

# Lecture 23: Databases

36-350

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## Agenda

- What databases are, and why
- SQL
- Interfacing R and SQL

Reading: Spector, chapter 3; handout on class website

## Databases

- A **record** is a collection of **fields**
- A **table** is a collection of records which all have the same fields (with different values)
- A **database** is a collection of tables

## Databases vs. Dataframes

- R's dataframes are actually tables

R jargon	Database jargon
column	field
row	record
dataframe	table
types of the columns	table <b>schema</b>
bunch of related dataframes	database

## So, Why Do We Need Database Software?

- *Size*
  - R keeps its dataframes in memory
  - Industrial databases can be much bigger
  - Work with selected subsets
- *Speed*
  - Clever people have worked very hard on getting just what you want fast
- *Concurrency*
  - Many users accessing the same database simultaneously)
  - Lots of potential for trouble (two users want to change the same record at once)

## The Client-Server Model

- Databases live on a **server**, which manages them
- Users interact with the server through a **client** program
- Lets multiple users access the same database simultaneously

## SQL

- **SQL (structured query language)** is the standard for database software
- Mostly about **queries**, which are like doing a selection in R

```
debt[debt$Country=="France",c("growth","ratio")]  
with(debt,debt[Country=="France",c("growth","ratio")])  
subset(x=debt,subset=(Country=="France"),select=c("growth","ratio"))
```

- Let's look at how SQL does stuff like this

## SELECT

```
SELECT columns or computations  
FROM table  
WHERE condition  
GROUP BY columns  
HAVING condition  
ORDER BY column [ASC|DESC]  
LIMIT offset,count;
```

- **SELECT** is the first word of a query, then modifiers say which fields/columns to use, and what conditions records/rows must meet, from which tables
- The final semi-colon is obligatory

## SELECT

```
SELECT PlayerID,yearID,AB,H FROM Batting;
```

Four columns from table `Batting`

```
SELECT * FROM Salaries;
```

All columns from table `Salaries`

```
SELECT * FROM Salaries ORDER BY Salary;
```

As above, but by ascending value of `Salary`

```
SELECT * FROM Salaries ORDER BY Salary DESC;
```

Descending order

```
SELECT * FROM Salaries ORDER BY Salary DESC LIMIT 10;
```

top 10 salaries

## SELECT

Picking out rows meeting a condition

```
SELECT PlayerID,yearID,AB,H FROM Batting WHERE AB > 100 AND H > 0;
```

vs.

```
Batting[Batting$AB>100 & Batting$H > 0, c("PlayerID","yearID","AB","H")]
```

## Calculated Columns

- SQL knows about some simple summary statistics:

```
SELECT MIN(AB), AVG(AB), MAX(AB) FROM Batting;
```

- It can do arithmetic

```
SELECT AB,H,H/CAST(AB AS REAL) FROM Batting;
```

Because `AB` and `H` are integers, and it won't give you a fractional part by default

- Calculated columns can get names:

```
SELECT PlayerID,yearID,H/CAST(AB AS REAL) AS BattingAvg FROM Batting  
ORDER BY BattingAvg DESC LIMIT 10;
```

## Aggregating

We can do calculations on value-grouped subsets, like in `aggregate` or `dply`

```
SELECT playerID, SUM(salary) FROM Salaries GROUP BY playerID
```

## Selecting Again

- First cut of records is with `WHERE`
- Aggregation of records with `GROUP BY`
- Post-aggregation selection with `HAVING`

```
SELECT playerID, SUM(salary) AS totalSalary FROM Salaries GROUP BY playerID  
HAVING totalSalary > 200000000
```

## JOIN

- So far FROM has just been one table
- Sometimes we need to combine information from many tables

patient_last	patient_first	physician_id	complaint
Morgan	Dexter	37010	insomnia
Soprano	Anthony	79676	malaise
Swearengen	Albert	NA	healthy as a goddam horse
Garrett	Alma	90091	nerves
Holmes	Sherlock	43675	nicotine-patch addiction

physician_last	physician_first	physicianID	plan
Meridian	Emmett	37010	UPMC
Melfi	Jennifer	79676	BCBS
Cochran	Amos	90091	UPMC
Watson	John	43675	VA

## JOIN

- Suppose we want to know which doctors are treating patients for insomnia
- Complaints are in one table
- Physicians are in the other
- In R, we'd use `merge` to link the tables up by `physicianID`
- Here, `physician_id` or `physicianID` is acting as the **key** or **unique identifier**

## JOIN

- SQL doesn't have `merge`, it has JOIN as a modifier to FROM

```
SELECT physician_first, physician_last FROM patients INNER JOIN physicians ON patients.physician_id == p
```

Creates a (virtual) table linking records where `physician_id` in one table matches `physicianID` in the other

- If the names were the same in the two tables, we could write (e.g.)

```
SELECT nameLast,nameFirst,yearID,AB,H FROM Master INNER JOIN Batting
  USING(playerID);
```

INNER JOIN ... USING links records with the same value of `playerID`

- There are some syntax variants here; see the handout

## JOIN

- LEFT OUTER JOIN includes records from the first table which don't match any record in the 2nd
  - The “extra” records get NA in the 2nd table's fields
- RIGHT OUTER JOIN is just what you'd think
  - so is FULL OUTER JOIN

## Updated Translation Table

R jargon	Database jargon
column	field
row	record
dataframe	table
types of the columns	table schema
bunch of dataframes	database
selections, subset	SELECT ... FROM ... WHERE ... HAVING
aggregate, d*ply	GROUP BY
merge	JOIN
order	ORDER BY

## Connecting R to SQL

- SQL is a language; database management systems (DBMS) actually implement it and do the work
  - MySQL, SQLite, etc., etc.
- They all have somewhat different conventions
- The R package DBI is a unified interface to them
- Need a separate “driver” for each DBMS

## Connecting R to SQL

```
install.packages("DBI", dependencies = TRUE) # Install DBI
install.packages("RSQLite", dependencies = TRUE) # Install driver for SQLite
library(RSQLite)
drv <- dbDriver('SQLite')
con <- dbConnect(drv, dbname="baseball.db")
```

con is now a persistent connection to the database `baseball.db`

## Connecting R to SQL

```
dbListTables(con)          # Get tables in the database (returns vector)
dbListFields(con, name)    # List fields in a table
dbReadTable(con, name)     # Import a table as a data frame
```

## Connecting R to SQL

```
dbGetQuery(conn, statement)
df <- dbGetQuery(con, paste(
  "SELECT nameLast,nameFirst,yearID,salary",
  "FROM Master NATURAL JOIN Salaries"))
```

## Connecting R to SQL

Usual workflow: - Load the driver, connect to the right database - R sends an SQL query to the DBMS - SQL **executes the query**, sending back a manageably small dataframe - R does the actual statistics - Close the connection when you're done

## Going the Other Way

- The `sqldf` package lets you use SQL commands on dataframes
- Mostly useful if you already know SQL better than R...

## Summary

- A database is basically a way of dealing efficiently with lots of potentially huge dataframes
- SQL is the standard language for telling databases what to do, especially what queries to run
- Everything in an SQL query is something we've practiced already in R
  - subsetting/selection, aggregation, merging, ordering
- Connect R to the database, send it an SQL query, analyse the returned dataframe