Selecting Data with SQL

Introduction

As a data scientist, the SQL query you'll likely use most often is **SELECT**. This lesson introduces how to use **SELECT** to subset and transform the columns of a database table.

Objectives

You will be able to:

- Retrieve a subset of columns from a table
- Create an alias in a SQL query
- Use SQL CASE statements to transform selected columns
- Use built-in SQL functions to transform selected columns

The Data

Below, we connect to a SQLite database using the Python sqlite3 library (documentation here):

```
# import module
import sqlite3
# create a connection
conn = sqlite3.connect('data.sqlite')
```

The database that you've just connected to is the same database you have seen previously, containing data about orders, employees, etc. Here's an overview of the database:

For this first section we'll be focusing on the employees table.

If we want to get all information about the employee records, we might do something like this (* means all columns):

```
# module
import pandas as pd
# pandas-sql-query
pd.read_sql("""SELECT * FROM employees;""", conn)
   employeeNumber
                    lastName firstName extension \
0
             1002
                                 Diane
                                           x5800
                      Murphy
1
             1056
                   Patterson
                                  Mary
                                           x4611
2
             1076
                   Firrelli
                                  Jeff
                                           x9273
3
                   Patterson
                               William
                                           x4871
             1088
```

4	1102	Bondur	Gerard	x5408			
5 6	1143 1165	Bow Jennings	Anthony Leslie	x5428 x3291			
7	1166	Thompson	Leslie	×4065			
8	1188	Firrelli	Julie	x2173			
9	1216	Patterson	Steve	x4334			
10	1286	Tseng	Foon Yue	x2248			
11	1323	Vanauf	George	x4102			
12 13	1337 1370	Bondur Hernandez	Loui Gerard	x6493 x2028			
14	1401	Castillo	Pamela	x2759			
15	1501	Bott	Larry	x2311			
16	1504	Jones	Barry	×102			
17	1611	Fixter	Andy	×101			
18	1612	Marsh	Peter	×102			
19	1619	King	Tom	x103			
20	1621	Nishi	Mami	×101			
21	1625	Kato	Yoshimi	x102			
22	1702	Gerard	Martin	x2312			
			email offi	.ceCode repo	ortsTo		
jobTitle				•			
	hy@cla	ssicmodelca	rs.com	1			
President				_	1000		
	so@cla	ssicmodelca	rs.com	1	1002		
<pre>VP Sales 2 ifirrel</pre>	li@cla	ssicmodelca	rs com	1	1002		VP
Marketing	116610	1331cmode tea	13100111	-	1002		VI
3 wpatters		ssicmodelca	rs.com	6	1056	Sales	
Manager (APA				_			
_		ssicmodelca	rs.com	4	1056	Sale	
Manager (EME 5 ab		ssicmodelca	rc com	1	1056	Sales	
Manager (NA)	owacta	1331CIIIOGE CCA	13.0011	1	1050	Jaces	
6 ljennin	gs@cla	ssicmodelca	rs.com	1	1143		
Sales Rep							
	on@cla	ssicmodelca	rs.com	1	1143		
Sales Rep	liacla	ssicmodelca	rc com	2	1143		
<pre>8 jfirrel Sales Rep</pre>	ri@cra	issicillodetca	i S. Colli	Z	1145		
	on@cla	ssicmodelca	rs.com	2	1143		
Sales Rep	0.1.6.0.10.			_			
	ng@cla	ssicmodelca	rs.com	3	1143		
Sales Rep							
	uf@cla	ssicmodelca	rs.com	3	1143		
Sales Rep 12 lbond	uracla	ccicmodol co	rc com	1	1100		
Sales Rep	uructa	ssicmodelca	i S. CUIII	4	1102		
	de@cla	ssicmodelca	rs.com	4	1102		
25 gilerilan				•	1102		

Sales Rep			
14 pcastillo@classicmode	elcars.com	4	1102
Sales Rep			
<pre>15 lbott@classicmode</pre>	elcars.com	7	1102
Sales Rep			
16 bjones@classicmode	elcars.com	7	1102
Sales Rep	_	_	
17 afixter@classicmode	elcars.com	6	1088
Sales Rep	2		1000
18 pmarsh@classicmode	elcars.com	6	1088
Sales Rep	.1		1000
19 tking@classicmode	elcars.com	6	1088
Sales Rep	oleans com	5	1056
20 mnishi@classicmode Sales Rep	ettal S. Com	5	1056
21 ykato@classicmode	alcars com	5	1621
Sales Rep	ccar 3. com	J	1021
22 mgerard@classicmode	elcars.com	4	1102
Sales Rep	3 (6) (6)	•	1102
54 165 1.6p			

Quick Note on String Syntax

When working with strings, you may have previously seen a 'string', a "string", a "''string'', or a """string"". While all of these are strings, the triple quotes have the added functionality of being able to use multiple lines within the same string as well as to use single quotes within the string. Sometimes, SQL queries can be much longer than others, in which case it's helpful to use new lines for readability. Here's the same example, this time with the string spread out onto multiple lines:

```
pd.read_sql("""
             SELECT
             FROM
                 employees;
             """, conn
   employeeNumber
                     lastName firstName extension
0
              1002
                                               x5800
                        Murphy
                                    Diane
1
              1056
                     Patterson
                                     Mary
                                               x4611
2
              1076
                     Firrelli
                                     Jeff
                                               x9273
3
              1088
                     Patterson
                                 William
                                               x4871
4
              1102
                        Bondur
                                   Gerard
                                               x5408
5
              1143
                           Bow
                                  Anthony
                                               x5428
6
              1165
                      Jennings
                                   Leslie
                                               x3291
7
              1166
                     Thompson
                                   Leslie
                                               x4065
8
                      Firrelli
                                    Julie
              1188
                                               x2173
9
              1216
                     Patterson
                                    Steve
                                               x4334
10
              1286
                         Tseng
                                Foon Yue
                                               x2248
```

11 12 13 14 15 16 17 18 19 20 21 22		1323 1337 1370 1401 1501 1504 1611 1612 1619 1621 1625 1702	Vanauf Bondur Hernandez Castillo Bott Jones Fixter Marsh King Nishi Kato Gerard	George Loui Gerard Pamela Larry Barry Andy Peter Tor Mami Yoshimi	i x64 d x20 d x27 d x23 d x1 d x1 d x1 i x1	93 28 59 11 92 91 92 93 91		
				email of	fficeCode	reportsTo		
jobTit	16			Ciliate	TICCCOUC	reportsio		
0		v@cla	ssicmodelca	ars com	1			
Presid		y we ca.	331cmode ee	11 3 1 COIII	_			
		o@cla	ssicmodelca	ars.com	1	1002		
VP Sal	•	oge ca.	331cmode ce	11 5 1 COIII	_	1002		
		i@cla	ssicmodelca	ars.com	1	1002		VP
Market		10000	331cmode ce	31 51 COIII		1002		V .
	_	n@cla	ssicmodelca	ars com	6	1056	Sales	
	er (APAC		331cmode ee	11 3 1 COIII	U	1030	Jaces	
4			ssicmodelca	ars com	4	1056	Sale	
	er (EMEA	_	331cmode ee	11 3 1 COIII	-	1030	Sacc	
5			ssicmodelca	ars com	1	1056	Sales	
	er (NA)	wee ca.	331cmode ce	21 3 1 COIII	_	1050	54 (65	
		s@cla	ssicmodelca	ars com	1	1143		
Sales		Jec ca.	331cmode ce	21 3 1 COIII	_	11.13		
		n@cla	ssicmodelca	ars.com	1	1143		
Sales			55±0545±0.	a. 5. co	_	11.0		
	•	i@cla	ssicmodelca	ars.com	2	1143		
Sales					_			
		n@cla:	ssicmodelca	ars.com	2	1143		
Sales					_			
10		g@cla:	ssicmodelca	ars.com	3	1143		
Sales		J C						
11		f@cla	ssicmodelca	ars.com	3	1143		
Sales	_	-						
12	lbondu	r@cla	ssicmodelca	ars.com	4	1102		
Sales		-						
		e@cla	ssicmodelca	ars.com	4	1102		
Sales		•						
		o@cla:	ssicmodelca	ars.com	4	1102		
Sales								
15		t@cla:	ssicmodelca	ars.com	7	1102		
Sales								
16		s@cla	ssicmodelca	ars.com	7	1102		
Sales	Rep							

17	afixter@classicmodelcars.com	6	1088
Sales 18	Rep pmarsh@classicmodelcars.com	6	1088
Sales	Rep		
19 Sales	tking@classicmodelcars.com	6	1088
20	mnishi@classicmodelcars.com	5	1056
Sales 21	Rep ykato@classicmodelcars.com	5	1621
Sales		J	1021
22	mgerard@classicmodelcars.com	4	1102
Sales	кер		

Unlike in Python, whitespace indentation in SQL is not used to indicate scope or any other important information. Therefore this:

```
SELECT *
  FROM employees;
```

(with two spaces in front of FROM)

is identical to this:

```
SELECT *
FROM employees;
```

(with zero spaces in front of FROM)

as far as SQL is concerned. However we will be aligning the right edge of the SQL keywords, using a "river" of whitespace down the center to improve legibility in this lesson, following this style guide. You will see multi-line SQL written with various different indentation styles, and you will want to check with your employer to learn what their style guide is.

Retrieving a Subset of Columns

Once we know what the column names are for a given table, we can select specific columns rather than using * to select all of them. This is achieved by replacing the * with the names of the columns, separated by commas.

For example, if we just wanted to select the last and first names of the employees:

```
lastName firstName

Murphy Diane

Patterson Mary

Firrelli Jeff

Patterson William

Bondur Gerard
```

We can also specify the columns in a different order than they appear in the database, in order to reorder the columns in the resulting dataframe:

```
pd.read sql("""
    SELECT
          firstName, lastName
    FROM
          employees;
""", conn
).head()
  firstName
               lastName
0
      Diane
                 Murphy
1
       Mary
              Patterson
2
       Jeff
               Firrelli
3
    William
             Patterson
4
     Gerard
                 Bondur
```

Additionally, we can use aliases (AS keyword) to change the column names in our query result:

```
pd.read_sql("""
    SELECT
         firstName AS name
    FROM
        employees;
""", conn
).head()
      name
0
     Diane
1
      Mary
2
      Jeff
3
   William
    Gerard
```

Note: the AS keyword is technically optional when assigning an alias in SQL, so you may see examples that don't include it. In other words, you could just say SELECT firstName name and it would work the same as SELECT firstName AS name. However we recommend being more explicit and including the AS, so that it's clearer what your code is doing.

Using SQL CASE Statements

CASE statements appear very frequently in SQL technical interview questions. They are a type of conditional statement, similar to if statements in Python. Whereas Python uses the keywords if, elif, and else, SQL uses CASE, WHEN, THEN, ELSE, and END.

CASE indicates that a conditional statement has begun, and END indicates that it has ended.

WHEN is similar to if, and then instead of a colon and an indented block, THEN indicates what should happen if the condition is true. After the first THEN has executed, it skips to the end, so each subsequent WHEN is more like elif in Python.

ELSE is essentially the same as else in Python.

CASE to Bin Column Values

One of the most common use cases for CASE statements is to bin the column values. This is true for both numeric and categorical columns.

In the example below, we use the jobTitle field to bin all employees into role categories based on whether or not their job title is "Sales Rep":

```
# Trial-test # brian-added
pd.read_sql("""SELECT firstName, lastName, jobTitle From employees
WHERE jobTitle = 'Sales Rep';""", conn).head()
  firstName
              lastName
                         jobTitle
0
     Leslie
              Jennings
                        Sales Rep
1
     Leslie
              Thompson
                        Sales Rep
2
                        Sales Rep
      Julie
              Firrelli
3
      Steve
             Patterson
                        Sales Rep
   Foon Yue
                        Sales Rep
                 Tseng
pd.read sql("""
SELECT firstName, lastName, jobTitle,
       WHEN jobTitle = "Sales Rep" THEN "Sales Rep"
       ELSE "Not Sales Rep"
       END AS role
  FROM employees;
""", conn).head(10)
  firstName
              lastName
                                     jobTitle
                                                         role
                                    President
0
      Diane
                                               Not Sales Rep
                Murphy
1
       Mary
             Patterson
                                     VP Sales
                                               Not Sales Rep
2
       Jeff
             Firrelli
                                 VP Marketing
                                               Not Sales Rep
3
    William Patterson
                        Sales Manager (APAC)
                                               Not Sales Rep
4
     Gerard
                Bondur
                         Sale Manager (EMEA)
                                               Not Sales Rep
5
    Anthony
                   Bow
                          Sales Manager (NA)
                                               Not Sales Rep
6
     Leslie
              Jennings
                                    Sales Rep
                                                   Sales Rep
7
     Leslie
              Thompson
                                    Sales Rep
                                                   Sales Rep
```

CASE to Make Values Human-Readable

Another typical way to use CASE is to translate the column values into something that your eventual audience will understand. This is especially true of data that is entered into the database as a "code" or "ID" rather than a human-readable name.

In the example below, we use a CASE statement with multiple WHEN's in order to transform the officeCode column into an office column that uses a more meaningful name for the office:

```
pd.read sql("""
SELECT firstName, lastName, officeCode,
       WHEN officeCode = "1" THEN "San Francisco, CA"
       WHEN officeCode = "2" THEN "Boston, MA"
       WHEN officeCode = "3" THEN "New York, NY"
       WHEN officeCode = "4" THEN "Paris, France"
       WHEN officeCode = "5" THEN "Tokyo, Japan"
       END AS office
  FROM employees;
""", conn).head(10)
              lastName officeCode
                                               office
  firstName
0
      Diane
                Murphy
                                1 San Francisco, CA
1
             Patterson
                                1 San Francisco, CA
       Mary
2
       Jeff
             Firrelli
                                1 San Francisco, CA
3
    William Patterson
                                6
                                                 None
4
                                4
     Gerard
                Bondur
                                        Paris, France
5
    Anthony
                                1 San Francisco, CA
                   Bow
6
     Leslie
                                1 San Francisco, CA
              Jennings
7
     Leslie
              Thompson
                                1 San Francisco, CA
                                2
8
      Julie
              Firrelli
                                           Boston, MA
9
                                2
      Steve
             Patterson
                                           Boston, MA
```

Note that because **we did not specify a name for officeCode** "6", and did not include an ELSE, the associated office value for William Patterson is NULL (represented as None in Python).

There is also a shorter syntax possible if all of the WHENs are just checking if a value is equal to another value (e.g. in this case where we are repeating officeCode = over and over). Instead we can specify officeCode right after CASE, then only specify the potential matching values:

```
WHEN "4" THEN "Paris, France"
       WHEN "5" THEN "Tokyo, Japan"
       END AS office
  FROM employees;
""", conn).head(10)
  firstName
              lastName officeCode
                                               office
0
      Diane
                Murphy
                                1 San Francisco, CA
1
       Mary
             Patterson
                                1 San Francisco, CA
2
       Jeff
                                1 San Francisco, CA
              Firrelli
3
    William
             Patterson
                                6
                                                 None
4
     Gerard
                Bondur
                                4
                                        Paris, France
5
    Anthony
                                1 San Francisco, CA
                   Bow
6
                                1 San Francisco, CA
     Leslie
              Jennings
7
     Leslie
              Thompson
                                1 San Francisco, CA
8
              Firrelli
                                 2
      Julie
                                           Boston, MA
9
      Steve Patterson
                                 2
                                           Boston, MA
```

Using Built-in SQL Functions

Similar to the Python built-in functions, SQL also has built-in functions. The available functions will differ somewhat by the type of SQL you are using, but in general you should be able to find functions for:

- String manipulation
- Math operations
- Date and time operations

For SQLite in particular, if you are looking for a built-in function, start by checking the core functions page, mathematical functions page, and/or date and time functions page.

Built-in SQL Functions for String Manipulation

length

Let's start with an example of a SQL built-in function that is very similar to one we have in Python: length (documentation here). This works very similarly to the len built-in function in Python. For a string, it returns the number of characters.

If we wanted to find the length of the first names of all employees, that would look like this:

upper

Now let's say we wanted to return all of the employee names in all caps. Similar to the Python string method, this SQL function is called upper (documentation here). However, since it's a built-in function and not a method, the syntax looks like:

```
upper(column_name)
```

```
and not column name.upper().
```

As you get more comfortable with Python and SQL, distinctions like this will get more intuitive, but for now don't worry if you have to look it up every time!

Here is an example using upper:

```
pd.read sql("""
  SELECT
        firstName, upper(firstName) AS name in CAPS
  FROM
        employees;
""", conn).head()
  firstName name in CAPS
0
      Diane
                    DIANE
1
       Mary
                     MARY
2
       Jeff
                     JEFF
3
    William
                  WILLIAM
4
     Gerard
                   GERARD
```

substr

Another form of string manipulation you might need is finding a substring (subset of a string). In Python, we do this with string slicing. In SQL, there is a built-in function that does this instead. For SQLite specifically, this is called substr (documentation here).

Let's say we wanted just the first initial (first letter of the first name) for each employee:

```
firstName first_initial Length_of_Substr
0
      Diane
                                              1
1
       Mary
                          М
                                              1
2
                                              1
       Jeff
                          J
3
                                              1
    William
                          W
4
     Gerard
                          G
                                              1
```

If we wanted to add a . after each first initial, we could use the SQLite | | (concatenate) operator. This works similarly to + with strings in Python:

We can also combine multiple column values, not just string literals. For example, below we combine the first and last name:

Hmm, that looks a bit odd. Let's concatenate those column values with a space (" ") string literal:

```
employees;
""", conn).head()

full_name

Diane Murphy

Mary Patterson

Jeff Firrelli

William Patterson

Gerard Bondur
```

That looks better!

Built-in SQL Functions for Math Operations

For these examples, let's switch over to using the orderDetails table:

```
pd.read sql("""SELECT * FROM orderDetails;""", conn).head()
  orderNumber productCode quantityOrdered priceEach orderLineNumber
0
        10100
                  S18 1749
                                         30
                                               136.00
                                                                     2
1
        10100
                  S18 2248
                                         50
                                                55.09
2
                 S18 4409
                                         22
                                                75.46
                                                                     4
        10100
3
                                                                     1
        10100
                  S24 3969
                                         49
                                                35.29
4
                  S18 2325
                                         25
                                               108.06
        10101
```

round

Let's say we wanted to round the price to the nearest dollar. We could use the SQL round function (documentation here), which is very similar to the the Python round:

```
pd.read sql("""
  SELECT
        priceEach, round(priceEach) AS rounded price
  FROM
        orderDetails;
""", conn).head()
  priceEach
            rounded price
     136.00
                      136.0
1
      55.09
                       55.0
2
      75.46
                       75.0
3
      35.29
                       35.0
4
     108.06
                      108.0
```

CAST

The previous result looks ok, but it's returning floating point numbers. What if we want integers instead?

In Python, we might apply the int built-in function. In SQLite, we can use a CAST expression (documentation here):

```
pd.read sql("""
SELECT
      CAST(round(priceEach) AS INTEGER) AS rounded price int
FROM
      orderDetails;
""", conn).head()
   rounded price int
0
                  136
1
                   55
2
                   75
3
                   35
4
                  108
```

Basic Math Operations

Just like when performing math operations with Python, you don't always need to use a function. Sometimes all you need is an operator like +, -, /, or *. For example, below we multiply the price times the quantity ordered to find the total price:

```
pd.read sql("""
SELECT
      priceEach * quantityOrdered AS Total Price
FROM
      orderDetails;
""", conn).head()
# alternative
## CAST((priceEach * quantityOrdered) AS INTEGER) AS Total Price
   Total Price
0
       4080.00
1
       2754.50
2
       1660.12
3
       1729.21
       2701.50
```

Built-in SQL Functions for Date and Time Operations

For these examples, we'll look at yet another table within the database, this time the orders table:

```
pd.read_sql("""SELECT * FROM orders;""", conn).head()
  orderNumber orderDate requiredDate shippedDate status \
     10100 2003-01-06 2003-01-13 2003-01-10 Shipped
```

```
1
        10101 2003-01-09
                           2003-01-18 2003-01-11
                                                   Shipped
                           2003-01-18 2003-01-14 Shipped
2
        10102 2003-01-10
3
        10103 2003-01-29
                           2003-02-07 2003-02-02 Shipped
4
        10104 2003-01-31
                           2003-02-09 2003-02-01 Shipped
                comments customerNumber
0
                                    363
1
  Check on availability.
                                    128
2
                                    181
3
                                    121
4
                                    141
```

What if we wanted to know how many days there are between the requiredDate and the orderDate for each order? Intuitively you might try something like this:

```
pd.read_sql("""
    SELECT
           requiredDate - orderDate
    FROM
           orders;
   , conn)
     requiredDate - orderDate
0
                               0
1
                               0
2
                               0
3
                               0
4
                               0
. .
321
                               0
322
                               0
323
                               0
324
                               0
325
                               0
[326 rows x 1 columns]
```

Clearly that didn't work.

It turns out that we need to specify that we want the difference in *days*. One way to do this is using the julianday function (documentation here):

```
days from order to required
0
1
                                   9
2
                                   8
3
                                   9
                                   9
4
321
                                   8
322
                                  12
323
                                   6
324
                                   8
325
                                   7
[326 rows x 1 columns]
```

If we wanted to select the order dates as well as dates 1 week after the order dates, that would look like this:

```
pd.read sql("""
    SELECT
          orderDate, date(orderDate, "+7 days") AS one week later,
date(orderDate, "+14 days") AS two week later
        orders;
""", conn)
      orderDate one_week_later two_week_later
0
     2003-01-06
                     2003-01-13
                                    2003-01-20
1
     2003-01-09
                     2003-01-16
                                    2003-01-23
2
     2003-01-10
                     2003-01-17
                                    2003-01-24
3
     2003-01-29
                     2003-02-05
                                    2003-02-12
4
     2003-01-31
                     2003-02-07
                                    2003-02-14
321
     2005-05-29
                     2005-06-05
                                    2005-06-12
322
     2005-05-30
                     2005-06-06
                                    2005-06-13
323
     2005-05-30
                     2005-06-06
                                    2005-06-13
324
     2005-05-31
                     2005-06-07
                                    2005-06-14
     2005-05-31
                     2005-06-07
                                    2005-06-14
325
[326 rows x 3 columns]
```

You can also use the **strftime** function, which is very similar to the Python version. This is useful if you want to split apart a date or time value into different sub-parts. For example, here we extract the year, month, and day of month from the order date:

```
pd.read_sql("""
    SELECT orderDate,
        strftime("%m", orderDate) AS month,
        strftime("%Y", orderDate) AS year,
```

```
strftime("%d", orderDate) AS day
    FROM orders;
""", conn).head()
    orderDate month
                     year day
  2003-01-06
                 01
                     2003
                           06
1
  2003-01-09
                 01
                     2003
                           09
  2003-01-10
                 01
                     2003
                          10
3
  2003-01-29
                 01
                     2003
                           29
  2003-01-31
                 01
                     2003
                           31
```

Now that we are finished with our queries, we can close the database connection.

```
conn.close()
```

Summary

In this lesson, you saw how to execute several kinds of SQL SELECT queries. First, there were examples of specifying the selection of particular columns, rather than always using SELECT * to select all columns. Then you saw some examples of how to use CASE to transform column values using conditional logic. Finally, we walked through how to use built-in SQL functions, particularly for string, numeric, and date/time fields.