

Lecture 9: The `%>%` Pipe Operator

STAT 450, Fall 2021

`dplyr` functions for data wrangling:

- `select()` take a subset of the columns (variables)
- `filter()` take a subset of the rows (observations)
- `arrange()` reorder the rows
- `mutate()` creates new variables that are functions of existing variables

The function names are *verbs* that describe the type of action each function performs on the data.

Last time we discussed how to use these function one-at-a-time. Today we will discuss the pipe operator `%>%` which can be used to combine a sequence of `dplyr` operations.

Flights Data Set

The `flights` data set contains information on all the flights that departed from New York City in 2013.

```
library(tidyverse)
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 <dtm>
```

Type `help(flights)` to read the documentation on this data set.

Exercise 1 (review): Use `filter()` to find all flights that

- Had an arrival delay of two or more hours
- Were operated by United (UA), American (AA), or Delta (DL)

Exercise 2: How many flights have a missing value for `dep_time`? What other variables have missing values? What might these rows represent? (Hint: use `is.na()`)

Using the pipe %>%

The pipe %>% allows us to combine a sequence of operations using dplyr.

Select columns by name:

```
flights %>% select(year:day, origin, dest)
```

```
## # A tibble: 336,776 x 5
##   year month   day origin dest
##   <int> <int> <int> <chr>  <chr>
## 1  2013     1     1 EWR    IAH
## 2  2013     1     1 LGA    IAH
## 3  2013     1     1 JFK    MIA
## 4  2013     1     1 JFK    BQN
## 5  2013     1     1 LGA    ATL
## 6  2013     1     1 EWR    ORD
## 7  2013     1     1 EWR    FLL
## 8  2013     1     1 LGA    IAD
## 9  2013     1     1 JFK    MCO
## 10 2013     1     1 LGA    ORD
## # ... with 336,766 more rows
```

Subset all flights on Dec 25:

```
flights %>% filter(month == 12, day == 25)
```

```
## # A tibble: 719 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    12    25     456           500          -4     649           651
## 2  2013    12    25     524           515           9     805           814
## 3  2013    12    25     542           540           2     832           850
## 4  2013    12    25     546           550          -4    1022          1027
## 5  2013    12    25     556           600          -4     730           745
## 6  2013    12    25     557           600          -3     743           752
## 7  2013    12    25     557           600          -3     818           831
## 8  2013    12    25     559           600          -1     855           856
## 9  2013    12    25     559           600          -1     849           855
## 10 2013    12    25     600           600           0     850           846
## # ... with 709 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 <dtm>
```

Select columns and then subset flights on Dec 25:

```
flights %>%
  select(year:day, origin, dest) %>%
  filter(month == 12, day == 25)
```

```
## # A tibble: 719 x 5
##   year month   day origin dest
##   <int> <int> <int> <chr>  <chr>
## 1  2013    12    25 EWR    CLT
## 2  2013    12    25 EWR    IAH
## 3  2013    12    25 JFK    MIA
## 4  2013    12    25 JFK    BQN
## 5  2013    12    25 LGA    ORD
## 6  2013    12    25 LGA    DTW
## 7  2013    12    25 LGA    ATL
## 8  2013    12    25 LGA    FLL
## 9  2013    12    25 EWR    FLL
## 10 2013    12    25 JFK    MCO
## # ... with 709 more rows
```

Select columns and then use `mutate()` to add a new column with the `speed` of the aircraft in miles per hour.

```
flights %>%
  select(year:day, distance, air_time) %>%
  mutate(speed = distance / air_time * 60)
```

```
## # A tibble: 336,776 x 6
##   year month   day distance air_time speed
##   <int> <int> <int>     <dbl>   <dbl> <dbl>
## 1  2013     1     1     1400     227  370.
## 2  2013     1     1     1416     227  374.
## 3  2013     1     1     1089     160  408.
## 4  2013     1     1     1576     183  517.
## 5  2013     1     1      762     116  394.
## 6  2013     1     1      719     150  288.
## 7  2013     1     1     1065     158  404.
## 8  2013     1     1      229      53  259.
## 9  2013     1     1      944     140  405.
## 10 2013     1     1      733     138  319.
## # ... with 336,766 more rows
```

Exercise 3:

- (a) Run the following code. Which three variables get selected by `contains("dep")`, and how are they related?

```
flights %>%  
  select(year:day, carrier, contains("dep")) %>%  
  filter(carrier == "UA")
```

- (b) Next, add another pipe with the `arrange()` function to identify the UA flights with the longest departure delays.
- (c) Similarly, identify the UA flights with the longest arrival delays.