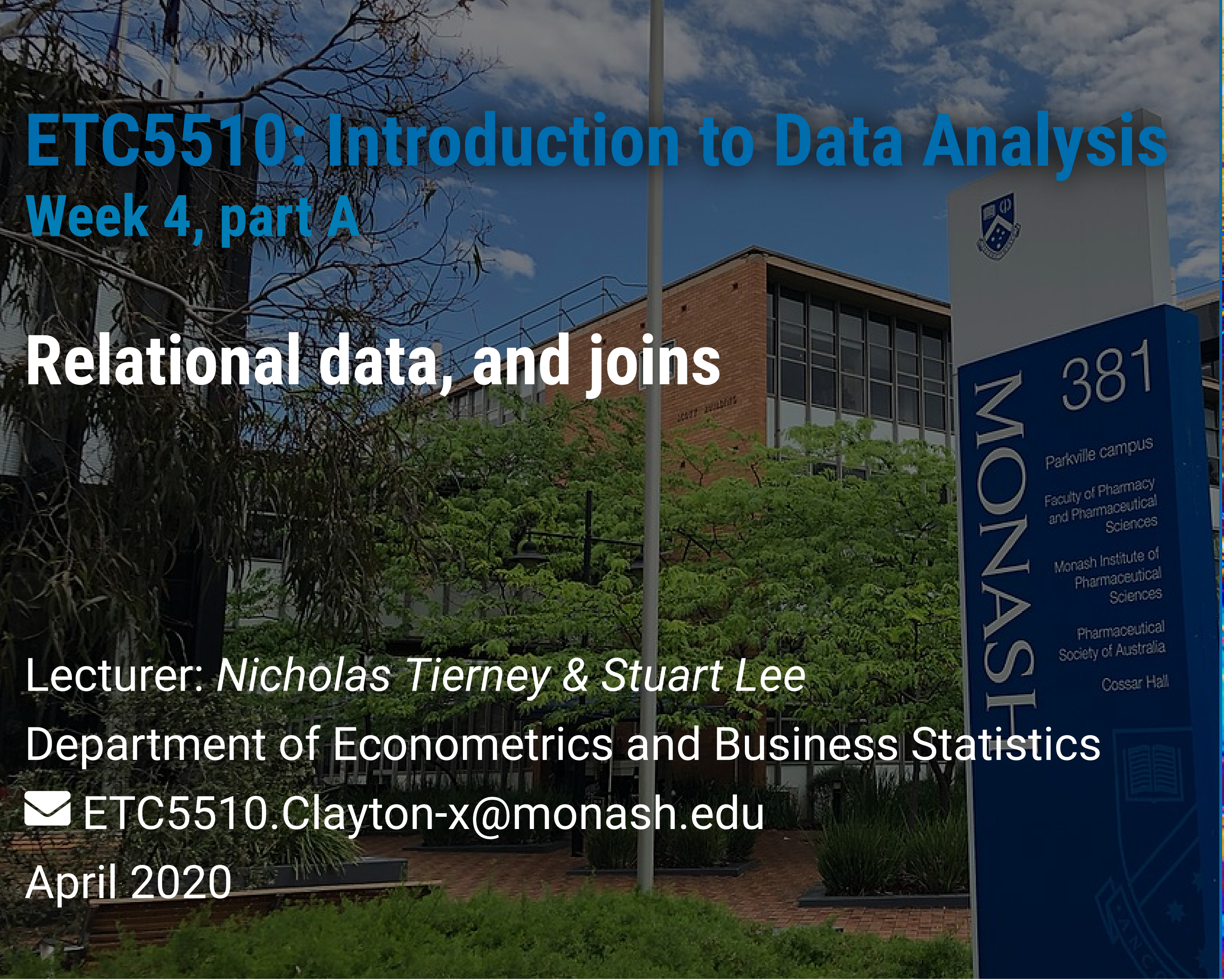
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MONASH
Parkville campus
Faculty of Pharmacy and Pharmaceutical Sciences
Monash Institute of Pharmaceutical Sciences
Pharmaceutical Society of Australia
Cossar Hall

ETC5510: Introduction to Data Analysis

Week 4, part A

Relational data, and joins

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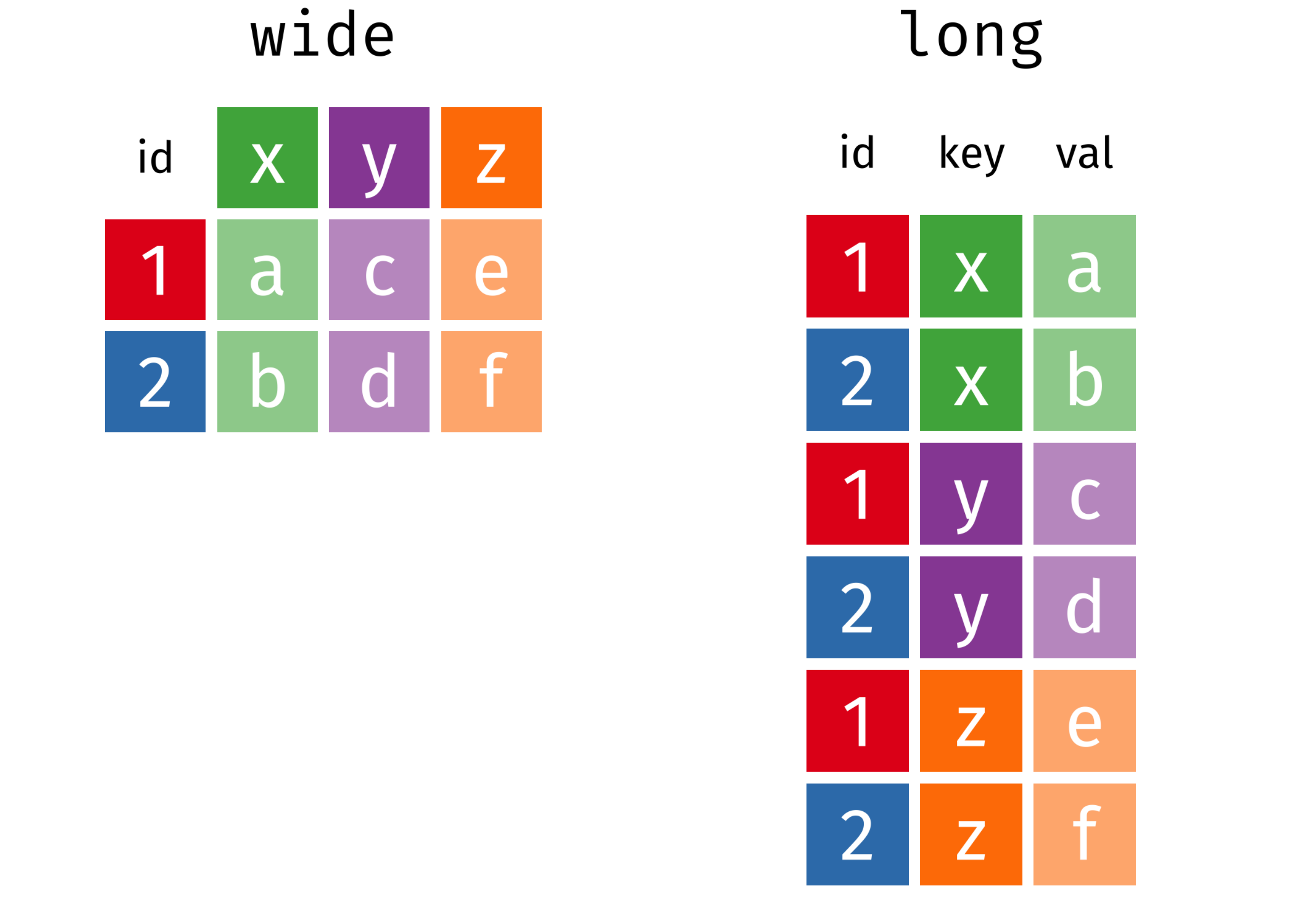
Recap

- consultation hours
- assignment 1
- tidy data

Recap: Tidy data



wide

id	x	y	z
1	a	c	e
2	b	d	f

long

id	key	val
1	x	a
2	x	b
1	y	c
2	y	d
1	z	e
2	z	f

Overview

- What is relational data?
- Keys
- Different sorts of joins
- Using joins to follow an aircraft flight path

Relational data

- Data analysis **rarely involves** only a single table of data.
- To answer questions you generally need to combine many tables of data
- Multiple tables of data are called *relational data*
- It is the **relations**, not just the individual datasets, that are important.

nycflights13

- Data set of flights that departed NYC in 2013 from https://www.transtats.bts.gov - a public database of all USA commercial airline flights. It has five tables:

  1. flights
  2. airlines
  3. airports
  4. planes
  5. weather

flights

```
library(nycflights13)
flights
## # A tibble: 336,776 x 19
##     year month   day dep_time sched_dep_time dep_delay arr_time sched_arr_time
##    <int> <int> <int>    <int>          <int>     <dbl>    <int>          <int>
##  1  2013     1     1      517            515         2      830            819
##  2  2013     1     1      533            529         4      850            830
##  3  2013     1     1      542            540         2      923            850
##  4  2013     1     1      544            545        -1     1004           1022
##  5  2013     1     1      554            600        -6      812            837
##  6  2013     1     1      554            558        -4      740            728
##  7  2013     1     1      555            600        -5      913            854
##  8  2013     1     1      557            600        -3      709            723
##  9  2013     1     1      557            600        -3      838            846
## 10  2013     1     1      558            600        -2      753            745
## # … with 336,766 more rows, and 11 more variables: arr_delay <dbl>,
## #   carrier <chr>, flight <int>, tailnum <chr>, origin <chr>, dest <chr>,
## #   air_time <dbl>, distance <dbl>, hour <dbl>, minute <dbl>, time_hour <dttm>
```

airlines

```
airlines
## # A tibble: 16 x 2
##    carrier name
##    <chr>   <chr>
##  1 9E      Endeavor Air Inc.
##  2 AA      American Airlines Inc.
##  3 AS      Alaska Airlines Inc.
##  4 B6      JetBlue Airways
##  5 DL      Delta Air Lines Inc.
##  6 EV      ExpressJet Airlines Inc.
##  7 F9      Frontier Airlines Inc.
##  8 FL      AirTran Airways Corporation
##  9 HA      Hawaiian Airlines Inc.
## 10 MQ      Envoy Air
## 11 OO      SkyWest Airlines Inc.
## 12 UA      United Air Lines Inc.
## 13 US      US Airways Inc.
## 14 VX      Virgin America
## 15 WN      Southwest Airlines Co.
## 16 YV      Mesa Airlines Inc.
```

airports

```
airports
## # A tibble: 1,458 x 8
##    faa   name                        lat    lon   alt    tz dst   tzone
##    <chr> <chr>                     <dbl>  <dbl> <dbl> <dbl> <chr> <chr>
##  1 04G   Lansdowne Airport          41.1  -80.6  1044    -5 A     America/New_Yo…
##  2 06A   Moton Field Municipal A…   32.5  -85.7   264    -6 A     America/Chicago
##  3 06C   Schaumburg Regional        42.0  -88.1   801    -6 A     America/Chicago
##  4 06N   Randall Airport            41.4  -74.4   523    -5 A     America/New_Yo…
##  5 09J   Jekyll Island Airport      31.1  -81.4    11    -5 A     America/New_Yo…
##  6 0A9   Elizabethton Municipal …   36.4  -82.2  1593    -5 A     America/New_Yo…
##  7 0G6   Williams County Airport    41.5  -84.5   730    -5 A     America/New_Yo…
##  8 0G7   Finger Lakes Regional A…   42.9  -76.8   492    -5 A     America/New_Yo…
##  9 0P2   Shoestring Aviation Air…   39.8  -76.6  1000    -5 U     America/New_Yo…
## 10 0S9   Jefferson County Intl      48.1 -123.    108    -8 A     America/Los_An…
## # … with 1,448 more rows
```

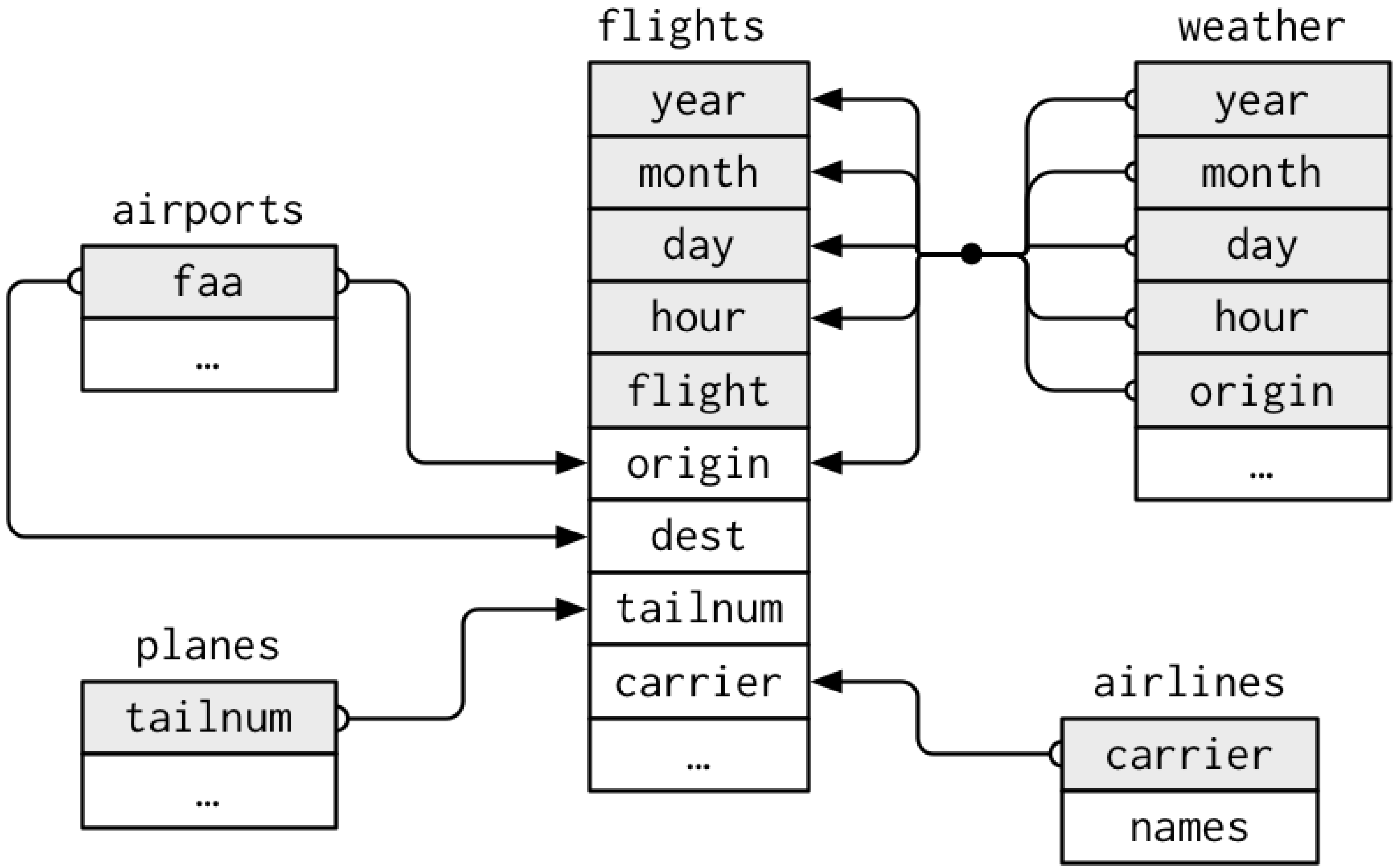
planes

```
planes
## # A tibble: 3,322 x 9
##    tailnum  year type          manufacturer   model  engines seats speed engine
##    <chr>   <int> <chr>         <chr>          <chr>    <int> <int> <int> <chr>
##  1 N10156   2004 Fixed wing m… EMBRAER        EMB-1…       2    55    NA Turbo-…
##  2 N102UW   1998 Fixed wing m… AIRBUS INDUST… A320-…       2   182    NA Turbo-…
##  3 N103US   1999 Fixed wing m… AIRBUS INDUST… A320-…       2   182    NA Turbo-…
##  4 N104UW   1999 Fixed wing m… AIRBUS INDUST… A320-…       2   182    NA Turbo-…
##  5 N10575   2002 Fixed wing m… EMBRAER        EMB-1…       2    55    NA Turbo-…
##  6 N105UW   1999 Fixed wing m… AIRBUS INDUST… A320-…       2   182    NA Turbo-…
##  7 N107US   1999 Fixed wing m… AIRBUS INDUST… A320-…       2   182    NA Turbo-…
##  8 N108UW   1999 Fixed wing m… AIRBUS INDUST… A320-…       2   182    NA Turbo-…
##  9 N109UW   1999 Fixed wing m… AIRBUS INDUST… A320-…       2   182    NA Turbo-…
## 10 N110UW   1999 Fixed wing m… AIRBUS INDUST… A320-…       2   182    NA Turbo-…
## # … with 3,312 more rows
```

weather

```
weather
## # A tibble: 26,115 x 15
##    origin  year month   day  hour  temp  dewp humid wind_dir wind_speed
##    <chr>  <int> <int> <int> <int> <dbl> <dbl> <dbl>    <dbl>      <dbl>
##  1 EWR     2013     1     1     1  39.0  26.1  59.4      270      10.4
##  2 EWR     2013     1     1     2  39.0  27.0  61.6      250       8.06
##  3 EWR     2013     1     1     3  39.0  28.0  64.4      240      11.5
##  4 EWR     2013     1     1     4  39.9  28.0  62.2      250      12.7
##  5 EWR     2013     1     1     5  39.0  28.0  64.4      260      12.7
##  6 EWR     2013     1     1     6  37.9  28.0  67.2      240      11.5
##  7 EWR     2013     1     1     7  39.0  28.0  64.4      240      15.0
##  8 EWR     2013     1     1     8  39.9  28.0  62.2      250      10.4
##  9 EWR     2013     1     1     9  39.9  28.0  62.2      260      15.0
## 10 EWR     2013     1     1    10  41    28.0  59.6      260      13.8
## # … with 26,105 more rows, and 5 more variables: wind_gust <dbl>, precip <dbl>,
## #   pressure <dbl>, visib <dbl>, time_hour <dttm>
```

Concept map of tables and joins from the text



Keys 🔑

- Keys = variables used to connect records in one table to another.
- In the `nycflights13` data,
  - `flights` connects to `planes` by a single variable `tailnum`
  - `flights` connects to `airlines` by a single variable `carrier`
  - `flights` connects to `airports` by two variables, `origin` and `dest`
  - `flights` connects to `weather` using multiple variables, `origin`, and `year`, `month`, `day` and `hour`.

Your turn: go to Rstudio and open today's exercises

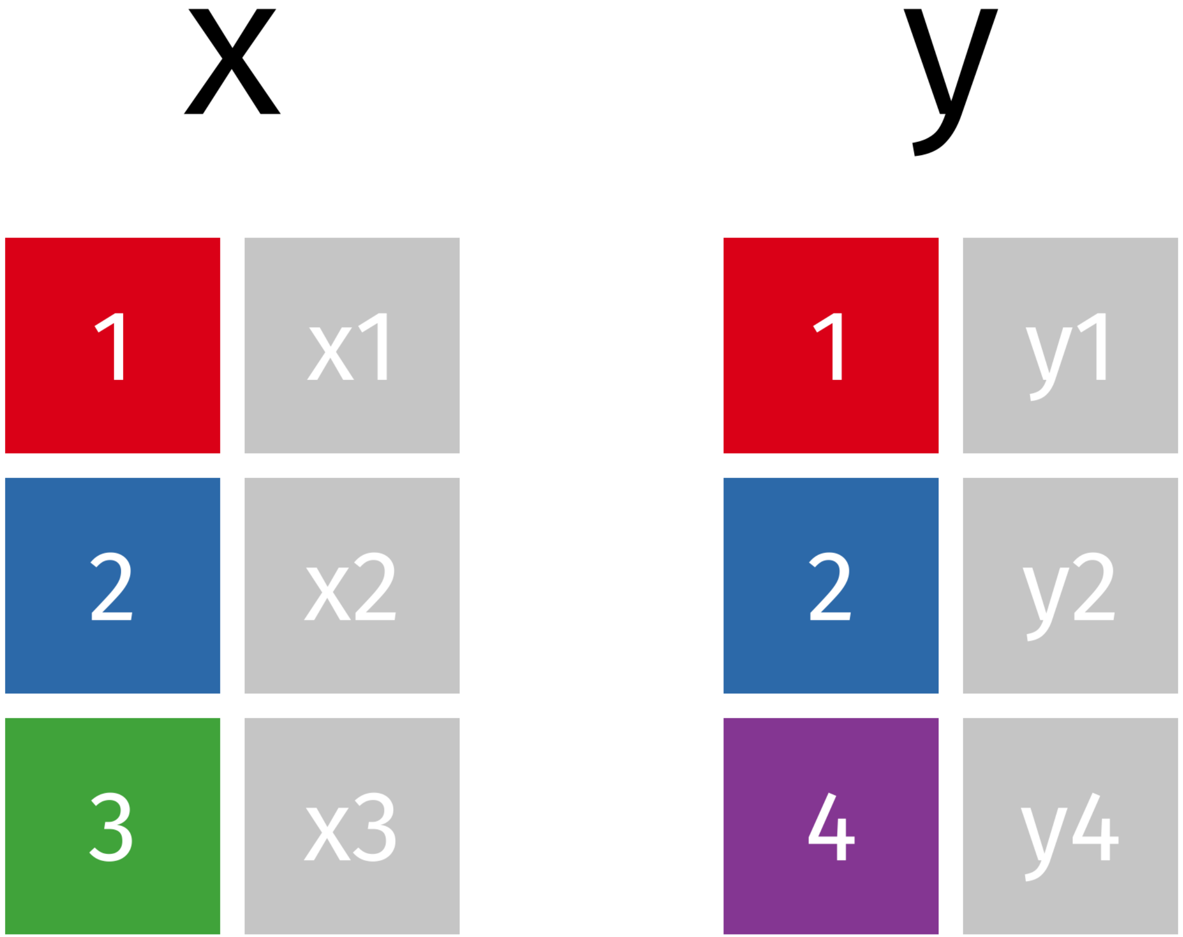
- Load the `Lahman` package, which contains multiple tables of baseball data.
- What key(s) connect the batting table with the salary table?
- Can you draw out a diagram of the connections amongst the tables?

04:00

Joins

- "mutating joins", add variables from one table to another.
- There is always a decision on what observations are copied to the new table as well.
- Let's discuss how joins work using some lovely animations provided by Garrick Aden-Buie.

Example data



x

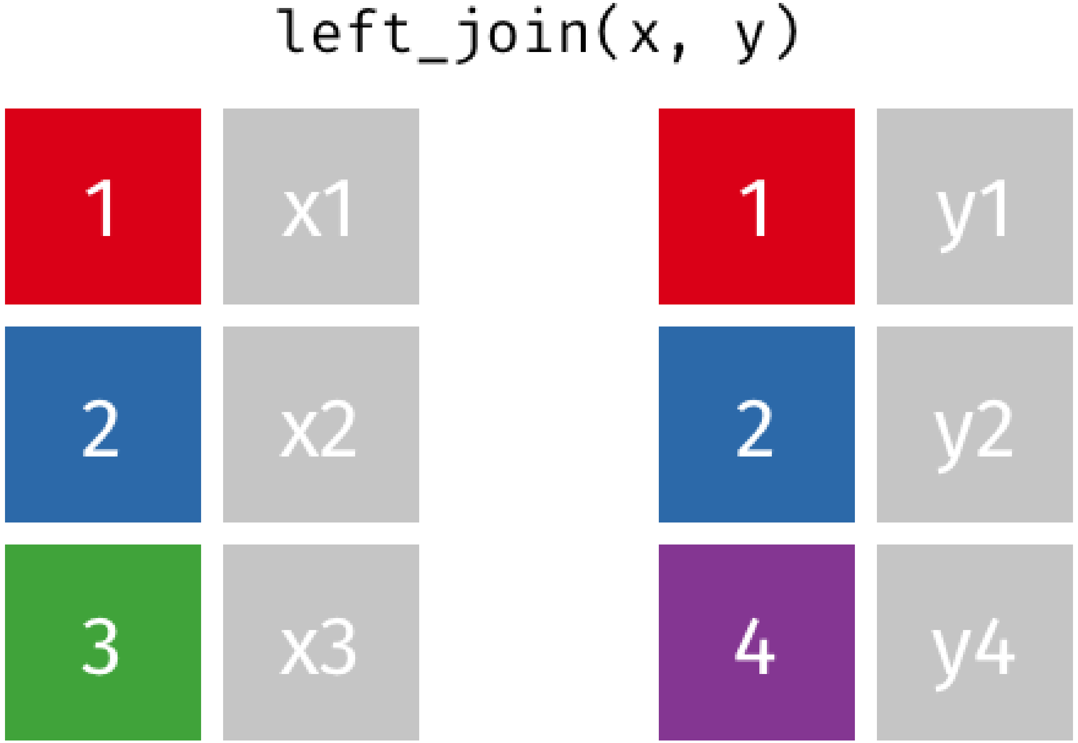
1	x1
2	x2
3	x3

y

1	y1
2	y2
4	y4

Left Join (Generally the one you want to use)

All observations from the "left" table, but only the observations from the "right" table that match those in the left.



```
left_join(x, y)
```

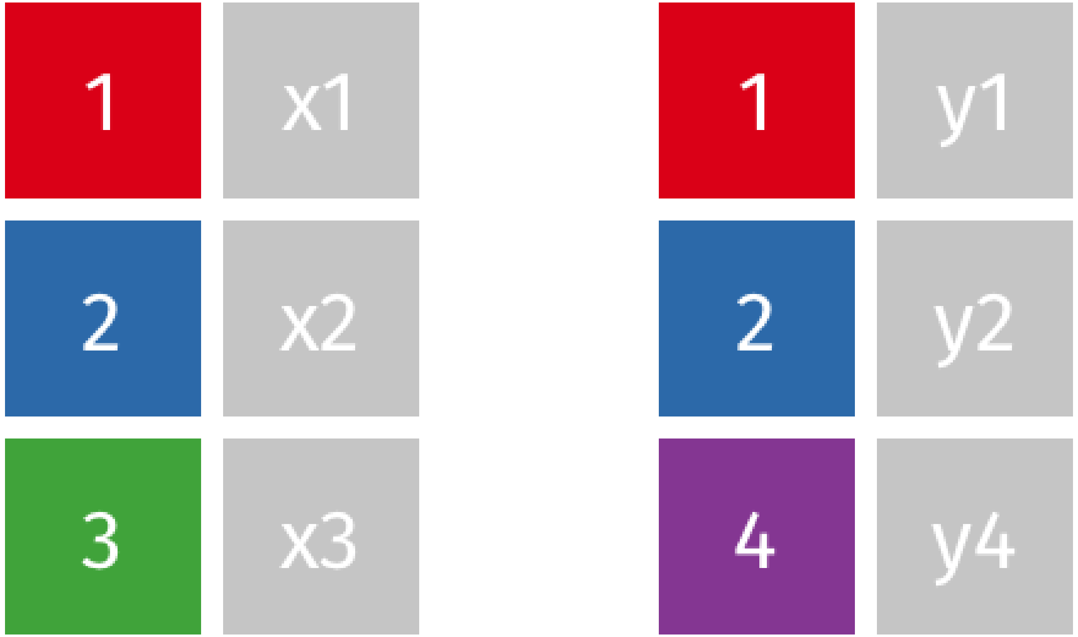
1	x1
2	x2
3	x3

1	y1
2	y2
4	y4

Right Join

Same as left join, but in reverse.

```
right_join(x, y)
```



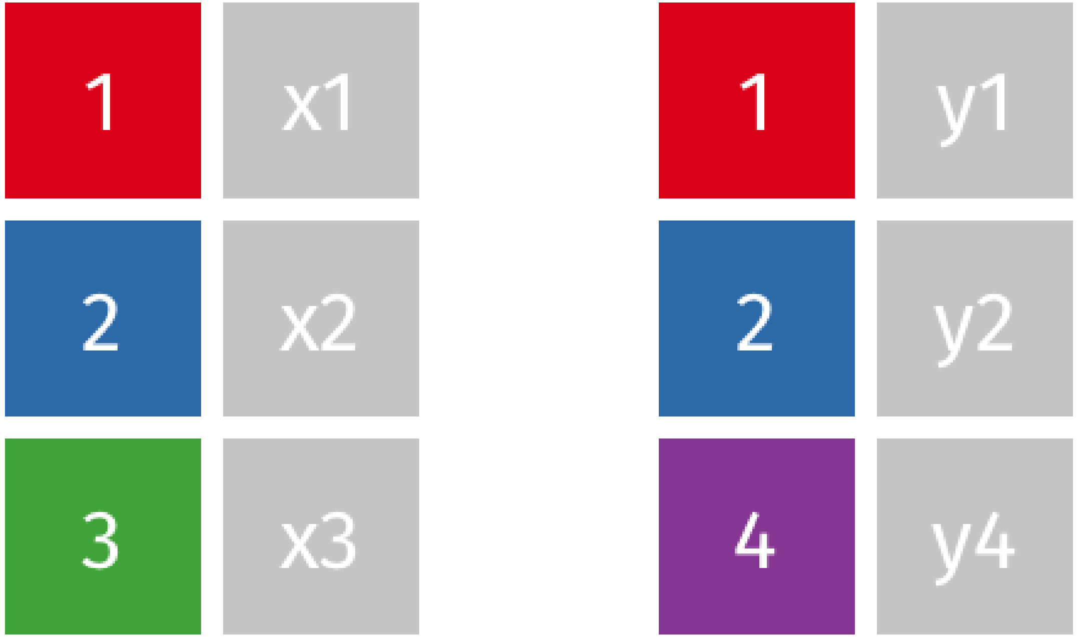
1	x1
2	x2
3	x3

1	y1
2	y2
4	y4

Inner join

Intersection between the two tables, only the observations that are in both

`inner_join(x, y)`



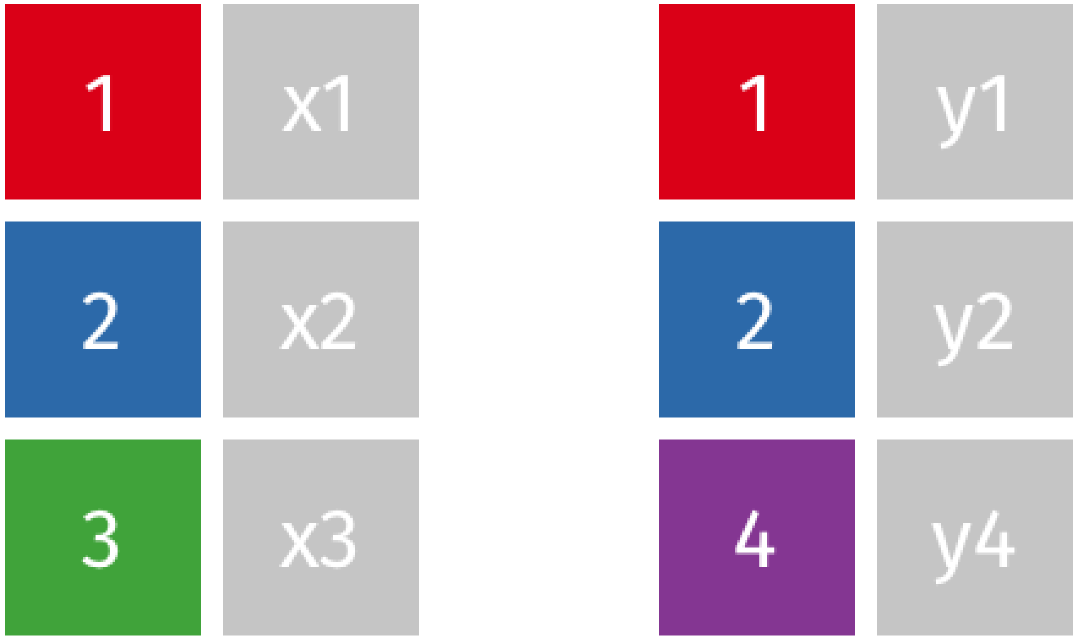
1	x1
2	x2
3	x3

1	y1
2	y2
4	y4

Outer (full) join

Union of the two tables, all observations from both, and missing values might get added

```
full_join(x, y)
```



1	x1
2	x2
3	x3

1	y1
2	y2
4	y4

Combine full airline name with flights data?

```
flights
## # A tibble: 336,776 x 19
##     year month   day dep_time sched_dep_time dep_delay arr_time sched_arr_time
##    <int> <int> <int>    <int>          <int>     <dbl>    <int>          <int>
##  1  2013     1     1      517            515         2      830            819
##  2  2013     1     1      533            529         4      850            830
##  3  2013     1     1      542            540         2      923            850
##  4  2013     1     1      544            545        -1     1004           1022
##  5  2013     1     1      554            600        -6      812            837
##  6  2013     1     1      554            558        -4      740            728
##  7  2013     1     1      555            600        -5      913            854
##  8  2013     1     1      557            600        -3      709            723
##  9  2013     1     1      557            600        -3      838            846
## 10  2013     1     1      558            600        -2      753            745
## # … with 336,766 more rows, and 11 more variables: arr_delay <dbl>,
## #   carrier <chr>, flight <int>, tailnum <chr>, origin <chr>, dest <chr>,
## #   air_time <dbl>, distance <dbl>, hour <dbl>, minute <dbl>, time_hour <dttm>
```

Combine full airline name with flights data?

```
airlines
## # A tibble: 16 x 2
##    carrier name
##    <chr>   <chr>
##  1 9E      Endeavor Air Inc.
##  2 AA      American Airlines Inc.
##  3 AS      Alaska Airlines Inc.
##  4 B6      JetBlue Airways
##  5 DL      Delta Air Lines Inc.
##  6 EV      ExpressJet Airlines Inc.
##  7 F9      Frontier Airlines Inc.
##  8 FL      AirTran Airways Corporation
##  9 HA      Hawaiian Airlines Inc.
## 10 MQ      Envoy Air
## 11 OO      SkyWest Airlines Inc.
## 12 UA      United Air Lines Inc.
## 13 US      US Airways Inc.
## 14 VX      Virgin America
## 15 WN      Southwest Airlines Co.
## 16 YV      Mesa Airlines Inc.
```

Combine airlines & flights using left_join()

```r
flights %>%
  left_join(airlines,
            by = "carrier") %>%
  glimpse()
```

```
## Observations: 336,776
## Variables: 20
## $ year           <int> 2013, 2013, 2013, 2013, 2
## $ month          <int> 1, 1, 1, 1, 1, 1, 1, 1, 1
## $ day            <int> 1, 1, 1, 1, 1, 1, 1, 1, 1
## $ dep_time       <int> 517, 533, 542, 544, 554,
## $ sched_dep_time <int> 515, 529, 540, 545, 600,
## $ dep_delay      <dbl> 2, 4, 2, -1, -6, -4, -5,
## $ arr_time       <int> 830, 850, 923, 1004, 812,
## $ sched_arr_time <int> 819, 830, 850, 1022, 837,
## $ arr_delay      <dbl> 11, 20, 33, -18, -25, 12,
## $ carrier        <chr> "UA", "UA", "AA", "B6", "
## $ flight         <int> 1545, 1714, 1141, 725, 46
## $ tailnum        <chr> "N14228", "N24211", "N619
## $ origin         <chr> "EWR", "LGA", "JFK", "JFK
## $ dest           <chr> "IAH", "IAH", "MIA", "BQN
## $ air_time       <dbl> 227, 227, 160, 183, 116,
## $ distance       <dbl> 1400, 1416, 1089, 1576, 7
## $ hour           <dbl> 5, 5, 5, 5, 6, 5, 6, 6, 6
```

Example: flights joining to airports

```r
flights %>%
  left_join(
    airports,
    by = c("origin" = "faa")) %>
  glimpse()
```

```
## Observations: 336,776
## Variables: 26
## $ year           <int> 2013, 2013, 2013, 2013, 20
## $ month          <int> 1, 1, 1, 1, 1, 1, 1, 1, 1
## $ day            <int> 1, 1, 1, 1, 1, 1, 1, 1, 1
## $ dep_time       <int> 517, 533, 542, 544, 554,
## $ sched_dep_time <int> 515, 529, 540, 545, 600,
## $ dep_delay      <dbl> 2, 4, 2, -1, -6, -4, -5,
## $ arr_time       <int> 830, 850, 923, 1004, 812,
## $ sched_arr_time <int> 819, 830, 850, 1022, 837,
## $ arr_delay      <dbl> 11, 20, 33, -18, -25, 12,
## $ carrier        <chr> "UA", "UA", "AA", "B6", "
## $ flight         <int> 1545, 1714, 1141, 725, 46
## $ tailnum        <chr> "N14228", "N24211", "N619
## $ origin         <chr> "EWR", "LGA", "JFK", "JFK
## $ dest           <chr> "IAH", "IAH", "MIA", "BQN
## $ air_time       <dbl> 227, 227, 160, 183, 116,
## $ distance       <dbl> 1400, 1416, 1089, 1576, 7
## $ hour           <dbl> 5, 5, 5, 5, 6, 5, 6, 6, 6
## $ minute         <dbl> 15, 29, 40, 45, 0, 58, 0,
```

Airline travel, ontime data

```
plane_N4YRAA <- read_csv("data/plane_N4YRAA.csv")

glimpse(plane_N4YRAA)
## Observations: 145
## Variables: 8
## $ FL_DATE  <date> 2017-05-26, 2017-05-02, 2017-05-05, 2017-05-11, 2017-05-03,…
## $ CARRIER  <chr> "AA", "AA", "AA", "AA", "AA", "AA", "AA", "AA", "AA", "AA", …
## $ FL_NUM   <dbl> 2246, 2276, 2278, 2287, 2288, 2291, 2297, 2297, 2297, 2297, …
## $ ORIGIN   <chr> "CVG", "DFW", "DFW", "STL", "IND", "CHS", "DFW", "DFW", "MKE…
## $ DEST     <chr> "DFW", "IND", "OKC", "ORD", "DFW", "DFW", "MKE", "MKE", "DFW…
## $ DEP_TIME <chr> "0748", "2020", "0848", "0454", "0601", "0807", "0700", "065…
## $ ARR_TIME <chr> "0917", "2323", "0941", "0600", "0719", "0947", "0905", "090…
## $ DISTANCE <dbl> 812, 761, 175, 258, 761, 987, 853, 853, 853, 853, 447, 447, …
```

Airline travel, airport location

```
airport_raw <- read_csv("data/airports.csv")

airport_raw %>%
  select(AIRPORT,
         LATITUDE,
         LONGITUDE,
         AIRPORT_STATE_NAME) %>%
  glimpse()
## Observations: 13,094
## Variables: 4
## $ AIRPORT            <chr> "01A", "03A", "04A", "05A", "06A", "07A", "08A", "…
## $ LATITUDE           <dbl> 58.10944, 65.54806, 68.08333, 67.57000, 57.74528, …
## $ LONGITUDE          <dbl> -152.90667, -161.07167, -163.16667, -148.18389, -1…
## $ AIRPORT_STATE_NAME <chr> "Alaska", "Alaska", "Alaska", "Alaska", "Alaska", …
```

Our Turn: Joining the two tables to show flight movements

- Go to Rstudio and open "flight-movements.Rmd" and complete exercise - the aim is to show flight movement on the map
- Next: Open "nycflights.Rmd"

Learning more

- The coat explanation of joins: Different types of joins explained using a person and a coat, by Leight Tami

References

- Chapter 13 of R4DS