# FOR INSTRUCTOR PURPOSES ONLY

# **INSTRUCTOR NOTES**

▶ Insert Text Here

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# **MATERIALS**

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# FOR INSTRUCTOR PURPOSES ONLY

# **PRE-WORK**

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# **LEARNING OBJECTIVES**

- ▶ Understanding of the uses and differences of databases
- ► Accessing databases from Pandas

# **COURSE**

# PRE-WORK

# PRE-WORK REVIEW

- There will be multiple ways to run the exercises:
  - **▶** Using Postgres Exercises
  - ▶ Setting up local Postgres
    - Install Postgres. If brew is installed, this should be as simple as brew install postgres
  - ▶ Use Wagon
    - Create an account at <a href="https://www.wagonhq.com/">https://www.wagonhq.com/</a> and download the software.

# **OPENING**

- Today's lesson will be on databases and the SQL query language.
- Databases are the standard solution for data storage. They're far more robust than text and CSV files.
- They come in many flavors, but we'll explore the most common: *relational databases*.



- Relational databases also come in different varieties, but almost all use SQL as a basis for querying (i.e. retrieving) data.
- ▶ Most analyses typically involve pulling data from a database.

# INTRODUCTION

- ▶ Databases are computer systems that manage the storage and querying of datasets.
- They provide a way to organize the data on disk (i.e. hard drive) and efficient methods to retrieve information. Databases allow a user to create rules that ensure proper data management and verification.
- ▶ Typically, retrieval is performed using a query language, a mini programming language with a few basic operators for data transformation.
- ▶ The most common query language is SQL (Structured Query Language).

- ▶ A *relational database* is based on links between data entities or concepts.
- ▶ Typically, a relational databases is organized into *tables*.
- ▶ Each table should correspond to one entity or concept. Each table is similar to a single CSV file or Pandas dataframe.
- For example, consider an application like Twitter. Our two main entities are Users and Tweets. For each of these, we would have a separate table.

- A table is made up of rows and columns, similar to a Pandas dataframe or Excel spreadsheet.
- ▶ Each table has a specific *schema*, a set of rules for what goes in each table. These specify which columns are contained in the table and what *type* of data is in each column (e.g. text, integers, decimals, etc).

Users Table Schema			
user_id	char		
user_sign_up_date	date		
user_follower_count	int		

- ▶ This means you can't add text data to an integer column in that database.
- The additional *type* information make this constraint stronger than the header of a CSV file.
- ▶ For this reason and many others, databases allow for stronger consistency of the data and are often a better solution for data storage.

- ▶ Each table typically has a *primary* key column. This column has a unique value per row and serves as the identifier for the row.
- A table can have many *foreign keys* as well. A *foreign key* is a column that contains values to link the table to the other tables.
- ▶ These keys that link the table together define the relational database.

- ▶ For example, the tweets table may have as columns:
  - ▶tweet\_id the primary key tweet identifier
  - ▶tweet\_text
  - ▶user\_id a foreign key to the users table

Users Table Schema		Tweets Table Schema	
user_id	char	tweet_id	int
user_sign_up_date	date	tweet_text	char
user_follower_count	int	user_id	int

- MySQL and Postgres are popular variants of relational databases and are widely used. Both are open-source and available for free.
- Alternatively, many companies use proprietary software such as Oracle or Microsoft SQL databases.
- ▶ While these databases offer many of the same features and use the same SQL language, the latter two offer some maintenance features and support that large companies find useful.

- Once we start organizing our data into tables, we start to separate it into normalized and denormalized setups.
- Normalized structures have a single table per entity and use many foreign keys or link tables to connect the entities.
- ▶ Denormalized structures have fewer tables that combine different entities.

▶ With our Twitter example, a *normalized* structure would place users and tweets in different tables.

Users Table Schema		Tweets Table Schema	
user_id	char	tweet_id	int
user_sign_up_date	date	tweet_text	char
user_follower_count	int	user_id	int

▶ A *denormalized* structure would put them both in one table.

Twitter Table Schema			
tweet_id	int		
tweet_text	char		
user_id	int		
user_sign_up_date	date		
user_follower_count	int		

#### **Denormalized structures:**

- ▶ Duplicates a lot of information
- Makes data easy to access since it's all in one table

#### **Normalized structures:**

- Save storage space by separating information
- Requires joining of table to access information about two different entities, a slow operation

# **ALTERNATIVE DATABASES**

- ▶ While relational databases are the most popular and broadly used, specific applications may require different data organization.
- ▶ You don't need to know every variety, but it's good to know some overall themes.

# **KEY-VALUE STORES**

- ▶ Key-Value databases are nothing more than very large and very fast hashmaps or dictionaries.
- ▶ These are useful for storing key based data, e.g. a count of things per user or customer, a last visit per customer.
- ▶ Every entry in these databases has two values, a key and a value. We can retrieve any value based upon its key.

# **KEY-VALUE STORES**

- This is exactly like a python dictionary, but it can be larger than your memory (i.e. RAM). So these systems use smart caching algorithms to ensure frequently or recently accessed items are quickly accessible.
- ▶ Popular key-value stores include *Cassandra* and *MemcacheDB* (pronounced mem-cash-dee-bee).

- "NoSQL" databases are those that don't rely on a traditional relational table setup and more flexible in their data organization.
- ▶ Typically they actually **do** have SQL querying abilities but model their data differently.

#### ▶ Relational Structure

user_id	user_name	user_hobby_1	user_hobby_2	user_age
13123	robby_g	guitar	cars	25
18423	jt1235	football		31

# ► NoSQL Data Structure

```
"user_id": 13123,

"user_name": "robby_g",

"user_hobbies": ["guitar",

"cars"],

"user_age": 25
```

```
{
    "user_id": 19423,
        "user_name": "jt1235",
        "user_hobbies": ["football"],
        "user_age": 31
}
```

- They may organize data on an entity level, but often have denormalized and nested data setups.
- ▶ This nested data layout is often similar to that in JSON documents.
- ▶ Popular databases include *MongoDB* and *CouchDB*.



#### Relational data model

Highly-structured table organization with rigidly-defined data formats and record structure.



### Document data model

Collection of complex documents with arbitrary, nested data formats and varying "record" format.

▶ The following is an example of the storage document for a tweet.

# **ACTIVITY: KNOWLEDGE CHECK**

#### **ANSWER THE FOLLOWING QUESTIONS**



In the following examples, which might be the best storage or database solution? Why?

- 1. An application where a user can create a profile
- 2. An online store
- 3. Storing the last visit date of a user

#### **DELIVERABLE**

Answers to the above questions

# **ACTIVITY: KNOWLEDGE CHECK**

#### ANSWER THE FOLLOWING QUESTIONS

Consider a dataset from Uber with the following fields:



- User ID
- User Name
- Driver ID
- Drive Name
- Ride ID
- Ride Time

- Pickup Latitude
- Pickup Longitude
- Pickup Location Entity
- Dropoff Longitude
- Dropoff Latitude

- Dropoff Location Entity
- Miles
- Travel Time
- Fare
- CC Number

In a group, discuss how you would design a relational database to support this data? List the tables you would create, the fields they would contain, and how they would link to other tables.

#### **DELIVERABLE**

Your database schema design

# ACCESSING DATABASES FROM PANDAS

# **ACCESSING DATABASES FROM PANDAS**

- ▶ While databases provide many analytical capabilities, often it's useful to pull the data back into Python for more flexible programming.
- Large, fixed operations would be more efficient in a database, but Pandas allows for interactive processing.
- For example, if you just want to aggregate login or sales data to present a report or dashboard, this operation is operating on a large dataset and not often changing.
- ▶ This would run very efficiently in a database vs connecting to Python.

# **ACCESSING DATABASES FROM PANDAS**

▶ However, if we want to investigate the login or sales data further and ask more interactive questions, then using Python would come in very handy.

```
import pandas as pd
from pandas.io import sql
```

▶ Pandas can be used to connect to most relational databases.

# **ACCESSING DATABASES FROM PANDAS**

- In this demonstration, we will create and connect to a SQLite database. SQLite creates portable relational databases saved in a single file.
- These databases are stored in a very efficient manner and allow fast querying, making them ideal for small databases or databases that need to be moved across machines.

### **ACCESSING DATABASES FROM PANDAS**

▶ We can create a SQLite databases as follows.

```
import sqlite3
conn = sqlite3.connect('dat-test.db')
```

This creates a file, dat-test.db, which will act as a relational/SQL database.

### WRITING DATA INTO A DATABASE

- ▶ Data in Pandas can be loaded into a relational database. For the most part, Pandas can use the databases column information to infer the schema for the table it creates.
- Let's return to the Rossmann sales data and load it into our database.

```
import pandas as pd

data = pd.read_csv('.../.../lesson-15/assets/dataset/rossmann.csv',
    low_memory=False)
    data.head()
```

### WRITING DATA INTO A DATABASE

- Data is moved to the database with the to\_sql command, similar to the to\_csv command.
- ▶to\_sql takes several arguments.
  - ▶name the table name to create
  - ▶con a connection to a database
  - ▶index whether to input the index column
  - >schema if we want to write a custom schema for the new table
  - \if\_exists what to do if the table already exists. We can overwrite it, add to it, or fail

### WRITING DATA INTO A DATABASE

▶ The following code loads the Rossmann sales data to our database.

### READING FROM A DATABASE

- ▶ If we already have data in the database, we can use Pandas to query our database.
- ▶ Querying is done through the read\_sql command in the sql module.

```
import pandas as pd
from pandas.io import sql

sql.read_sql('select * from rossmann_sales limit 10', con=conn)
```

This runs the query passed in and returns a dataframe with the results.

### **ACTIVITY: KNOWLEDGE CHECK**

#### **ANSWER THE FOLLOWING QUESTIONS**



1. Load the Rossmann Store metadata in rossmann-stores.csv and create a table in the database with it.

### **DELIVERABLE**

Created table for store metadata

# SQL SYNTAX: SELECT, WHERE, GROUP BY, JOIN

### **SQL OPERATORS: SELECT**

- ▶ Every query should start with SELECT. SELECT is followed by the names of the columns in the output.
- ▶ SELECT is always paired with FROM, which identifies the table to retrieve data from.

```
SELECT <columns>
FROM
```

▶ SELECT \* denotes returning *all* of the columns.

### **SQL OPERATORS: SELECT**

▶ Rossmann Stores example:

```
SELECT Store, Sales
FROM rossmann_sales;
```

### **ACTIVITY: KNOWLEDGE CHECK**

### **ANSWER THE FOLLOWING QUESTIONS**



1. Write a query for the Rossmann Sales data that returns Store, Date, and Customers.

#### **DELIVERABLE**

The requested query

### **SQL OPERATORS: WHERE**

WHERE is used to filter a table using a specific criteria. The WHERE clause follows the FROM clause.

```
SELECT <columns>
FROM 
WHERE <condition>
```

▶ The condition is some filter applied to the rows, where rows that match the condition will be output.

### **SQL OPERATORS: WHERE**

### ▶ Rossmann Stores example:

```
SELECT Store, Sales
FROM rossmann_sales
WHERE Store = 1;

SELECT Store, Sales
FROM rossmann_sales
WHERE Store = 1 and Open = 1;
```

### **ACTIVITY: KNOWLEDGE CHECK**

#### **ANSWER THE FOLLOWING QUESTIONS**



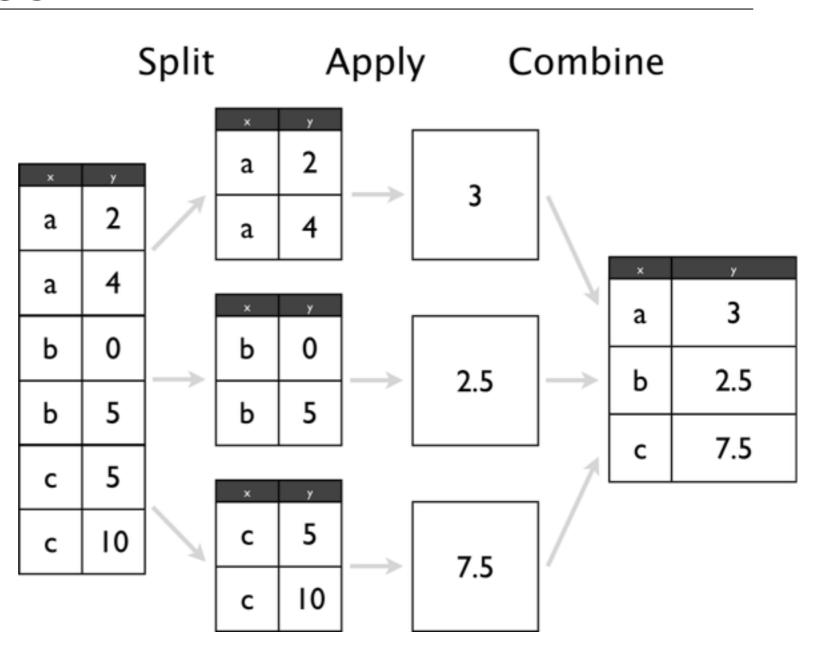
1. Write a query for the Rossmann Sales data that returns Store, Date, and Customers for stores that were open and running a promotion.

#### **DELIVERABLE**

The requested query

### **SQL OPERATORS: GROUP BY**

- GROUP BY allows us to aggregate over any field in the table by applying the concept of Split Apply Combine.
- We identify some key with which we want to segment the rows. Then, we roll up or compute some statistics over all of the rows that match that key.



### **SQL OPERATORS: GROUP BY**

- GROUP BY *must* be paired with an aggregate function, the statistic we want to compute in the rows, in the SELECT statement.
- ▶ COUNT (\*) denotes counting up all of the rows. Other aggregate functions commonly available are AVG (average), MAX, MIN, and SUM.
- If we want to aggregate over the entire table, without results specific to any key, we can use an aggregate function in the SELECT clause and ignore the GROUP BY clause.

### **SQL OPERATORS: GROUP BY**

▶ Rossmann Stores example:

```
SELECT Store, SUM(Sales), AVG(Customers)
FROM rossmann_sales
GROUP BY Store
WHERE Open = 1;
```

### **ACTIVITY: KNOWLEDGE CHECK**

### **ANSWER THE FOLLOWING QUESTIONS**



1. Write a query that returns the total sales on the promotion and non-promotion days.

### **DELIVERABLE**

The requested query

### **SQL OPERATORS: ORDER BY**

▶ ORDER BY is used to sort the results of a query.

```
SELECT <columns>
FROM 
WHERE <condition>
ORDER BY <columns>
```

You can order by multiple columns in ascending (ASC) or descending (DESC) order.

### **SQL OPERATORS: ORDER BY**

▶ Rossmann Stores example:

```
SELECT Store, SUM(Sales) as total_sales, AVG(Customers)
FROM rossmann_sales
GROUP BY Store
WHERE Open = 1;
ORDER BY total_sales desc;
```

▶COUNT(\*) AS cnt renames the COUNT(\*) value to cnt so we can refer to it later in the ORDER BY clause.

### **SQL OPERATORS: JOIN**

▶ JOIN allows us to access data across many tables. We specify how a row in one table links to another.

```
SELECT a.Store, a.Sales, s.CompetitionDistance
FROM rossmann_sales a
JOIN rossmann_stores s
ON a.Store = s.Store
```

▶ Here, ON denotes an *inner* join.

### **SQL OPERATORS: JOIN**

- ▶ By default, most joins are an *Inner Join*, which means only when there is a match in both tables does a row appear in the results.
- If we want to keep the rows of one table even if there is no matching counterpart, we can perform an Outer Join.
- Outer joins can be LEFT, RIGHT, or FULL, meaning keep all of the left rows, all the right rows, or all the rows, respectively.

### INDEPENDENT PRACTICE

# PANDAS AND SQL

### **ACTIVITY: PANDAS AND SQL**

# EXERCISE

### **DIRECTIONS (40 minutes)**

- 1. Load the Walmart sales and store features data.
- 2. Create a table for each of those datasets.
- 3. Select the store, date and fuel price on days it was over 90 degrees.
- 4. Select the store, date and weekly sales and temperature.
- 5. What were average sales on holiday vs. non-holiday sales?
- 6. What were average sales on holiday vs. non-holiday sales when the temperature was below 32 degrees?

#### **DELIVERABLE**

Answers to the above questions

### INDEPENDENT PRACTICE

# EXTRASQL PRACTICE

### **ACTIVITY: EXTRA SQL PRACTICE**

#### **DIRECTIONS**



There are many options for extra SQL practice.

- **1. PG-Exercises**: The website pgexercises.com is a very good site for Postgres exercsises. Go <a href="here">here</a> to get started. Complete 3-5 questions in each of the following.
  - a. Simple SQL Queries
  - b. Aggregation
  - c. <u>Joins and Subqueries</u>

### **ACTIVITY: EXTRA SQL PRACTICE**

#### **DIRECTIONS**



**EXERCISE** 

There are many options for extra SQL practice.

- **2. Wagon:** This requires signing up for the Wagon service and downloading their application. It gives access to some sample databases.
  - a. Display all tracks on which Jimmy Page was the composer.
  - b. Who were the top five composers by number of tracks?
  - c. Who were the top five composers by length of tracks?
  - d. Select all of the albums from Led Zeppelin.
  - e. Count the number of albums per artist, and display the top 10.
  - f. Display the track name and album name from all Led Zeppelin albums.
  - g. Compute how many songs and how long (in minutes) each Led Zeppelin album was.

# INSTALLING POSTGRES

### **INSTALLING POSTGRES**

- ▶On a Mac, brew install postgres
- ▶brew will provide a few commands to make sure postgres runs on startup.
- ▶ If this is done, you can use the Postgres command line tool.

## **POSTGRES SHELL**

```
▶ Starting Postgres: psql
▶ Listing tables: \dt
▶ Creating a table
CREATE TABLE example(
    id int,
    name varchar(140),
    value float
```

);

### **POSTGRES SHELL**

▶ Inserting a row

```
INSERT INTO example VALUES(1, 'general assembly', 3.14);
```

▶ Querying the table

```
SELECT *
FROM example;
```

### CONCLUSION

# TOPIC REVIEW

### CONCLUSION

- While this was a brief introduction, databases are often at the core of any data analysis. Most analysis starts with retrieving data from a database.
- SQL is a key language that any data scientist should understand.
  - ▶SELECT: Used in every query to define the resulting columns
  - ▶WHERE: Filters rows based on a given condition
  - ▶ GROUP BY: Groups rows for aggregation
  - ▶JOIN: Combines two tables based upon a given condition

### CONCLUSION

- ▶ Pandas can be used to access data from databases as well. The result of the queries will end up in a Pandas dataframe.
- There is much more to learn about query optimization if one dives further!

# BEFORE NEXT CLASS

### **BEFORE NEXT CLASS**

# **DUE DATE**

▶ **Project**: Final Project, Part 3

### **LESSON**

# CREDITS

### **LESSON**

# Q&A

### **LESSON**

# EXIT TICKET

DON'T FORGET TO FILL OUT YOUR EXIT TICKET

## **THANKS!**

# NAME

- Optional Information:
- Email?
- Website?
- Twitter?