OPEN DATA SCIENCE CONFERENCE

Burlingame I November 2nd 2017

@ODSC

@WinVectorLLC

@RStudio

Nov 02
2:00 PM
Room T2

Modeling big data with R, sparklyr, and Apache Spark

BIG DATARINTERMEDIATE

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Part 2: Exercises: manipulating data locally.

RStudio Shiny Server Pro Accounts

- Distribute credentials and get everyone started with RStudio Server Pro.
- Server time generously donated by RStudio.

RStudio Server Pro

RStudio for the Enterprise

RStudio is the premier IDE for R. RStudio Server lets you access RStudio from anywhere using a web browser. RStudio Server Pro delivers the team productivity, security, centralized management, metrics, and commercial support that professional data science teams need to develop at scale.

From: https://www.rstudio.com/products/rstudio-server-pro/

What are we going to do?

- Become `dplyr` masters.

Work through markdowns in a bit

- Exercises/01-Universal-tools.Rmd
- Exercises/02-Big-Data.Rmd
- slides/lazyeval.Rmd

If you have trouble: ask your neighbors, flag me and the TAs, or peek in Exercises/solutions. These exercises are a memory aid, not a test.

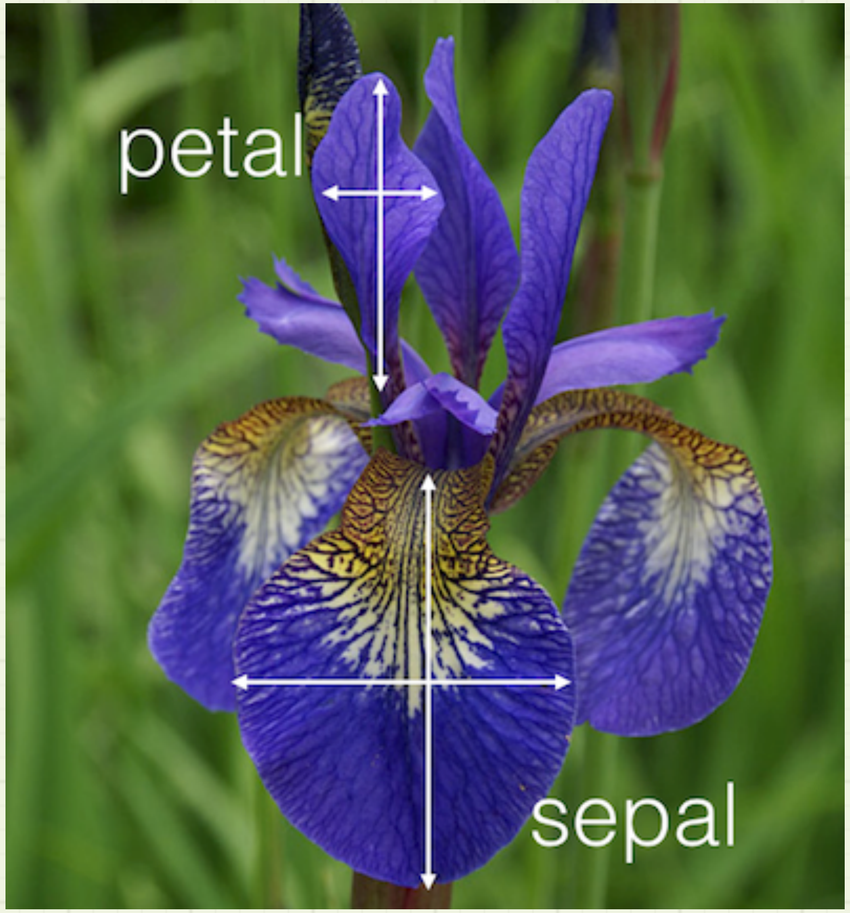
Exercises/01-Universal-tools.Rmd Exercise 1

- Let’s start with Exercises/01-Universal-tools.Rmd

Exercises/01-Universal-tools.Rmd
Exercise 2

iris

On average, which species has the greatest difference in *petal width* and *petal length*?



Exercises/01-Universal-tools.Rmd
Exercise 2

1. **Group** iris by Species
2. For each group, return:
   - avg_width = mean Petal.Width
   - avg_length = mean Petal.Length
3. For each row, calculate diff = avg_length - avg_width
4. Return the **row** whose diff == the max diff
5. Return the **columns** above

Hint 1: save each results in a different variable such as iris1, iris2, … and examine your results before moving to the next step.

Hint 2: Consider save the max diff *value* into a temp table before trying to find which row has that value.

Target start: 2:04pm

Exercises/01-Universal-tools.Rmd
Exercise 2 solution

```
library(dplyr)


iris1 <- group_by(iris, Species)
iris2 <- summarise(iris1,
  avg_width = mean(Petal.Width),
  avg_length = mean(Petal.Length))
iris3 <- mutate(iris2, diff = avg_length - avg_width)
iris4 <- filter(iris3, diff == max(diff))
select(iris4, Species, avg_width, avg_length)
```

Mini-topic
pipes and pipelines

%>%

Ceci n'est pas une pipe.

The pipe operator %>%

These all do the same thing

Try it!

```
filter(iris, Sepal.Length == max(Sepal.Length))
```

```
iris %>% filter(., Sepal.Length == max(Sepal.Length))
```

```
iris %>% filter(Sepal.Length == max(Sepal.Length))
```


```
iris          filter(_______, Sepal.Length == max(Sepal.Length))
```

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Exercises/01-Universal-tools.Rmd
Exercise 3

Use %>% to turn your code from that last exercise into a single long pipe.

Target start: 2:13pm

Exercises/01-Universal-tools.Rmd
Exercise 3 solution:
Take Exercise 2 solution

```
library(dplyr)


iris1 <- group_by(iris, Species)
iris2 <- summarise(iris1,
  avg_width = mean(Petal.Width),
  avg_length = mean(Petal.Length))
iris3 <- mutate(iris2, diff = avg_length - avg_width)
iris4 <- filter(iris3, diff == max(diff))
select(iris4, Species, avg_width, avg_length)
```

Exercises/01-Universal-tools.Rmd
Exercise 3 solution:
and *mechanically* translate it into a pipeline

```
iris %>%
  group_by(Species) %>%
  summarise(
    avg_width = mean(Petal.Width),
    avg_length = mean(Petal.Length)) %>%
  mutate(diff = avg_length - avg_width) %>%
  filter(diff == max(diff)) %>%
  select(Species, avg_width, avg_length)
```

pipeline debugging hint

- Break the pipeline early and look at intermediate results.
- Use “->” (“write arrow”) to save result to “.” (dot).
    - Since right arrow is forbidden by most style guides you can search for it to make sure you have not left debugging code in!

Debugging example

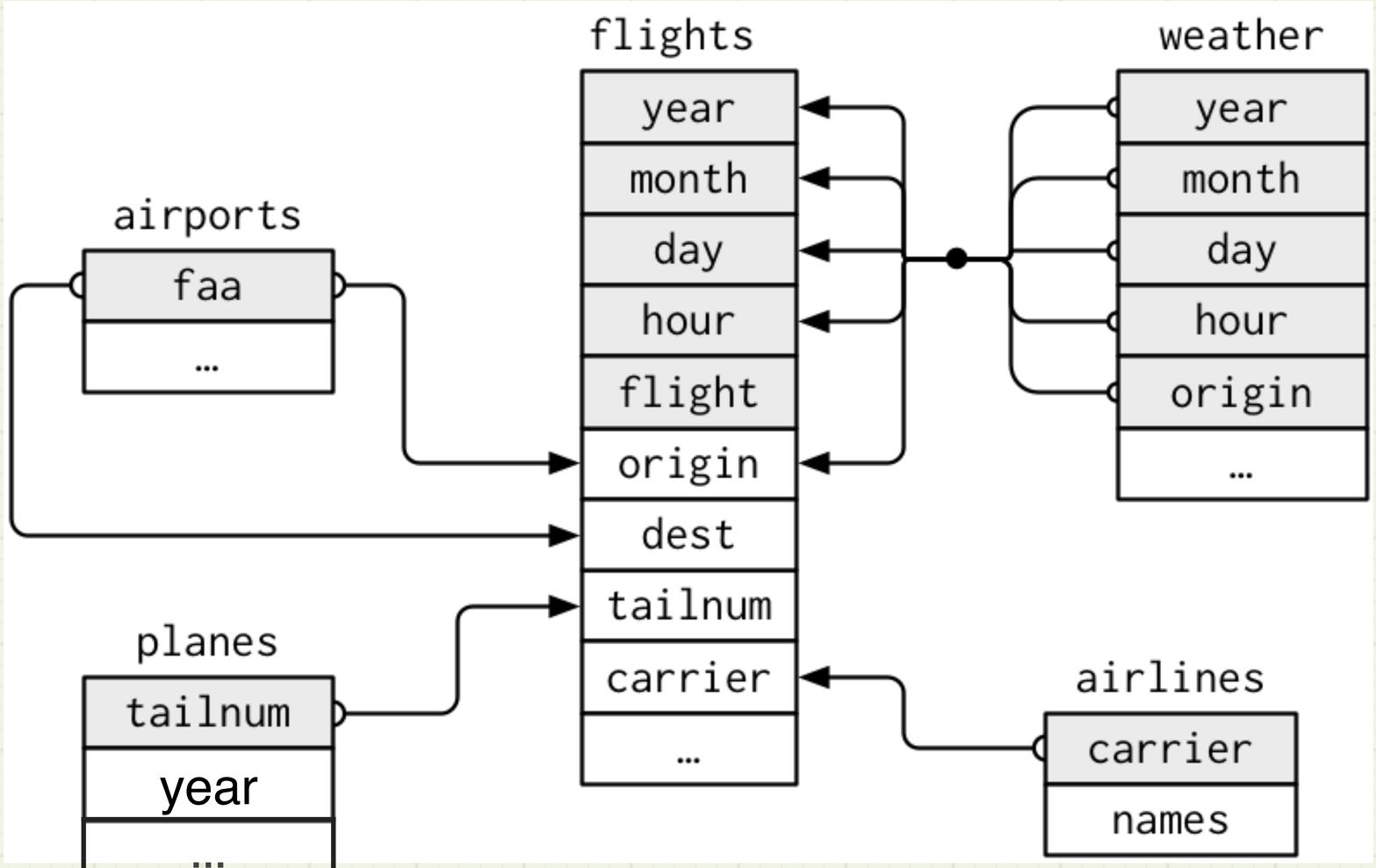
```
iris %>%
  group_by(Species) %>%
  summarise(
    avg_width = mean(Petal.Width),
    avg_length = mean(Petal.Length)) ->.

print(.)

  . %>% mutate(diff = avg_length - avg_width) %>%
  filter(diff == max(diff)) %>%
  select(Species, avg_width, avg_length)
```

Exercises/01-Universal-tools.Rmd
Exercise 5

nycflights13

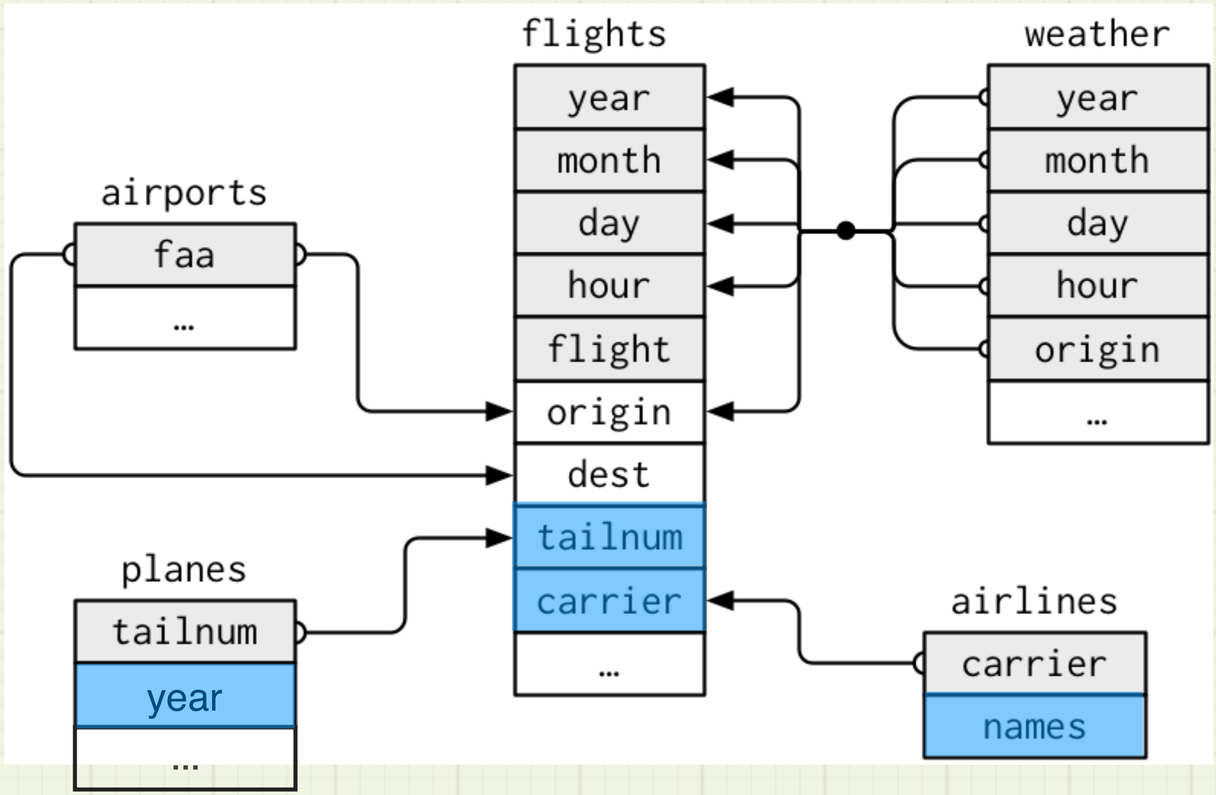
Data on every flight that departed La Guardia, JFK, or Newark airports in 2013

nycflights13



nycflights13

On average, which airline has the newest planes (assigned to the NYC area)?



Exercises/01-Universal-tools.Rmd
Exercise 5

Determine which airline has the newest planes. Please start with the code in the notebook.

```
flights %>%
  distinct(carrier, tailnum) %>%
  _____________________________ %>%
  _____________________________ %>%
  _____________________________ …
```

Target start: 2:19 pm

Exercises/01-Universal-tools.Rmd
Exercise 5 answer

name	avg	n	nas
Hawaiian Airlines Inc.	2011.77	14	1
Virgin America	2008.71	53	1
Frontier Airlines Inc.	2008.00	26	3
Alaska Airlines Inc.	2007.84	84	1
JetBlue Airways	2006.50	193	6
SkyWest Airlines Inc.	2005.86	28	0
Endeavor Air Inc.	2004.71	204	2
Mesa Airlines Inc.	2003.56	58	1
ExpressJet Airlines Inc.	2002.44	316	8
AirTran Airways Corporation	2002.21	129	17

Exercises/01-Universal-tools.Rmd
Exercise 5 solution

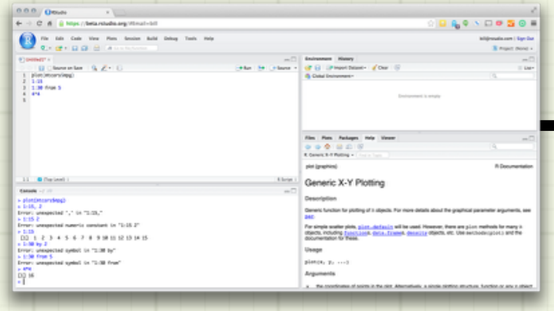
```
flights %>%
  # selects distinct combinations of carrier and tailnum
  distinct(carrier, tailnum) %>%
  # join to planes to get year manufactured
  # (which column should you join on?)
  left_join(planes, by = "tailnum") %>%
  # group by carrier (e.g. the airline)
  group_by(carrier) %>%
  # calculate by carrier:
  #   1. avg - the mean year (with na.rm = TRUE)
  #   2. n - the total number of planes
  #   3. nas - the number of planes with unknown year (year == NA)
  summarise(avg = mean(year, na.rm = TRUE),
            n = n(), nas = sum(is.na(year))) %>%
  # join to airlines to get full airline name
  # (which column should you join on?)
  left_join(airlines, by = "carrier") %>%
  # select just the name, avg, n, and nas variables in that order
  select(name, avg, n, nas) %>%
  # order the results by avg with the newest planes at the top
  arrange(desc(avg))
```

Mini topic databases

User Browser

Server

Database



RStudio Server Pro

Redshift

dplyr

Airlines

Win-Vector LLC

dplyr driver functions

Package	DBMS
src_sqlite()	SQLite
src_mysql()	MySQL, MariaDB
src_postgres()	PostgreSQL
library(bigrquery) src_bigquery()	Google BigQuery

https://cran.r-project.org/web/packages/dplyr/vignettes/databases.html

`dplyr` adapts to databases

- Many common base-R task must be translated `dplyr` to work on databases.

- Most `dplyr` verbs work with many data sources.

Base `R`	`dplyr`
`x$col`	`select(x, col)`

dplyr database workflow

1. **Create a connection**

   ```
   con <- dplyr::src_sqlite(":memory:", create = TRUE)
   ```

2. **Create a reference**

   ```
   tab <- copy_to(con, data, 'tablename')
   ```

3. **Manipulate the reference**

   ```
   query <- tab %>% filter(x > 1) %>% select(x, y, z)
   ```

4. **Collect the results**

   ```
   results <- collect(query)
   ```

5. **Close the connection**

   ```
   rm(con); gc()
   ```

1. **Create a connection**

```
con <- dplyr::src_sqlite(“:memory:", create = TRUE)
```

Save to use

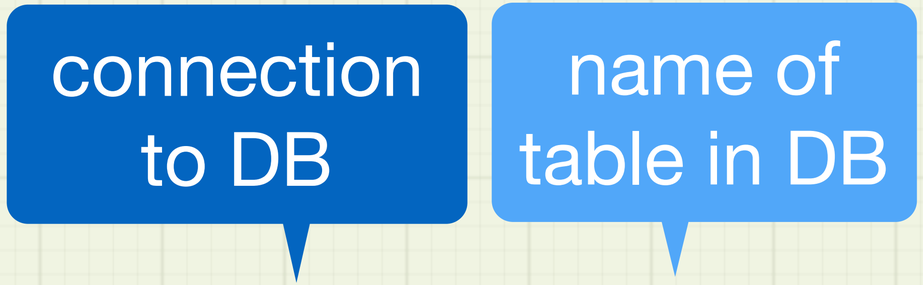
src_ driver function

driver specific args

Lists tables in DB

```
src_tbls(con)
## "iris"   "iris2"   "iris3"
```

2. **Create a table reference**




```
tab <- tbl(con, "table_name")
```

Use `tbl()` to create objects that refer to tables in the database

3. **Manipulate the reference**

Treat the reference as if it were a table in R. dplyr will translate your code to SQL and execute it in the DBMS.*

```
flights <- tbl(air, "flights")
flights %>%
  distinct(uniquecarrier, tailnum) %>%
  …
```

4. **Collect the results**

Use `collect()` to import the entire set of results into R.

```
q6 <- flights %>%
    filter(year > 2007, depdelay > 15) %>%
    filter(depdelay == 240) %>%
    collect()
```

5. **Close the connection**

```
rm(air)
gc()
```

dplyr automatically closes connections when you remove the connection object *and then run the garbage collector,* `gc()`*.*

Exercises/02-Big-Data.Rmd

Determine which airline has the newest planes using `dplyr` to control data in our practice database.

```
flights %>%
  distinct(carrier, tailnum) %>%
  _____________________________ %>%
  _____________________________ %>%
  _____________________________ …
```

Hint: **CHEAT!!!**

Target start: 2:32pm

Mini-topic: lazy evaluation

"Just in Time" delivery of results.

slides/lazyeval.Rmd

Lazy Execution 1

```
q1 <- filter(flights, year < 2007)
q2 <- filter(q1, depdelay > 15)
q3 <- filter(q2, depdelay < 240)
q4 <- select(q3, arrdelay, depdelay, year)
q4
```

Lazy Execution 1

```
q1 <- filter(flights, year < 2007)
q2 <- filter(q1, depdelay > 15)
q3 <- filter(q2, depdelay < 240)
q4 <- select(q3, arrdelay, depdelay, year)
q4
```

dplyr will not retrieve data until last possible moment. It combines all necessary work into a single optimized query.

```
show_query(q4)
##  <SQL>
##  SELECT "arrdelay" AS "arrdelay",
##    "depdelay" AS "depdelay",
##    "year" AS "year"
##  FROM "flights"
##  WHERE "year" > 2007.0
##    AND "depdelay" > 15.0
##    AND "depdelay" < 240.0
```

collapse()

Forces execution in DBMS

```
q5 <- flights %>%
    mutate(adjdelay = depdelay - 15) %>%
    collapse() %>%
    filter(adjdelay > 0)
```

collapse() turns the preceding queries into a table expression

remaining queries are run against the table described in the collapsed expression

Next:
`Spark` and `sparklyr`