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BIDD330  
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Module 03 Writeup - Intro to Jupyter Notebook

## Overview

This week, we focused on establishing and manipulating databases directly from Jupyter Notebook. We managed databases hosted on a class-linked server, involving data imports, procedural queries, and database exports.

## Development Environment Setup

To streamline our data analysis and database management tasks, we utilized Jupyter Notebook for its robust, interactive computing capabilities. Below are the steps and configurations implemented:

### Imports:

We started by importing necessary Python libraries to facilitate database connectivity, data manipulation, and visualization:

- pyodbc: For connecting to the SQL Server.
- pandas: Essential for data manipulation and querying in a tabular format.
- SQLAlchemy: Utilized for database connection pooling and providing a higher abstraction for database manipulations.
- matplotlib.pyplot: For plotting graphs directly from the data.
- warnings: To suppress deprecation warnings that could clutter the notebook's output.

### Database Connections:

Two methods were used to connect to our SQL Server, ensuring robust access and query capabilities:

- Direct Connection using pyodbc: This traditional method provides detailed control over database operations.
- Enhanced Connection using SQLAlchemy: Offers a more flexible and powerful approach to manage SQL operations and transactions.

### Imports

```
[1]: import pyodbc
import pandas as pd
from sqlalchemy import create_engine, text
import matplotlib.pyplot as plt
import warnings
warnings.filterwarnings("ignore", category=DeprecationWarning)
```

## Connections

Note: Connect via BigIP and SQL Server first

```
[11]: # Connect using pyodbc
conn = pyodbc.connect('DRIVER={SQL Server};SERVER=uwc-studentsql.continuum.uw.edu\\uwcbidsql;DATABASE=Crimson_Katharine;')

[13]: # Connect using SQLAlchemy
engine = create_engine('mssql+pyodbc://@uwc-studentsql.continuum.uw.edu\\uwcbidsql/Crimson_Katharine?driver=SQL+Server')

# Example to test the connection and execute a query
with engine.connect() as connection:
    # Wrap the SQL command in text() for explicit execution
    result = connection.execute(text("SELECT TOP 10 * FROM [dbo].[FactUnemployment]"))
    for row in result:
        print(row) # This will print the rows from the FactUnemployment table

(1, 'Alabama', '2020-01-04', 4578, '2019-12-28', 18523, 1923741, 0.96)
(2, 'Alabama', '2020-01-11', 3629, '2020-01-04', 21143, 1923741, 1.1)
(3, 'Alabama', '2020-01-18', 2483, '2020-01-11', 17402, 1923741, 0.9)
(4, 'Alabama', '2020-01-25', 2129, '2020-01-18', 18390, 1923741, 0.96)
(5, 'Alabama', '2020-02-01', 2170, '2020-01-25', 17284, 1923741, 0.9)
(6, 'Alabama', '2020-02-08', 2176, '2020-02-01', 16745, 1923741, 0.87)
(7, 'Alabama', '2020-02-15', 1981, '2020-02-08', 16571, 1923741, 0.86)
(8, 'Alabama', '2020-02-22', 1735, '2020-02-15', 16059, 1923741, 0.83)
(9, 'Alabama', '2020-02-29', 1575, '2020-02-22', 14721, 1923741, 0.77)
(10, 'Alabama', '2020-03-07', 1663, '2020-02-29', 13657, 1923741, 0.71)
```

## Query Execution and Dataframe Creation:

Utilizing both connection methods, I executed queries to fetch data from `FactUnemployment` and `FactCovid` tables. This was done using both raw SQL execution via pyodbc and pandas integration with SQLAlchemy.

### SQL Queries

#### SQL Query - FactUnemployment

```
[20]: cursor = conn.cursor()
cursor.execute("SELECT TOP 10 * FROM [dbo].[FactUnemployment]")
for i in cursor:
    print(i)

(1, 'Alabama', '2020-01-04', 4578, '2019-12-28', 18523, 1923741, 0.96, 'USA')
(2, 'Alabama', '2020-01-11', 3629, '2020-01-04', 21143, 1923741, 1.1, 'USA')
(3, 'Alabama', '2020-01-18', 2483, '2020-01-11', 17402, 1923741, 0.9, 'USA')
(4, 'Alabama', '2020-01-25', 2129, '2020-01-18', 18390, 1923741, 0.96, 'USA')
(5, 'Alabama', '2020-02-01', 2170, '2020-01-25', 17284, 1923741, 0.9, 'USA')
(6, 'Alabama', '2020-02-08', 2176, '2020-02-01', 16745, 1923741, 0.87, 'USA')
(7, 'Alabama', '2020-02-15', 1981, '2020-02-08', 16571, 1923741, 0.86, 'USA')
(8, 'Alabama', '2020-02-22', 1735, '2020-02-15', 16059, 1923741, 0.83, 'USA')
(9, 'Alabama', '2020-02-29', 1575, '2020-02-22', 14721, 1923741, 0.77, 'USA')
(10, 'Alabama', '2020-03-07', 1663, '2020-02-29', 13657, 1923741, 0.71, 'USA')
```

#### SQL Query - FactCovid

```
[21]: cursor = conn.cursor()
cursor.execute("SELECT TOP 10 * FROM [dbo].[FactCovid]")
for i in cursor:
    print(i)

(1, 77765279, '2020-10-22', 512, 10, 25, 0, 0, 0, '42.17028', '-76.3063', 'US', 'USA', 'United States', 'New York', 'US-NY', 'Tioga County')
(2, 78096938, '2020-10-23', 549, 37, 25, 0, 0, 0, '42.17028', '-76.3063', 'US', 'USA', 'United States', 'New York', 'US-NY', 'Tioga County')
(3, 78440992, '2020-10-24', 572, 23, 25, 0, 0, 0, '42.17028', '-76.3063', 'US', 'USA', 'United States', 'New York', 'US-NY', 'Tioga County')
(4, 78767346, '2020-10-25', 589, 17, 26, 1, 0, 0, '42.17028', '-76.3063', 'US', 'USA', 'United States', 'New York', 'US-NY', 'Tioga County')
(5, 79068611, '2020-10-26', 610, 21, 26, 0, 0, 0, '42.17028', '-76.3063', 'US', 'USA', 'United States', 'New York', 'US-NY', 'Tioga County')
(6, 79419515, '2020-10-27', 629, 19, 27, 1, 0, 0, '42.17028', '-76.3063', 'US', 'USA', 'United States', 'New York', 'US-NY', 'Tioga County')
(7, 79674733, '2020-10-28', 653, 24, 27, 0, 0, 0, '42.17028', '-76.3063', 'US', 'USA', 'United States', 'New York', 'US-NY', 'Tioga County')
(8, 80116653, '2020-10-29', 672, 19, 30, 3, 0, 0, '42.17028', '-76.3063', 'US', 'USA', 'United States', 'New York', 'US-NY', 'Tioga County')
(9, 80467166, '2020-10-30', 689, 17, 32, 2, 0, 0, '42.17028', '-76.3063', 'US', 'USA', 'United States', 'New York', 'US-NY', 'Tioga County')
(10, 80802061, '2020-10-31', 699, 10, 33, 1, 0, 0, '42.17028', '-76.3063', 'US', 'USA', 'United States', 'New York', 'US-NY', 'Tioga County')
```

Next, I ran a Jupyter Magic command- this was required for the exercise.

## Jupyter Magic Command Demo

```
[22]: !pip install pretty

Requirement already satisfied: pretty in c:\users\test\appdata\local\programs\python\python312\lib\site-packages (0.1)
```

## Setting PANDAS dataframes

### Create PANDAS dataframe - FactUnemployment

```
[24]: dataframeUnemployment = pd.read_sql("SELECT U.[State], SUM(U.[Initial Claims]) [ClaimsYTD] FROM [dbo].[FactUnemployment] U GROUP BY U.[State];", conn)
print(dataframeUnemployment)
```

	State	ClaimsYTD
0	Puerto Rico	1398560
1	Illinois	10768238
2	Oklahoma	2968662
3	Wisconsin	4581194
4	Maine	751146
5	Pennsylvania	9953896
6	Massachusetts	6038466
7	Ohio	10027234
8	Arizona	3378458
9	Oregon	3261200
10	Wyoming	322152
11	Louisiana	3400916
12	Hawaii	1458728
13	New Hampshire	842658
14	Virginia	4275742
15	New York	16341314
16	South Dakota	231972
17	Run Date: 3/22/2024	0
18	Maryland	3529852
19	California	41148682
20	Missouri	3687392
21	Minnesota	4101638
22	District of Columbia	717422

At the end of this query, I encountered this error:

46	Utah	1154140
47	Alabama	3196876
48	Tennessee	3411958
49	Rhode Island	1348646
50	Arkansas	1527088
51	Nevada	2707192
52	Kansas	1839014
53	Kentucky	4133742
54	New Mexico	1165336

```
C:\Users\test\AppData\Local\Temp\ipykernel_10868\1269065339.py:1: UserWarning: pandas only supports SQLAlchemy connectable (engine/connection) or databa
e string URI or sqlite3 DBAPI2 connection. Other DBAPI2 objects are not tested. Please consider using SQLAlchemy.
dataframeUnemployment = pd.read_sql("SELECT U.[State], SUM(U.[Initial Claims]) [ClaimsYTD] FROM [dbo].[FactUnemployment] U GROUP BY U.[State];", conn)
```

So then I recreated the dataframe using SQLAlchemy:

The code block below uses the SQLAlchemy engine to execute a query and load the result into a DataFrame.

```
[19]: # Using SQL Alchemy engine to execute a SQL query and load the result into a DataFrame
sql_query = "SELECT U.[State], SUM(U.[Initial Claims]) AS ClaimsYTD FROM [dbo].[FactUnemployment] U GROUP BY U.[State];"
dataframeUnemployment = pd.read_sql(sql_query, engine)
print(dataframeUnemployment)
```

	State	ClaimsYTD
0	Puerto Rico	699280
1	Illinois	5384119
2	Oklahoma	1484331
3	Wisconsin	2290597
4	Maine	375573
5	Pennsylvania	4976948
6	Massachusetts	3019233
7	Ohio	5013617
8	Arizona	1689229
9	Oregon	1630600
10	Wyoming	161076
11	Louisiana	1700458
12	Hawaii	729364
13	New Hampshire	421329
14	Virginia	2137871
15	New York	8170657
16	South Dakota	115986
17	Run Date: 3/22/2024	0
18	Maryland	1764926
19	California	20574341
20	Missouri	1843696
21	Minnesota	2050819

## Data Visualization:

With the data successfully queried and loaded into pandas DataFrames, I utilized `matplotlib` to create visual representations (not shown in the code snippet but mentioned for completeness).

### Query The Dataframes Using Pandas

#### Query FactUnemployment with Pandas

```
[12]: #Show me the top 3 rows in unemployment table
unemployment_data = (dataframeUnemployment)
unemployment_data.head(3)

[12]:
```

	State	ClaimsYTD
0	Puerto Rico	699280
1	Illinois	5384119
2	Oklahoma	1484331

#### Query FactCovid Table with Pandas

```
[13]: bing = (dataframeFactCovid)
bing.head(3)

[13]:
```

	FactCovid_Key	ID	Updated	Confirmed	Confirmed_Change	Deaths	Deaths_Change	Recovered	Recovered_Change	Latitude	Longitude	Iso2	Iso3	Country_Region	Admin_Region_1	Iso_Subdivision	Admin_Region_2
0	1	77765279	2020-10-22	512	10	25	0	0	0	42.17028	-76.3063	US	USA	United States	New York	US-NY	Tioga County
1	2	78096938	2020-10-23	549	37	25	0	0	0	42.17028	-76.3063	US	USA	United States	New York	US-NY	Tioga County
2	3	78440992	2020-10-24	572	23	25	0	0	0	42.17028	-76.3063	US	USA	United States	New York	US-NY	Tioga County

### Query the Dataframes using SQLAlchemy:

SQL Alchemy does not have a "head" function, and doesn't return a pretty formatted table like PANDAS. But, we can use a combination of SQLAlchemy to query the database, Pandas Dataframe (pd) to read the SQL query, which does return the headers and number lines.

#### FactUnemployment Dataframe Query with SQLAlchemy

```
[25]: # SQL Query to fetch top 3 rows of the FactUnemployment table
sql_query = "SELECT TOP 3 U.[State], SUM(U.[Initial Claims]) AS ClaimsYTD FROM [dbo].[FactUnemployment] U GROUP BY U.[State];"

# Use pandas to load the SQL query result into a Dataframe
dataframeUnemployment = pd.read_sql_query(sql_query, engine)

# Display the Dataframe
print(dataframeUnemployment)

State ClaimsYTD
0 Puerto Rico 699280
1 Illinois 5384119
2 Oklahoma 1484331
```

#### FactCovid Dataframe Query with SQLAlchemy

```
[24]: # SQL Query to fetch top 3 rows
sql_query = "SELECT TOP 3 * FROM [dbo].[FactCovid];"

# Use pandas to load the SQL query result into a Dataframe
dataframeFactCovid = pd.read_sql_query(sql_query, engine)

# Display the Dataframe
print(dataframeFactCovid)
```

	FactCovid_Key	ID	Updated	Confirmed	Confirmed_Change	Deaths	Deaths_Change	Recovered	Recovered_Change	Latitude	Longitude	Iso2	Iso3	Country_Region	Admin_Region_1	Iso_Subdivision	Admin_Region_2
0	1	77765279	2020-10-22	512	10	25	0	0	0	42.17028	-76.3063	US	USA	United States	New York	US-NY	Tioga County
1	2	78096938	2020-10-23	549	37	25	0	0	0	42.17028	-76.3063	US	USA	United States	New York	US-NY	Tioga County
2	3	78440992	2020-10-24	572	23	25	0	0	0	42.17028	-76.3063	US	USA	United States	New York	US-NY	Tioga County

## Learning and Utility:

- Jupyter Magic Commands: I explored and applied Jupyter-specific commands (`!pip install package\_name`) demonstrating the notebook's capability to handle package management internally.
- Practical Application: The integration of SQLAlchemy and pandas significantly streamlined the process of querying, which was evident when handling data retrieval and manipulation tasks. This setup not only facilitated a more profound understanding of these tools but also enhanced the workflow efficiency.

## Reflection:

This module enriched my understanding and proficiency with pandas and SQLAlchemy. It highlighted the practical applications and efficiencies these tools bring to data handling and analysis within a Jupyter Notebook environment.