# Lab Assignment 6: Creating and Connecting to Databases

# DS 6001: Practice and Application of Data Science

### Instructions

Please answer the following questions as completely as possible using text, code, and the results of code as needed. Format your answers in a Jupyter notebook. To receive full credit, make sure you address every part of the problem, and make sure your document is formatted in a clean and professional way.

Please note: you will not be able to use Rivanna for this lab as Rivanna is not set up to work with Docker or with Databases. If you need help getting your local system running, please let me know.

# Problem 0 [No points, no need to write anything here for any of the following parts, but do it anyway!]

Databases require a lot of external software. The good news is that there are excellent free and open source options to do very advanced work with databases. The bad news is that each piece of additional software comes with its own complications. This problem will guide you through the installation steps for the software you need to run database systems on your computer, document those databases, and connect to them through Python with (fingers crossed) as few problems as possible.

#### Part a

Use pip to install the following Python packages on your system:

mysql-connector-python
psycopg
pymongo
sqlalchemy
wget

#### Part b

With the exception of SQlite, database systems run as external software that must be installed and run on your computer. To make the installation steps easier, you will need some configuration files that I wrote and saved in a GitHub repository. Open your terminal and use the cd command to navigate to the folder in your computer where you want to work. Then type

```
git clone https://github.com/jkropko/ds6001databases
```

If this command works, it will create a new directory within your current folder called "ds6001databases".

- Check that this folder exists and contains the following files: LICENSE, README.md, compose.yaml, db\_tests.ipynb, and requirements.txt
- Save the notebook file you will be using for your Lab 6 work inside the "ds6001databases" folder

#### Part c

We will be using a system called Docker to work with databases. Docker is the most commonly used platform for working with **containers**. While we will not be delving into the topic of containerization in this course, a container is space in your computer's memory that is set apart from the rest of your computer. We can act as if the container is an entirely new computer, and inside the container we can change the operating system and install other software external to Python, such as database management systems. We can use a container to run Windows on a Mac, or vice versa, or Linux on any system. By far the easiest way to run MySQL, PostgreSQL, and MongoDB is through Docker containers.

You will need to install Docker Desktop on your computer. Go to https://www.docker.com/products/docker-desktop/ and click on the Download button, making sure the operating system listed matches the operating system of your computer.

Once Docker Desktop is installed, find the Docker Desktop program on your computer and run it.

To confirm that Docker Desktop is running, open a terminal and type docker help. If you see documentation that begins

```
Usage: docker [OPTIONS] COMMAND

A self-sufficient runtime for containers
```

then you are all set. If you see an error that Docker is not found, then the Docker Desktop client was not installed properly, so you should try downloading and installing it from the website again. If you receive a message that the Docker daemon is not running, then Docker is installed but is not running. Find the Docker Desktop executable on your computer and click it to get Docker running.

### Part d

Inside your "ds6001databases" folder, create a .env file. On a Mac, type touch .env then open .env to create and open the file. On Windows, open a new file on Notepad and go to "Save As", then save it in your "ds6001databases" folder -- make sure to set "File As Type" to "All files" and name the file ".env".

Inside the .env file you need to choose passwords for the MySQL, PostgreSQL, and MongoDB databases, so type

```
MYSQL_ROOT_PASSWORD=redlobstercheddarbiscuits
POSTGRES_PASSWORD=outbackbloominonion
MONGO_INITDB_ROOT_PASSWORD=olivegardenunlimitedbreadsticks
MONGO_INITDB_ROOT_USERNAME=mongo
mongo_init_db = mongodb
MYSQL_DATABASE=mysql
```

Change the passwords on the first three lines to whatever you want, but DON'T USE THE @ SYMBOL as that will cause problems. Leave the fourth, fifth, and sixth lines alone, as well as the names of each environmental variable.

#### Part e

In the terminal, make sure you are in the "ds6001databases" folder (you can check by typing pwd . If not, then use cd to navigate to the "ds6001databases" folder). Then type

```
docker compose up
```

This command launches all of the databases. If successful, you will see a long stream of output with messages that begin ds6001databases-postgres-1, ds6001databases-mysql-1, and ds6001databases-mongo-1. If not, we will need to debug together, but the issue likely has to do with something preventing parts a, b, c, or d from being completed successfully.

#### Part f

To confirm that the databases are running on your system, open the "db\_tests.ipynb" notebook file, which should be saved in you "ds6001databases" folder. Run everything in this notebook and make sure there are no errors.

#### Part g

In addition to the databases, we will be using dbdocs.io to create documentation for our databases and post them online with a stable URL. But to get dbdocs running, you first need to install NodeJS on your computer: https://nodejs.org/en

Then to install dbdocs, follow the instructions here: https://dbdocs.io/docs

#### Part h

Finally, create a notebook inside your "ds6001databases" folder for your work on this lab. Import the following libraries, and load the •env file where you store your passwords.

```
In [1]: import dotenv.main
        import numpy as np
        import pandas as pd
        import wget
        import sqlite3
        import requests
        import json
        import os
        import sys
        import dotenv
        from dotenv.main import load_dotenv
        import mysql.connector
        import pymongo
        from sqlalchemy import create_engine
        dotenv.main.load dotenv()
        POSTGRES_PASSWORD = os.getenv('POSTGRES_PASSWORD')
        MONGO_INITDB_ROOT_USERNAME = os.getenv('MONGO_INITDB_ROOT_USERNAME')
        MONGO_INITDB_ROOT_PASSWORD = os.getenv('MONGO_INITDB_ROOT_PASSWORD')
        mongo init db = os.getenv('mongo init db')
        MYSQL_ROOT_PASSWORD = os.getenv('MYSQL_ROOT_PASSWORD')
```

### Problem 1

### This problem requires you to create Markdown tables

To create a table in a markdown cell, I recommend using the markdown table generator here:

https://www.tablesgenerator.com/markdown\_tables. This interface allows you to choose the number of rows and columns, fill in those rows and columns, and push the "generate" button. The website will display markdown table code that looks like:

| ļ     | Day       |       | Temp | ļ     | Rain |         |
|-------|-----------|-------|------|-------|------|---------|
| -<br> | Monday    | -<br> | 74   | -<br> | No   | ۱-<br>ا |
| ĺ     | Tuesday   | ĺ     | 58   | ĺ     | Yes  | ĺ       |
|       | Wednesday |       | 76   |       | No   |         |

Copy the markdown code and paste it into a markdown cell in your notebook. Markdown will read the code and display a table that looks like this:

| Day       | Temp | Rain |
|-----------|------|------|
| Monday    | 74   | No   |
| Tuesday   | 58   | Yes  |
| Wednesday | 76   | No   |

Suppose that we have (fake) data on people who were hospitalized and received at least one prescription for a medication. Here are ten records in the data:

(If this table gets cut off in the PDF, please look at the .ipynb notebook file on the module 6 page on Canvas)

| patient_name          | date_of_birth | prescribed_drug | prior_conditions         | patient_sex | patient_insurance | drug_maker    | drug_cos |
|-----------------------|---------------|-----------------|--------------------------|-------------|-------------------|---------------|----------|
| Nkemdilim<br>Arendonk | 2/21/1962     | Amoxil          | [Pneumonia,<br>Diabetes] | М           | Aetna             | USAntibiotics | 14.62    |
| Nkemdilim<br>Arendonk | 2/21/1962     | Micronase       | [Pneumonia,<br>Diabetes] | М           | Aetna             | Pfizer        | 20.55    |

| patient_name       | date_of_birth | prescribed_drug | prior_conditions                           | patient_sex | patient_insurance         | drug_maker                  | drug_cos |
|--------------------|---------------|-----------------|--|-------------|---------------------------|-----------------------------|----------|
| Raniero<br>Coumans | 8/15/1990     | Zosyn           | [Appendicitis,<br>Crohn's disease]         | М           | Cigna                     | Baxter<br>International Inc | 394.00   |
| Raniero<br>Coumans | 8/15/1990     | Humira          | [Appendicitis,<br>Crohn's disease]         | М           | Cigna                     | Abbvie                      | 7000.00  |
| Mizuki<br>Debenham | 3/12/1977     | Inlyta          | [Kidney Cancer]                            | F           | Kaiser Permanente         | Pfizer                      | 21644.00 |
| Zoë De Witt        | 11/23/1947    | Atenolol        | [Cardiomyopathy,<br>Diabetes,<br>Sciatica] | F           | Medicare                  | Mylan<br>Pharmaceuticals    | 10.58    |
| Zoë De Witt        | 11/23/1947    | Micronase       | [Cardiomyopathy,<br>Diabetes,<br>Sciatica] | F           | Medicare                  | Pfizer                      | 20.55    |
| Zoë De Witt        | 11/23/1947    | Demerol         | [Cardiomyopathy,<br>Diabetes,<br>Sciatica] | F           | Medicare                  | Pfizer                      | 37.50    |
| Bonnie<br>Hooper   | 7/4/1951      | Xeloda          | [Pancreatic<br>Cancer, Sciatica]           | F           | Blue Cross Blue<br>Shield | Genentech                   | 860.00   |
| Bonnie<br>Hooper   | 7/4/1951      | Demerol         | [Pancreatic<br>Cancer, Sciatica]           | F           | Blue Cross Blue<br>Shield | Pfizer                      | 37.50    |

#### The columns in this dataset are:

- patient\_name: The patient's name
- date\_of\_birth: The patient's date of birth
- prescribed\_drug: The brand name of the medication that patient has been prescribed
- prior\_conditions: A list of the conditions that the patient had been diagnosed with prior to the patient's hospitalization
- patient\_sex: The patient's sex
- patient\_insurance: The company responsible for the patient's health insurance coverage
- drug\_maker: The company that manufactures the prescribed drug
- drug\_cost: The cost of the prescribed drug
- attending\_physician: The name of the attending physician for the patient
- AP\_medschool: The name of the school where the attending physician got a medical degree
- AP\_years\_experience: The attending physician's number of years of experience post-residency
- hospital: The hospital where the attending physicial is employed
- hospital\_location: The location of the hospital

### For this problem, assume that

- 1. No two rows in this table share both the same patient and the same prescribed drug.
- 2. Some patients in the data share the same name, but no two patients in the data share the same name and date of birth.
- 3. No two different drugs share the same brand name.
- 4. No two attending physicians have the same name, and every attending physician is employed at only one hospital.
- 5. No two hospitals share the same name, and every hospital exists at only one location.
- 6. Each patient has only one attending physician. (In real-world applications we may want to design a database that allows for multiple hospitalizations for some patients, but here we'll keep it simpler by assuming each patient has one hospitalization with one attending physician.)

In [3]: data\_df = pd.DataFrame(data)
data\_df

Out[3]:

| : |   | patient_name          | date_of_birth | prescribed_drug | prior_conditions                           | patient_sex | patient_insurance         | drug_maker                  | drug_c |
|---|---|-----------------------|---------------|-----------------|--|-------------|---------------------------|-----------------------------|--------|
|   | 0 | Nkemdilim<br>Arendonk | 2/21/1962     | Amoxil          | [Pneumonia,<br>Diabetes]                   | М           | Aetna                     | USAntibiotics               | 14     |
|   | 1 | Nkemdilim<br>Arendonk | 2/21/1962     | Micronase       | [Pneumonia,<br>Diabetes]                   | М           | Aetna                     | Pfizer                      | 20     |
|   | 2 | Raniero<br>Coumans    | 8/15/1990     | Zosyn           | [Appendicitis,<br>Crohn's disease]         | М           | Cigna                     | Baxter<br>International Inc | 394    |
|   | 3 | Raniero<br>Coumans    | 8/15/1990     | Humira          | [Appendicitis,<br>Crohn's disease]         | М           | Cigna                     | Abbvie                      | 7000   |
|   | 4 | Mizuki<br>Debenham    | 3/12/1977     | Inlyta          | [Kidney Cancer]                            | F           | Kaiser Permanente         | Pfizer                      | 21644  |
|   | 5 | Zoë De Witt           | 11/23/1947    | Atenolol        | [Cardiomyopathy,<br>Diabetes,<br>Sciatica] | F           | Medicare                  | Mylan<br>Pharmaceuticals    | 10     |
|   | 6 | Zoë De Witt           | 11/23/1947    | Micronase       | [Cardiomyopathy,<br>Diabetes,<br>Sciatica] | F           | Medicare                  | Pfizer                      | 2(     |
|   | 7 | Zoë De Witt           | 11/23/1947    | Demerol         | [Cardiomyopathy,<br>Diabetes,<br>Sciatica] | F           | Medicare                  | Pfizer                      | 37     |
|   | 8 | Bonnie<br>Hooper      | 7/4/1951      | Xeloda          | [Pancreatic<br>Cancer, Sciatica]           | F           | Blue Cross Blue<br>Shield | Genentech                   | 86(    |
|   | 9 | Bonnie<br>Hooper      | 7/4/1951      | Demerol         | [Pancreatic<br>Cancer, Sciatica]           | F           | Blue Cross Blue<br>Shield | Pfizer                      | 37     |

# Part a

Rearrange the data into a group of data tables that together meet the requirements of first normal form. [2 points]

```
In [4]: patients = data_df.loc[:,('patient_name', 'date_of_birth', 'patient_sex', 'prior_conditions','patient_insura
    patients['patient_id'] = patients.index + 1
    patients
```

| Out[4]: |   | patient_name          | date_of_birth | patient_sex | prior_conditions                        | patient_insurance         | prescribed_drug | patient_id |
|---------|---|-----------------------|---------------|-------------|---|---------------------------|-----------------|------------|
|         | 0 | Nkemdilim<br>Arendonk | 2/21/1962     | М           | [Pneumonia, Diabetes]                   | Aetna                     | Amoxil          | 1          |
|         | 1 | Nkemdilim<br>Arendonk | 2/21/1962     | М           | [Pneumonia, Diabetes]                   | Aetna                     | Micronase       | 2          |
|         | 2 | Raniero<br>Coumans    | 8/15/1990     | М           | [Appendicitis, Crohn's disease]         | Cigna                     | Zosyn           | 3          |
|         | 3 | Raniero<br>Coumans    | 8/15/1990     | М           | [Appendicitis, Crohn's disease]         | Cigna                     | Humira          | 4          |
|         | 4 | Mizuki<br>Debenham    | 3/12/1977     | F           | [Kidney Cancer]                         | Kaiser Permanente         | Inlyta          | 5          |
|         | 5 | Zoë De Witt           | 11/23/1947    | F           | [Cardiomyopathy,<br>Diabetes, Sciatica] | Medicare                  | Atenolol        | 6          |
|         | 6 | Zoë De Witt           | 11/23/1947    | F           | [Cardiomyopathy,<br>Diabetes, Sciatica] | Medicare                  | Micronase       | 7          |
|         | 7 | Zoë De Witt           | 11/23/1947    | F           | [Cardiomyopathy,<br>Diabetes, Sciatica] | Medicare                  | Demerol         | 8          |
|         | 8 | Bonnie Hooper         | 7/4/1951      | F           | [Pancreatic Cancer,<br>Sciatica]        | Blue Cross Blue<br>Shield | Xeloda          | 9          |
|         | 9 | Bonnie Hooper         | 7/4/1951      | F           | [Pancreatic Cancer,<br>Sciatica]        | Blue Cross Blue<br>Shield | Demerol         | 10         |

```
In [5]: drugs = data_df.loc[:,('prescribed_drug', 'drug_maker', 'drug_cost')]
    drugs['drug_id'] = drugs.index + 1
    drugs
```

| Out[5]: |   | prescribed_drug | drug_maker               | drug_cost | drug_id |
|---------|---|-----------------|--------------------------|-----------|---------|
|         | 0 | Amoxil          | USAntibiotics            | 14.62     | 1       |
|         | 1 | Micronase       | Pfizer                   | 20.55     | 2       |
|         | 2 | Zosyn           | Baxter International Inc | 394.00    | 3       |
|         | 3 | Humira          | Abbvie                   | 7000.00   | 4       |
|         | 4 | Inlyta          | Pfizer                   | 21644.00  | 5       |
|         | 5 | Atenolol        | Mylan Pharmaceuticals    | 10.58     | 6       |
|         | 6 | Micronase       | Pfizer                   | 20.55     | 7       |
|         | 7 | Demerol         | Pfizer                   | 37.50     | 8       |
|         | 8 | Xeloda          | Genentech                | 860.00    | 9       |
|         | 9 | Demerol         | Pfizer                   | 37.50     | 10      |

```
In [6]: doctors = data_df.loc[:,('attending_physician', 'AP_medschool', 'AP_years_experience')]
doctors['physician_id'] = doctors.index + 1
doctors
```

```
Out[6]:
             attending_physician
                                                   AP_medschool AP_years_experience physician_id
                                     University of California (Irvine)
          0
                      Earnest Caro
                                                                                        14
                                                                                                        1
                                     University of California (Irvine)
                                                                                                        2
          1
                      Earnest Caro
                                                                                        14
          2
                    Pamela English
                                             University of Michigan
                                                                                        29
                                                                                                        3
          3
                    Pamela English
                                             University of Michigan
                                                                                        29
                                                                                                        4
          4
                       Lewis Conti
                                     North Carolina State University
                                                                                         8
                                                                                                        5
          5
                 Theresa Dahlmans
                                      Lake Erie College of Medicine
                                                                                        17
                                                                                                        6
                                                                                                        7
          6
                 Theresa Dahlmans
                                      Lake Erie College of Medicine
                                                                                        17
          7
                 Theresa Dahlmans
                                      Lake Erie College of Medicine
                                                                                        17
                                                                                                        8
          8
                    Steven Garbutt
                                              Ohio State University
                                                                                        36
                                                                                                        9
          9
                    Steven Garbutt
                                              Ohio State University
                                                                                        36
                                                                                                        10
```

```
In [7]: hospitals = data_df.loc[:,('hospital', 'hospital_location')]
hospitals['hospital_id'] = hospitals.index + 1
hospitals
```

```
hospital hospital_location hospital_id
0
     UPMC Presbyterian Shadyside
                                       Pittsburgh, PA
                                                                 1
1
     UPMC Presbyterian Shadyside
                                       Pittsburgh, PA
                                                                 2
2 Northwestern Memorial Hospital
                                                                 3
                                          Chicago, IL
   Northwestern Memorial Hospital
                                          Chicago, IL
                                                                 4
4
       Houston Methodist Hospital
                                         Houston, TX
                                                                 5
5
              Mount Sinai Hospital
                                        New York, NY
                                                                 6
6
                                                                 7
              Mount Sinai Hospital
                                        New York, NY
7
              Mount Sinai Hospital
                                        New York, NY
                                                                 8
8
             UCSF Medical Center
                                   San Francisco, CA
                                                                 9
9
             UCSF Medical Center
                                   San Francisco, CA
                                                                10
```

Out[7]:

```
In [8]: prescriptions = data_df[['patient_name', 'prescribed_drug', 'attending_physician', 'hospital']]
In [9]: prescriptions = prescriptions.merge(patients[['patient_name', 'patient_id']], on='patient_name')
    prescriptions = prescriptions.merge(drugs[['prescribed_drug', 'drug_id']], on='prescribed_drug')
    prescriptions = prescriptions.merge(doctors[['attending_physician', 'physician_id']], on='attending_physician'
    prescriptions = prescriptions.merge(hospitals[['hospital', 'hospital_id']], on='hospital')
    prescriptions
```

|     | patient_name          | prescribed_drug | attending_physician | hospital                          | patient_id | drug_id | physician_id | hospital_id |
|-----|-----------------------|-----------------|---------------------|-----------------------------------|------------|---------|--------------|-------------|
| 0   | Nkemdilim<br>Arendonk | Amoxil          | Earnest Caro        | UPMC<br>Presbyterian<br>Shadyside | 1          | 1       | 1            | 1           |
| 1   | Nkemdilim<br>Arendonk | Amoxil          | Earnest Caro        | UPMC<br>Presbyterian<br>Shadyside | 1          | 1       | 1            | 2           |
| 2   | Nkemdilim<br>Arendonk | Amoxil          | Earnest Caro        | UPMC<br>Presbyterian<br>Shadyside | 1          | 1       | 2            | 1           |
| 3   | Nkemdilim<br>Arendonk | Amoxil          | Earnest Caro        | UPMC<br>Presbyterian<br>Shadyside | 1          | 1       | 2            | 2           |
| 4   | Nkemdilim<br>Arendonk | Amoxil          | Earnest Caro        | UPMC<br>Presbyterian<br>Shadyside | 2          | 1       | 1            | 1           |
| ••• |                       |                 |                     |                                   |            |         |              |             |
| 195 | Bonnie Hooper         | Demerol         | Steven Garbutt      | UCSF<br>Medical<br>Center         | 10         | 8       | 10           | 10          |
| 196 | Bonnie Hooper         | Demerol         | Steven Garbutt      | UCSF<br>Medical<br>Center         | 10         | 10      | 9            | 9           |
| 197 | Bonnie Hooper         | Demerol         | Steven Garbutt      | UCSF<br>Medical<br>Center         | 10         | 10      | 9            | 10          |
| 198 | Bonnie Hooper         | Demerol         | Steven Garbutt      | UCSF<br>Medical<br>Center         | 10         | 10      | 10           | 9           |
| 199 | Bonnie Hooper         | Demerol         | Steven Garbutt      | UCSF<br>Medical<br>Center         | 10         | 10      | 10           | 10          |

200 rows × 8 columns

# Part b

Out[9]:

Rearrange the data on the five patients into a group of data tables that together meet the requirements of second normal form. [2 points]

```
In [10]: second_nf_df = prescriptions[['patient_id', 'drug_id', 'physician_id', 'hospital_id']]
second_nf_df
```

| Out[10]: |     | patient_id | drug_id | physician_id | hospital_id |
|----------|-----|------------|---------|--------------|-------------|
|          | 0   | 1          | 1       | 1            | 1           |
|          | 1   | 1          | 1       | 1            | 2           |
|          | 2   | 1          | 1       | 2            | 1           |
|          | 3   | 1          | 1       | 2            | 2           |
|          | 4   | 2          | 1       | 1            | 1           |
|          |     |            |         |              |             |
|          | 195 | 10         | 8       | 10           | 10          |
|          | 196 | 10         | 10      | 9            | 9           |
|          | 197 | 10         | 10      | 9            | 10          |
|          | 198 | 10         | 10      | 10           | 9           |
|          | 199 | 10         | 10      | 10           | 10          |

200 rows × 4 columns

#### Part c

Rearrange the data into a group of data tables that together meet the requirements of third normal form. [2 points]

```
In [11]: insurance = data_df[['patient_insurance']].drop_duplicates().reset_index(drop=True)
insurance['insurance_id'] = insurance.index + 1
insurance
```

```
Out[11]:
                 patient_insurance insurance_id
           0
                              Aetna
                                                1
                                                2
           1
                              Cigna
           2
                  Kaiser Permanente
                                                3
           3
                          Medicare
                                                4
           4 Blue Cross Blue Shield
                                                5
```

```
In [12]: patients = patients.merge(insurance, on='patient_insurance')
    patients = patients[['patient_id', 'patient_name', 'date_of_birth', 'prior_conditions', 'patient_sex', 'insu
In [13]: prescriptions = data_df.loc[:,('patient_name', 'prescribed_drug', 'attending_physician', 'hospital')]
In [14]: prescriptions = prescriptions.merge(patients[['patient_name', 'patient_id']], on='patient_name')
    prescriptions = prescriptions.merge(drugs[['prescribed_drug', 'drug_id']], on='prescribed_drug')
    prescriptions = prescriptions.merge(doctors[['attending_physician', 'physician_id']], on='attending_physician'
    prescriptions = prescriptions.merge(hospitals[['hospital', 'hospital_id']], on='hospital')
In [15]: third_nf_df = prescriptions.loc[:,('patient_id', 'drug_id', 'physician_id', 'hospital_id')]
third_nf_df
```

| $\cap$ |   | + | Γ | 1 | 5 | 1 |  |
|--------|---|---|---|---|---|---|--|
| v      | u | L | L | _ | J | л |  |

|     | patient_id | drug_id | physician_id | hospital_id |
|-----|------------|---------|--------------|-------------|
| 0   | 1          | 1       | 1            | 1           |
| 1   | 1          | 1       | 1            | 2           |
| 2   | 1          | 1       | 2            | 1           |
| 3   | 1          | 1       | 2            | 2           |
| 4   | 2          | 1       | 1            | 1           |
| ••• |            | •••     |              |             |
| 195 | 10         | 8       | 10           | 10          |
| 196 | 10         | 10      | 9            | 9           |
| 197 | 10         | 10      | 9            | 10          |
| 198 | 10         | 10      | 10           | 9           |
| 199 | 10         | 10      | 10           | 10          |

200 rows × 4 columns

## Problem 2

For this problem, create ER diagrams of the database you created in problem 1, part c using https://dbdocs.io/. Make sure you install DBDocs on your system by following these instructions: https://dbdocs.io/docs

### Part a

Write code using the database markup language (DBML) that represents all of the tables in this database and the connections between the tables. Paste your DBML code in a markdown cell in your notebook, contained within three backticks to begin and end the code snippet, as shown in the cell below.

Two good resources to help you:

- 1. The example on the Getting Started page on dbdocs.io: https://dbdocs.io/docs
- 2. The full syntax guide for DBML: https://dbml.dbdiagram.io/docs/#project-definition

#### A few notes:

- Make sure to specify the data type for each column in each table. Use varchar for strings/text, int for integers, and float for numeric data with decimals.
- You will probably find it useful to alias each table with one or two letters, such as: Table PRESCRIPTIONS as PR. That will allow you to use PR to refer to the PRESCRIPTIONS table, for example, in the Reference statements to link tables together.
- Use the syntax [pk] after a column name and data type to designate the columns that are primary keys in each table.
- To draw the lines linking one table to another, use the Ref: syntax.
  - If many rows from the left table match to one row in the right table, use the "many to one" symbol >
  - If one row from the left table matches to many rows in the right table, use the "one to many" symbol <
  - If one row from the left table matches to one row in the right table, use the "one to one" symbol -
  - If many rows from the left table match to many rows in the right table, use the "many to many" symbol <>>

# [2 points]

```
Project hospitalizations {
  database_type: 'PostgreSQL'
  Note: '''
    # fake data on people who were hospitalized and received at least one prescription for a medication.
    **Database created for DS6001 July 2024**
    ***Interval to the project of th
```

```
Table patients {
    patient_id int [pk, increment]
    patient_name varchar
    date of birth date
    prior_conditions varchar
    patient sex char
    insurance_id int
}
Table drugs {
    drug_id int [pk, increment]
    prescribed_drug varchar
    drug_maker varchar
    drug_cost float
}
Table doctors {
    physician_id int [pk, increment]
    attending_physician varchar
    AP medschool varchar
    AP_years_experience int
}
Table hospitals {
    hospital_id int [pk, increment]
    hospital varchar
    hospital_location varchar
}
Table insurance {
    insurance_id int [pk, increment]
    patient_insurance varchar
}
Table prescriptions {
    prescription_id int [pk, increment]
    patient_id int
    drug_id int
    physician_id int
    hospital_id int
}
Ref: prescriptions.patient_id > patients.patient_id
Ref: prescriptions.drug_id > drugs.drug_id
Ref: prescriptions.physician_id > doctors.physician_id
Ref: prescriptions.hospital_id > hospitals.hospital_id
Ref: patients.insurance_id > insurance.insurance_id
```

# Part b

Use the instructions on DBDocs.io (https://dbdocs.io/docs) to create a website for your ER diagram. Type the URL for your website in a markdown cell here. [1 point]

https://dbdocs.io/kianadane/hospitalizations

### Problem 3

For this problem, you will download the individual CSV files that comprise a relational database on album reviews from Pitchfork Magazine, collected via webscraping by Nolan B. Conaway, and use them to initialize local databases using SQlite, MySQL, and PostgreSQL.

The following code of code will download the CSV files. Please run this as is:

```
In [16]: url = "https://github.com/nolanbconaway/pitchfork-data/raw/master/pitchfork.db"
    pfork = wget.download(url)
    pitchfork = sqlite3.connect(pfork)
    for t in ['artists','content','genres','labels','reviews','years']:
        datatable = pd.read_sql_query("SELECT * FROM {tab}".format(tab=t), pitchfork)
        datatable.to_csv("{tab}.csv".format(tab=t))
```

Note: this code downloaded a SQlite database and extracted the tables, saving each one as a CSV. That seems backwards, as the purpose of this exercise is to create databases. But the point here is to practice creating databases from individual data frames. Next we load the CSVs to create the data frames in Python:

```
In [17]: reviews = pd.read_csv("reviews.csv")
    artists = pd.read_csv("artists.csv")
    content = pd.read_csv("content.csv")
    genres = pd.read_csv("genres.csv")
    labels = pd.read_csv("labels.csv")
    years = pd.read_csv("years.csv")
```

#### Part a

Initialize a new database using SQlite and the sqlite3 library. Add the six dataframes to this database. Then issue the following query to the database

```
SELECT title, artist, score FROM reviews WHERE score=10
```

using two methods: first, using the <code>.cursor()</code> method, and second using <code>pd.read\_sql\_query()</code> . Finally, commit your changes to the database and close the database. (If you get a warning about spaces in the column names, feel free to ignore it this time.) [2 points]

```
In [18]: reviews.to_sql('reviews', pitchfork, if_exists='replace', index=False)
    artists.to_sql('artists', pitchfork, if_exists='replace', index=False)
    content.to_sql('content', pitchfork, if_exists='replace', index=False)
    genres.to_sql('genres', pitchfork, if_exists='replace', index=False)
    labels.to_sql('labels', pitchfork, if_exists='replace', index=False)
    years.to_sql('years', pitchfork, if_exists='replace', index=False)
```

Out[18]: 19108

# Issuing the query using the .cursor() method

```
In [19]: cursor = pitchfork.cursor()
   query = '''SELECT title, artist, score FROM reviews WHERE score=10'''
   cursor.execute(query)
   rows = cursor.fetchall()

   rows[:5]

Out[19]: [('metal box', 'public image ltd', 10.0),
        ('blood on the tracks', 'bob dylan', 10.0),
        ('another green world', 'brian eno', 10.0),
        ('songs in the key of life', 'stevie wonder', 10.0),
```

# Issuing query using pd.read\_sql\_query

('in concert', 'nina simone', 10.0)]

```
In [20]: pandas_query = pd.read_sql_query(query, pitchfork)
pandas_query.head(5)
```

```
1
                 blood on the tracks
                                           bob dylan
                                                        10.0
                 another green world
           2
                                            brian eno
                                                        10.0
           3 songs in the key of life
                                        stevie wonder
                                                        10.0
                          in concert
                                         nina simone
                                                        10.0
           pitchfork.commit()
In [21]:
           pitchfork.close()
```

### Part b

Out [25]: 19108

title

metal box public image ltd

artist score

10.0

Out[20]:

0

Follow the instructions in the Jupyter notebook for this module to install MySQL and mysql.connector on your computer.

Make sure the MySQL server is running. Then import mysql.connector and do all of the tasks listed for part a using a MySQL database (including committing changes and closing the database connection). Take steps to hide your password - do not let it display in your notebook. [3 points]

```
In [22]: MYSQL_ROOT_PASSWORD = os.getenv('MYSQL_ROOT_PASSWORD')
In [23]: dbserver = mysql.connector.connect(
             user='root',
             password=MYSQL ROOT PASSWORD,
             host='localhost',
             port='3306',
             database= 'hpz db'
             #auth plugin=MYSQL ROOT PASSWORD
         if dbserver.is_connected():
             print("Connected to the MySQL database")
             print("Failed to connect to the MySQL database")
        Connected to the MySQL database
In [24]: hpz db = 'hpz db'
         engine = create_engine(f'mysql+mysqlconnector://root:{MYSQL_ROOT_PASSWORD}@localhost:3306/{hpz db}')
In [25]: reviews.to_sql('reviews', con=engine, chunksize=1000, if_exists='replace', index=False)
         artists.to_sql('artists', con=engine, chunksize=1000, if_exists='replace', index=False)
         content.to_sql('content', con=engine, chunksize=1000, if_exists='replace', index=False)
         genres.to_sql('genres', con=engine, chunksize=1000, if_exists='replace', index=False)
         labels.to_sql('labels', con=engine, chunksize=1000, if_exists='replace', index=False)
         years.to_sql('years', con=engine, chunksize=1000, if_exists='replace', index=False)
```

# Issuing the query using the .cursor() method

```
In [26]: cursor2 = dbserver.cursor()
   query = '''SELECT title, artist, score FROM reviews WHERE score=10'''
   cursor2.execute(query)
   results = cursor2.fetchall()
```

# Issuing the query using pd.read\_sql\_query

```
In [27]: pandas_query = pd.read_sql_query(query, con=engine)
pandas_query.head(10)
```

| artist score  |      | artist                 | title               |   | t[27]: |
|---|------|------------------------|---------------------|---|--------|
|   | 10.0 | public image Itd       | metal box           | 0 |        |
|   | 10.0 | bob dylan              | blood on the tracks | 1 |        |
| 1 blood on the tracks bob dylan 10.0 2 another green world brian eno 10.0 3 songs in the key of life stevie wonder 10.0 4 in concert nina simone 10.0 5 tonight's the night neil young 10.0 |      |                        |                     |   |        |
|   |      |                        |                     |   |        |
|   |      | nina simone            | in concert          | 4 |        |
|   | 10.0 | neil young             | tonight's the night | 5 |        |
|   | 10.0 | kate bush              | hounds of love      | 6 |        |
| hounds of love kate bush 10.0   |      | sign "o" the times     | 7                   |   |        |
|   | 10.0 | prince                 | 1999                | 8 |        |
|   | 10.0 | prince, the revolution | purple rain         | 9 |        |
|   |      |                        |                     |   |        |

```
In [28]: dbserver.commit()
dbserver.close()
```

# Part c

Follow the instructions in the Jupyter notebook for this module to install PostgreSQL and psycopg2 on your computer. Then import psycopg2 and do all of the tasks listed for part a using a PostgreSQL database (including committing changes and closing the database connection). Take steps to hide your password - do not let it display in your notebook. [3 points]

genres.to\_sql('genres', con=engine2, if\_exists='replace', index=False)

```
labels.to_sql('labels', con=engine2, if_exists='replace', index=False)
years.to_sql('years', con=engine2, if_exists='replace', index=False)
```

Out [32]: 108

# Issuing the query using the .cursor() method

# Issuing the query using pandas

```
In [34]: filtered = reviews[reviews['score'] == 10]
filtered = filtered.loc[:, ['title', 'artist', 'score']]
filtered.head(10)
```

| Out[34]: |     | title                    | artist                 | score |
|----------|-----|--------------------------|------------------------|-------|
|          | 191 | metal box                | public image Itd       | 10.0  |
|          | 200 | blood on the tracks      | bob dylan              | 10.0  |
|          | 355 | another green world      | brian eno              | 10.0  |
|          | 451 | songs in the key of life | stevie wonder          | 10.0  |
|          | 530 | in concert               | nina simone            | 10.0  |
|          | 654 | tonight's the night      | neil young             | 10.0  |
|          | 706 | hounds of love           | kate bush              | 10.0  |
|          | 857 | sign "o" the times       | prince                 | 10.0  |
|          | 858 | 1999                     | prince                 | 10.0  |
|          | 861 | purple rain              | prince, the revolution | 10.0  |

In [35]: conn.close()

### Problem 4

Colin Mitchell is a web-developer and artist who has a bunch of cool projects that play with what data can do on the internet.

One of his projects is Today in History, which provides an API to access all the Wikipedia pages for historical events that happened on this day in JSON format. The records in this JSON are stored in the ['data']['events'] path. Here's the first listing for today:

```
In [36]: import requests
In [37]: history = requests.get("https://history.muffinlabs.com/date", verify=False)
```

/Users/kianadane/Documents/GitHub/ds6001databases/.conda/lib/python3.12/site-packages/urllib3/connectionpool.py:1099: InsecureRequestWarning: Unverified HTTPS request is being made to host 'history.muffinlabs.com'. Add ing certificate verification is strongly advised. See: https://urllib3.readthedocs.io/en/latest/advanced-usage.html#tls-warnings.warn(

For this problem, you will use MongoDB and the **pymongo** library to create a local document store NoSQL database containing these historical events.

Follow the instructions in the Jupyter notebook for this module to install MongoDB and pymongo on your computer. Make sure the local MongoDB server is running. Then import pymongo, connect to the local MongoDB client, create a database named "history" and a collection within that database named "today". Insert all of the records in events into this collection. Then issue the following query to find all of the records whose text contain the word "Virginia":

```
query = {
    "text":{
        "$regex": 'Virginia'
    }
}
```

If there are no results that contain the word "Virginia", choose a different work like "England" or "China". Display the count of the number of documents that match this query, display the output of the query, and generate a JSON formatted variable containing the output. [3 points]

```
In [38]: from pymongo import MongoClient
         load_dotenv
Out[38]: <function dotenv.main.load dotenv(dotenv path: Union[str, ForwardRef('os.PathLike[str]'), NoneType] = None,
          stream: Optional[IO[str]] = None, verbose: bool = False, override: bool = False, interpolate: bool = True,
         encoding: Optional[str] = 'utf-8') -> bool>
In [39]: mongo_init_db = os.getenv('MONGO_INIT_DB')
         mongo_user = os.getenv('MONGO_USER')
         mongo_password = os.getenv('MONGO_PASSWORD')
In [40]: client = MongoClient(f"mongodb://{MONGO_INITDB_ROOT_USERNAME}:{MONGO_INITDB_ROOT_PASSWORD}@localhost:27017/"
         trv:
             databases = client.list_database_names()
             print("Connected to MongoDB!")
             print("Databases:", databases)
         except Exception as e:
             print("Could not connect to MongoDB:", e)
        Connected to MongoDB!
        Databases: ['admin', 'config', 'local', 'mongo_init_db']
In [41]: history_db = client['History']
In [42]: today = history db['Events']
In [43]: import requests
         history = requests.get("https://history.muffinlabs.com/date")
         history
Out[43]: <Response [200]>
In [44]: history = json.loads(history.text)
In [52]: today = history['data']['Events']
         today = pd.DataFrame(today)
In [57]: query = 'China'
         results = today[today['text'].str.contains(query, case=False, na=False)]
         results
```

| Out[57]: |    | year | text   | html   | no_year_html                                   | links   |
|----------|----|------|--|--|--|---|
|          | 35 | 1941 | World War II: In response to the Japanese occu | 1941 - World War II: In response to the Japane | World War II: In response to the Japanese occu | [{'title': 'French<br>Indochina', 'link': 'https: |

In [58]: client.close()

# Problem 5 [No points, no need to write anything here, but do it anyway!]

When you are done working, go to the same terminal window you used to launch the databases with docker compose up . Here, type CONTROL + C on your keyboard to shut down the container.

Next type

docker compose down

This step removes extra database software, networks, volumes, etc. running on your computer. If you don't need them, don't clog your computer.

Whenever you need to work with databases, return to the terminal, navigate to this folder and type

docker compose up

to bring all these resources back online.