

DATA 100, Week 3 (A)

Recall `library(tidyverse)` should appear in the beginning of every .qmd (or .rmd) file we have from now on

3.3.4. `filter()`

Get flights going out on a given day. **Note** that the column names `month` and `day` are not in quotation marks

```
filter(flights, month == 6, day == 30) # June 30th
```

DATA 100, Week 3 (A)

```
#> # A tibble: 918 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     6    30      12          2231          101     352           226
#> 2  2013     6    30      21          2300           81     116            8
#> 3  2013     6    30      23          2055          208     123          2230
#> 4  2013     6    30      25          2359           26     413           350
#> 5  2013     6    30      43          2250          113     150            14
#> 6  2013     6    30      56          2245          131     201            3
#> # i 912 more rows
#> # i 11 more variables: arr_delay <dbl>, carrier <chr>, flight <int>, ...
#filter(flights, month == 2, day == 30) # Non-existent February 30th
```

DATA 100, Week 3 (A)

```
# use pipe
flights |>
filter(month == 6, day == 30)
```

```
#> # A tibble: 918 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     6    30      12           2231         101     352           226
#> 2  2013     6    30      21           2300          81     116            8
#> 3  2013     6    30      23           2055         208     123          2230
#> 4  2013     6    30      25           2359          26     413           350
#> 5  2013     6    30      43           2250         113     150            14
#> 6  2013     6    30      56           2245         131     201            3
#> # i 912 more rows
#> # i 11 more variables: arr_delay <dbl>, carrier <chr>, flight <int>, ...
```

Since the input dataframe is not modified, if the filtered result is needed later, should assign it to a new variable to keep the filtered result.

DATA 100, Week 3 (A)

Comment: The name of the variable should reflect the meaning of its contents. If possible, should avoid *update in-place* as much as possible.

```
Jun30 <- flights |> filter(month == 6, day == 30)
```

```
Jun30
```

```
# A tibble: 918 × 19
  year month   day dep_time sched_dep_time dep_delay arr_time sched_arr_time arr_delay carrier flight
  <int> <int> <int>   <int>         <int>         <dbl>   <int>         <int>         <dbl> <chr>   <int>
1  2013     6    30      12           2231          101     352           226           86 B6      1203
2  2013     6    30      21           2300           81     116            8      68 B6       718
3  2013     6    30      23           2055          208     123           2230          173 WN       579
4  2013     6    30      25           2359           26     413           350           23 B6       745
5  2013     6    30      43           2250          113     150            14          96 B6      2002
6  2013     6    30      56           2245          131     201            3         118 B6       486
7  2013     6    30     116           2359           77     451           344           67 B6      1503
8  2013     6    30     153           2245          188     422           135          167 B6       623
9  2013     6    30     217           2359          138     545           340          125 B6       839
10 2013     6    30     525           500           25     703           640           23 US      1431
# i 908 more rows
# i 8 more variables: tailnum <chr>, origin <chr>, dest <chr>, air_time <dbl>, distance <dbl>,
#   hour <dbl>, minute <dbl>, time_hour <dtm>
# i Use `print(n = ...)` to see more rows
```

DATA 100, Week 3 (A)

Surrounding the assign statement by a pair of round parenthesis () accomplishes assignment and print-out in one go.

```
(September <- flights |> filter(month == 9))
```

```
#> # A tibble: 27,574 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     9     1         9         2359         10       343           340
#> 2  2013     9     1        117         2245        152       218          2359
#> 3  2013     9     1        508         516         -8       717           800
#> 4  2013     9     1        537         545         -8       849           855
#> 5  2013     9     1        537         545         -8       906           921
#> 6  2013     9     1        549         600        -11       815           850
#> # i 27,568 more rows
#> # i 11 more variables: arr_delay <dbl>, carrier <chr>, flight <int>, ...
```

DATA 100, Week 3 (A)

```
# get flights during the first half of the months  
flights |> filter(day %in% 1:15)
```

```
#> # A tibble: 166,192 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  
#> # i 166,186 more rows  
#> # i 11 more variables: arr_delay <dbl>, carrier <chr>, flight <int>, ...
```

DATA 100, Week 3 (A)

Rules of logical operations:

- DeMorgan's Laws

- $\neg(x \mid y) = (\neg x) \ \& \ (\neg y)$

- $\neg(x \ \& \ y) = (\neg x) \mid (\neg y)$

- Distributivity

- $x \ \& \ (y \mid z) = (x \ \& \ y) \mid (x \ \& \ z)$

- $x \mid (y \ \& \ z) = (x \mid y) \ \& \ (x \mid z)$

DATA 100, Week 3 (A)

The following commands give the same list of flights

- In filter's parameter list: and can be represented by , and & in the filtering conditions.

```
flights |> filter(!(arr_delay > 120 | dep_delay > 120))
```

```
flights |> filter(arr_delay <= 120 & dep_delay <= 120)
```

```
flights |> filter(arr_delay <= 120, dep_delay <= 120)
```


DATA 100, Week 3 (A)

Filter out (for) missing values

- `== NA` or `!= NA` do not work
- ... since as we saw `NA == NA` is still `NA`, (as is `NA != NA`)
- Use `is.na()`

DATA 100, Week 3 (A)

The flights that departed:

```
glimpse(flights |> filter(!is.na(dep_time)))
```

DATA 100, Week 3 (A)

```
#> Rows: 328,521
#> Columns: 19
#> $ year      <int> 2013, 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, 1~
#> $ day       <int> 1, 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, 558, 55~
#> $ sched_dep_time <int> 515, 529, 540, 545, 600, 558, 600, 600, 600, 600, 60~
#> $ dep_delay <dbl> 2, 4, 2, -1, -6, -4, -5, -3, -3, -2, -2, -2, -2, -2, ~
#> $ arr_time  <int> 830, 850, 923, 1004, 812, 740, 913, 709, 838, 753, 8~
#> $ sched_arr_time <int> 819, 830, 850, 1022, 837, 728, 854, 723, 846, 745, 8~
#> $ arr_delay <dbl> 11, 20, 33, -18, -25, 12, 19, -14, -8, 8, -2, -3, 7, ~
#> $ carrier   <chr> "UA", "UA", "AA", "B6", "DL", "UA", "B6", "EV", "B6"~
#> $ flight    <int> 1545, 1714, 1141, 725, 461, 1696, 507, 5708, 79, 301~
#> $ tailnum   <chr> "N14228", "N24211", "N619AA", "N804JB", "N668DN", "N~
#> $ origin    <chr> "EWR", "LGA", "JFK", "JFK", "LGA", "EWR", "EWR", "LG~
#> $ dest      <chr> "IAH", "IAH", "MIA", "BQN", "ATL", "ORD", "FLL", "IA~
#> $ air_time  <dbl> 227, 227, 160, 183, 116, 150, 158, 53, 140, 138, 149~
#> $ distance  <dbl> 1400, 1416, 1089, 1576, 762, 719, 1065, 229, 944, 73~
#> $ hour      <dbl> 5, 5, 5, 5, 6, 5, 6, 6, 6, 6, 6, 6, 6, 6, 5, 6, 6~
#> $ minute    <dbl> 15, 29, 40, 45, 0, 58, 0, 0, 0, 0, 0, 0, 0, 0, 0, 59~
#> $ time_hour <dtm> 2013-01-01 05:00:00, 2013-01-01 05:00:00, 2013-01-0~
#v.s.
```

`glimpse(flights |> filter(dep_time != NA))`

DATA 100, Week 3 (A)

Rows: 0

Columns: 19

```
$ year      <int>
$ month     <int>
$ day       <int>
$ dep_time  <int>
$ sched_dep_time <int>
$ dep_delay <dbl>
$ arr_time  <int>
$ 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 <dtm>
```

The flights that never arrived:

DATA 100, Week 3 (A)

`glimpse(flights |> filter(is.na(arr_time)))`

```
#> Rows: 8,713
#> Columns: 19
#> $ year      <int> 2013, 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, 1~
#> $ day       <int> 1, 1, 1, 1, 1, 2, 2, 2, 2, 2, 2, 2, 2, 2, 3, 3, 3~
#> $ dep_time  <int> 2016, NA, NA, NA, NA, 2041, 2145, NA, NA, NA, NA, NA~
#> $ sched_dep_time <int> 1930, 1630, 1935, 1500, 600, 2045, 2129, 1540, 1620,~
#> $ dep_delay <dbl> 46, NA, NA, NA, NA, -4, 16, NA, NA, NA, NA, NA, NA, ~
#> $ arr_time  <int> NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, ~
#> $ sched_arr_time <int> 2220, 1815, 2240, 1825, 901, 2359, 33, 1747, 1746, 1~
#> $ arr_delay <dbl> NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, ~
#> $ carrier   <chr> "EV", "EV", "AA", "AA", "B6", "B6", "UA", "EV", "EV"~
#> $ flight    <int> 4204, 4308, 791, 1925, 125, 147, 1299, 4352, 4406, 4~
#> $ tailnum   <chr> "N14168", "N18120", "N3EHAA", "N3EVAA", "N618JB", "N~
#> $ origin    <chr> "EWR", "EWR", "LGA", "LGA", "JFK", "JFK", "EWR", "EW~
#> $ dest      <chr> "OKC", "RDU", "DFW", "MIA", "FLL", "RSW", "RSW", "CV~
#> $ air_time  <dbl> NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, ~
#> $ distance  <dbl> 1325, 416, 1389, 1096, 1069, 1074, 1068, 569, 319, 2~
#> $ hour      <dbl> 19, 16, 19, 15, 6, 20, 21, 15, 16, 13, 14, 13, 15, 1~
#> $ minute    <dbl> 30, 30, 35, 0, 0, 45, 29, 40, 20, 55, 20, 21, 45, 30~
#> $ time_hour <dtm> 2013-01-01 19:00:00, 2013-01-01 16:00:00, 2013-01-0~
```

The flight that departed but never arrived:

DATA 100, Week 3 (A)

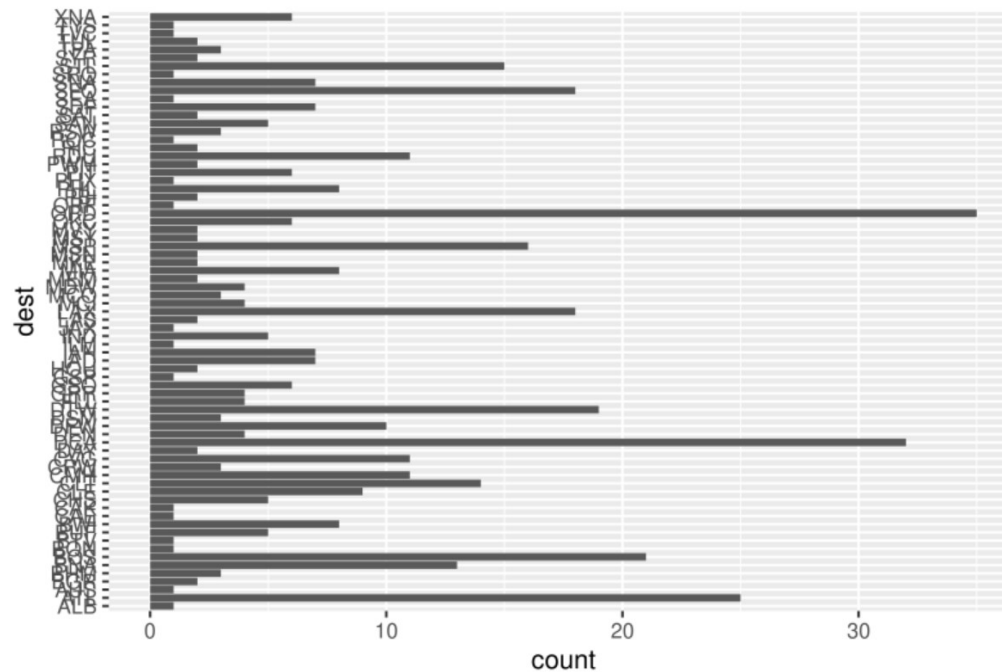
`glimpse(flights |> filter(!is.na(dep_time) & is.na(arr_time)))`

```
Rows: 458
Columns: 19
$ year      <int> 2013, 2013, 2013, 2013, 2013, 2013, 2013, 2013, 2013, 2013, 2013, 2013, 2013, 2...
$ month     <int> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 10, 10, 10, 10, 10, 10, 10, 10, 10...
$ day       <int> 1, 2, 2, 9, 9, 11, 13, 13, 16, 25, 25, 25, 29, 30, 31, 3, 11, 13, 14, 15, 16, 1...
$ dep_time  <int> 2016, 2041, 2145, 615, 2042, 1344, 1907, 2239, 837, 1452, 1457, 2010, 1559, 102...
$ sched_dep_time <int> 1930, 2045, 2129, 615, 2040, 1350, 1634, 2159, 840, 1500, 1505, 1630, 1605, 635...
$ dep_delay <dbl> 46, -4, 16, 0, 2, -6, 153, 40, -3, -8, -8, 220, -6, 230, 4, 119, 7, 1, 1, 32, 2...
$ arr_time  <int> NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA,...
$ sched_arr_time <int> 2220, 2359, 33, 855, 2357, 1518, 1837, 30, 1030, 1619, 1637, 1807, 1925, 934, 1...
$ arr_delay <dbl> NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA,...
$ carrier   <chr> "EV", "B6", "UA", "9E", "B6", "EV", "EV", "EV", "MQ", "US", "9E", "EV", "9E", "E...
$ flight    <int> 4204, 147, 1299, 3856, 677, 4171, 4411, 4519, 4521, 2179, 3393, 4702, 3325, 983...
$ tailnum   <chr> "N14168", "N630JB", "N12221", "N161PQ", "N807JB", "N15985", "N11535", "N17196",...
$ origin    <chr> "EWR", "JFK", "EWR", "JFK", "JFK", "EWR", "EWR", "EWR", "LGA", "LGA", "JFK", "E...
$ dest      <chr> "OKC", "RSW", "RSW", "ATL", "LAX", "MSN", "MEM", "BWI", "RDU", "DCA", "DCA", "G...
$ air_time  <dbl> NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA, NA,...
$ distance  <dbl> 1325, 1074, 1068, 760, 2475, 799, 946, 169, 431, 214, 213, 445, 1391, 1010, 187...
$ hour      <dbl> 19, 20, 21, 6, 20, 13, 16, 21, 8, 15, 15, 16, 16, 6, 14, 11, 8, 8, 8, 6, 20, 21...
$ minute    <dbl> 30, 45, 29, 15, 40, 50, 34, 59, 40, 0, 5, 30, 5, 35, 45, 5, 32, 25, 29, 25, 3, ...
$ time_hour <dtm> 2013-01-01 19:00:00, 2013-01-02 20:00:00, 2013-01-02 21:00:00, 2013-01-09 06:0...
```

`dep_no_arr <- flights |>`

DATA 100, Week 3 (A)

```
filter(!is.na(dep_time), is.na(arr_time)) ggplot(data = dep_no_arr, mapping =  
aes(x = dest)) +  
geom_bar() + coord_flip()
```



DATA 100, Week 3 (A)

filter **basic syntax:**

`filter(DATAFRAME, CONDITIONS separated by comma)`

or using pipe

`DATAFRAME |> filter(CONDITIONS separated by comma)`

use `?filter` to learn more

DATA 100, Week 3 (A)

filter mistakes:

Mix up == and =: one is comparison (==) which gives TRUE or FALSE, the other is assignment (=) which puts a value into a variable

```
flights |>
```

```
filter(month == 6, day == 30)
```

```
#> # A tibble: 918 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     6    30      12           2231        101     352           226
#> 2  2013     6    30      21           2300         81     116             8
#> 3  2013     6    30      23           2055        208     123          2230
#> 4  2013     6    30      25           2359         26     413           350
#> 5  2013     6    30      43           2250        113     150            14
#> 6  2013     6    30      56           2245        131     201             3
#> # i 912 more rows
#> # i 11 more variables: arr_delay <dbl>, carrier <chr>, flight <int>, ...
# v.s.
```

DATA 100, Week 3 (A)

```
flights |>
```

```
filter(month == 6, day = 30)
```

```
#> Error in `filter()`:
```

```
#> ! We detected a named input.
```

```
#> i This usually means that you've used `=` instead of `==`.
```

```
#> i Did you mean `day == 30`?
```

DATA 100, Week 3 (A)

`arrange()`

Get data in order according to the values in some variables (columns)

- Usually in *ascending* order
- Use `desc()` to specify in descending order

Why ordering with respect to a collection of variables?

- Quickly see the range of data directly, combined with `glimpse()` for example
- Have some idea about the extreme cases

DATA 100, Week 3 (A)

```
flights |>  
arrange(dep_delay)
```

```
#> # 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    12     7    2040         2123      -43      40         2352  
#> 2  2013     2     3    2022         2055     -33     2240         2338  
#> 3  2013    11    10    1408         1440     -32     1549         1559  
#> 4  2013     1    11    1900         1930     -30     2233         2243  
#> 5  2013     1    29    1703         1730     -27     1947         1957  
#> 6  2013     8     9     729          755     -26     1002          955  
#> # i 336,770 more rows  
#> # i 11 more variables: arr_delay <dbl>, carrier <chr>, flight <int>, ...
```

DATA 100, Week 3 (A)

```
flights |> arrange(desc(arr_delay))
```

```
#> # 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     9     641           900      1301     1242           1530
#> 2  2013     6    15    1432          1935      1137     1607           2120
#> 3  2013     1    10    1121          1635      1126     1239           1810
#> 4  2013     9    20    1139          1845      1014     1457           2210
#> 5  2013     7    22     845          1600      1005     1044           1815
#> 6  2013     4    10    1100          1900       960     1342           2211
#> # i 336,770 more rows
#> # i 11 more variables: arr_delay <dbl>, carrier <chr>, flight <int>, ...
```

DATA 100, Week 3 (A)

Multiple columns: sorted from left to right

```
arrange(flights, dep_delay, desc(arr_delay))
```

```
arrange(flights, dep_delay, desc(arr_delay))
#> # 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    12     7    2040           2123      -43        40           2352
#> 2  2013     2     3    2022           2055      -33       2240           2338
#> 3  2013    11    10    1408           1440      -32       1549           1559
#> 4  2013     1    11    1900           1930      -30       2233           2243
#> 5  2013     1    29    1703           1730      -27       1947           1957
#> 6  2013     8     9     729            755      -26       1002            955
#> # i 336,770 more rows
#> # i 11 more variables: arr_delay <dbl>, carrier <chr>, flight <int>, ...
```

DATA 100, Week 3 (A)

Where did NA go?

- NA values in dep_delay are all arranged to the end no matter what order we use.
- tail gives the last given number of rows
- head gives the first given number of rows
- check them out by ?tail or ?head

Page 75

DATA 100, Week 3 (A)

The following comparison illustrates some benefits (in terms of readability) of using pipe:

```
tail(arrange(flights, dep_delay, desc(arr_delay)), 30)
```

```
tail(arrange(flights, dep_delay, desc(arr_delay)), 30)
#> # A tibble: 30 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     9    24      NA           1625      NA      NA           1750
#> 2  2013     9    25      NA           1259      NA      NA           1507
#> 3  2013     9    25      NA            845      NA      NA           1018
#> 4  2013     9    25      NA           1755      NA      NA           1932
#> 5  2013     9    25      NA            600      NA      NA            716
#> 6  2013     9    25      NA            836      NA      NA            944
#> # i 24 more rows

#> # i 11 more variables: arr_delay <dbl>, carrier <chr>, flight <int>, ...
# v.s.
```


DATA 100, Week 3 (A)

```
flights |>  
  arrange(dep_delay, desc(arr_delay)) |>  
  tail(30)
```

```
#> # A tibble: 30 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     9    24      NA          1625      NA        NA          1750  
#> 2  2013     9    25      NA          1259      NA        NA          1507  
#> 3  2013     9    25      NA           845      NA        NA          1018  
#> 4  2013     9    25      NA          1755      NA        NA          1932  
#> 5  2013     9    25      NA           600      NA        NA           716  
#> 6  2013     9    25      NA           836      NA        NA           944  
#> # i 24 more rows  
#> # i 11 more variables: arr_delay <dbl>, carrier <chr>, flight <int>, ...
```

DATA 100, Week 3 (A)

Also, formatting the code well makes it easy to try and see:

```
flights |>
  arrange(
    desc(is.na(dep_delay)),
    #dep_delay, # try desc()
    desc(arr_delay)
  )
```

```
#> # 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     NA           1630         NA        NA           1815
#> 2  2013     1     1     NA           1935         NA        NA           2240
#> 3  2013     1     1     NA           1500         NA        NA           1825
#> 4  2013     1     1     NA            600         NA        NA            901
#> 5  2013     1     2     NA           1540         NA        NA           1747
#> 6  2013     1     2     NA           1620         NA        NA           1746
#> # i 336,770 more rows
#> # i 11 more variables: arr_delay <dbl>, carrier <chr>, flight <int>, ...
```

DATA 100, Week 3 (A)

arrange **basic syntax:**

DATAFRAME |> arrange(COLUMNS to be reordered separated by comma)

Use desc(COLUMN) when necessary. ?arrange() for more.

Can arrange by computed values as well (# 3 in 2e, 4.2.5 Exercises).

Use desc(is.na(COLUMN)) appropriately to bring NA up top.

DATA 100, Week 3 (A)

`distinct()`

- ✓ As the name suggests, the output of `distinct()` **contains rows that are distinct**, maybe only in values of a collection of columns that are specified in the parameters.
- ✓ Sometimes **whole rows may be duplicated during manipulation of data** and this functions can be very handy.

DATA 100, Week 3 (A)

remove duplicated rows, if there are any

```
flights |>  
distinct() #|>
```

```
#> # 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  
#> # i 336,770 more rows  
#> # i 11 more variables: arr_delay <dbl>, carrier <chr>, flight <int>, ...  
# View()
```

DATA 100, Week 3 (A)

remove only duplications of values in some columns

flights |>

distinct(dest, sched_dep_time) #|>

```
# remove only duplications of values in some columns
```

```
flights |>
```

```
  distinct(dest, sched_dep_time) #|>
```

```
#> # A tibble: 11,302 x 2
```

```
#>   dest   sched_dep_time
```

```
#>   <chr>         <int>
```

```
#> 1 IAH           515
```

```
#> 2 IAH           529
```

```
#> 3 MIA           540
```

```
#> 4 BQN           545
```

```
#> 5 ATL           600
```

```
#> 6 ORD           558
```

```
#> # i 11,296 more rows
```

```
# View()
```

DATA 100, Week 3 (A)

Notice that when columns are specified, the *default* is to only keep the specified columns in the output.

- To keep all the columns, use `.keep_all = TRUE`. Some internal algorithms are used to determine which rows are included in the result

DATA 100, Week 3 (A)

```
flights |>
distinct(dest, sched_dep_time, .keep_all = TRUE) |>
glimpse()
```

```
#> Rows: 11,302
#> Columns: 19
#> $ year      <int> 2013, 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, 1~
#> $ day       <int> 1, 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, 558, 55~
#> $ sched_dep_time <int> 515, 529, 540, 545, 600, 558, 600, 600, 600, 600, 60~
#> $ dep_delay <dbl> 2, 4, 2, -1, -6, -4, -5, -3, -3, -2, -2, -2, -2, -2, ~
#> $ arr_time  <int> 830, 850, 923, 1004, 812, 740, 913, 709, 838, 753, 8~
#> $ sched_arr_time <int> 819, 830, 850, 1022, 837, 728, 854, 723, 846, 745, 8~
#> $ arr_delay <dbl> 11, 20, 33, -18, -25, 12, 19, -14, -8, 8, -2, -3, 7, ~
#> $ carrier   <chr> "UA", "UA", "AA", "B6", "DL", "UA", "B6", "EV", "B6"~
#> $ flight    <int> 1545, 1714, 1141, 725, 461, 1696, 507, 5708, 79, 301~
#> $ tailnum   <chr> "N14228", "N24211", "N619AA", "N804JB", "N668DN", "N~
#> $ origin    <chr> "EWR", "LGA", "JFK", "JFK", "LGA", "EWR", "EWR", "LG~
#> $ dest      <chr> "IAH", "IAH", "MIA", "BQN", "ATL", "ORD", "FLL", "IA~
#> $ air_time  <dbl> 227, 227, 160, 183, 116, 150, 158, 53, 140, 138, 149~
#> $ distance  <dbl> 1400, 1416, 1089, 1576, 762, 719, 1065, 229, 944, 73~
#> $ hour      <dbl> 5, 5, 5, 5, 6, 5, 6, 6, 6, 6, 6, 6, 6, 6, 5, 6, 6~
#> $ minute    <dbl> 15, 29, 40, 45, 0, 58, 0, 0, 0, 0, 0, 0, 0, 0, 59~
#> $ time_hour <dtm> 2013-01-01 05:00:00, 2013-01-01 05:00:00, 2013-01-0~
```


DATA 100, Week 3 (A)

count()

Related to distinct(), using count() can get the *number* of rows that has the same values in given columns.

count how many flights are scheduled at a given time of the day to a given destination flights |>

count(dest, sched_dep_time) #|>

```
#> # A tibble: 11,302 x 3
#>   dest sched_dep_time     n
#>   <chr>         <int> <int>
#> 1 ABQ             1630     1
#> 2 ABQ             1959     4
#> 3 ABQ             2000    37
#> 4 ABQ             2001   128
#> 5 ABQ             2007    76
#> 6 ABQ             2025     8
#> # i 11,296 more rows
# View()
```

DATA 100, Week 3 (A)

`mutate()`

Analysis means to answer questions, which implies to **create new data from existing ones**

- the average speed in air
- the gain in air time
- the number of flights that were delayed on any given day
- the number of flights that were delayed going to any given destination
- ... etc

DATA 100, Week 3 (A)

Operations are generally *vectorized*, i.e. automatically done for all observations / rows

- the formula **distance/air_time** would compute the **average speed** for all flights
- the formula **arr_delay - dep_delay** would compute the **gain in air time** for all flights

To keep the computed results in the dataframe, use **mutate()**

- with meaningful names for the new variables

DATA 100, Week 3 (A)

```
flights_speed <- flights |>
mutate(
  gain = arr_delay - dep_delay, hours = air_time / 60,
  speed = distance / air_time * 60, # average speed per hour, could use `/ hour`
  gain_per_hour = gain / hours )
```

Take a look at the new dataframe

```
glimpse(flights_speed)
```

```
#> Rows: 336,776
#> Columns: 23
#> $ year      <int> 2013, 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, 1, 1~
#> $ day       <int> 1, 1, 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, 558, 55~
#> $ sched_dep_time <int> 515, 529, 540, 545, 600, 558, 600, 600, 600, 600, 60~
```

DATA 100, Week 3 (A)

```
#> $ 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, 753, 8~
#> $ sched_arr_time <int> 819, 830, 850, 1022, 837, 728, 854, 723, 846, 745, 8~
#> $ arr_delay      <dbl> 11, 20, 33, -18, -25, 12, 19, -14, -8, 8, -2, -3, 7, ~
#> $ carrier        <chr> "UA", "UA", "AA", "B6", "DL", "UA", "B6", "EV", "B6"~
#> $ flight         <int> 1545, 1714, 1141, 725, 461, 1696, 507, 5708, 79, 301~
#> $ tailnum        <chr> "N14228", "N24211", "N619AA", "N804JB", "N668DN", "N~
#> $ origin         <chr> "EWR", "LGA", "JFK", "JFK", "LGA", "EWR", "EWR", "LG~
#> $ dest           <chr> "IAH", "IAH", "MIA", "BQN", "ATL", "ORD", "FLL", "IA~
#> $ air_time       <dbl> 227, 227, 160, 183, 116, 150, 158, 53, 140, 138, 149~
#> $ distance       <dbl> 1400, 1416, 1089, 1576, 762, 719, 1065, 229, 944, 73~
#> $ hour           <dbl> 5, 5, 5, 5, 6, 5, 6, 6, 6, 6, 6, 6, 6, 6, 5, 6, 6~
#> $ minute         <dbl> 15, 29, 40, 45, 0, 58, 0, 0, 0, 0, 0, 0, 0, 0, 0, 59~
#> $ time_hour      <dtm> 2013-01-01 05:00:00, 2013-01-01 05:00:00, 2013-01-0~
#> $ gain           <dbl> 9, 16, 31, -17, -19, 16, 24, -11, -5, 10, 0, -1, 9, ~
#> $ hours          <dbl> 3.7833333, 3.7833333, 2.6666667, 3.0500000, 1.9333333~
#> $ speed          <dbl> 370.0441, 374.2731, 408.3750, 516.7213, 394.1379, 28~
#> $ gain_per_hour  <dbl> 2.3788546, 4.2290749, 11.6250000, -5.5737705, -9.827~
# or
#View(flights_speed)
```

DATA 100, Week 3 (A)

More detailed control of the columns in the new dataframe, using parameters with . in their names.

Again, use ?mutate() for more information.

- Position: like relocate()
 - Default position of the newly created columns are at the (right) end
 - .before add the newly created columns right before (to the left of) a certain column
 - .after add the newly created columns right after (to the right of) a certain column

DATA 100, Week 3 (A)

- Choice: like `select()`
 - `.keep`: can have value
 - ✓ all (default),
 - ✓ used (only the ones listed when calling the function),
 - ✓ unused (remove the ones listed when calling the function) and
 - ✓ none (only the newly created columns)

DATA 100, Week 3 (A)

```
flights |>
mutate(
  gain = arr_delay - dep_delay, hours = air_time / 60,
  speed = distance / air_time * 60, # average speed per hour, could use `/ hour`
  gain_per_hour = gain / hours,
  .before = 4,
  # .after = tailnum,
  # .keep = "used"
) #|>
```


DATA 100, Week 3 (A)

```
#> # A tibble: 336,776 x 23
#>   year month   day gain hours speed gain_per_hour dep_time sched_dep_time
#>   <int> <int> <int> <dbl> <dbl> <dbl>         <dbl>     <int>         <int>
#> 1  2013     1     1     9  3.78  370.         2.38       517           515
#> 2  2013     1     1    16  3.78  374.         4.23       533           529
#> 3  2013     1     1    31  2.67  408.        11.6       542           540
#> 4  2013     1     1   -17  3.05  517.        -5.57       544           545
#> 5  2013     1     1   -19  1.93  394.        -9.83       554           600
#> 6  2013     1     1    16  2.5   288.         6.4       554           558
#> # i 336,770 more rows
#> # i 14 more variables: dep_delay <dbl>, arr_time <int>, ...
# View()
```

DATA 100, Week 3 (A)

`select()`

Too many columns to get *quickly* to the variable that we are interested in

- `select()` operates on the *names* of the variable, or equivalently, the column head
 - Thus can use string functions for comparison and matching

Compare to `filter()` and `arrange()` that operates on the *values* of the variables, i.e. on the *rows*

DATA 100, Week 3 (A)

glimpse(flights)

```
#> Rows: 336,776
#> Columns: 19
#> $ year      <int> 2013, 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, 1, 1~
#> $ day       <int> 1, 1, 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, 558, 55~
#> $ sched_dep_time <int> 515, 529, 540, 545, 600, 558, 600, 600, 600, 600, 60~
#> $ dep_delay <dbl> 2, 4, 2, -1, -6, -4, -5, -3, -3, -2, -2, -2, -2, -2, ~
#> $ arr_time  <int> 830, 850, 923, 1004, 812, 740, 913, 709, 838, 753, 8~
#> $ sched_arr_time <int> 819, 830, 850, 1022, 837, 728, 854, 723, 846, 745, 8~
#> $ arr_delay <dbl> 11, 20, 33, -18, -25, 12, 19, -14, -8, 8, -2, -3, 7, ~
#> $ carrier   <chr> "UA", "UA", "AA", "B6", "DL", "UA", "B6", "EV", "B6"~
#> $ flight    <int> 1545, 1714, 1141, 725, 461, 1696, 507, 5708, 79, 301~
#> $ tailnum   <chr> "N14228", "N24211", "N619AA", "N804JB", "N668DN", "N~
#> $ origin    <chr> "EWR", "LGA", "JFK", "JFK", "LGA", "EWR", "EWR", "LG~
#> $ dest      <chr> "IAH", "IAH", "MIA", "BQN", "ATL", "ORD", "FLL", "IA~
#> $ air_time  <dbl> 227, 227, 160, 183, 116, 150, 158, 53, 140, 138, 149~
#> $ distance  <dbl> 1400, 1416, 1089, 1576, 762, 719, 1065, 229, 944, 73~
#> $ hour      <dbl> 5, 5, 5, 5, 6, 5, 6, 6, 6, 6, 6, 6, 6, 6, 5, 6, 6~
#> $ minute    <dbl> 15, 29, 40, 45, 0, 58, 0, 0, 0, 0, 0, 0, 0, 0, 0, 59~
#> $ time_hour <dtm> 2013-01-01 05:00:00, 2013-01-01 05:00:00, 2013-01-0~
```

DATA 100, Week 3 (A)

select **the fields related to how fast the flight goes**

The **order matters** inside the bracket of select().

The output dataframe will have the columns in the order the way they are specified in select()

- all the **unselected** fields are dropped from the output

COLUMN_NAME: COLUMN_NAME: takes all columns between (and including) the two columns given.

- Just like 1:15 takes all integers between 1 and 15, including 1 and 15

DATA 100, Week 3 (A)

```
flights |>
select(carrier, flight, year:dep_time, origin:distance) |>
glimpse()
```

```
#> Rows: 336,776
#> Columns: 10
#> $ carrier   <chr> "UA", "UA", "AA", "B6", "DL", "UA", "B6", "EV", "B6", "AA"~
#> $ flight    <int> 1545, 1714, 1141, 725, 461, 1696, 507, 5708, 79, 301, 49, ~
#> $ year      <int> 2013, 2013, 2013, 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, 1, 1, 1, 1~
#> $ day       <int> 1, 1, 1, 1, 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, 558, 558, 558~
#> $ origin    <chr> "EWR", "LGA", "JFK", "JFK", "LGA", "EWR", "EWR", "LGA", "J~
#> $ dest      <chr> "IAH", "IAH", "MIA", "BQN", "ATL", "ORD", "FLL", "IAD", "M~
#> $ air_time  <dbl> 227, 227, 160, 183, 116, 150, 158, 53, 140, 138, 149, 158,~
#> $ distance  <dbl> 1400, 1416, 1089, 1576, 762, 719, 1065, 229, 944, 733, 102~
```

DATA 100, Week 3 (A)

Can select by type of values using where(is.___).

```
flights |>
```

```
select( where(is.numeric),
```

```
# where(is.character),
```

```
# where(is.Date)
```

```
)
```

```
#> # A tibble: 336,776 x 14
```

```
#>   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
```

```
#> # i 336,770 more rows
```

```
#> # i 6 more variables: arr_delay <dbl>, flight <int>, air_time <dbl>, ...
```

DATA 100, Week 3 (A)

Removing some columns that won't be used

Use `!c(...)` (or `-c(...)`) to remove columns in the parenthesis, i.e. select *against* them.

- **Cannot** mix select *for* and select *against*
- i.e.: `flights |> select(dep_delay, !c(hour:time_hour))` is **not expected to work**

DATA 100, Week 3 (A)

```
flights |> select(contains("delay"))
```

```
# A tibble: 336,776 × 2
```

	dep_delay	arr_delay
	<dbl>	<dbl>
1	2	11
2	4	20
3	2	33
4	-1	-18
5	-6	-25
6	-4	12
7	-5	19
8	-3	-14
9	-3	-8
10	-2	8

```
# i 336,766 more rows
```


DATA 100, Week 3 (A)

- Other useful string functions, as included in 2e, section 4.3.2
 - `starts_with()`
 - `matches()`: regular expression, which will be discussed later with `stringr`
- and choice functions such as `one_of()`
- `num_range("x", 1:3)` can be useful, but hopefully not in `select()` – bad variable name, IMO.

All from package `tidy-select` ... try `?contains()` in the Console

DATA 100, Week 3 (A)

Reusability, or try not to repeat

Suppose that a decision is made to always keep the columns carrier, flight and tailnum in all the analysis

- while only keep necessary variables for various different tasks

Could do

```
# select a bunch of columns and work with them
```

```
flights |>
```

```
select(carrier, flight, tailnum, origin, dest, air_time)
```

DATA 100, Week 3 (A)

```
#> # A tibble: 336,776 x 6
#>   carrier flight tailnum origin dest  air_time
#>   <chr>      <int> <chr>   <chr> <chr>    <dbl>
#> 1 UA          1545 N14228  EWR   IAH      227
#> 2 UA          1714 N24211  LGA   IAH      227
#> 3 AA          1141 N619AA  JFK   MIA      160
#> 4 B6           725 N804JB  JFK   BQN      183
#> 5 DL           461 N668DN  LGA   ATL      116
#> 6 UA          1696 N39463  EWR   ORD      150
#> # i 336,770 more rows
####
## Many lines later
####
```

DATA 100, Week 3 (A)

flights |>

select(carrier, flight, tailnum, dep_delay, arr_delay, dep_time)

```
#> # A tibble: 336,776 x 6
#>   carrier flight tailnum dep_delay arr_delay dep_time
#>   <chr>     <int> <chr>         <dbl>     <dbl>     <int>
#> 1 UA         1545 N14228             2         11         517
#> 2 UA         1714 N24211             4         20         533
#> 3 AA         1141 N619AA             2         33         542
#> 4 B6          725 N804JB            -1        -18         544
#> 5 DL          461 N668DN            -6        -25         554

#> 6 UA         1696 N39463            -4         12         554
#> # i 336,770 more rows
####
## More lines later
####
```

DATA 100, Week 3 (A)

flights |>

select(carrier, flight, tailnum, year, month, day)

```
#> # A tibble: 336,776 x 6
#>   carrier flight tailnum  year month  day
#>   <chr>      <int> <chr>   <int> <int> <int>
#> 1 UA          1545 N14228   2013     1     1
#> 2 UA          1714 N24211   2013     1     1
#> 3 AA          1141 N619AA   2013     1     1
#> 4 B6           725 N804JB   2013     1     1
#> 5 DL           461 N668DN   2013     1     1
#> 6 UA          1696 N39463   2013     1     1
#> # i 336,770 more rows
####
## Many days pass
####
```

DATA 100, Week 3 (A)

```
must_have <- c("carrier", "flight", "tailnum")
```

```
####
```

```
## whatever happens ####
```

```
flights |>
```

```
select(all_of(must_have), origin, dest, air_time)
```

```
#> # A tibble: 336,776 x 6
```

```
#>   carrier flight tailnum origin dest  air_time
```

```
#>   <chr>    <int> <chr>   <chr> <chr>    <dbl>
```

```
#> 1 UA        1545 N14228 EWR    IAH      227
```

```
#> 2 UA        1714 N24211 LGA    IAH      227
```

```
#> 3 AA        1141 N619AA JFK    MIA      160
```

```
#> 4 B6         725 N804JB JFK    BQN      183
```

```
#> 5 DL         461 N668DN LGA    ATL      116
```

```
#> 6 UA        1696 N39463 EWR    ORD      150
```

```
#> # i 336,770 more rows
```

```
####
```

```
## Many lines later
```

DATA 100, Week 3 (A)

```
flights |>
```

```
select(any_of(must_have), dep_delay, arr_delay, dep_time)
```

```
#> # A tibble: 336,776 x 6
#>   carrier flight tailnum dep_delay arr_delay dep_time
#>   <chr>     <int> <chr>         <dbl>     <dbl>     <int>
#> 1 UA         1545 N14228             2         11         517
#> 2 UA         1714 N24211             4         20         533
#> 3 AA         1141 N619AA             2         33         542
#> 4 B6          725 N804JB            -1        -18         544
#> 5 DL          461 N668DN            -6        -25         554
#> 6 UA         1696 N39463            -4         12         554
#> # i 336,770 more rows
# Check the difference between `all_of` and `any_of` using `?`
####
## More lines later
####
```

DATA 100, Week 3 (A)

`all_of()` requires that all specified variable names exist in the data frame;

if some of the specified variables may not exist, and you want to select those that do exist, you should use `any_of()`

For example

```
must_have1 <- c("carrier", "flight", "tailnum", "meal")
```

```
flights |> select(all_of(must_have1), dep_delay, arr_delay, dep_time)
```

```
Error in `select()`:  
i In argument: `all_of(must_have1)`.  
Caused by error in `all_of()`:  
! Can't subset elements that don't exist.  
✖ Element `meal` doesn't exist.  
Run `rlang::last_trace()` to see where the error occurred.
```


DATA 100, Week 3 (A)

```
flights |> select(any_of(must_have1), dep_delay, arr_delay, dep_time)
```

```
# A tibble: 336,776 × 6
  carrier flight tailnum dep_delay arr_delay dep_time
  <chr>    <int> <chr>      <dbl>    <dbl>    <int>
1 UA        1545 N14228         2        11      517
2 UA        1714 N24211         4        20      533
3 AA        1141 N619AA         2        33      542
4 B6         725 N804JB        -1       -18      544
5 DL         461 N668DN        -6       -25      554
6 UA        1696 N39463        -4        12      554
7 B6         507 N516JB        -5        19      555
8 EV        5708 N829AS        -3       -14      557
9 B6          79 N593JB        -3        -8      557
10 AA        301 N3ALAA        -2         8      558
# i 336,766 more rows
# i Use `print(n = ...)` to see more rows
```

DATA 100, Week 3 (A)

flights |>

select(all_of(must_have), year, month, day)

```
#> # A tibble: 336,776 x 6
#>   carrier flight tailnum  year month  day
#>   <chr>    <int> <chr>   <int> <int> <int>
#> 1 UA        1545 N14228   2013     1     1
#> 2 UA        1714 N24211   2013     1     1
#> 3 AA        1141 N619AA   2013     1     1
#> 4 B6         725 N804JB   2013     1     1
#> 5 DL         461 N668DN   2013     1     1
#> 6 UA        1696 N39463   2013     1     1
#> # i 336,770 more rows
####
## Many days pass
####
```

DATA 100, Week 3 (A)

Keep the changes

- Use <- to assign the revised dataframe to a new variable

```
flights_sml <- flights |>
select( year:day,
ends_with("delay"),
distance, air_time )
```

flights_sml

```
# A tibble: 336,776 × 7
  year month   day dep_delay arr_delay distance air_time
<int> <int> <int>   <dbl>   <dbl>   <dbl>   <dbl>
1  2013     1     1         2        11    1400     227
2  2013     1     1         4        20    1416     227
3  2013     1     1         2        33    1089     160
4  2013     1     1        -1       -18    1576     183
5  2013     1     1        -6       -25     762     116
6  2013     1     1        -4        12     719     150
7  2013     1     1        -5        19    1065     158
8  2013     1     1        -3       -14     229      53
9  2013     1     1        -3        -8     944     140
10 2013     1     1        -2         8     733     138
# i 336,766 more rows
# i Use `print(n = ...)` to see more rows
```

DATA 100, Week 3 (A)

`rename()`

Change column (variable) names.

It is useful for creating meaningful names, or making variable names conform to certain protocol.

- **Very** useful when dealing with data *in the wild*
 - “Say what you mean and mean what you say”
 - “Meaningful name is good name”
- We do not want to guess reading our own code
- We want others (read: **our future selves**) read our code with minimal amount of help

DATA 100, Week 3 (A)

The following changes tailnum into tail_num — the *new* name is on the left side of the = sign.

```
flights |>
rename(tail_num = tailnum) #|>
```

```
# A tibble: 336,776 × 19
  year month   day dep_time sched_dep_time dep_delay arr_time sched_arr_time arr_delay carrier flight tail_num origin
  <int> <int> <int>   <int>         <int>         <dbl>   <int>         <int>         <dbl> <chr>   <int> <chr>   <chr>
1  2013     1     1     517             515           2       830             819           11 UA      1545 N14228 EWR
2  2013     1     1     533             529           4       850             830           20 UA      1714 N24211 LGA
3  2013     1     1     542             540           2       923             850           33 AA      1141 N619AA JFK
4  2013     1     1     544             545          -1      1004            1022          -18 B6       725 N804JB JFK
5  2013     1     1     554             600          -6       812             837          -25 DL       461 N668DN LGA
6  2013     1     1     554             558          -4       740             728           12 UA      1696 N39463 EWR
7  2013     1     1     555             600          -5       913             854           19 B6       507 N516JB EWR
8  2013     1     1     557             600          -3       709             723          -14 EV      5708 N829AS LGA
9  2013     1     1     557             600          -3       838             846           -8 B6        79 N593JB JFK
10 2013     1     1     558             600          -2       753             745            8 AA       301 N3ALAA LGA
# i 336,766 more rows
# i 6 more variables: dest <chr>, air_time <dbl>, distance <dbl>, hour <dbl>, minute <dbl>, time_hour <dtm>
# i Use `print(n = ...)` to see more rows
```

DATA 100, Week 3 (A)

Can also rename *while* select. Compare the difference from above.

```
flights |>  
select(tail_num = tailnum) #|>
```

```
#> # A tibble: 336,776 x 1  
#>   tail_num  
#>   <chr>  
#> 1 N14228  
#> 2 N24211  
#> 3 N619AA  
#> 4 N804JB  
#> 5 N668DN  
#> 6 N39463  
#> # i 336,770 more rows  
# View()
```