Relational Databases and SQL

A database is a collection of data with information about how the data are organized (meta-data). A database server is like a web server, but responds to requests for data rather than web pages.

We'll talk about relational database management systems (RDBMS) and how to communicate with them using the structured query language (SQL).

Why use a database?

- Coordinate synchronized access to data
- Change continually; give immediate access to live data
- Centralize data for backups
- Control access to the data

A RDBMS had three main parts

- Data definition
- Data access
- Privilege management

We will concentrate on data access, assuming the database is already available and we have the needed privileges.

Topics:

- use SQL to extract information from an RDBMS
- relate these SQL statements back to similar tasks in R
- use SQL from within R

There are tradeoffs in terms of what we choose to do using SQL and what we do in R.

- A database is made up of one or more two dimensional tables, usually stored as files on the server.
- A table is a rectangular arrangement of values, where rows represent cases, and columns represent variables (just like data frames in R).
- Another term for a table is a relation. Rows are also referred to as tuples and the columns as attributes.

- SQL allows us to interactively query the database to reduce the data by subsetting, grouping, or aggregation.
- Each database program tends to have its own version of SQL, but they all support the same basic SQL statements. (We say statements rather than commands because SQL is referred to as a declarative rather than an imperative language.)
- The SQL statement for retrieving data is the SELECT statement. This operates on one or more tables. The result will always be another table.

We have a table called chips, with data about the CPU development of PCs over time

The simplest possible query gives back everything:

SELECT * FROM chips;

processor	l date	I			microns	I	clockspeed	I	width	I	mips
	+	+-		+-		+-		+-	+		
8080	l 1974	- 1	6000	-	6	-	2		8	1	0.64
8088	l 1979	- 1	29000	-	3	-	5	-	16	1	0.33
80286	1 1982	- 1	134000	-	1.5	1	6	-	16	1	1
80386	l 1985	- 1	275000	-	1.5	1	16	1	32	I	5
80486	l 1989	- 1	1200000	-	1	1	25	-	32	I	20
Pentium	l 1993	-	3100000	-	0.8	1	60	-	32	1	100
PentiumII	l 1997	- 1	7500000	-	0.35	1	233	-	32	1	300
PentiumIII	l 1999	- 1	9500000	-	0.25	1	450	-	32	1	510
Pentium4	1 2000	I	42000000	1	0.18	1	1500		32	I	1700

• By convention, we display SQL statements in upper case. Statements must end with a semicolon.

Attributes / Variables

 Recall that in R, we can select particular variables (columns) by name.

```
chips[ , c('mips', 'microns')]
```

- The order of the variable names determines the order in which they are returned in the resulting data frame.
- The corresponding SQL query is

```
SELECT mips, microns FROM chips;
```

SQL Syntax

- Similar to a sentence in English, except that there's less flexibility in the order of the words.
- Sentence ends with a;
- Use blanks and "," and "=" and "()" as delimiters
- We will only look at SELECT statements, i.e., statements that begin with the term SELECT

Examples of SELECT statements

```
SELECT * FROM chips;
SELECT mips, microns FROM chips;

SELECT * FROM chips
WHERE processor = 'Pentium' OR
    processor = 'PentiumII';
```

Selecting Tuple/Row

In R we can select rows that match a condition:

```
chips[chips$processor == 'Pentium' |
    chips$processor == 'PentiumII', ]
```

The corresponding SQL statement is

```
SELECT * FROM chips
WHERE processor = 'Pentium' OR
processor = 'PentiumII';
```

Note: Whitespace can be used freely in SQL statements. We often separate and indent lines for clarity. The statement isn't evaluated until the semicolon is entered.

- In R, it usually does not matter whether you use single or double quotes to surround character strings. In SQL, the standard is to use single quotes so we will do this throughout for both R and SQL.
- The WHERE clause can also be used with Boolean operators. The keyword NOT negates a condition, and parentheses can be used to clarify order of evaluation.

```
SELECT * FROM chips WHERE date > 1990;

SELECT * FROM chips WHERE NOT width = 8;

SELECT * FROM chips WHERE NOT (width = 8 OR width = 16);
```

In both R and SQL, we can do both types of subsetting at once, i.e., subset columns and rows

In R:

```
chips[chips$processor == 'Pentium' |
      chips$processor == 'PentiumII',
      c('mips', 'microns')]
```

In SQL:

```
SELECT mips, microns FROM chips
WHERE processor = 'Pentium' OR
    processor = 'PentiumII';
```

General Syntax

```
SELECT attribute(s) FROM table(s)
[WHERE constraints];——[optional]
```

How would we pull the years of all 32-bit processors that execute fewer than 250 million instructions per second (mips),

```
In R,
chips[chips$mips < 250 & chips$width == 32, "date"]
In SQL
SELECT date FROM chips
  WHERE mips < 250 AND width = 32;</pre>
```

SQL offers limited features for summarizing data -- some aggregate functions that operate over the rows of a table and some mathematical functions that operate on individual values in a row.

The aggregate functions are

- COUNT number of rows
- SUM total of all values for an attribute
- AVG average value for an attribute
- MIN minimum value for an attribute
- MAX maximum value for an attribute

SELECT attribute(s) FROM table(s) [WHERE constraints];

can also be aggregate functions of attributes

Additional clauses: GROUP BY

 The GROUP BY clause makes the aggregate functions in SQL more useful. It enables the aggregates to be applied to *subsets* of the rows in a table.

More than one attribute can be included in the GROUP BY clause.

Additional clauses: **HAVING**

 The WHERE clause can't contain an aggregate function, but the HAVING clause can be used to refer to the groups to be selected.

First the chips table is separated into sets of rows by width.
 For each set, MAX(mips) is calculated, and the set is
 discarded if MAX(mips) < 1. Finally, width and
 MAX(mips) are returned for each set.

A few other predicates and clauses

- DISTINCT forces values of an attribute in the results table to have unique values
- NOT negates conditions in WHERE or HAVING clause
- LIMIT limits the number of rows returned

```
SELECT * FROM chips LIMIT 3;
SELECT DISTINCT width FROM chips;
```

Order of Execution

The order of execution of the clauses in a **SELECT** statement is as follows:

- 1. FROM: The working table is constructed.
- 2. WHERE: The WHERE clause is applied to each row of the table, and only the rows that test TRUE are retained.
- 3. GROUP BY: The results are broken into groups of rows all with the same value of the GROUP BY clause.
- 4. HAVING: The HAVING clause is applied to each group and only those that test TRUE are retained.
- 5. **SELECT:** The attributes not in the list are dropped, aggregates are calculated, and options **DISTINCT**, **ORDER BY** and **LIMIT** are applied.

Exercises:

- 1) How many rows are in the chips table? SELECT COUNT(*) FROM chips;
- 2) How many chips have attribute width equal to 32?

```
SELECT COUNT(*) FROM chips WHERE width = 32;
```

3) What is the average clock speed for the chips in question 2?

```
SELECT AVG(clockspeed) FROM chips WHERE width = 32;
```

Now answer the same questions assuming chips is a data frame in R.

```
nrow(chips);
sum(chips$width == 32);
mean(chips$clockspeed[chips$width == 32)
```

Exercises:

4) What is the average clock speed for each unique value of width?

```
SELECT AVG(clockspeed) FROM chips GROUP BY width;
```

5) How many chips are in each width group (except width 8)?

SELECT COUNT(*) FROM chips

GROUP BY width HAVING width > 8;

Now answer the same questions assuming chips is a data frame in R.

Exercises:

- 6) For those chips models developed before 2000, what is the average clock speed per width?
- A) SELECT AVG(clockspeed) FROM chips GROUP BY width HAVING date < 2000;
- B) SELECT AVG(clockspeed) FROM chips WHERE date < 2000 GROUP BY width;

Now answer the same questions assuming chips is a data frame in R.

```
with(chips[chips$date < 2000],
     tapply(clockspeed, width, mean)</pre>
```

Using SQL with R

The DBI, RSQLite libraries in R allow you to connect to an SQL database, submit a query, and receive the results as a data frame.

```
library(RSQLite)
```

Set up an interface to SQLite

```
drv = dbDriver("SQLite")
```

Connect to DBMS (no security SQLite)

Using SQL with R

The RMySQL library also allows you to connect to a mySQL database.

```
library(RMySQL)
```

Set up an interface to MySQL

```
drv = dbDriver("MySQL")
```

Connect to DBMS with security

```
con = dbConnect(drv, dbname = "lahman",
  user = "s133", password = "s133",
  host = "radagast.berkeley.edu")
```

Sending an SQL query to the database with R A data frame is returned

We can of course import the whole table, then extracting what we want using R commands.

But, it may be more efficient to use SELECT to extract and import only the results with which we want to work

If the results are large, we don't have to pull them all over into R at once.

Submit an SQL statement; keep results in database Use dbSendQuery to do this rather than dbGetQuery

Retrieve the first 5 rows in the results

```
fetch(rs, n = 5)
```

Retrieve the next 3 rows

```
fetch(rs, n = 3)
```

Retrieve the remaining rows

```
fetch(rs, n = -1)
```

Close the query dbClearResult(rs)

Can close without retrieving all rows

Close the connection and turn off the driver

Disconnect from the database.

```
dbDisconnect(con)
[1] TRUE
```

Unload the Driver

```
dbUnloadDriver(drv)
```

[1] TRUE