders_df['time'].dt.second

#group the orders by hour, minute, and second and count the number of orders for each timestamp
order_frequency=orders_df.groupby(['hour', 'minute', 'second']).size()

#plot the order frequency over time
plt.figure(figsize=(12, 6))
order_frequency.plot(kind='line', color='purple')
plt.title('Order Frequency Over Time')
plt.xlabel('Time')
plt.ylabel('Number of Orders')
plt.grid(True)
plt.show()

```

* The result:



* Conclusion: The graph shows order frequency throughout the day, revealing busy peak hours and quieter times. This insight aids in resource planning for better customer service.

14. I used python and the jupyter notebook extension to display the result more clearly

* The code:

```
import pandas as pd

#load the orders CSV into a pandas DataFrame
orders_df=pd.read_csv(r"D:\internship_diconium\tables\orders.csv")

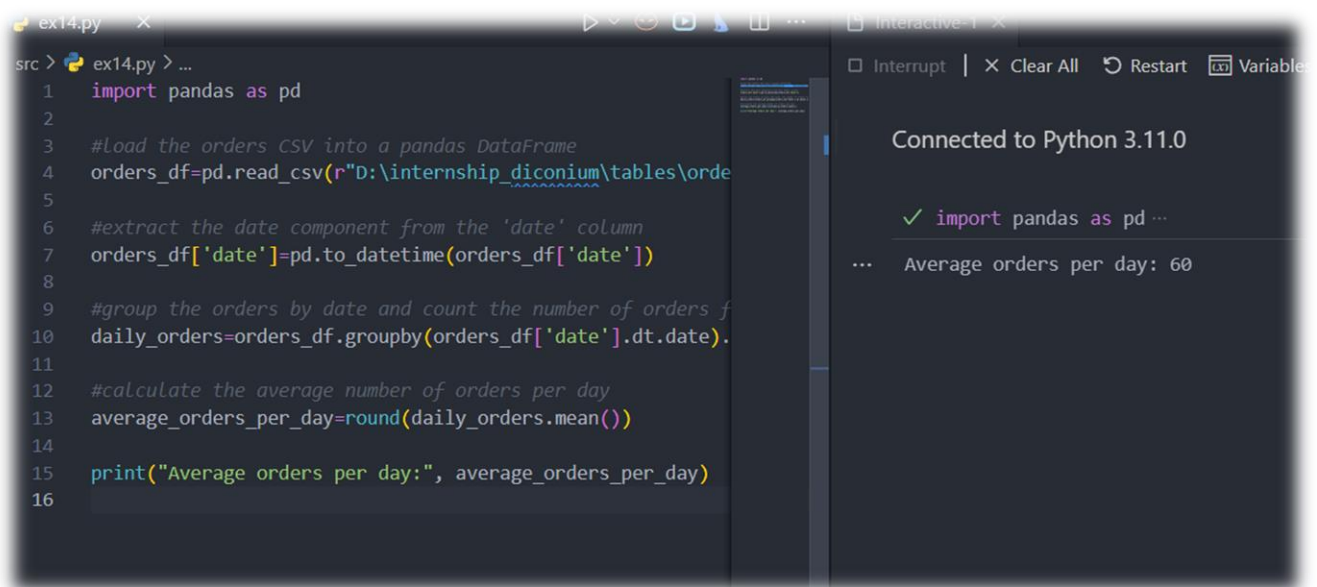
#extract the date component from the 'date' column
orders_df['date']=pd.to_datetime(orders_df['date'])

#group the orders by date and count the number of orders for each date
daily_orders=orders_df.groupby(orders_df['date'].dt.date).size()

#calculate the average number of orders per day
average_orders_per_day=round(daily_orders.mean())

print("Average orders per day:", average_orders_per_day)
```

* The result:



The screenshot shows a Jupyter Notebook interface with a dark theme. The left pane displays a Python script named 'ex14.py' with 16 lines of code. The code imports pandas, reads a CSV file, converts the date column to datetime, groups by date, and calculates the average number of orders per day. The right pane shows the output of the code, which is 'Average orders per day: 60'. The interface includes a toolbar with icons for running, saving, and other notebook functions, and a status bar at the bottom indicating the connection to Python 3.11.0.

```
src > ex14.py > ...
1  import pandas as pd
2
3  #Load the orders CSV into a pandas DataFrame
4  orders_df=pd.read_csv(r"D:\internship_diconium\tables\orde
5
6  #extract the date component from the 'date' column
7  orders_df['date']=pd.to_datetime(orders_df['date'])
8
9  #group the orders by date and count the number of orders f
10 daily_orders=orders_df.groupby(orders_df['date'].dt.date).
11
12 #calculate the average number of orders per day
13 average_orders_per_day=round(daily_orders.mean())
14
15 print("Average orders per day:", average_orders_per_day)
16
```

Connected to Python 3.11.0

✓ import pandas as pd ...

... Average orders per day: 60

15. I used python and the jupyter notebook extension to display the result more clearly

- * The code:

```
import pandas as pd
import matplotlib.pyplot as plt

#load the orders CSV into a pandas DataFrame
orders_df=pd.read_csv(r"D:\internship_diconium\tables\orders.csv")

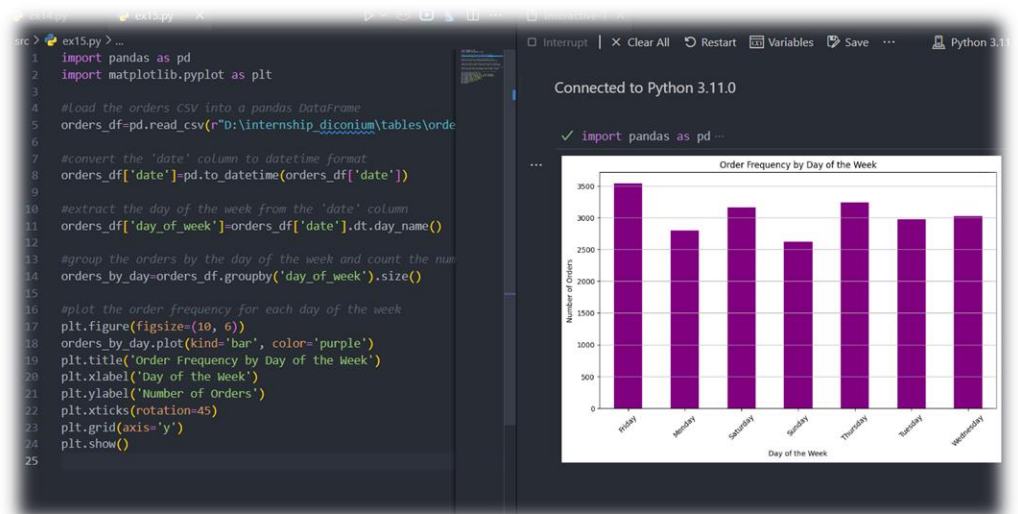
#convert the 'date' column to datetime format
orders_df['date']=pd.to_datetime(orders_df['date'])

#extract the day of the week from the 'date' column
orders_df['day_of_week']=orders_df['date'].dt.day_name()

#group the orders by the day of the week and count the number of orders for each day
orders_by_day=orders_df.groupby('day_of_week').size()

#plot the order frequency for each day of the week
plt.figure(figsize=(10, 6))
orders_by_day.plot(kind='bar', color='purple')
plt.title('Order Frequency by Day of the Week')
plt.xlabel('Day of the Week')
plt.ylabel('Number of Orders')
plt.xticks(rotation=45)
plt.grid(axis='y')
plt.show()
```

- * The result:



- * Conclusion: Based on the graph, it seems like Fridays are the most crowded days of the week. This could be because people tend to dine out more often as the workweek comes to a close, or perhaps it's influenced by social customs and traditions.

16. I used python and the jupyter notebook extension to display the result more clearly

- * The code:

```
import pandas as pd
from pandasql import sqldf
import matplotlib.pyplot as plt

#load the CSV files into pandas DataFrames
pizzas_df=pd.read_csv(r"D:\internship_diconium\tables\pizzas.csv")
orders_df=pd.read_csv(r"D:\internship_diconium\tables\orders.csv")
order_details_df=pd.read_csv(r"D:\internship_diconium\tables\order_details.csv")

#clean the date column in orders_df and handle potential issues
orders_df['date']=pd.to_datetime(orders_df['date'], format='%m/%d/%Y', errors='coerce')
orders_df=orders_df.dropna(subset=['date']) # Drop rows with invalid dates

#add a column for the day of the week to the orders_df
orders_df['day_of_week']=orders_df['date'].dt.day_name()

#define the SQL query to calculate revenue for each day of the week
query = """
SELECT o.day_of_week, COALESCE(SUM(od.quantity*p.price), 0) AS revenue
FROM order_details_df od
LEFT JOIN pizzas_df p ON p.pizza_id=od.pizza_id
LEFT JOIN orders_df o ON od.order_id=o.order_id
GROUP BY o.day_of_week
ORDER BY revenue DESC
"""

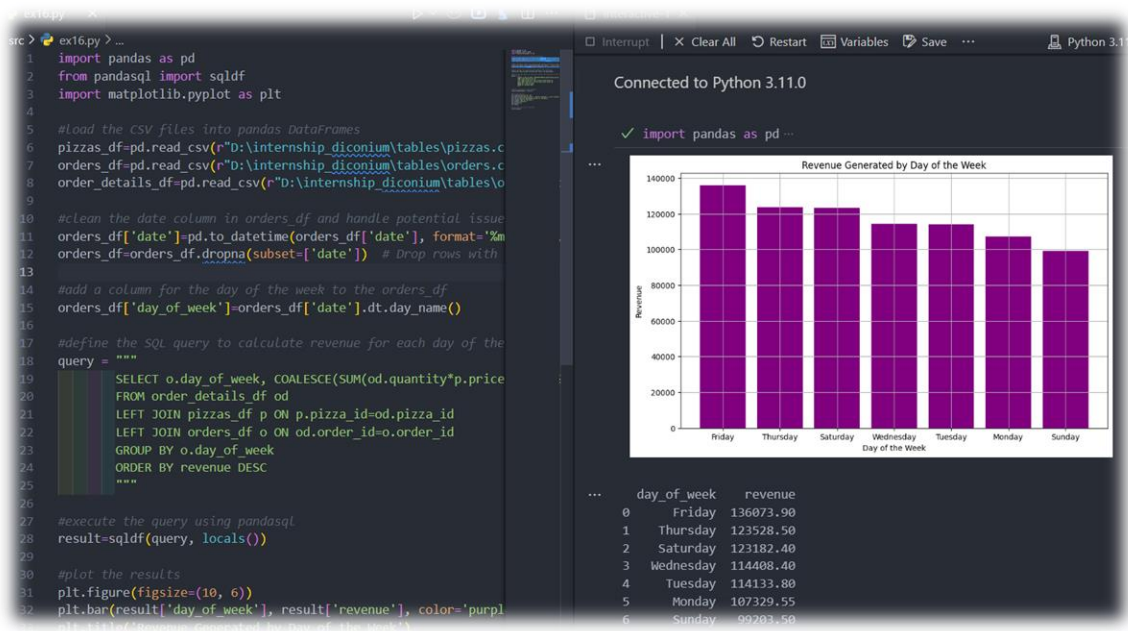
#execute the query using pandasql
result=sqldf(query, locals())

#plot the results
plt.figure(figsize=(10, 6))
plt.bar(result['day_of_week'], result['revenue'], color='purple')
plt.title('Revenue Generated by Day of the Week')
plt.xlabel('Day of the Week')
```

```
plt.ylabel('Revenue')
plt.grid(True)
plt.show()
```

```
#print the result for reference
print(result)
```

* The result:



* Conclusion: Based on the graph, you can observe which day of the week generates the highest revenue. This insight helps understand customer behavior and optimize restaurant operations. For instance, if Fridays show the highest revenue, it could be due to increased dining out as people celebrate the end of the workweek.

17. I used python and the jupyter notebook extension to display the result more clearly

* The code:

```
import pandas as pd
from pandasql import sqldf

#load the CSV files into pandas DataFrames
pizzas_df=pd.read_csv(r"D:\internship_diconium\tables\pizzas.csv")
orders_df=pd.read_csv(r"D:\internship_diconium\tables\orders.csv")
order_details_df=pd.read_csv(r"D:\internship_diconium\tables\order_details.csv")
```



```

#convert the 'date' column to datetime
orders_df['date']=pd.to_datetime(orders_df['date'], format='%m/%d/%Y')

#create a DataFrame with a full range of dates in 2015
full_date_range=pd.date_range(start='2015-01-01', end='2015-12-31')
full_date_df=pd.DataFrame(full_date_range, columns=['date'])

#define the SQL queries to join the tables and calculate daily revenue
query="""
    SELECT fd.date, COALESCE(SUM(od.quantity*p.price), 0) AS daily_revenue
    FROM full_date_df fd
    LEFT JOIN orders_df o ON fd.date=o.date
    LEFT JOIN order_details_df od ON o.order_id=od.order_id
    LEFT JOIN pizzas_df p ON od.pizza_id=p.pizza_id
    GROUP BY fd.date
    """

#execute the query using pandasql
full_revenue_df=sqldf(query, globals())

#identify the dates with no sales (daily_revenue is 0)
no_sales_dates=full_revenue_df[full_revenue_df['daily_revenue']==0]

#display the periods with no pizza sales
print("Periods with no pizza sales:")
no_sales_dates

```

* The result:

```

In [1]: ex17.py> ...
1 import pandas as pd
2 from pandasql import sqldf
3
4 #load the CSV files into pandas DataFrames
5 pizzas_df=pd.read_csv(r"D:\internship_diconium\tables\pizzas.csv")
6 orders_df=pd.read_csv(r"D:\internship_diconium\tables\orders.csv")
7 order_details_df=pd.read_csv(r"D:\internship_diconium\tables\order_details.csv")
8
9 #convert the 'date' column to datetime
10 orders_df['date']=pd.to_datetime(orders_df['date'], format='%m/%d/%Y')
11
12 #create a DataFrame with a full range of dates in 2015
13 full_date_range=pd.date_range(start='2015-01-01', end='2015-12-31')
14 full_date_df=pd.DataFrame(full_date_range, columns=['date'])
15
16 #define the SQL queries to join the tables and calculate daily revenue
17 query="""
18     SELECT fd.date, COALESCE(SUM(od.quantity*p.price), 0) AS daily_revenue
19     FROM full_date_df fd
20     LEFT JOIN orders_df o ON fd.date=o.date
21     LEFT JOIN order_details_df od ON o.order_id=od.order_id
22     LEFT JOIN pizzas_df p ON od.pizza_id=p.pizza_id
23     GROUP BY fd.date
24     """
25
26 #execute the query using pandasql
27 full_revenue_df=sqldf(query, globals())
28
29 #identify the dates with no sales (daily_revenue is 0)
30 no_sales_dates=full_revenue_df[full_revenue_df['daily_revenue']==0]
31
32 #display the periods with no pizza sales
33 print("Periods with no pizza sales:")
34 no_sales_dates

```

Connected to Python 3.11.0

✓ import pandas as pd ...

... Periods with no pizza sales:

	date	daily_revenue
266	2015-09-24 00:00:00.000000	0.0
267	2015-09-25 00:00:00.000000	0.0
277	2015-10-05 00:00:00.000000	0.0
284	2015-10-12 00:00:00.000000	0.0
291	2015-10-19 00:00:00.000000	0.0
298	2015-10-26 00:00:00.000000	0.0
358	2015-12-25 00:00:00.000000	0.0

* Conclusion: The absence of pizza sales during certain periods could be attributed to factors like restaurant closures, changes in customer behavior, or external influences like holidays or events.