Combining Multiple Tables

Today we will start to look at the nycflights13 data tables.

You can think of this dataset as if it was a spreadsheet with multiple spreadsheets within a workbook.

Take a look at the *flights* table and the *airlines* table. Do the two tables have a common variable that can be used as a key to match the rows of the tables?

library(tidyverse)
library(nycflights13)
library(skimr)

flights %>% head()

## # A tibble: 6 x 19
## year month day dep\_time sched\_dep\_time dep\_delay arr\_time
## <int> <int> <int> <int> <int> <dbl> <int>
## 1 2013 1 1 517 515 2 830
## 2 2013 1 1 533 529 4 850
## 3 2013 1 1 542 540 2 923
## 4 2013 1 1 544 545 -1 1004
## 5 2013 1 1 554 600 -6 812
## 6 2013 1 1 554 558 -4 740
## # ... with 12 more variables: sched\_arr\_time <int>, 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 %>% head()

## # A tibble: 6 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.

airports %>% head()

## # A tibble: 6 x 8
## faa name lat lon alt tz dst tzone
## <chr> <chr> <dbl> <dbl> <int> <dbl> <chr> <chr>
## 1 04G Lansdowne Airport 41.1 -80.6 1044 -5 A America/New…
## 2 06A Moton Field Municipal … 32.5 -85.7 264 -6 A America/Chi…
## 3 06C Schaumburg Regional 42.0 -88.1 801 -6 A America/Chi…
## 4 06N Randall Airport 41.4 -74.4 523 -5 A America/New…
## 5 09J Jekyll Island Airport 31.1 -81.4 11 -5 A America/New…
## 6 0A9 Elizabethton Municipal… 36.4 -82.2 1593 -5 A America/New…

flightsJoined <- flights %>%
 inner\_join(airlines, by = c("carrier" = "carrier") )

flightsJoined %>% head()

## # A tibble: 6 x 20
## year month day dep\_time sched\_dep\_time dep\_delay arr\_time
## <int> <int> <int> <int> <int> <dbl> <int>
## 1 2013 1 1 517 515 2 830
## 2 2013 1 1 533 529 4 850
## 3 2013 1 1 542 540 2 923
## 4 2013 1 1 544 545 -1 1004
## 5 2013 1 1 554 600 -6 812
## 6 2013 1 1 554 558 -4 740
## # ... with 13 more variables: sched\_arr\_time <int>, 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>, name <chr>

glimpse(flightsJoined)

## Observations: 336,776
## Variables: 20
## $ year <int> 2013, 2013, 2013, 2013, 2013, 2013, 2013, 2013,...
## $ month <int> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,...
## $ day <int> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,...
## $ dep\_time <int> 517, 533, 542, 544, 554, 554, 555, 557, 557, 55...
## $ sched\_dep\_time <int> 515, 529, 540, 545, 600, 558, 600, 600, 600, 60...
## $ dep\_delay <dbl> 2, 4, 2, -1, -6, -4, -5, -3, -3, -2, -2, -2, -2...
## $ arr\_time <int> 830, 850, 923, 1004, 812, 740, 913, 709, 838, 7...
## $ sched\_arr\_time <int> 819, 830, 850, 1022, 837, 728, 854, 723, 846, 7...
## $ arr\_delay <dbl> 11, 20, 33, -18, -25, 12, 19, -14, -8, 8, -2, -...
## $ carrier <chr> "UA", "UA", "AA", "B6", "DL", "UA", "B6", "EV",...
## $ flight <int> 1545, 1714, 1141, 725, 461, 1696, 507, 5708, 79...
## $ tailnum <chr> "N14228", "N24211", "N619AA", "N804JB", "N668DN...
## $ origin <chr> "EWR", "LGA", "JFK", "JFK", "LGA", "EWR", "EWR"...
## $ dest <chr> "IAH", "IAH", "MIA", "BQN", "ATL", "ORD", "FLL"...
## $ air\_time <dbl> 227, 227, 160, 183, 116, 150, 158, 53, 140, 138...
## $ distance <dbl> 1400, 1416, 1089, 1576, 762, 719, 1065, 229, 94...
## $ hour <dbl> 5, 5, 5, 5, 6, 5, 6, 6, 6, 6, 6, 6, 6, 6, 6, 5,...
## $ minute <dbl> 15, 29, 40, 45, 0, 58, 0, 0, 0, 0, 0, 0, 0, 0, ...
## $ time\_hour <dttm> 2013-01-01 05:00:00, 2013-01-01 05:00:00, 2013...
## $ name <chr> "United Air Lines Inc.", "United Air Lines Inc....

A package that I like for summarizing the variables in a dataframe is the *skim* function from the *skimr* package. What do you think?

skim(flightsJoined)

## Skim summary statistics
## n obs: 336776
## n variables: 20
##
## ── Variable type:character ───────────────────────────────────────────────────────────────────
## variable missing complete n min max empty n\_unique
## carrier 0 336776 336776 2 2 0 16
## dest 0 336776 336776 3 3 0 105
## name 0 336776 336776 9 27 0 16
## origin 0 336776 336776 3 3 0 3
## tailnum 2512 334264 336776 5 6 0 4043
##
## ── Variable type:integer ─────────────────────────────────────────────────────────────────────
## variable missing complete n mean sd p0 p25 p50
## arr\_time 8713 328063 336776 1502.05 533.26 1 1104 1535
## day 0 336776 336776 15.71 8.77 1 8 16
## dep\_time 8255 328521 336776 1349.11 488.28 1 907 1401
## flight 0 336776 336776 1971.92 1632.47 1 553 1496
## month 0 336776 336776 6.55 3.41 1 4 7
## sched\_arr\_time 0 336776 336776 1536.38 497.46 1 1124 1556
## sched\_dep\_time 0 336776 336776 1344.25 467.34 106 906 1359
## year 0 336776 336776 2013 0 2013 2013 2013
## p75 p100 hist
## 1940 2400 ▁▁▃▇▆▆▇▆
## 23 31 ▇▇▇▇▆▇▇▇
## 1744 2400 ▁▁▇▆▆▇▆▂
## 3465 8500 ▇▅▂▃▂▁▁▁
## 10 12 ▇▅▇▃▅▇▅▇
## 1945 2359 ▁▁▂▇▆▇▇▆
## 1729 2359 ▁▃▇▆▆▇▇▂
## 2013 2013 ▁▁▁▇▁▁▁▁
##
## ── Variable type:numeric ─────────────────────────────────────────────────────────────────────
## variable missing complete n mean sd p0 p25 p50 p75 p100
## air\_time 9430 327346 336776 150.69 93.69 20 82 129 192 695
## arr\_delay 9430 327346 336776 6.9 44.63 -86 -17 -5 14 1272
## dep\_delay 8255 328521 336776 12.64 40.21 -43 -5 -2 11 1301
## distance 0 336776 336776 1039.91 733.23 17 502 872 1389 4983
## hour 0 336776 336776 13.18 4.66 1 9 13 17 23
## minute 0 336776 336776 26.23 19.3 0 8 29 44 59
## hist
## ▇▇▂▃▁▁▁▁
## ▇▁▁▁▁▁▁▁
## ▇▁▁▁▁▁▁▁
## ▆▇▂▂▁▁▁▁
## ▁▃▇▆▅▇▇▂
## ▇▂▃▃▅▂▃▅
##
## ── Variable type:POSIXct ─────────────────────────────────────────────────────────────────────
## variable missing complete n min max median
## time\_hour 0 336776 336776 2013-01-01 2013-12-31 2013-07-03
## n\_unique
## 6936

Check that the new column *name* has been added.

flightsJoined %>% select(carrier, name, flight, origin, dest) %>%
 head()

## # A tibble: 6 x 5
## carrier name flight origin dest
## <chr> <chr> <int> <chr> <chr>
## 1 UA United Air Lines Inc. 1545 EWR IAH
## 2 UA United Air Lines Inc. 1714 LGA IAH
## 3 AA American Airlines Inc. 1141 JFK MIA
## 4 B6 JetBlue Airways 725 JFK BQN
## 5 DL Delta Air Lines Inc. 461 LGA ATL
## 6 UA United Air Lines Inc. 1696 EWR ORD

How big are the dataframes? Note the base R finction *nrow* just give a number as output, using the dplyr function *count()* give a dataframe.

nrow(flights)

## [1] 336776

nrow(flightsJoined)

## [1] 336776

flights %>% count()

## # A tibble: 1 x 1
## n
## <int>
## 1 336776

flightsJoined %>% count()

## # A tibble: 1 x 1
## n
## <int>
## 1 336776

Suppose we are interested in the flights from NYC airports to the West Coast.

The Pacific Time Zone is time zone -8 in the airports table. (I do not know why there are more airports that given in the book.)

airportsPT <- airports %>% filter(tz == -8)

airportsPT %>% head()

## # A tibble: 6 x 8
## faa name lat lon alt tz dst tzone
## <chr> <chr> <dbl> <dbl> <int> <dbl> <chr> <chr>
## 1 0S9 Jefferson County I… 48.1 -123. 108 -8 A America/Los\_Ang…
## 2 1C9 Frazier Lake Airpa… 54.0 -125. 152 -8 A America/Vancouv…
## 3 1RL Point Roberts Airp… 49.0 -123. 10 -8 A America/Los\_Ang…
## 4 38W Lynden Airport 49.0 -122. 106 -8 A America/Los\_Ang…
## 5 49X Chemehuevi Valley 34.5 -114. 638 -8 A America/Los\_Ang…
## 6 55S Packwood 46.4 -121. 1057 -8 A America/Los\_Ang…

airportsPT %>% count()

## # A tibble: 1 x 1
## n
## <int>
## 1 178

Now if we are interested in the flights from NYC to the west coast, find the airports in the Pacific Time Zone and join the airportPT we will get the flights to the west coast.

nycDestPT <- flights %>% inner\_join(airportsPT, by = c("dest" = "faa") )

nycDestPT %>% count()

## # A tibble: 1 x 1
## n
## <int>
## 1 46324

If we *left\_join* we will get all of the rows in *flights*.

Check out the *map* function, it applies a function to all columns.

nycDest <- flights %>% left\_join(airportsPT, by = c("dest" = "faa") )

nycDest %>% count()

## # A tibble: 1 x 1
## n
## <int>
## 1 336776

nycDest %>% head()

## # A tibble: 6 x 26
## year month day dep\_time sched\_dep\_time dep\_delay arr\_time
## <int> <int> <int> <int> <int> <dbl> <int>
## 1 2013 1 1 517 515 2 830
## 2 2013 1 1 533 529 4 850
## 3 2013 1 1 542 540 2 923
## 4 2013 1 1 544 545 -1 1004
## 5 2013 1 1 554 600 -6 812
## 6 2013 1 1 554 558 -4 740
## # ... with 19 more variables: sched\_arr\_time <int>, 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>, name <chr>, lat <dbl>, lon <dbl>, alt <int>,
## # tz <dbl>, dst <chr>, tzone <chr>

nycDest %>% map(~sum(is.na(.)))

## $year
## [1] 0
##
## $month
## [1] 0
##
## $day
## [1] 0
##
## $dep\_time
## [1] 8255
##
## $sched\_dep\_time
## [1] 0
##
## $dep\_delay
## [1] 8255
##
## $arr\_time
## [1] 8713
##
## $sched\_arr\_time
## [1] 0
##
## $arr\_delay
## [1] 9430
##
## $carrier
## [1] 0
##
## $flight
## [1] 0
##
## $tailnum
## [1] 2512
##
## $origin
## [1] 0
##
## $dest
## [1] 0
##
## $air\_time
## [1] 9430
##
## $distance
## [1] 0
##
## $hour
## [1] 0
##
## $minute
## [1] 0
##
## $time\_hour
## [1] 0
##
## $name
## [1] 290452
##
## $lat
## [1] 290452
##
## $lon
## [1] 290452
##
## $alt
## [1] 290452
##
## $tz
## [1] 290452
##
## $dst
## [1] 290452
##
## $tzone
## [1] 290452