MTH 3270 Notes 10

15 Database Querying Using SQL (15)

15.1 Introduction

- Computers have two ways of storing data:
 - Memory (or RAM): Only a few Gb of data can be stored here, but it can be quickly accessed.
 Actively running programs (like R) store data they're currently working on (like objects in the R Workspace) here.
 - Hard Disk: Hundreds or thousands of Gb of data can be stored here, but accessing it is much slower.
 Saved files (that aren't being worked on by an actively running program) are stored here.
- Relatively small data sets can be stored in in memory. But not larger ones. For example, a data set with **100 columns** and **1 million rows** takes up about **three-quarters** of a **Gb** of memory:

```
n <- 100 * 1000000

# Generate a matrix of random numbers uniformly distributed between 0 and 1:
x <- matrix(runif(n), nrow = 1000000, ncol = 100)

print(object.size(x), units = "Mb")

## 762.9 Mb</pre>
```

An R data frame with 10 million rows would hog memory, resulting in slowed computations, and one with 100 million rows wouldn't fit in memory.

- For large data sets, instead of trying to read *all* the data into R, it's preferable to store it on a hard disk and be allowed to *access* only *portions* of the data.
- Two commonly used forms of data storage:
 - A *flat file* is a data file (.txt, .csv, .xlsx, etc.) stored in a *table* format, i.e. containing just **rows** and **columns** (possibly with a **header** row).
 - A *relational database* (or just *database*) is a collection of *linkable tables* (each of which contains just **rows** and **columns**).
- SQL ("structured query language") is a programming language for data stored as a database.
- There are a few different implementations of SQL, also called database management systems or DBMSs:
 - Oracle
 - Microsoft SQL Server
 - SQLite
 - MySQL
 - PostgreSQL

Some of them use different dialects of the SQL language.

• **SQL** uses a *client-server* model – data are stored in a **database** on the **hard disk** of a *server* computer somewhere, and there's a *server program* running on that computer. You can connect to that server over the internet from a *client* computer using one of several *client programs*.

(In some cases, the server and client computers might be the same, in which case the internet isn't needed.)

• For example, recall (Class Notes 3) that the "nycflights13" package contained a flights data frame with 336,776 rows, each one a flight that departed New York in 2013.

But the *full* **flights** data set is **much larger**, with more than **169 million** flights going back to 1987 – each comprising a different row in the table. The data would occupy nearly **20 Gb** as a **.csv** file on a computer.

Instead, the data have been written (by the textbook authors) to a **table** (named **flights**) in a **database** (named **airlines**) on a **server**. We can use **SQL** to access *only* the rows that interest us.

• A *query* is a **request** for **data**. **Queries** in **SQL** are *statements* that start with the **SELECT** *keyword* and consist of several *clauses* involving *other* **keywords**, which have to be written in this order:

```
SELECT
              # Select columns to extract from a table.
                                                          Similar to select()
              # in "dpyr".
FROM
              # Specify the table where data are stored.
JOIN
              # Combine together two tables based on a key.
              # inner_join() and left_join() from "dplyr".
              # Filter rows according to some criteria. Analogous to filter()
WHF.R.F.
              # in "dplyr".
GROUP BY
              # Group together rows of a table according to values of a
              # categorical variable. Analogous to group_by() in "dplyr".
              # Like a WHERE clause that operates on the result set--not the
HAVING
              # rows themselves. Analogous to applying a second filter()
              # command in "dplyr", after the rows have already been aggregated.
ORDER BY
              # Specifies a condition for ordering the rows. Analogous to
              # arrange() in "dplyr".
              # Restrict the number of rows in the output. Similar to head()
LIMIT
              # and slice() in R.
```

- **SQL queries** can be executed from **within R** by:
 - 1. Connecting to the database.
 - 2. Sending the query.
 - 3. Disconnecting from the database.

These tasks are carried out using the following functions from the "DBI" package.

```
dbConnect()  # Connect to a database stored on a server.
dbGetQuery()  # Execute a query statement on a connected database.
dbDisconnect()  # Disconnect from the database.
```

A list of the tables in a database can be obtained using the following function (also from "DBI").

```
	ext{dbListTables}() # Returns a list of the tables in a connected database.
```

15.2 Connecting to a Server Using dbConnect() and Querying Using dbGetQuery()

• The first step to conducting a database query is to connect to the server on which the database is stored.

In dbConnect(), you'll specify an **SQL** implementation (RSQLite, MySQL, PostgreSQL, etc.) via the drv (*driver*) argument, and you'll specify authentication information via dbname, host, user, password, and sometimes port.

The airlines database was set up using MySQL (by the textbook authors) on an AWS server. Connecting to a MySQL database requires the MySQL() function from the "RMySQL" package.

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```
MySQL() # Allows you to connect to a MySQL database.
```

Below, we open a **connection** to the **airlines database**.

• Once we're connected, we can view a list of the tables in the database using the following:

```
# This will return a list of the tables in the database:
dbListTables(db_con)
```

- SQL queries (and other SQL statements) can be carried out using dbGetQuery(), which takes arguments conn, an open connection to a server, and statement, the SQL query (or other SQL statement) to perform on the database or specific table.
- For example, DESCRIBE will return a *description* of the variables ("fields") in a table, similar to str() in R. Descriptions of the flights and airports tables (from the flights database) are obtained via:

- It's customary for keywords in SQL statements to be written in capital letters and for each statement to end with a semicolon.
- The last step in running a query is to disconnect from the server.

A **connection** will **time out** (automatically disconnect) if it's inactive for several minutes. To **disconnect** manually, use:

```
# Disconnect from the server:
dbDisconnect(conn = db_con)
```

Section 15.2 Exercises Exercise 1 Load the following packages. library(DBI) library(RMySQL)

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Now open a **connection** to the **airlines database** using the command involving dbConnect() (from "DBI") given above. Save the **connection** as, say, db_con.

- a) What is the class of the object db_con? Use class().
- b) How many tables are in the airlines database and what are their names? Use dbListTables().
- c) Recall that DESCRIBE is used to get a description of the contents of a table. Using dbGetQuery(), with statement = "DESCRIBE airports;", how many variables ("fields") are in the airports table?
- d) Using dbGetQuery(), with statement = "DESCRIBE flights;", how many variables ("fields") are in the flights table?

15.3 SELECT ... FROM

• As mentioned, a query is a request for data. We carry out **SQL** queries in R using dbGetQuery(), which returns the results as a *data frame*.

Every SQL query must contain, at a minimum, SELECT and FROM clauses.

- SELECT is used to select variables (columns) to extract from a table, similar to select() in "dpyr".
- FROM is used to specify the table where the data are stored.
- For example, to select specific variables (columns) from flights, specify their names after SELECT:

(Output not shown.)

Warning: LIMIT 0, 6 limits the query to just the *first six* observations (rows). If we *didn't* specify LIMIT 0, 6, the command would attempt to retrieve *all* 196 million observations (rows).

• To select all variables (columns) from flights, type:

(Output not shown.)

Above, the asterisk indicates all variables should be selected.

• We can also form and select new columns from existing ones, similar to using mutate() in "dplyr".

For example, below, we form a *new column* containing the travel times of the flights (arr_time minus dep_time), using AS to give it the name trvl_time:

(Output not shown.)

As another example, below, CONCAT combines (or "concatenates") two existing columns (lat and lon) from airports to form a new one, coords, analogous to using unite() in "tidyr":

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(Output not shown.)

Section 15.3 Exercises

Exercise 2 Use dbGetQuery() with SELECT and FROM to query the flights table to retrieve just the carrier and tailnum variables (columns).

Warning: Make sure to use LIMIT to limit the number of rows returned in your query.

Report your R command(s) (or just your SQL statement).

Exercise 3 Now use dbGetQuery() with SELECT and FROM to form a query to retrieve all the variables (columns) from the carriers table.

This time, save the data returned as, say, my.carriers in R.

(There's no need to LIMIT the number of rows returned by the query for the carriers table).

a) Confirm that the data set returned by dbGetQuery() is a data frame by typing:

```
is.data.frame(my.carriers)
```

b) You can find out how much memory an object occupies in R using the object.size() function and its print() method.

How much memory does my.carriers occupy? Find out by typing:

```
print(object.size(my.carriers), units = "Kb")
or
print(object.size(my.carriers), units = "Mb")
```

Exercise 4 Now use dbGetQuery() with SELECT and FROM to form a query to retrieve all the variables (columns) from the airports table.

Save the data returned as, say, my.airports in R.

(There's no need to LIMIT the number of rows returned by the query for the airports table).

- a) Each observation (row) in the my.airports data frame is an airport. How many rows does the data frame have? Use str() or nrow() or dim().
- b) **How many columns** (variables) does the data frame have?

Exercise 5 Recall that we can use dbGetQuery() with SELECT and FROM to form and select new columns from existing ones in a table, similar to using mutate() in "dplyr".

Form a new column containing the travel speeds (in mph) of the flights (distance divided by air_time, then multiplied by 60), using AS to give it the name trvl_speed.

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Warning: Make sure to use LIMIT to limit the number of rows returned in your query.

Report your **R** command(s) (or just your **SQL** statement).

15.4 WHERE

• WHERE is used to *filter* observations (rows) according to some criteria. It's analogous to the filter() function in "dplyr".

Logical operators AND, OR, and NOT can be used with WHERE to specify the condition the retrieved rows should meet.

• For example, to *filter* rows corresponding to flights on June, 26, 2013 from *Bradley International Airport* (BDL) from the **flights table**, type:

(Output not shown.)

• We can simultaneously select columns and filter rows in the same query, for example:

(Output not shown.)

• As mentioned, AND, OR, and NOT can be used in a WHERE clause.

These operations aren't performed left to right. Rather, the ANDs are performed before the ORs.

Parentheses can be used, though, to control the order of these operations.

For example, below COUNT is used to count rows. This first command (DON'T RUN IT) returns a count of 557.874:

It's equivalent to this:

On the other hand, the following command only returns a count of 2,542:

• We can use a BETWEEN clause or an IN clause to filter rows using a *range* or *set* of values, respectively. For example, the following commands (DON'T RUN THEM) *both* return flights on the 27th, 28th, and 29th of the June:

• WHERE can operate on **functions** of columns (i.e. values *computed* from columns).

For example, below, we filter on travel times of the flights (computed as arr_time minus dep_time), retrieving only those flights whose travel times were more than 10 hours (600 minutes):

Section 15.4 Exercises

Exercise 6 Report R command(s) (or just the SQL statements) involving dbGetQuery() and the logical operators (AND, OR, and NOT) with the flights table to retrieve flights meeting the following conditions.

Warning: Make sure to use LIMIT to limit the number of rows returned in your query.

- a) Had an arrival delay of more than hours (arr_delay more than 120 minutes).
- b) Flew to Houston (i.e. had a dest of IAH or HOU).
- c) Were operated by United, American, or Delta (i.e. had a carrier of UA, AA, or DL).
- d) Departed in summer (July, August, or September, i.e. month 7, 8, or 9).
- e) Departed between midnight and 6:00 AM, inclusive (dep_time at 2400 or between 0 and 600, inclusive).
- f) Were operated by United (carrier UA), departed in July (month 7), and had an arrival delay of more than two hours (arr_delay more than 120 minutes).

15.5 GROUP BY

• GROUP BY will *group* together (*aggregate*) rows of a **table** according to values of a categorical variable and then reduce each *group* of rows to a *single* row by computing a summary statistic for each *group*.

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It's analogous to group_by() followed by summarize() in "dplyr".

COUNT(), SUM(), and AVG(), STDEV(), MIN(), and MAX(), etc. can be used to specify the summary statistic to be computed.

• For example, we know that there were **65** flights that left *Bradley Airport* on June 26th, 2013, but how many belonged to each airline carrier?

To answer this, we use GROUP BY to group the individual flights based on who the carrier was and COUNT() to count the flights:

(Output not shown.)

GROUP BY can summarize *more than one variable* at a time. How many flights were there *and* what was the average departure delay for each carrier?

(Output not shown.)

• We can group by values of more than one categorical variable, for example by carrier and destination below:

(Output not shown.)

Section 15.5 Exercises

Exercise 7 We know there were **65** flights that left *Bradley Airport* on June 26th, 2013, but what was the shortest departure delay for each airline carrier? What was the longest?

- a) Use dbGetQuery() with GROUP BY and MIN() to retrieve the *shortest* departure delay for each carrier.

 Report your R command(s) (or just your SQL statement).
- b) GROUP BY can summarize more than one variable at a time. Modify your command(s) from part a to use MIN() and MAX() to retrieve the shortest and longest departure delays for each carrier.

Report your R command(s) (or just your SQL statement).

Exercise 8 This problem concerns the GROUP BY and AVG() functions. It uses the flights table.

a) Explain in words what the following command does (recall that dest is the destination of the flight):

15.6 ORDER BY

- ORDER BY is used to specify a condition for ordering the rows. It's analogous to arrange() in "dplyr".
- For example, to order flights from Bradley Airport on June, 26, 2013 by departure delay, type:

(Output not shown.)

ASC and DESC can be used to specify ascending or descending order. For example, to order the flights descending by departure delay, use DESC:

• Combining GROUP BY with an ORDER BY clause will bring the most interesting results to the top.

For example, there were 22,258 flights that left *Bradley Airport* in the year 2013. Which destinations are most common from *Bradley Airport* in 2013?

Note that *derived* columns like numFlights above *can* be referenced in the ORDER BY clause because ORDER BY operates on the *retrieved* data, not on the rows of the original data.

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As another example, which of those destinations had the lowest average arrival delay time?

The output (not shown) tells us Cleveland had the lowest average arrival delay – more than 13 minutes ahead of schedule – among destinations of flights from *Bradley* in 2013.

Section 15.6 Exercises

Exercise 9 This problem concerns ORDER BY combined with GROUP BY and AVG(). It uses the flights table.

There were 22,258 flights that left Bradley Airport in the year 2013.

- a) Use GROUP BY with AVG() and ORDER BY to determine the destination dest for which the average travel time (air_time) from *Bradley Airport* was *shortest* in 2013. Report the (abbreviated) destination (dest) name.
- b) Now use GROUP BY with AVG() and ORDER BY to determine the destination dest for which the average travel time (air_time) from *Bradley Airport* was *longest* in 2013. Report the (abbreviated) destination (dest) name.

Exercise 10 This problem concerns ORDER BY combined with GROUP BY and COUNT(). It uses the flights table.

There were 22,258 flights that left *Bradley Airport* in the year 2013.

- a) Use GROUP BY with COUNT() and ORDER BY to determine to determine which dest (i.e. which destination) was flown to the *most* times out of *Bradley Airport* in 2013. Report the (abbreviated) destination (dest) name.
- b) Use GROUP BY with COUNT() and ORDER BY to determine which tailnum (i.e. which individual airplane) flew the *most* times out of *Bradley Airport* in 2013. Report the tailnum value of the airplane.

15.7 HAVING

• HAVING is like a WHERE clause, but it retrieves a subset of rows from a table that's *already* been summarized by group (i.e. after the *full table's* rows have been *aggregated*).

It's analogous to applying a filter() command in "dplyr" after the rows have already been grouped and summarized using group_by() and summarize().

• For example, although flights to Cleveland had the lowest average arrival delay, there were only 57 flights that went to from *Bradley* to Cleveland in all of 2013.

It probably makes more sense to consider only those destinations that had, say, at least two flights per day.

We can filter the result set from the query made at the end of Section 15.6 using a HAVING clause:

Note that above, the HAVING clause needs numflights, which is is only available in the **result set** retrieved by the SELECT, FROM, WHERE, and GROUP BY clauses.

Section 15.7 Exercises

Exercise 11 There were 22,258 flights that left Bradley Airport in the year 2013.

There were 25 destinations of those flights from *Bradley*:

a) Alter the command above using HAVING so that it only returns destinations for which the average arrival delay is positive (avg_arr_delay > 0).

Report your R command(s) (or just your SQL statement).

b) Now alter your command from part a so that it only returns destinations for which the average arrival delay is positive (avg_arr_delay > 0) and the total number of flights was more than 1,000 (numFlights > 1000).

Report your R command(s) (or just your SQL statement).

Exercise 12 There were 22,258 flights that left Bradley Airport in the year 2013.

Those flights from *Bradley* were made by 12 airline carriers.

Use dbGetQuery() with SELECT, FROM, WHERE, GROUP BY, and HAVING to query the flights table to:

- 1. Group the flights from *Bradley* in 2013 by airline carrier (carrier).
- 2. Find the average departure delay for each carrier (use AVG() with dep_delay).
- 3. Retrieve just the carriers whose average departure delays were longer than 10 minutes.

Report your **R** command(s) (or just your **SQL** statement).

15.8 LIMIT

• We've seen that LIMIT is used to restrict the number of rows in the output, similar to using head() in R.

But LIMIT can also be used to select a specified number of rows starting from a specific row, similar to

```
slice() in "dplyr".
```

For example, this query will return just the 4th-7th flight destinations returned by the previous query.

Section 15.8 Exercises

Exercise 13 This exercise concerns LIMIT clauses.

Recall that this query returns just the 4th-7th flight destinations returned by a previous query:

How would you modify the query above so that it returns instead the 6th-9th flight destinations?

15.9 **JOIN**

- Recall that **SQL** is a **relational** database management system the **relations** between *linkable* **tables** allow for **queries** to tie together information from **multiple tables**.
- JOIN is used to **combine together** two (or more) tables based on a **key** variable (column) in each table, analogous to inner_join() and left_join() from "dplyr".
- There are four pieces of information you need to specify in order to join two tables:
 - The **name** of the **first table** that you want to join.
 - (Optional) The **type** of **join** that you want to use.
 - The name of the second table that you want to join.
 - The **key** columns in the two **tables**, i.e. the **condition(s)** under which you want the row in the **first table** to **match** the rows in the **second table**.
- In practice, the JOIN syntax varies among **SQL implementations**. In **MySQL**, OUTER JOINs (aka **full joins**) are *not* available, but the following **join types** are:
 - JOIN: includes all of the rows that are present in **both tables** and **match**.
 - LEFT JOIN: includes all of the rows that are present in the **first table**. Rows in the **first table** that have **no match** in the **second** are filled with NULLs.
 - RIGHT JOIN: include all of the rows that are present in the second table. This is the opposite of a LEFT JOIN.

- CROSS JOIN: the so-called **Cartesian product** of the two tables. Thus, **all possible combinations** of rows **matching** the **joining condition** are returned.
- For example, recall that in the flights table, the origin and destination of each flight are recorded.

(Output not shown.)

The flights table contains only the FAA three-character codes for the airports – not the full airport names.

It would be convenient to have the airport names in the flights table, but it would be storage-inefficient.

The solution is to store information about *airports* in the airports table (which only has 1,458 rows, not 169 million as flights has), along with the three-character codes – the *keys* – and to only store these *keys* in the flights table.

We can then use these **keys** to **join** the two **tables** together in our **query**.

• For our query, we specify the table (airports) we want to join onto flights and the condition by which we want to match rows in flights with rows in airports.

In this case, we want the *destination* airport code in the dest column of flights to be **matched** to the faa airport code in airports. These are the **key** columns.

We also specify that we want to see the full *airport* name (name column from the airports table) in the result set.

Note there are also codes in flights for the airline *carriers*. The full name of each *carrier* is stored in the carriers table.

We can **join** the **carriers** table to our **result set** from above to retrieve the full name of each *carrier*.

(Above, the name columns from airports and carriers were renamed dest_name and carrier_name, otherwise there would've been two name columns in the result set.)

Finally, to retrieve the name of the *originating airport* (in addition to the *destination airport*), we can join onto the airports table more than once.

Here, so-called *table aliases* (a1 and a2 below) are necessary:

Section 15.9 Exercises

Exercise 14 This exercise involves a JOIN clause.

There were 65 flights that left *Bradley Airport* on June 26th, 2013.

For this query, you'll need to join the airports table onto the flights table, matching the destination airport code (dest column) of flights to the airport code (faa column) in airports.

You'll also need to see the *destinations*, *flight numbers*, and airline *carrier* codes (dest, flight, and carrier columns) from the flights table and the full *airport* names (name column) from the airports table in the result set.

Use dbGetQuery() to answer the following problem. List the full name of the destination airport of flight EV 4714 from from Bradley on June 26th, 2013.

Exercise 15 This exercise involves a JOIN clause.

There were **65** flights that left *Bradley Airport* on June 26th, 2013.

For this query, you'll need to join the carriers table onto the flights table, matching the airline carrier code (carrier column) of flights to the carrier code (carrier column) in carriers.

You'll also need to see the *destinations* and *flight numbers* and (dest and flight columns) from the flights table and the full *carrier* names (name column) from the carriers table in the result set.

Use dbGetQuery() to answer the following problems.

- a) List the full airline *carrier* name and *flight number* for all flights between *Bradley Airport* (BDL) and MSP on June 26th, 2013.
- b) List the *destination airport* codes and *flight numbers* for the **three** flights from *Bradley* made by Mesa Airlines Inc. on June 26th, 2013.

15.9.1 **LEFT JOIN**

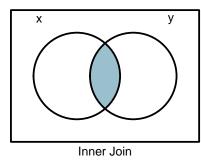
- JOIN (discussed above) returns *only* the rows that are present in **both tables** and **match**, i.e. it performs an *inner join*, the same as inner_join() in "dplyr".
- LEFT JOIN returns all of the rows in the first table, regardless of whether they have a match in the second. Rows in the first table that have no match in the second are filled with NULLs (converted NAs in R). It's equivalent to inner_join() in "dplyr".
- RIGHT JOIN does the **opposite** of LEFT JOIN. It returns *all* of the rows in the **second table** *regardless* of whether they have a match in the first.

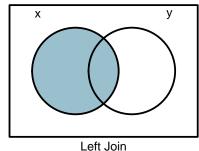
• For example, here's a **table** with **names** and **ages** of three people, and another **table** containing **names** and **heights** of three people, only two of which match the first **table**.

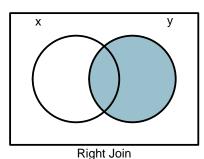
The Name variable is common to both tables, and is used as the key to match rows:

$\mathbf{Table} \ \mathbf{X}$		
Name	Age	
John	23	
Karen	27	
Ann	19	

Table Y			
Name	Height		
Karen	63		
Ann	65		
Karl	68		







JOIN \mathbf{X},\mathbf{Y}		
Name	Age	Height
Karen	27	63
Ann	19	65

LEFT	JOIN	X, Y
Name	Age	Height
John	23	NA
Karen	27	63
Ann	19	65

RIGHT	JOIN	X, Y
Name	Age	Height
Karen	27	63
Ann	19	65
Karl	NA	68

(Note the NAs produced by LEFT JOIN and by RIGHT JOIN.)

• As another example, there are flights in the **flights table** whose destination airport (SJU) has no corresponding entry in **airports**.

LEFT JOIN returns all 65 flights from *Bradley Airport*, *including* the flight to SJU, but with NA as its destination airport name for that flight:

JOIN doesn't include the flight to SJU in the result set at all:

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Section 15.9 Exercises

Exercise 16 This exercise concerns LEFT JOIN.

There were 50 flights that left Palm Beach International Airport ('PBI') on June 26th, 2013.

Explain in words why, below, LEFT JOIN returns all **50** flights (with NA as one of the *destination airport names*), but JOIN only returns **49**.

15.10 UNION

- Two separate SQL queries can be combined using a UNION clause. This is analogous to rbind() in R and bind_rows() in "dplyr".
- Each query should be enclosed in parentheses.

For example, in the **query** below, flights from Bradley (BDL) to $Minneapolis\ St\ Paul$ (MSP) are **combined** with ones from $John\ F\ Kennedy$ (JFK) to $Chicago\ Ohare$ (ORD) to comprise the **result set**:

Section 15.10 Exercises

Exercise 17 This exercise concerns the use of UNION.

Describe **in words** which sets of flights will be **combined** to comprise the **result set** of the following **query**, then check your answer.

15.11 Subqueries

- It's possible to query the result set of another query as if it were a table. The *initial* query is called a *subquery*.
- For example, does Bradley Airport have any flights going to Alaska or Hawaii?

We can retrieve a list of the **258** airports in Alaska and Hawaii by filtering the **airports table** using the time zone (tz) column:

Now we'll use the **result set** (list of Alaska and Hawaii airports) generated by the **query** above as a **subquery** in a WHERE clause to retrieve the flights from *Bradley* in 2013 whose destinations were in Alaska or Hawaii:

The output (not shown) indicates no flights were identified by the **query**, meaning there weren't any to Alaska or Hawaii.

Note (above) that the **subquery** needed to be enclosed in **parentheses**.

Section 15.11 Exercises

Exercise 18 In this section, we used a **subquery** to determine whether *Bradley Airport* had any flights to Alaska or Hawaii (time zone, tz, less than -8) in 2013.

Did *Bradley* have any flights to airports in the *Pacific time zone* (tz less than -7) in 2013? If so, which airports (three-letter codes) in the *Pacific time zone* were the flights' destinations?

Exercise 19 In this section, we used a **subquery** to determine whether *Bradley Airport* had any flights to Alaska or Hawaii (time zone, tz, less than -8) in 2013.

Did John F Kennedy Airport (JFK) have any flights to Alaska or Hawaii (time zone, tz, less than -8) in 2013? If so, which airport(s) (three-letter codes) in Alaska or Hawaii were the flights' destinations?

15.12 Database Querying With "dplyr" and "dbplyr"

- The "dplyr" package allows you to use **tables** in a remote **database** as if they're **in-memory** (on-your-computer) data frames by automatically translating "dplyr" code to **SQL**.
- "dplyr" automatically calls functions from the "dbplyr" package, which then translate "dplyr" commands into SQL queries .
- We'll use the following functions from "dplyr" and "dbplyr".

```
tbl()  # Maps a table in a remote database to an object in R
show_query()  # Show the SQL statement corresponding to an R query.
translate_sql()  # Translates an R expression to SQL.
collect()  # Saves data retrieved by a "dplyr" database query as a
  # data frame.
```

• The first step is again to connect to the remote server using dbConnect() (from the "DBI" package).

Here, we connect to the airlines database on the remote MySQL server (set up by the textbook authors):

• Now we can map to two tables using tbl():

```
flights <- tbl(db_con, "flights")
carriers <- tbl(db_con, "carriers")</pre>
```

Above, the tbl() function maps the flights table in the airlines database to an object in R, in this case also called flights. The same is done for the carriers table.

Now we can use flights and carriers as if they were data frames:

```
head(flights)
head(carriers)
```

(Output not shown.)

• As another example, this command runs a **query** to retrieve the **65** flights from *Bradley Airport* on June 6, 2013:

```
filter(.data = flights, year == 2013 & month == 6 & day == 6 & origin == "BDL")
```

(Output not shown.)

• A tbl is a special kind of object created by "dbplyr" that behaves like a data frame, but isn't, in fact, a data frame. Rather, they belong to the tbl_MySQLConnection class, and more generally, tbl_sql and tbl:

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```
class(flights)
class(carriers)
```

(Output not shown.)

Because flights is a tbl_sql and not a data frame, it resides on the remote server, not on our computer, and the computations are actually being done by MySQL on the server, not here on our computer.

The "dplyr" commands are simply translated into SQL and submitted to the server.

We can see the translation by passing our "dplyr" query through show_query():

```
my.query <- filter(.data = flights, year == 2013 & month == 6 & day == 6 & origin == "BDL")
show_query(my.query)

## <SQL>
## SELECT *
## FROM `flights`
## WHERE (`year` = 2013.0 AND `month` = 6.0 AND `day` = 6.0 AND `origin` = 'BDL')
```

Even though we wrote our query in R, it was translated to SQL by "dbplyr".

"dbplyr" does this automatically any time it encounters an object of class tbl_sql (such as flights in the query above).

• How is the translation done? "dplyr" calls the translate_sql() function (from "dbplyr"). For example:

```
library(dbplyr)
translate_sql(mean(arr_delay, na.rm = TRUE))
## <SQL> AVG(`arr_delay`) OVER ()
```

- Note that translate_sql() cannot translate every R command to SQL. (See the textbook for examples.)
- Note also that running a query using "dplyr" doesn't actually retrieve the data onto our computer.

Instead, it returns an object of class tbl_MySQLConnection, and more generally, tbl_sql and tbl:

This means the **result set** of the "dplyr" query still **resides** on the **remote server**, not on our computer.

If we want the results of the "dplyr" query pulled into a data frame here in R, we can use the collect() function. For example:

```
my.query <- filter(.data = flights, year == 2013 & month == 6 & day == 6 & origin == "BDL")
my.query.df <- collect(my.query)

class(my.query.df)
## [1] "tbl_df" "tbl" "data.frame"</pre>
```

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The collect() function breaks the connection to the MySQL server and returns a data frame (which is also a tbl_df).

Warning: Above, collect() retrieved a query result set that only had 65 rows. Do not attempt to use collect() to retrieve all 196 million rows of the flights table.

```
## [1] TRUE
```

Section 15.12 Exercises

Exercise 20 Make sure the following packages are loaded.

```
library(dplyr)
library(DBI)
library(RMySQL)
```

Now open a connection to the airlines database using dbConnect(), and save the connection as, say, db_con:

Next, map to the flights and carriers tables from the airlines database using tbl():

```
flights <- tbl(db_con, "flights")
carriers <- tbl(db_con, "carriers")</pre>
```

In Exercise 2, SELECT and FROM were used to query the flights table to retrieve *just* the carrier and tailnum variables (columns).

Write a command using "dplyr" that performs this same query. Do not use dbGetQuery().

Report your **R** command(s).

Exercise 21 If you're not still connected to the airlines database and mapped to the flights and carriers tables, reconnect and re-map.

In Excercise 6, the logical operators (AND, OR, and NOT) were used with the **flights table** to retrieve flights meeting the conditions listed below.

Report R command(s) involving "dplyr" and the logical operators & ("and"), | ("or"), and | ("not") with the flights table to map to flights meeting the conditions:

- a) Had an arrival delay of more than hours (arr_delay more than 120 minutes).
- b) Flew to Houston (i.e. had a dest of IAH or HOU).
- c) Were operated by United, American, or Delta (i.e. had a carrier of UA, AA, or DL).
- d) Departed in summer (July, August, or September, i.e. month 7, 8, or 9).
- e) Departed between midnight and 6:00 AM, inclusive (dep_time at 2400 or between 0 and 600, inclusive).
- f) Were operated by United (carrier UA), departed in July (month 7), and had an arrival delay of more than two hours (arr_delay more than 120 minutes).

15.13 Practicing SQL In-Memory (Without Connecting to a Server)

• Up to now, we've only **retrieved** data from a **SQL database** by running **queries**.

Modifying a database on a remote server requires permission from the manager of the database.

• For **practicing modifying** a **SQL database**, it's useful to be able to be able to create *your own* **database** on your **computer's memory**, *not* on a remote server.

This can be done by specifying dbname = ":memory:" in dbConnect().

Note that for this, we need to use **SQLite**. MySQL doesn't work.

This causes **SQLite** to create a temporary **in-memory database**.

Initially, there are **no tables** in the **database**:

```
dbListTables(db_con)
## character(0)
```

• We can use dbWriteTable() (from the "DBI" package) to write an R data frame into a table in the database stored on the computer's memory (or a remote server).

Other functions (from "DBI") can be used to modify or remove tables.

```
dbWriteTable()  # Copy a data frame to a table in a database.

dbListTables()  # List the tables in a database.

dbListFields()  # List field (column) names of a table in a database.

dbAppendTable()  # Insert rows into a table in a database.

dbRemoveTable()  # Remove a table from a database.
```

• dbWriteTable() takes arguments conn (a connection to computer's memory or a server), name, the name to be assigned to the newly created table in the database, and value, the data frame used to create the table.

Data Set: who and population

The who and population data sets (in "tidyr") contain a subset of data from the World Health Organization Global Tuberculosis (TB) Report, and accompanying global populations.

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```
who has 7,240 rows and the 60 variables:
                               Country name.
 country
                               2 & 3 letter ISO country codes.
 iso2, iso3
                               Year.
 year
                               Counts of new TB cases recorded by group. Column names
 new_sp_m014 - new_rel_f65
                               encode three variables that describe the group (see details
                               in the help page, ?who).
population has 4,060 rows and three variables:
               Country name.
 country
               Year.
 year
 population Population.
The help page (?who) has more information about these data sets.
```

• For example, below, we create an *in-memory* database containing two tables, the who and population *data* frames (from the "tidyr" package):

Now we can confirm that the database contains the two tables:

```
# Now the database contains the population and who tables:
dbListTables(db_con)
## [1] "popTable" "whoTable"
```

Note that the **two tables** aren't stored in the R Workspace:

```
whoTable
## Error in eval(expr, envir, enclos): object 'whoTable' not found
popTable
## Error in eval(expr, envir, enclos): object 'popTable' not found
```

Rather, they're in a separate $\mathbf{database}$ elsewhere in the computer's memory.

Once the *in-memory* database contains one or more tables, we can run queries using dbGetQuery().

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For example,

(Output not shown.)

Recall that dbGetQuery() returns the result set as a data frame:

```
is.data.frame(my.query)
## [1] TRUE
```

• The function dbAppendTable() can be used to insert rows into a table in a database.

dbAppendTable() takes arguments conn, an open connection to a database, name, the name of the table in the database, and value, a *data frame* containing the rows to be appended to the table.

For example, below, we create a data frame containing a single row and append it to the popTable table:

Section 15.13 Exercises

Exercise 22 Create an *in-memory* database, then add two tables, the who and population data frames (from the "dplyr" package):

a) Now use dbGetQuery() with SELECT and FROM to query the who table to retrieve just the country, year, and new_sp_m014 variables (columns).

Report your R command(s) (or just your SQL statement).

b) Now modify your query from part a using a WHERE clause so that it *only* retrieves rows corresponding to the U.S. ("United States of America") more recently than 2009 (year >= 2010).

Report your **R** command(s) (or just your **SQL** statement).

Exercise 23 If you're not still connected to the *in-memory* database containing the two tables, who and population, reconnect and reinsert the tables.

Use GROUP BY with AVG() and ORDER BY to determine the country for which the average number of new TB cases for the (new_sp_m014) group was lowest in the year 2013.

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Note: It's okay if you end up with NAs.

Report your R command(s) (or just your SQL statement).