Week 1 Internship Report: OUBT Bootcamp – Day 1 to Day 5 Learnings:

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Introduction

This document summarizes my learnings and activities during Week 1 of the OUBT Bootcamp Internship (Day 1 to Day 5). Over the course of the week, I focused on building foundational skills in AWS services, Python for data engineering, SQL, and GitHub version control. I also gained hands-on experience with **Git and GitHub**, including repository management, branching, and committing changes, which helped me track and manage my code effectively. Each day involved hands-on exercises, practical tasks, and mini-projects aimed at developing core competencies in cloud computing, data processing, and automation workflows. The following sections detail the tasks completed, concepts learned, and technical skills acquired throughout the week.

Project Repository Details:

All Week 1 code, scripts, and documentation have been maintained in my Git repository: https://github.com/likhith5697/Week1_OUBT_Learnings. The repository includes Python scripts, SQL queries, and AWS configuration files for tasks completed from Day 1 to Day 5.

NAME: LIKHITH SASANK UPPALAPATI VENKATA

DAY: 1

AWS IAM Setup:

Create IAM users, groups, and roles.

Assign permissions and policies.

Test login using IAM sign-in link.

Billing Alert and CloudWatch Alarm:

Enable billing alerts in the AWS Billing Console.

Create a CloudWatch alarm to track monthly charges.

Set up an SNS topic to send cost alerts via email.

Amazon S3 Bucket:

Create a new S3 bucket.

Upload files of different types (PDF, image, CSV).

Enable bucket versioning and upload the same file again to check multiple versions.

GitHub Setup:

Create a new repository for OUBT bootcamp tasks.

Clone the repository locally.

Add the Day 1 Word document and screenshots.

Commit and push the files to GitHub.

1. Introduction

This week focused on AWS fundamentals, IAM, S3, GitHub basics, and Python for Data Engineering. The goal was to understand access management, cloud storage, and data handling concepts.

2. IAM Concepts Summary

IAM Overview:
IAM – It is a security's service that will control who can access our AWS account

IAM – USERS:

Users means one individual person that needs an access

Example : Let's say we are a team of developers, admin, testers. Each person needs a particular access so we create users as Likhith.Dev, Sasank.Admin, Venkat.Tester and give them roles and permissions as shown below.

Likhith.Dev -> **Developer** -> S3, Glue, RDS, Lambda

Sasank.Admin -> **Admin** -> Full Access

Venkat.Tester -> **Tester** -> Read Only

Here we attach policies to Users.

IAM – GROUPS:

Groups are the collection of Users who share same permissions.

Here we do assign policies to groups not for users separately as shown below,

Developers -> Access to S3, Glue, RDS, Lambda -> Members (Likhith, Sasank)

Admin -> Full Access -> Members (Venkat)

Tester -> Read Only Access (Durga)

IAM ROLES:

Role is a temporary identity with specific permissions.

AWS service assumes a role
Users assume a role
Group assumes a role

Example : Suppose I have a microservice or may be a spring boot app deployed in EC2 that basically needs to read and write to RDS, So instead of hardcoding credentials, I can create a role and assign to our EC2 instance for access to RDS, So it gets permission automatically.

IAM Policy:

Policy defines permissions.

It is a json object that says what is allowed and what is not allowed.

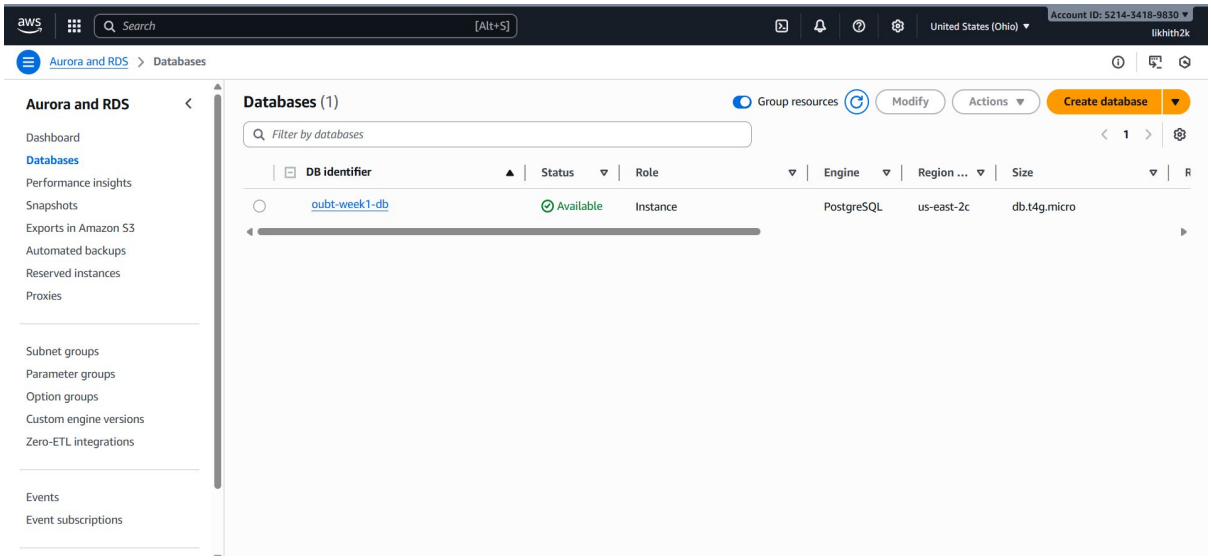
Billing Alerts:

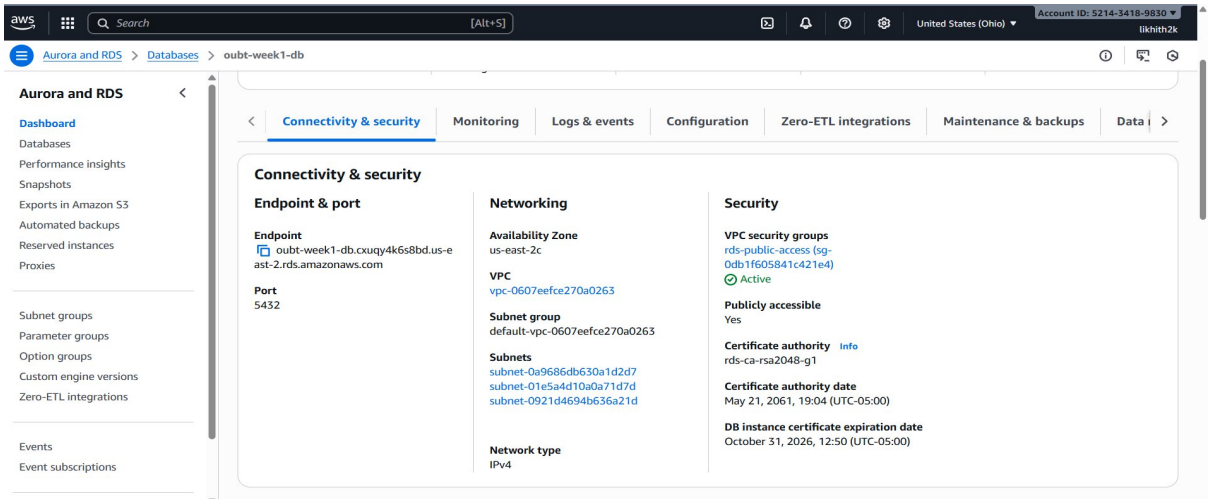
This helps to avoid surprise charges .

We can use cloudwatch alerts / Billing alarms to notify us when usage exceeds a limit.

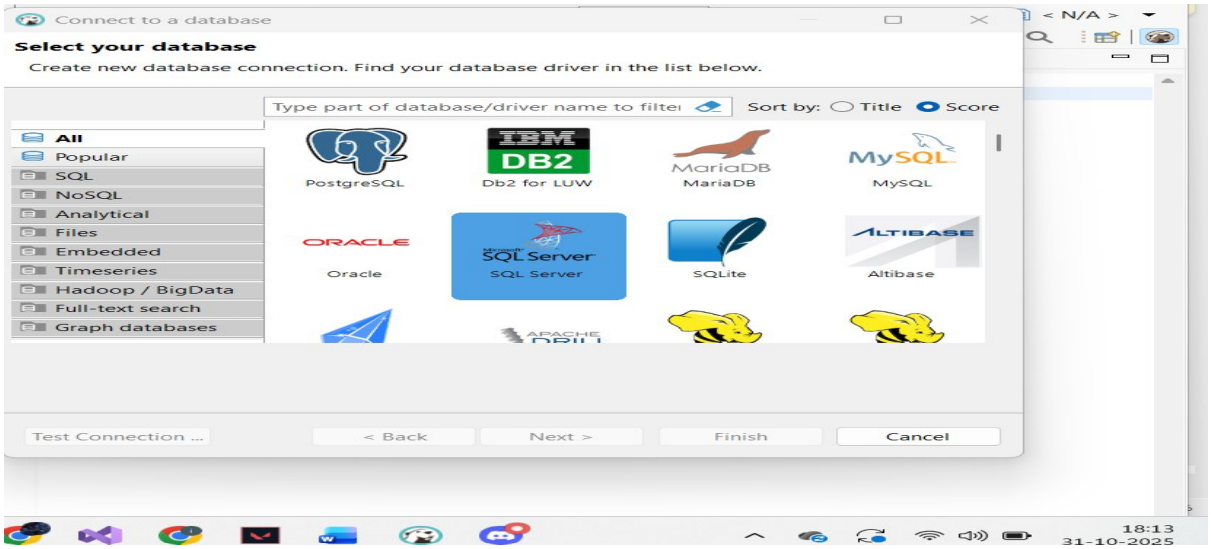
Here Am showing all my hand-ON PRACTICE,

Task 1: Created IAM User

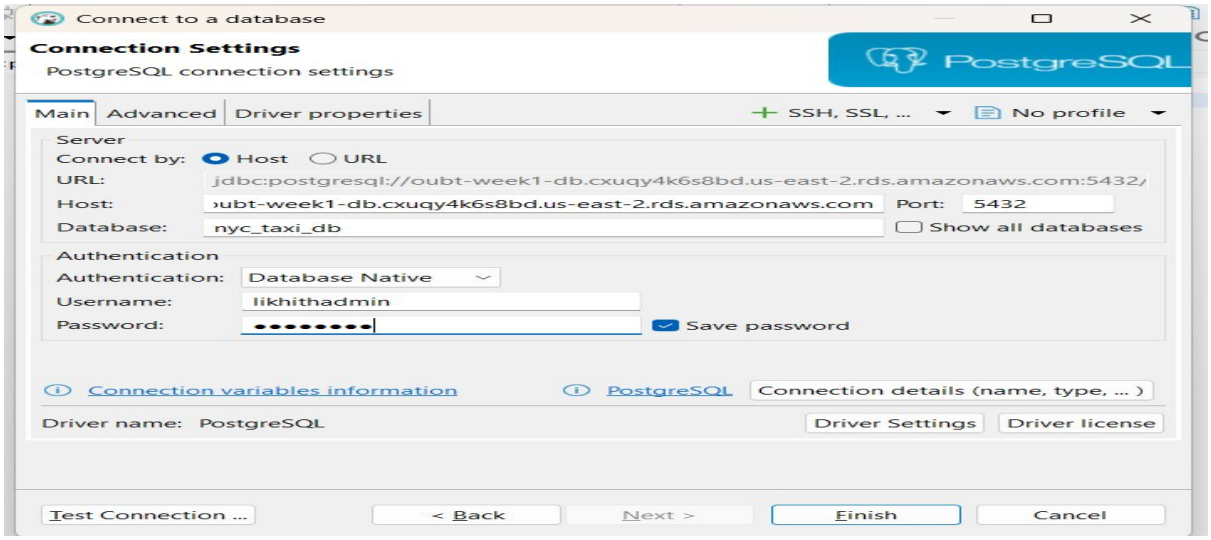


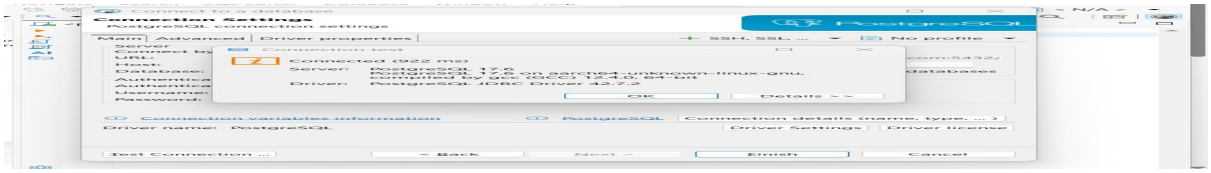


Successfully created user,

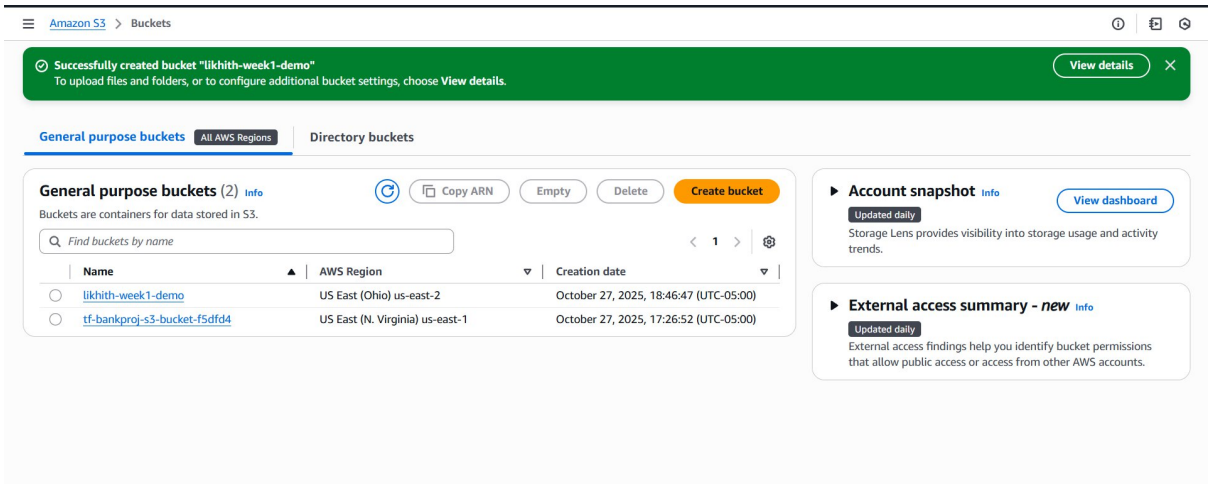


Task 2: Created S3 Bucket:



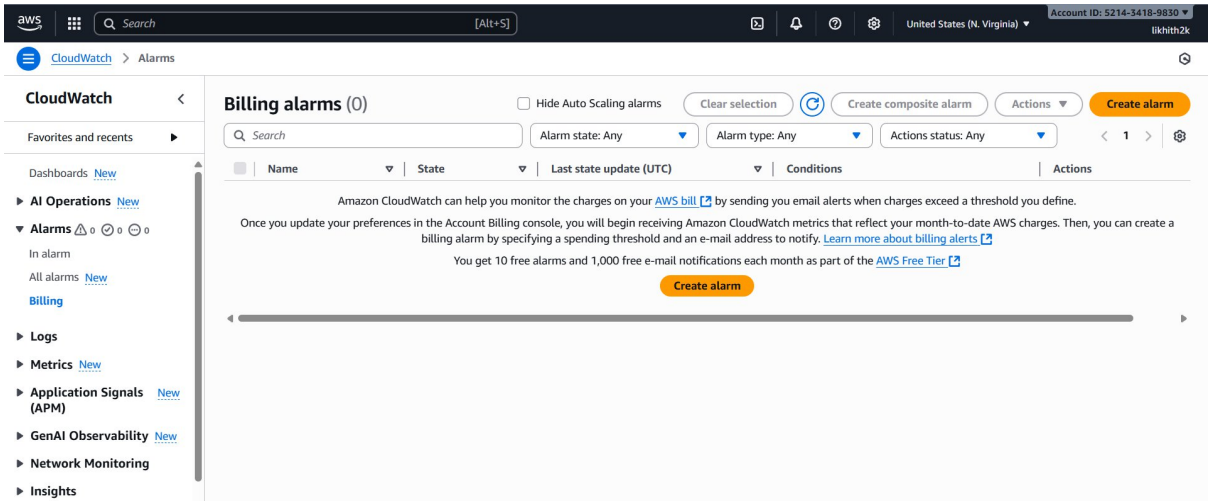


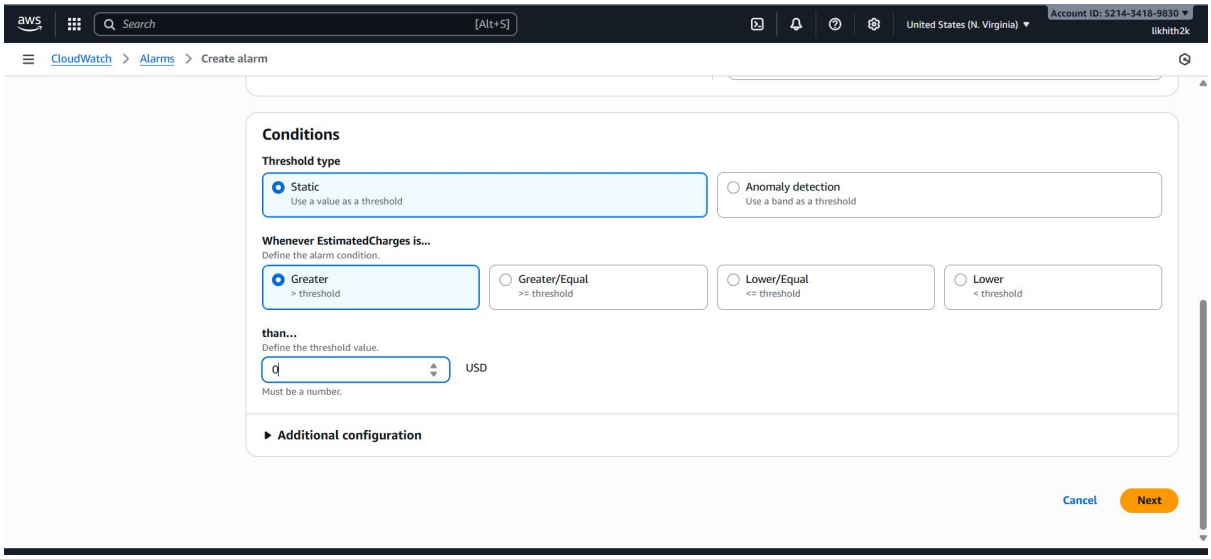
Successfully created S3-BUCKET,

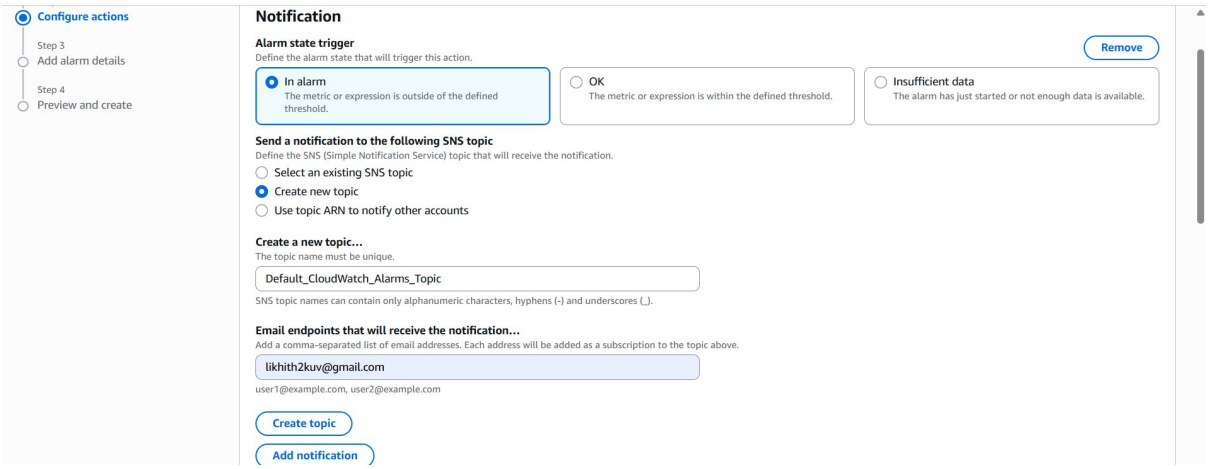


Task 3: Billing Alert

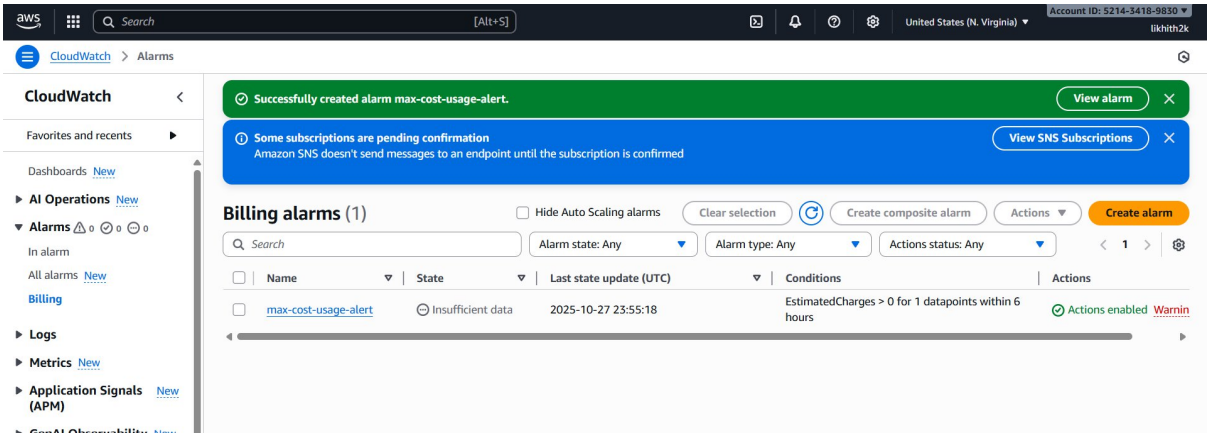
Set up CloudWatch Billing Alarm for $1 threshold with email alert.





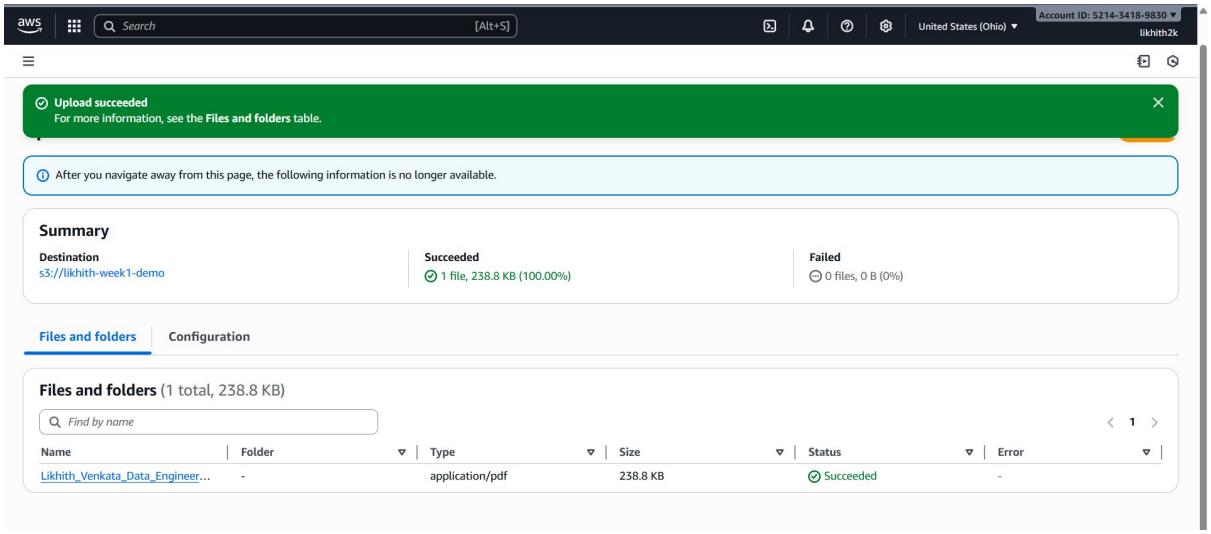


Finally created an alarm,



S3 - Section

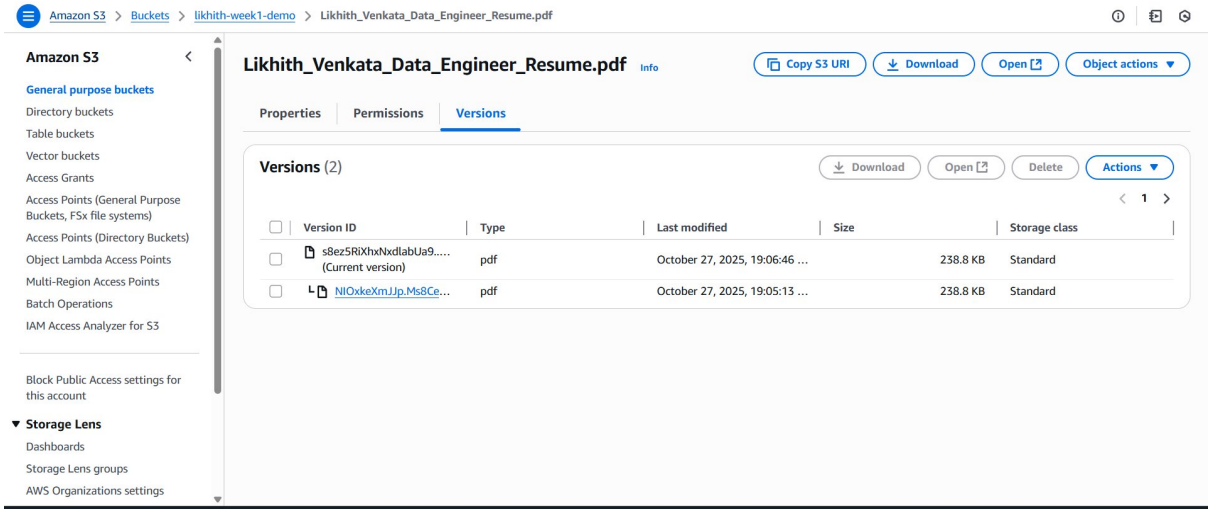
Here I have uploaded the files to S3,



S3 Bucket Versioning Test

I uploaded the same PDF file twice to my S3 bucket. With versioning enabled, both versions of the file are saved. The screenshot below shows the **current version** and the **previous version**. This confirms that versioning is working correctly and older files can be restored if needed.

Screenshot inserted below.



Summary :

Today I learned how AWS **Identity and Access Management (IAM)** helps control and secure user access across AWS services. I created **IAM users, groups, and roles**, assigned appropriate permissions, and understood the concept of least privilege for secure account management.

I also explored **Amazon S3**, where I created a bucket, uploaded multiple file types, and enabled **versioning** to maintain file history and data recovery. In addition, I configured **AWS CloudWatch billing alerts and alarms** to track usage and receive cost notifications, ensuring proactive budget management.

Finally, I set up a **GitHub repository** to store and manage all my bootcamp work, learned the basics of commits and pushing files, and organized my Day 1 deliverables systematically. Overall, I gained hands-on experience in foundational AWS services, account monitoring, and version control — key building blocks for cloud and data engineering.

OUBT Week 1 – Day 2–3: Python & Pandas & Boto3 SDK Practice

Author: Likhith Sasank Uppalapati Venkata

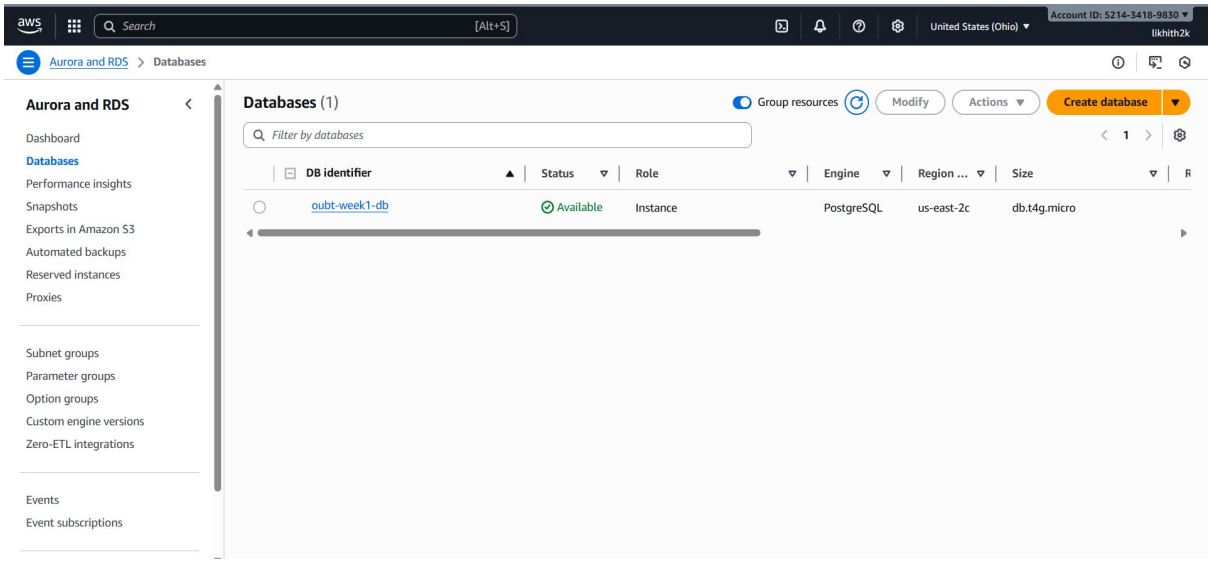
1. Python Practice

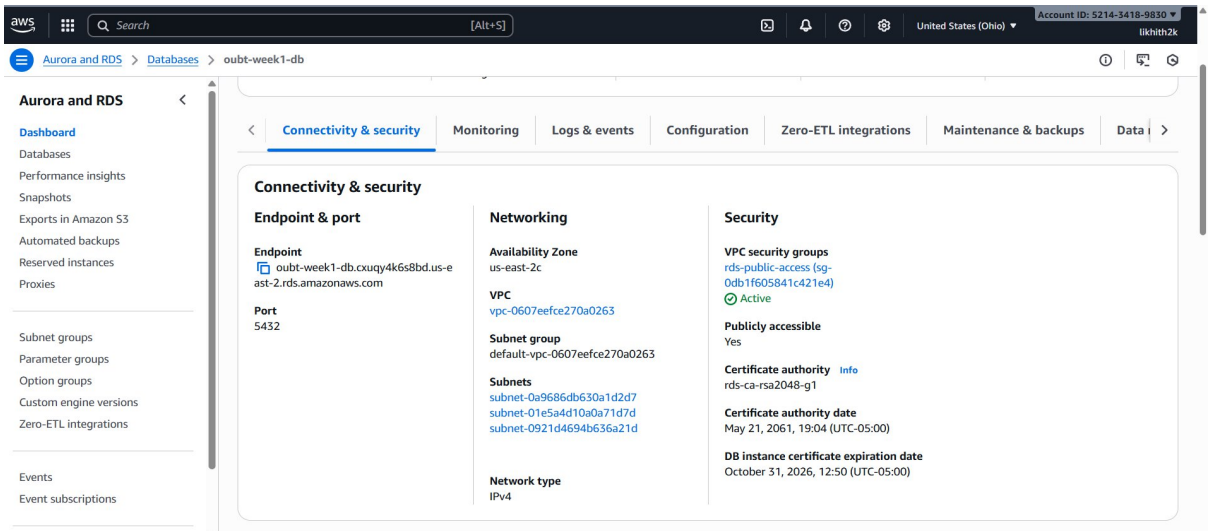
Practiced core Python concepts such as lists, dictionaries, loops, functions, and error handling.

Created and managed simple CSV files for student scores and invoice records.

Worked with data filtering and conditional logic using Python structures.

Focused on writing clean, reusable functions and handling exceptions gracefully.





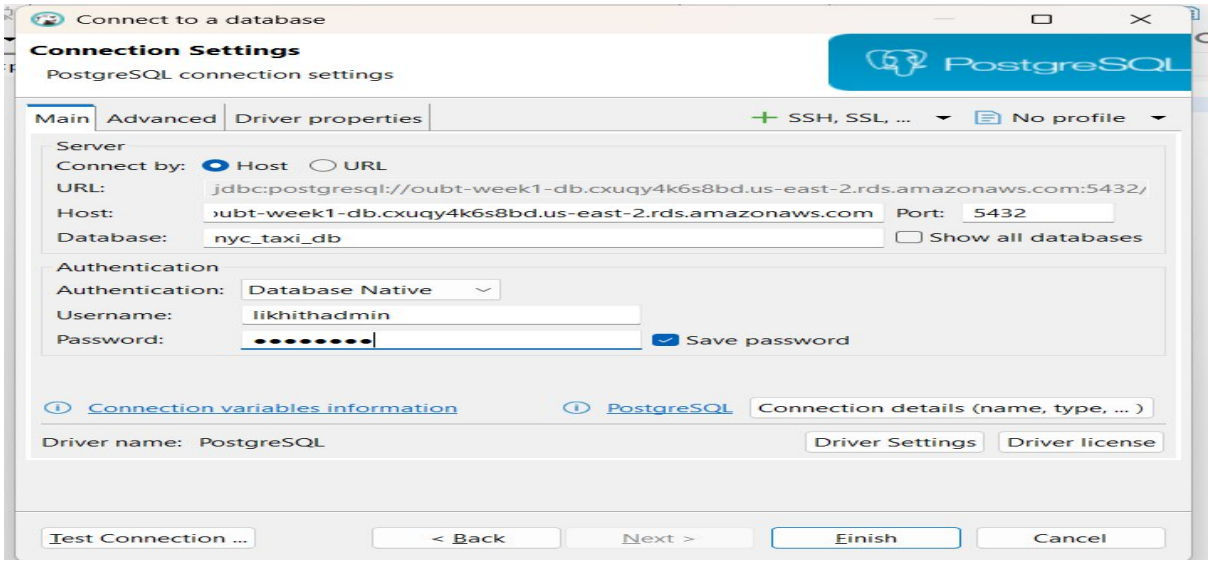
2. Pandas Practice

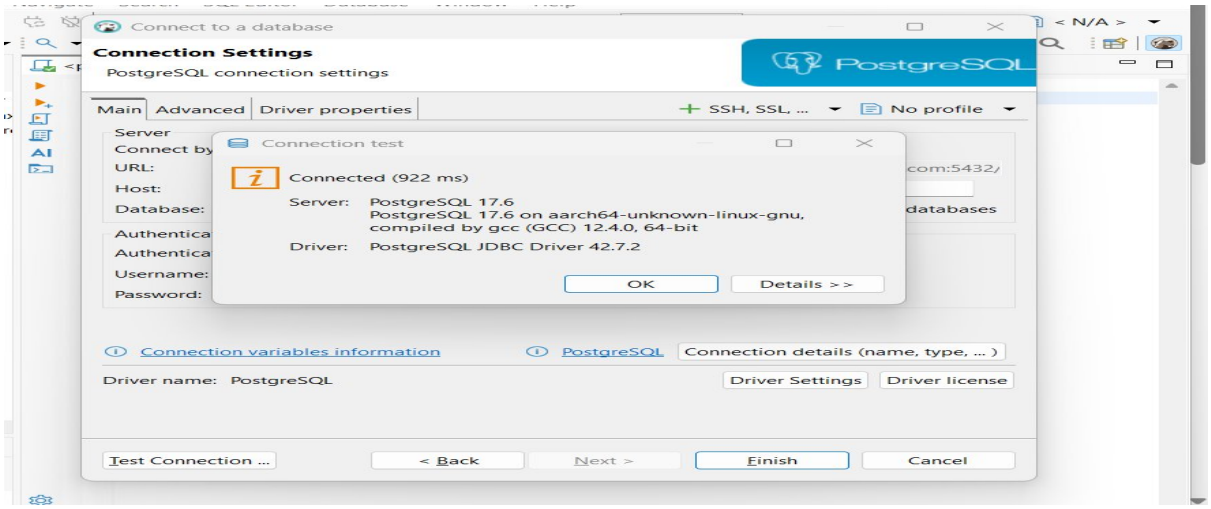
Learned to create, read, and manipulate datasets using Pandas.

Performed basic data cleaning by removing duplicates and adding calculated columns.

Aggregated and analyzed data to find total sales and quantities.

Combined multiple datasets into a single, cleaned file for further use.





```python
#Dataframes - merging

invoice_data = [
    {"InvoiceId" : "I101","Customer":"Likhith"},
    {"InvoiceId" :"I102","Customer":"Sasank"}
]

with open("data/raw/invoice.json","w") as f:
    json.dump(invoice_data,f)


invoice_df = pd.read_json("data/raw/invoice.json")

print("Invoice Data : ")
print(invoice_data,"\n")


merged_df = pd.merge(df,invoice_df,on="Customer",how="left")
print("Merged Sales and Invoice Data:")
print(merged_df.head(), "\n")

merged_df.to_csv("data/processed/cleaned_sales_data.csv",index=False)


print("Cleaned data saved to data/processed/cleaned_sales_data.csv\n")
print("Day 2-3 Pandas Practice Completed Successfully!")

sys.stdout.close()
```

3. AWS S3 (Boto3) Practice

Installed and configured Boto3 to connect Python with AWS S3.

Created and managed an S3 bucket for uploading, listing, and downloading files.

C: > Users > likhi > OUBT-TASKS > WEEK1 > Day2_3_Python_Boto3_Pandas_Likhith > scripts > boto3_s3_handson.py > ...

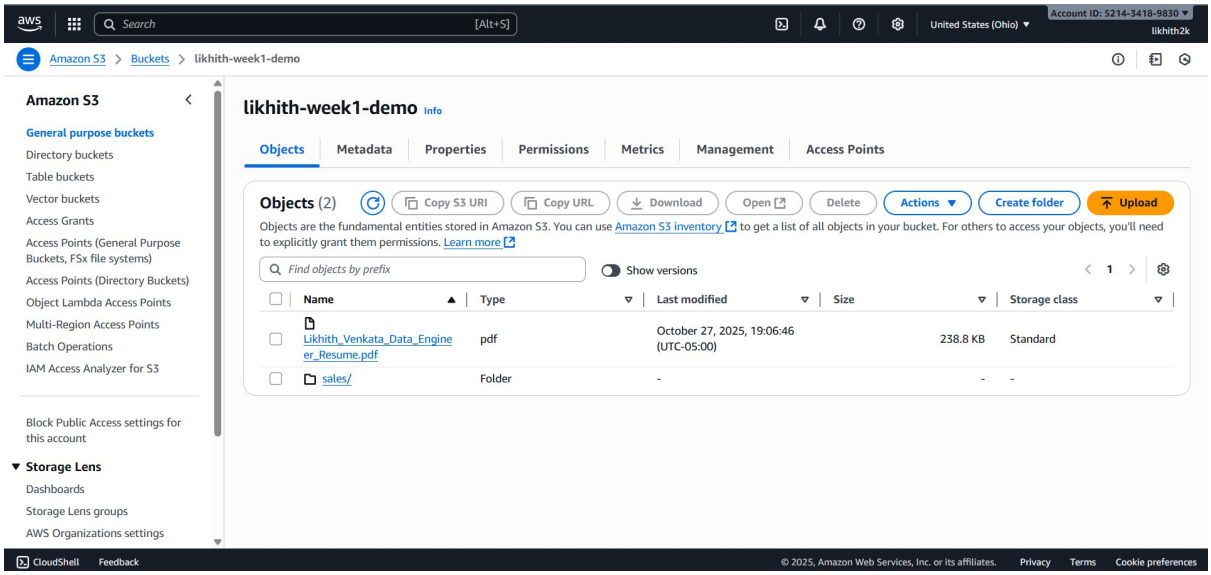
```python
import boto3
import pandas as pd
import sys
import os
from botocore.exceptions import ClientError


os.makedirs("output", exist_ok=True)
sys.stdout = open("output/boto3_s3_output.txt", "w")

s3 = boto3.client("s3", region_name="us-east-1")

bucket_name = "likhith-week1-demo"
file_path = "data/raw/local_sales.csv"
object_name = "sales/local_sales.csv"

sample_df = pd.DataFrame({
    "OrderId":[1,2,3],
    "Customer":["Likhith","Sasank","Siva"],
    "Amount":[120,160,280]
})

sample_df.to_csv(file_path,index=False)

print(f"Created local file: {file_path}\n")
```

Integrated Pandas with AWS S3 to upload processed data and retrieve it for analysis.

```python
#  Here I am sending the files to S3 BUCKET after Creating S3 bucket

try:
    s3.create_bucket(Bucket = bucket_name)
except ClientError as err:
    if(err.response["Error"]["Code"] == "BucketAlreadyOwnedByYou"):
        print(f"Bucket already exists: {bucket_name}")
    else:
        print("Error creating bucket:", err)


try:
    s3.upload_file(file_path,bucket_name,object_name)
    print(f"Uploaded {file_path} to s3://{bucket_name}/{object_name}")
except ClientError as err:
    print("Upload failed: ",err)

print("\n Files in S3 bucket:")
objects = s3.list_objects_v2(Bucket = bucket_name)
if "Contents" in objects:
    for obj in objects["Contents"]:
        print("  .",obj["Key"])
else:
    print(" (Bucket is empty) ")
```

```python
# Here I am downloading from s3
download_path = "data/s3_upload/sales_downloaded.csv"
os.makedirs("data/s3_upload", exist_ok=True)

try:
    s3.download_file(bucket_name,object_name,download_path)
    print(f"\n Downloaded file is in {download_path}")
except ClientError as err:
    print("Download failed:", err)



df = pd.read_csv(download_path)
print("\n Data Loaded Back from S3:")
print(df, "\n")
```

Practiced automating simple data workflows between local and cloud storage.



All Python Scripts outputs have been saved and organized in a separate folder for reference and documentation purposes.

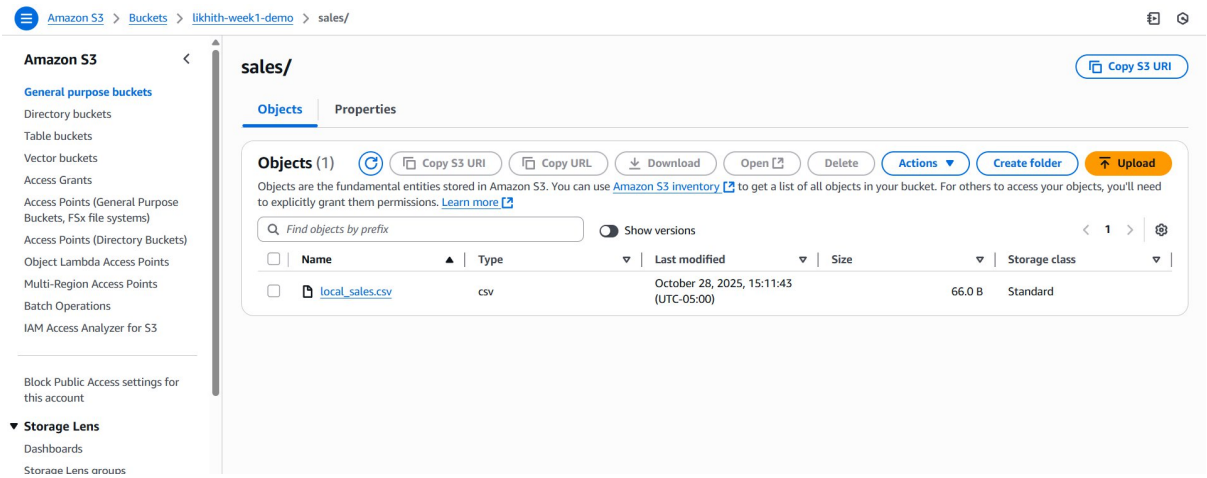
C: > Users > likhi > OUBT-TASKS > WEEK1 > Day2_3_Python_Boto3_Pandas_Likhith > output > boto3_s3_output.txt

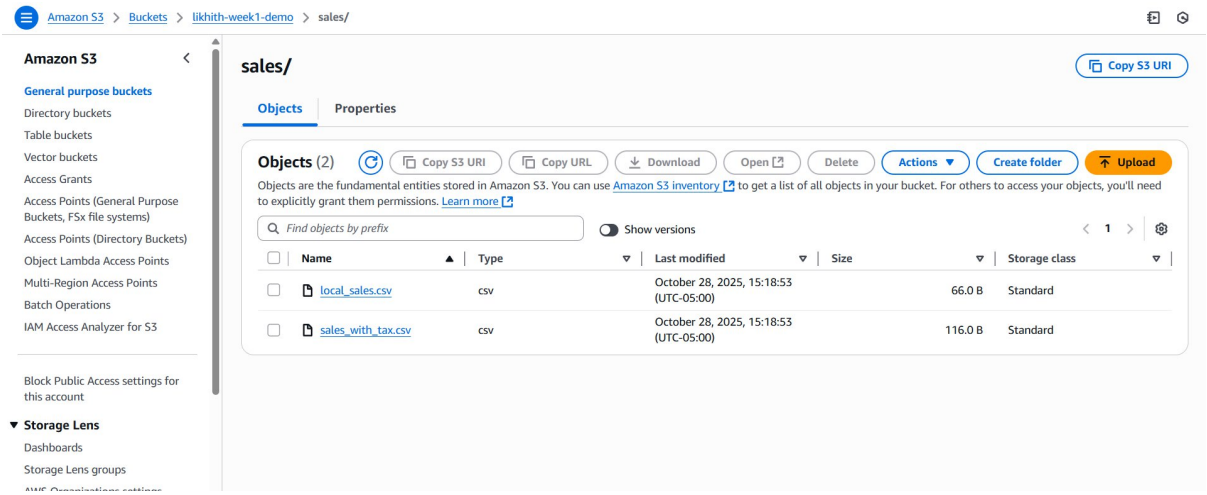
```
Created local file: data/raw/local_sales.csv

Error creating bucket: An error occurred (IllegalLocationConstraintException) when calling the CreateBucket operation: The unspecified location constraint is incompatible
Uploaded data/raw/local_sales.csv to s3://likhith-week1-demo/sales/local_sales.csv

 Files in S3 bucket:
  . Likhith_Venkata_Data_Engineer_Resume.pdf
  . sales/local_sales.csv

 Downloaded file is in data/s3_upload/sales_downloaded.csv

 Data Loaded Back from S3:
   OrderId Customer  Amount
0        1  Likhith     120
1        2   Sasank     160
2        3     Siva     280

After doing Tax calculation:
   OrderId Customer  Amount   Tax  TotalWithTax
0        1  Likhith     120  12.0         132.0
1        2   Sasank     160  16.0         176.0
2        3     Siva     280  28.0         308.0

Final data saved locally in data/s3_upload/sales_with_tax.csv
Uploaded final version to  s3://likhith-week1-demo/sales/sales_with_tax.csv

```

C: > Users > likhi > OUBT-TASKS > **WEEK1** > Day2_3_Python_Boto3_Pandas_Likhith > output > pandas_basics_output.txt

```
OUBT Week 1 - Day 2-3: Python for Data Engineering
Topic: Pandas for Data Manipulation

Sales Data:    orderId   Customer Product  Quantity  Price
0      101    Likhith  Laptop         1    800
1      102     Sasank  Mobile         2    500
2      103  Prathyush  Tablet         1    300
3      104    Anushka  Laptop         1    800
4      105     Sasank  Tablet         3    300
5      106    Likhith  Mobile         2    500
6      107  Prathyush  Laptop         1    800
7      108    Anushka  Tablet         2    300
8      109    Likhith  Mobile         1    800
9      110     Sasank  Laptop         1    500

Created sample sales_data.csv

Data read from csv:   orderId   Customer Product  Quantity  Price
0      101    Likhith  Laptop         1    800
1      102     Sasank  Mobile         2    500
2      103  Prathyush  Tablet         1    300
3      104    Anushka  Laptop         1    800
4      105     Sasank  Tablet         3    300
5      106    Likhith  Mobile         2    500
6      107  Prathyush  Laptop         1    800
7      108    Anushka  Tablet         2    300
8      109    Likhith  Mobile         1    800
9      110     Sasank  Laptop         1    500

```

```
Total Revenue per customer:
Customer
Anushka      1400
Likhith      2600
Prathyush    1100
Sasank       2400
Name: total, dtype: int64

Total Quantity per Product:
Product
Laptop    4
Mobile    5
Tablet    6
Name: Quantity, dtype: int64

Invoice Data :
[{'InvoiceId': 'I101', 'Customer': 'Likhith'}, {'InvoiceId': 'I102', 'Customer': 'Sasank'}]

Merged Sales and Invoice Data:
   orderId   Customer Product  Quantity  Price  total InvoiceId
0      101    Likhith  Laptop         1    800    800      I101
1      102     Sasank  Mobile         2    500   1000      I102
2      103  Prathyush  Tablet         1    300    300       NaN
3      104    Anushka  Laptop         1    800    800       NaN
4      105     Sasank  Tablet         3    300    900      I102

Cleaned data saved to data/processed/cleaned_sales_data.csv

Day 2 3 Pandas Practice Completed Successfully!
```





4. Summary

During these sessions, I improved my understanding of Python and file handling, worked on data cleaning and analysis using Pandas, and practiced using AWS S3 with Boto3 for cloud data management. I also organized my work with a proper folder structure, keeping raw, processed, and uploaded data separate, and saving all outputs and screenshots in their own folders. This helped me follow good project practices and build a smooth workflow that connects Python, Pandas, and AWS for managing and analyzing data.

OUBT Week-1 – Day 4: SQL & Data Modeling basics

NAME: LIKHITH SASANK UPPALAPATI VENKATA

Overview:
Day 4 focused on understanding database fundamentals and data modeling concepts. I learned about SQL basics, relational vs. non-relational databases, ACID properties, normalization, and schema evolution. I also studied how to design a proper data model using entities, attributes, and relationships, and explored star schema and Slowly Changing Dimensions (SCD Types 1 & 2). Along with this, I practiced SQL operations like joins, aggregations, and window functions on sample data to strengthen query skills before working with AWS RDS.

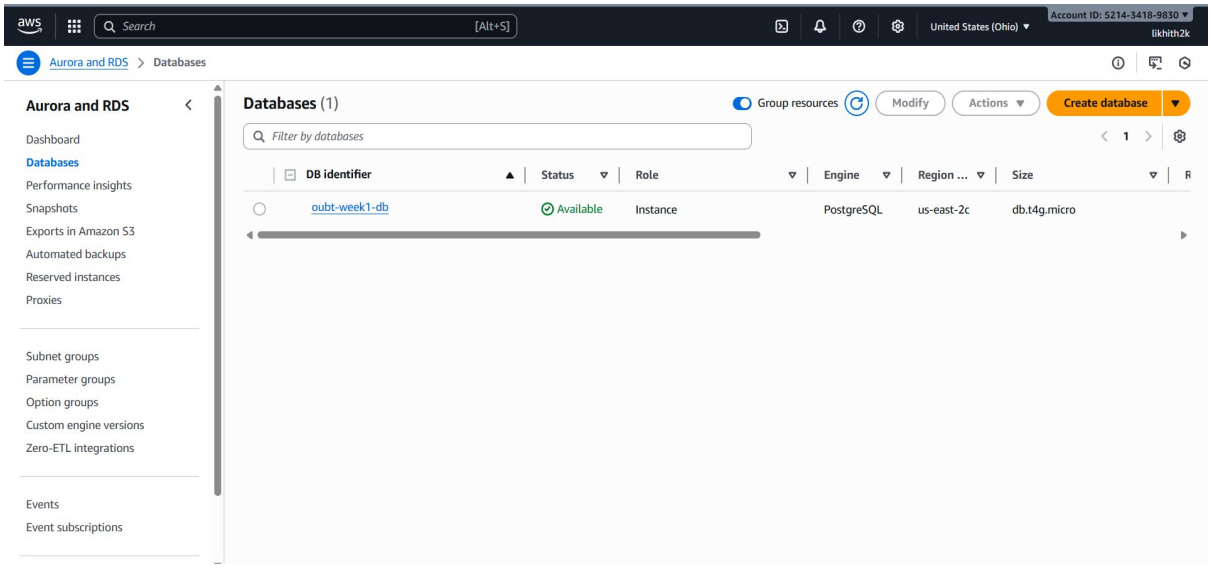
Hands-On Practice:

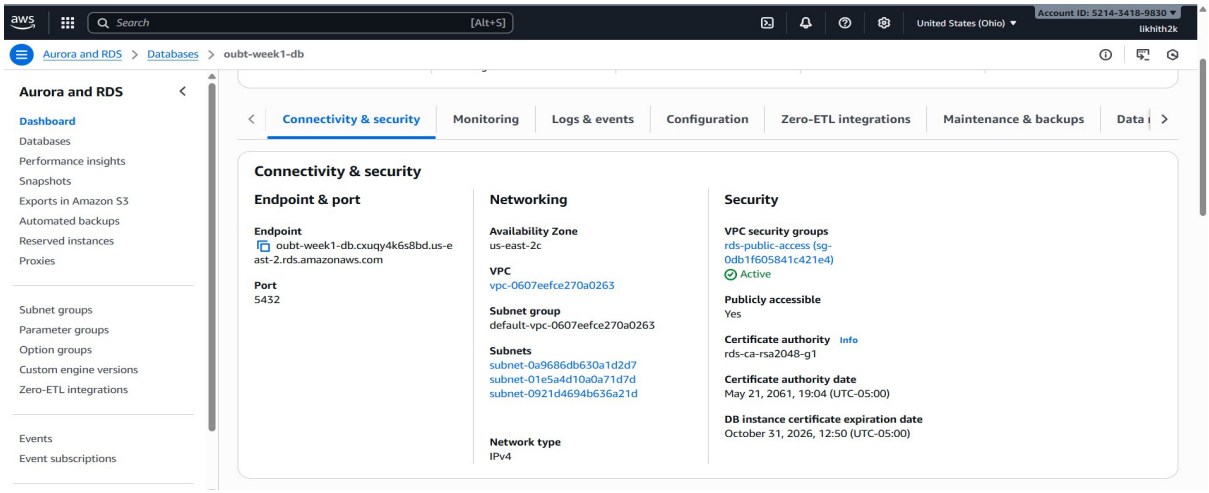
Created sample tables and inserted data for customers, products, and orders.

Practiced SQL operations such as:

Basic filtering using WHERE.
Aggregations with COUNT(), SUM(), and AVG().
Joins between multiple tables to fetch related information.
Grouping and ordering results.
Used the RANK() window function to rank customers by total spending.

Designed the schema in a **star-schema pattern** with Orders and OrderDetails as fact tables and Customers, Products as dimensions.





SQL – QUERIES:

SELECT * FROM Customers WHERE Country = 'USA';

-- List all products under $500

SELECT ProductName, Price FROM Products WHERE Price < 500;

-- Count total number of orders

SELECT COUNT(*) AS TotalOrders FROM Orders;

-- Get all orders for customer 'Likhith'

SELECT o.OrderID, o.OrderDate, o.TotalAmount FROM Orders o JOIN Customers c ON o.CustomerID = c.CustomerID

WHERE c.Name = 'Likhith';

```
-- Total spent by each customer
SELECT c.Name, SUM(o.TotalAmount) AS TotalSpent
FROM Customers c
JOIN Orders o ON c.CustomerID = o.CustomerID
GROUP BY c.Name;

-- List all orders with product names
SELECT o.OrderID, c.Name AS CustomerName, p.ProductName, od.Quantity
FROM Orders o
JOIN Customers c ON o.CustomerID = c.CustomerID
JOIN OrderDetails od ON o.OrderID = od.OrderID
JOIN Products p ON od.ProductID = p.ProductID;

-- Find top 2 customers by spending
SELECT c.Name, SUM(o.TotalAmount) AS TotalSpent
FROM Customers c
JOIN Orders o ON c.CustomerID = o.CustomerID
GROUP BY c.Name
ORDER BY TotalSpent DESC
LIMIT 2;

-- Find products never ordered
SELECT p.ProductName
FROM Products p
LEFT JOIN OrderDetails od ON p.ProductID = od.ProductID
WHERE od.ProductID IS NULL;
```

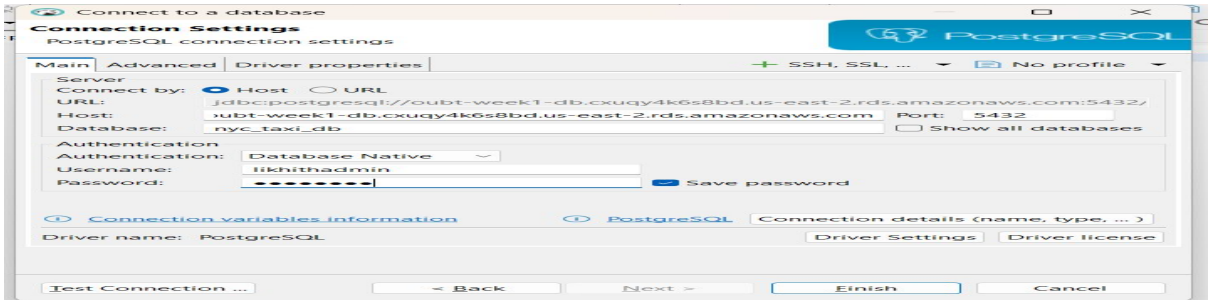
-- Average order amount per country

```
SELECT c.Country, AVG(o.TotalAmount) AS AvgOrderAmount
FROM Customers c
JOIN Orders o ON c.CustomerID = o.CustomerID
GROUP BY c.Country;
```

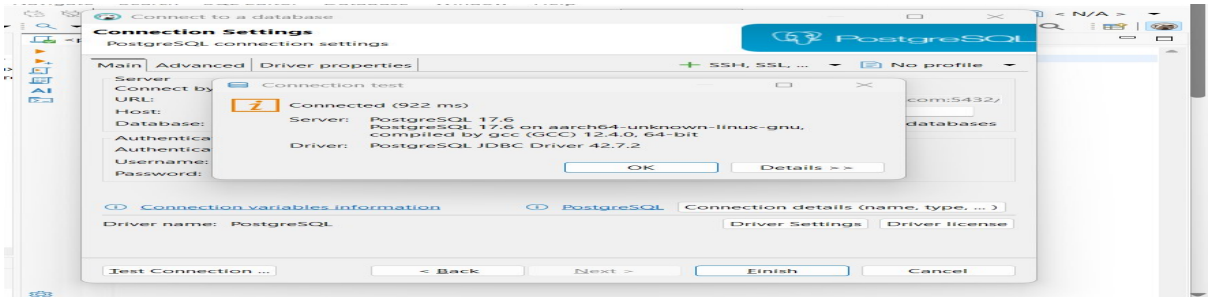
SQL Query results Screenshots:

Query 1: Fetch all customers from the USA

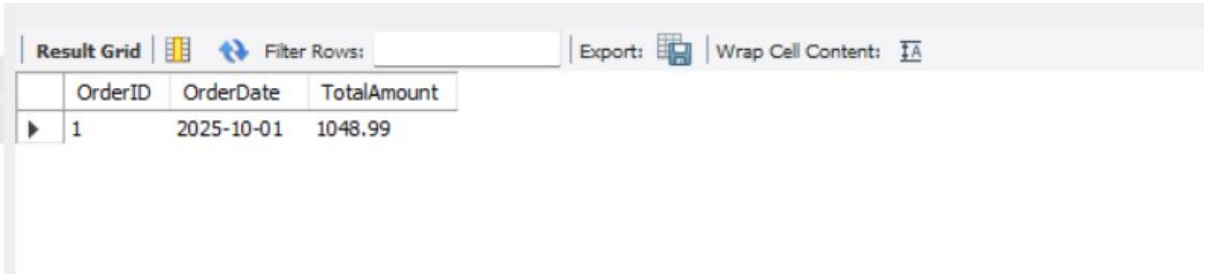
Query 2: List all products priced under $500



Query 3: Count total number of orders



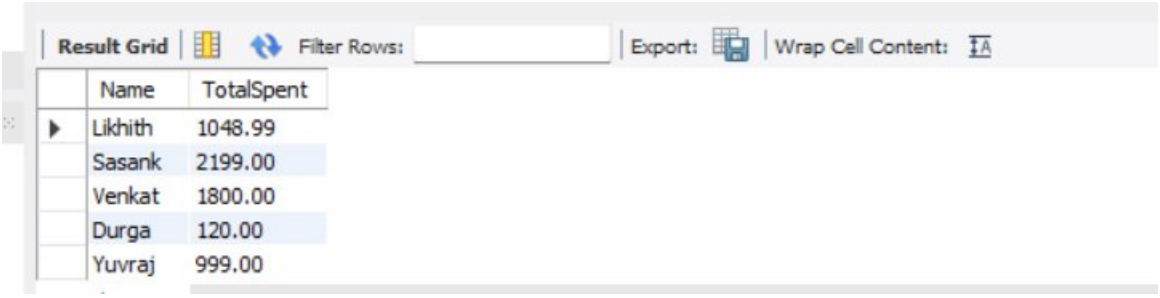
Query 4: Retrieve all orders made by customer “Likhith”



Result Grid | Filter Rows: | Export: | Wrap Cell Content:

OrderID	OrderDate	TotalAmount
1	2025-10-01	1048.99

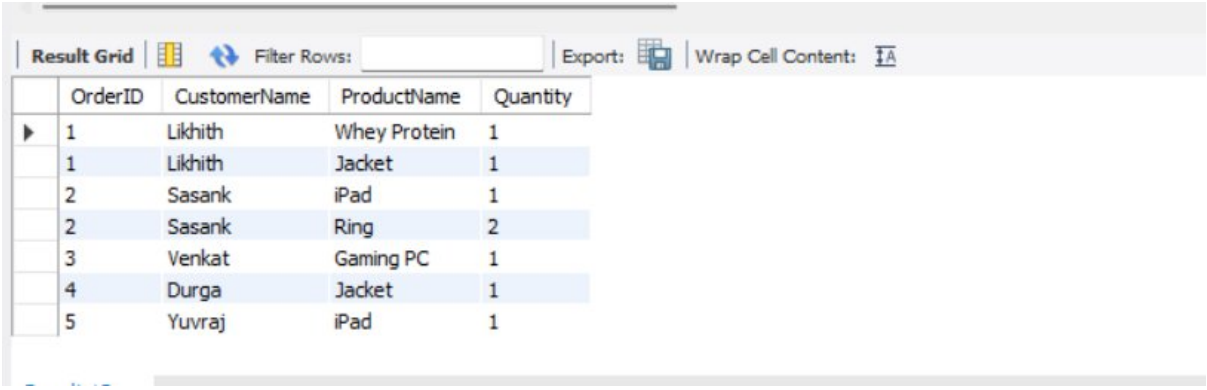
Query 5: Calculate total amount spent by each customer



Result Grid | Filter Rows: | Export: | Wrap Cell Content:

Name	TotalSpent
Likhith	1048.99
Sasank	2199.00
Venkat	1800.00
Durga	120.00
Yuvraj	999.00

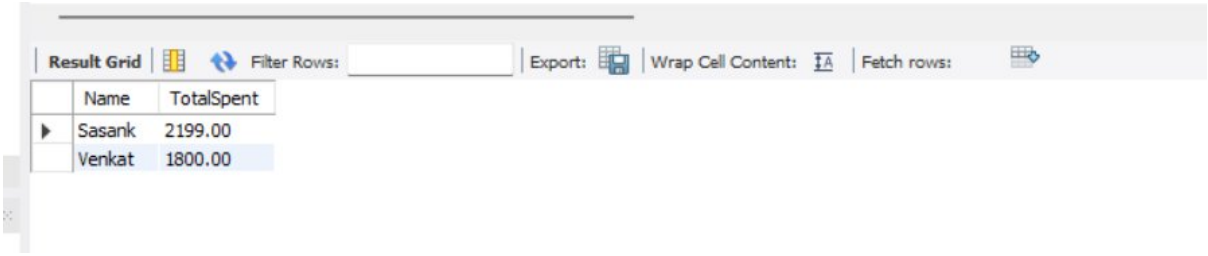
Query 6: Display all orders with product names and quantities



Result Grid | Filter Rows: | Export: | Wrap Cell Content:

OrderID	CustomerName	ProductName	Quantity
1	Likhith	Whey Protein	1
1	Likhith	Jacket	1
2	Sasank	iPad	1
2	Sasank	Ring	2
3	Venkat	Gaming PC	1
4	Durga	Jacket	1
5	Yuvraj	iPad	1

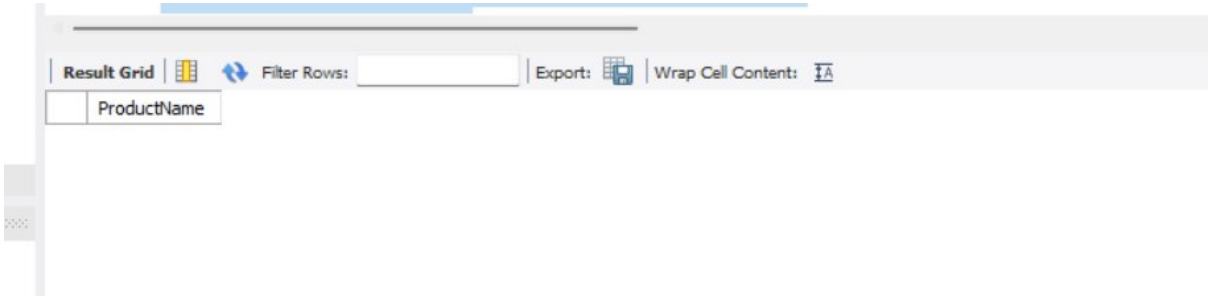
Query 7: Find top 2 customers based on total spending



Result Grid | Filter Rows: | Export: | Wrap Cell Content: | Fetch rows:

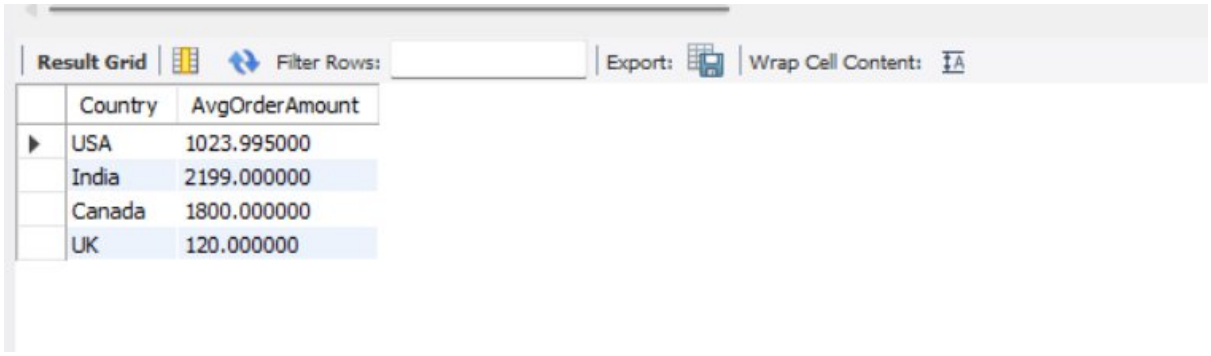
Name	TotalSpent
Sasank	2199.00
Venkat	1800.00

Query 8: Identify products that were never ordered



Query 9: Find average order amount by country

```
4      --  Average order amount per country
5 •    SELECT c.Country, AVG(o.TotalAmount) AS AvgOrderAmount
6      FROM Customers c
7      JOIN Orders o ON c.CustomerID = o.CustomerID
8      GROUP BY c.Country;
9
0
```



Country	AvgOrderAmount
USA	1023.995000
India	2199.000000
Canada	1800.000000
UK	120.000000

Key Learnings:

Learned how relational databases store and manage structured data.

Understood **ACID properties**, normalization, and schema design concepts.

Practiced real SQL operations – joins, aggregations, and ranking queries.

Learned how **star schema modeling** supports analytical queries.

Improved understanding of how to represent one-to-many and many-to-many relationships.

Summary:

Overall, Day 4 helped me connect theoretical database concepts with real SQL practice. Building the e-commerce schema gave me a clear idea of how relational data is structured, queried, and optimized. This day built a strong base for moving into AWS RDS and advanced data modeling in the next steps.

OUBT Week-1 – Day 5: SQL & Data Modeling (AWS RDS)

NAME : LIKHITH SASANK UPPALAPATI VENKATA

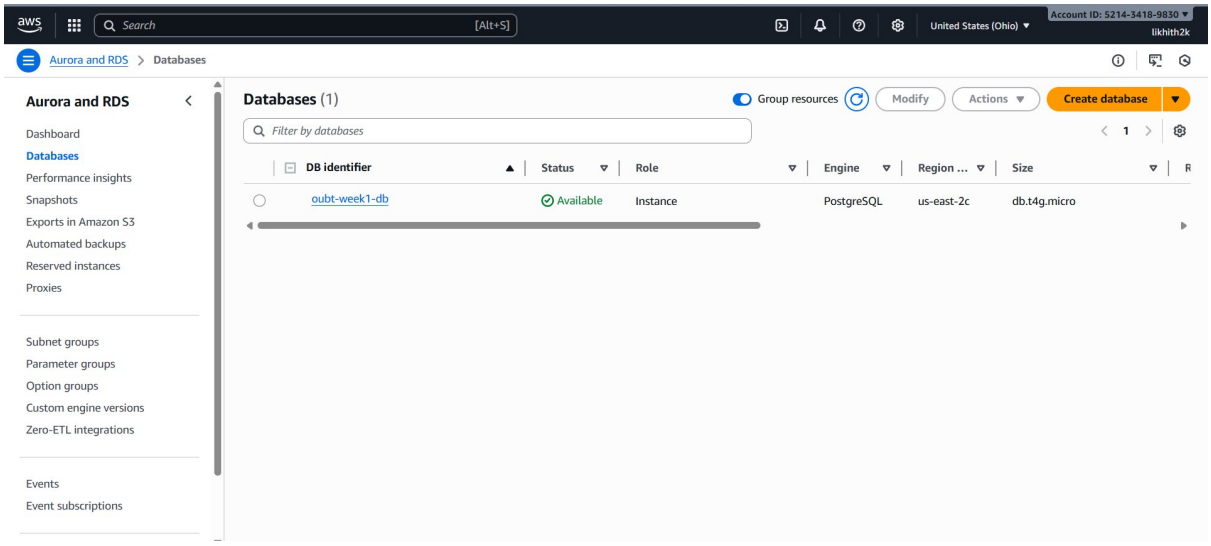
Overview:

Day 5 focused on setting up and working with a live PostgreSQL database on AWS RDS. I created the instance, configured IAM access, and connected through DBeaver to build tables and run SQL queries on sample taxi data. I also checked RDS metrics in CloudWatch to confirm query activity and understand how CPU and connections change when the database is in use. This helped me connect SQL practice with real AWS database monitoring and management

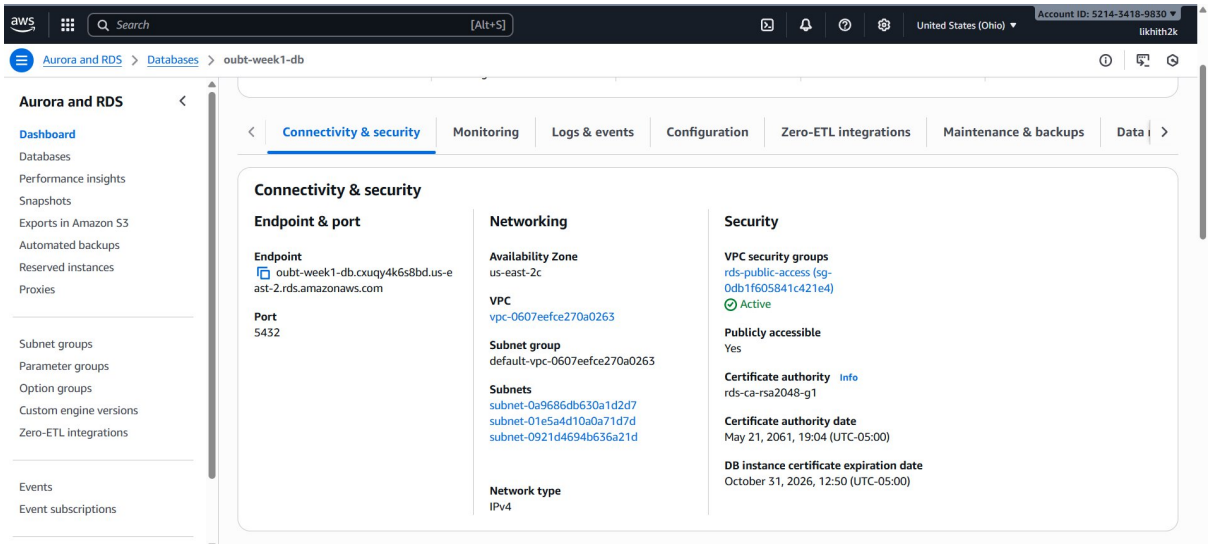
Objective:
Design a simple relational schema on AWS RDS PostgreSQL and practice SQL operations on sample NYC Taxi Trips data.

Steps Performed:

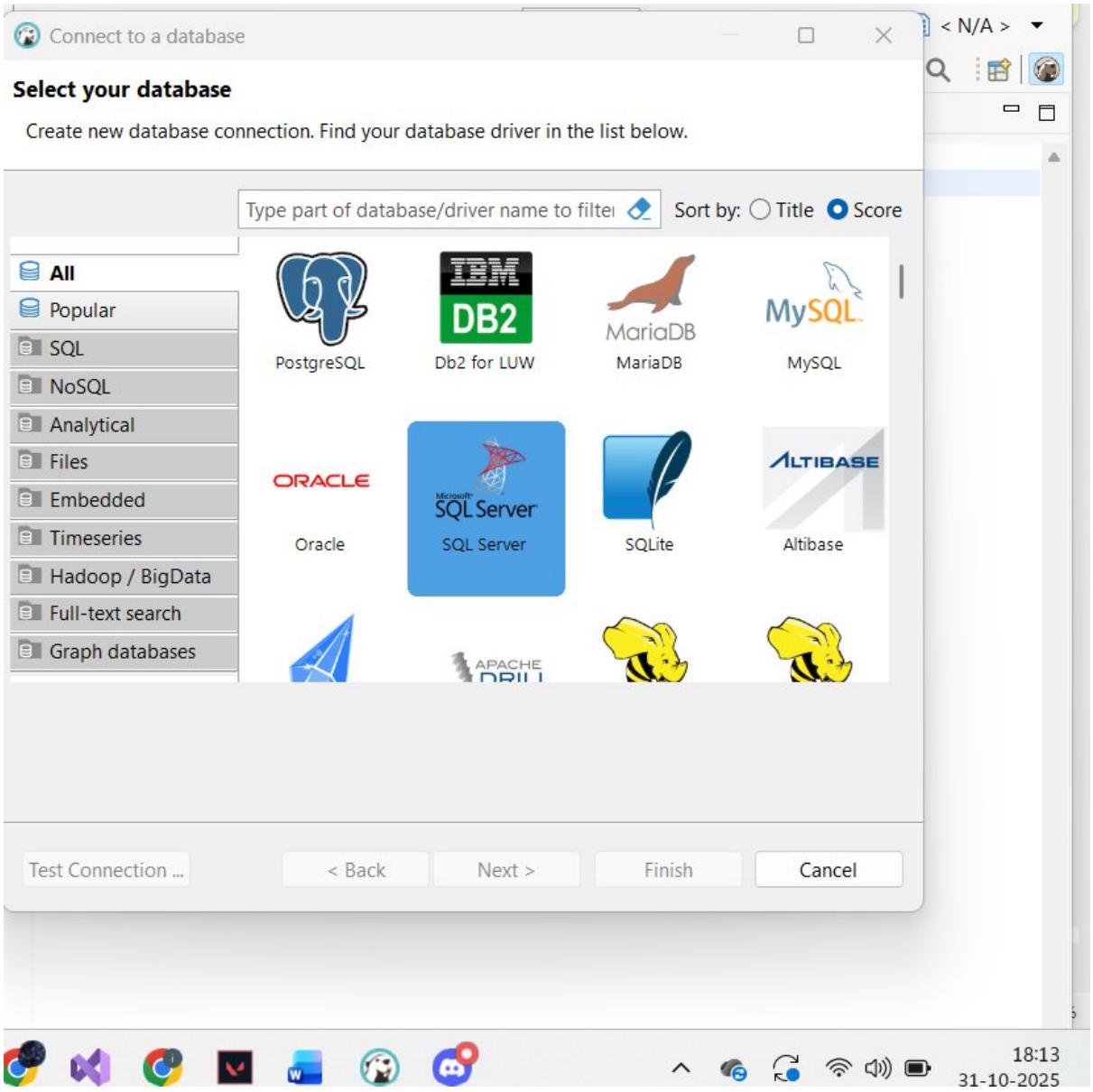
Created an **AWS RDS PostgreSQL** instance under Free Tier with public access enabled.

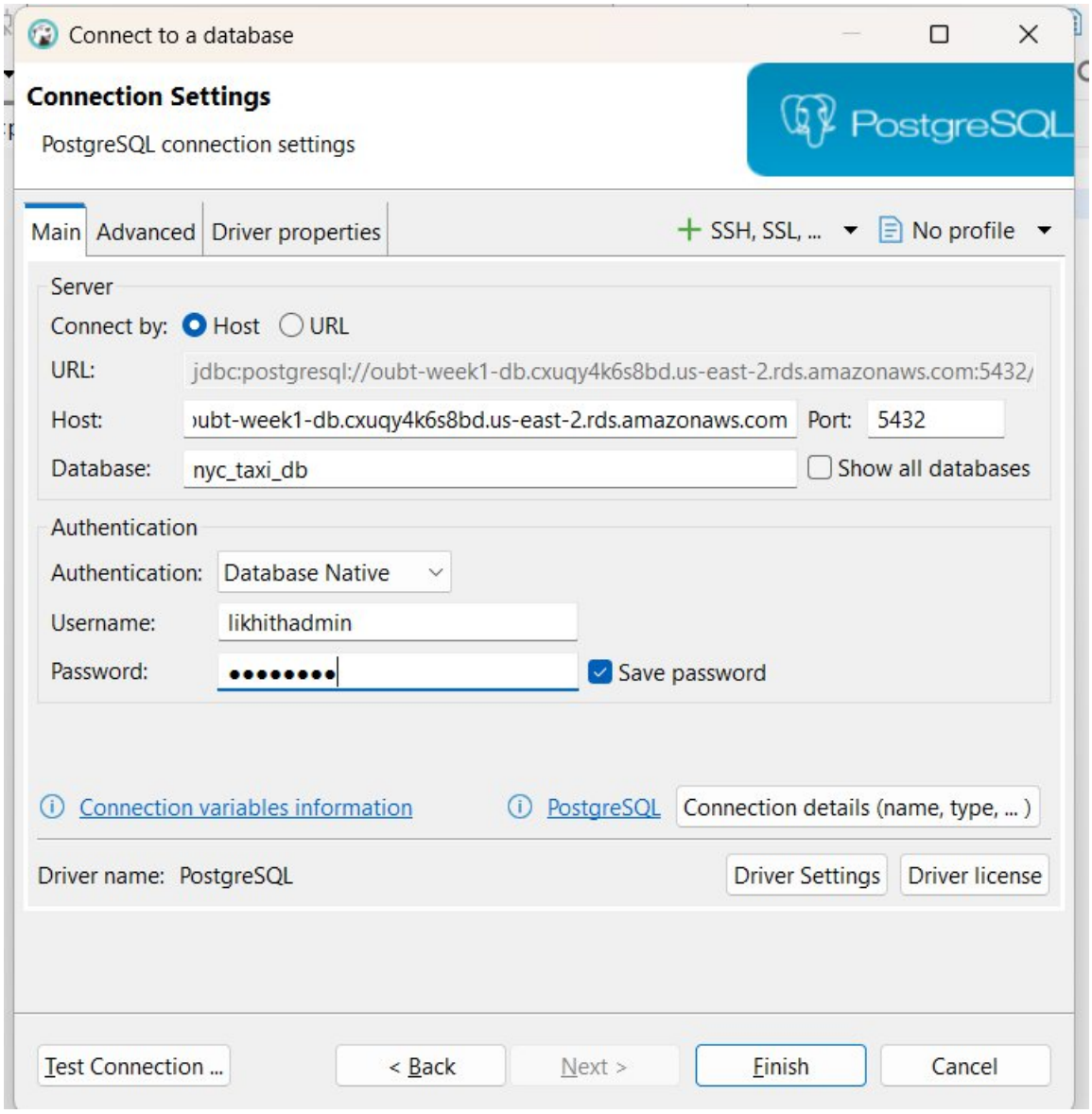


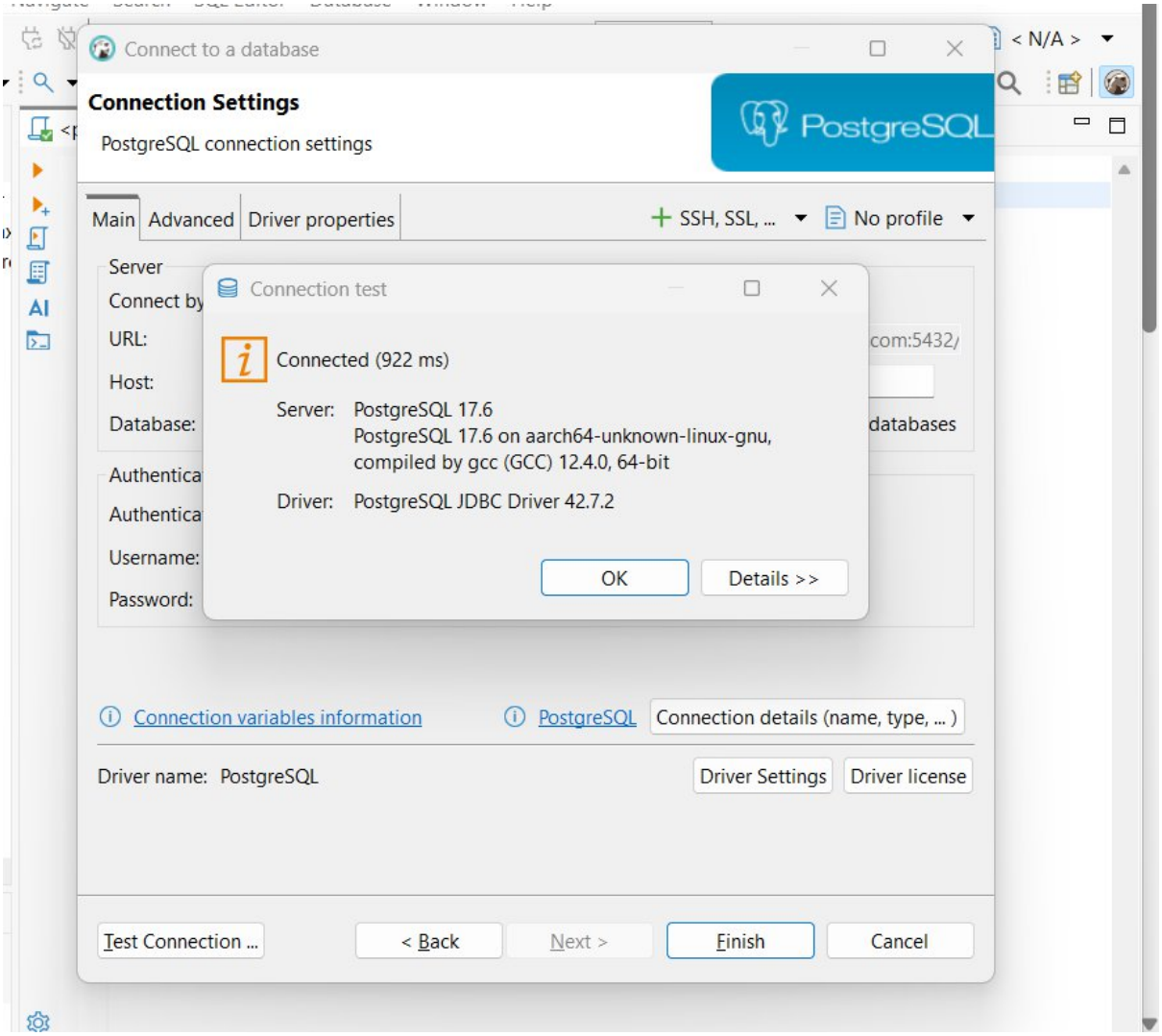
Configured inbound rules (port 5432) to allow access from local machine.



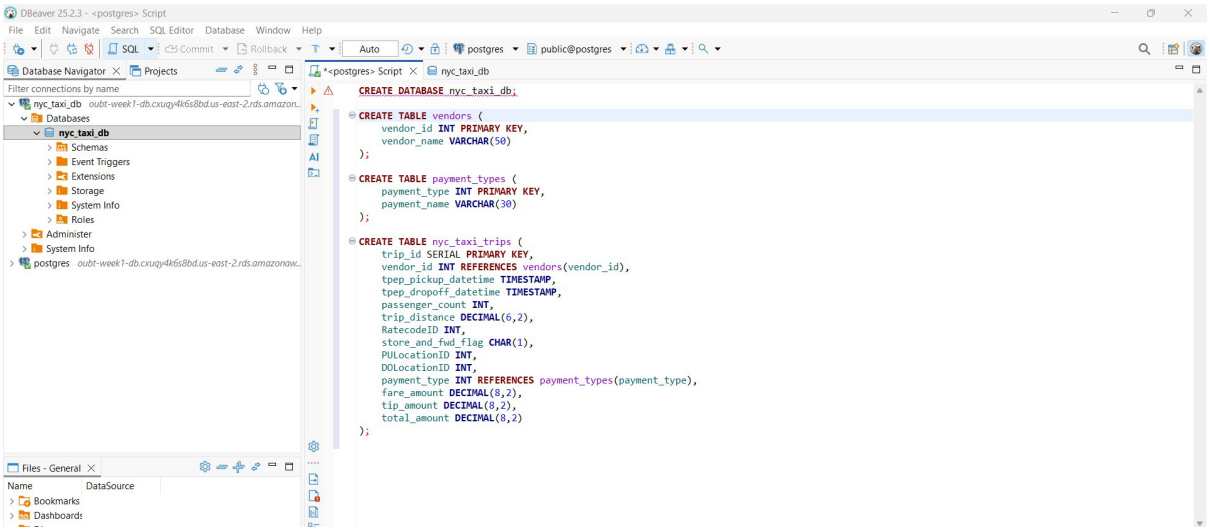
Connected to the RDS instance through **DBeaver** using the database endpoint.



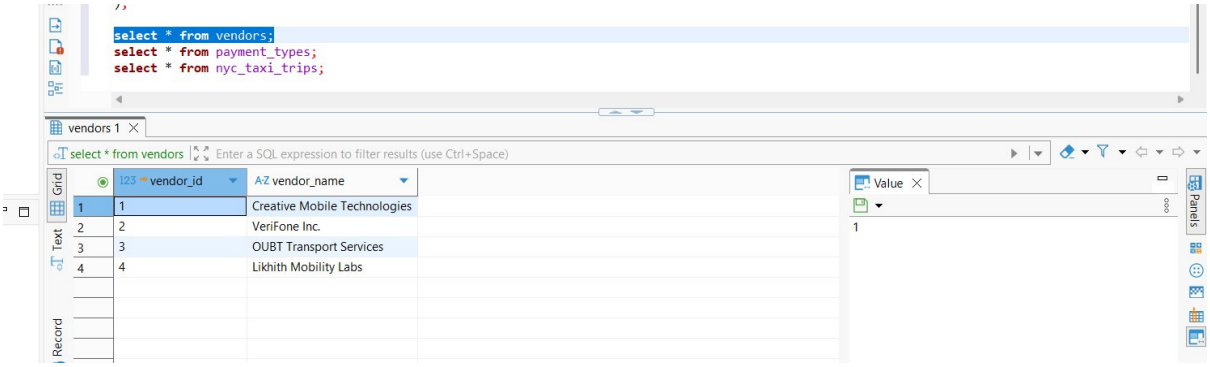

Connect to a database
Connection Settings
PostgreSQL connection settings
PostgreSQL
Main
Advanced
Driver properties
SSH, SSL, ...
No profile
Server
Connect by: Host URL
URL: jdbc:postgresql://oubt-week1-db.cxuqy4k6s8bd.us-east-2.rds.amazonaws.com:5432/
Host: ubt-week1-db.cxuqy4k6s8bd.us-east-2.rds.amazonaws.com
Port: 5432
Database: nyc_taxi_db
Show all databases
Authentication
Authentication: Database Native
Username: likhithadmin
Password: ••••••••
Save password
Connection variables information
PostgreSQL
Connection details (name, type, ...)
Driver name: PostgreSQL
Driver Settings
Driver license
Test Connection ...
< Back
Next >
Finish
Cancel

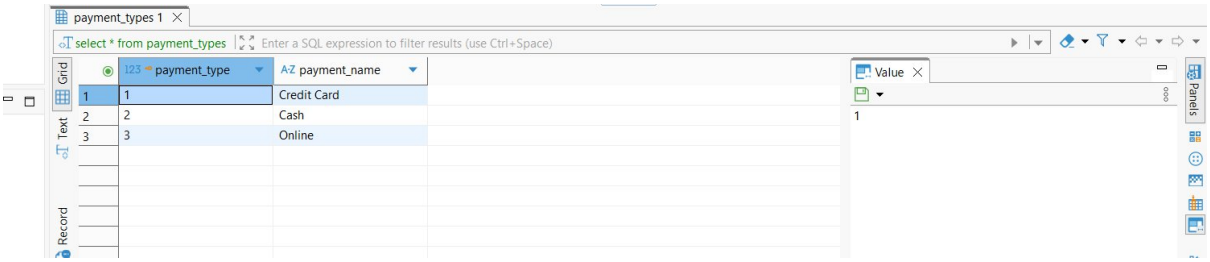


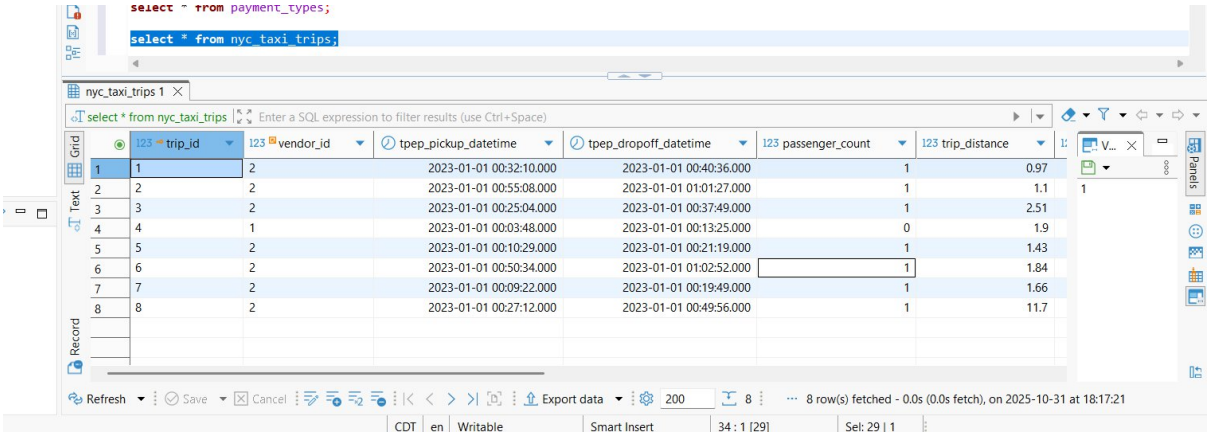
Created a new database nyc_taxi_db.



Designed schema with three tables — vendors, payment_types, and nyc_taxi_trips — following a **Star Schema** model.







Inserted sample trip data and reference data for vendors and payment types.

Executed multiple **SQL queries** covering JOIN, GROUP BY, aggregations, filtering, window functions, and CTEs to analyze trip records.

```
SELECT v.vendor_name,
COUNT(*) AS total_trips,
SUM(fare_amount) AS total_fare
    FROM nyc_taxi_trips t
JOIN vendors v ON t.vendor_id = v.vendor_id
GROUP BY v.vendor_name;
```

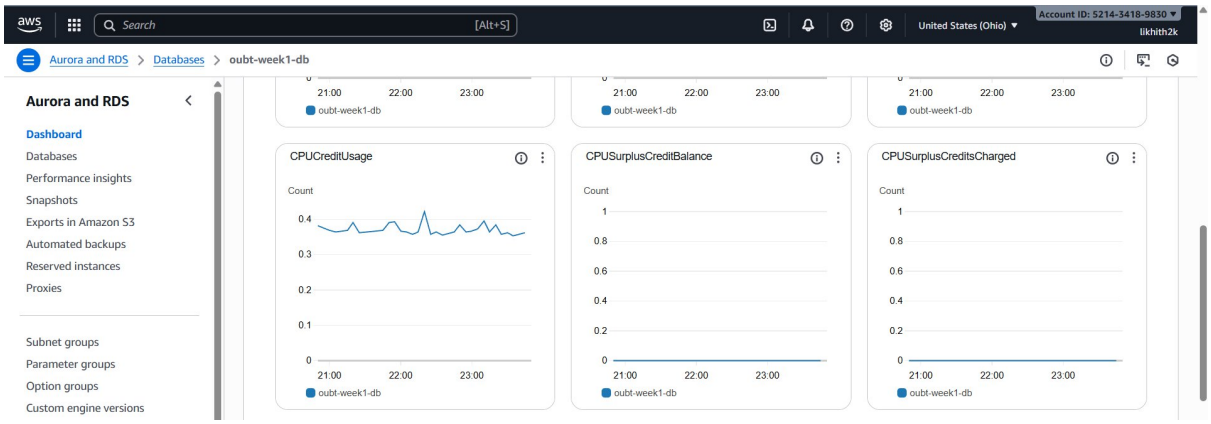
```sql
SELECT p.payment_name,
    ROUND(AVG(fare_amount),2) AS avg_fare
FROM nyc_taxi_trips t
JOIN payment_types p ON t.payment_type = p.payment_type
GROUP BY p.payment_name;


SELECT trip_id, vendor_id, fare_amount
FROM nyc_taxi_trips
ORDER BY fare_amount DESC
LIMIT 3;

SELECT trip_id, vendor_id, fare_amount,
RANK() OVER(ORDER BY fare_amount DESC) AS fare_rank
FROM nyc_taxi_trips;
```

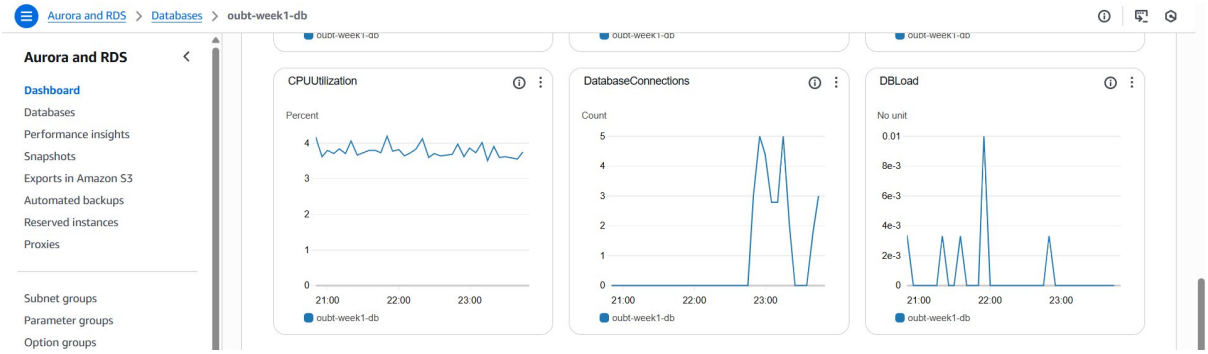
Verified query results and captured screenshots of successful execution in DBeave

CloudWatch & RDS Monitoring Test

To verify my AWS RDS instance activity, I ran multiple SQL queries from DBeaver and monitored performance in the RDS dashboard. The CloudWatch graphs showed clear spikes in **CPU utilization**, **DB load**, and **active connections** while queries were executing. This confirmed that my PostgreSQL instance was processing live workloads.

This exercise helped me understand how CloudWatch metrics can be used to monitor database health and query performance in real time.





Key Learnings:

Gained hands-on experience in **AWS RDS setup**, database connectivity, schema creation, and SQL querying through DBeaver.

Configured **IAM permissions and security groups** to enable secure database access.

Understood how **Star Schema modeling** supports analytical queries and relational design.

Practiced organizing and structuring data for better readability, consistency, and performance.