# Data Frames and dplyr

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#### Learning Objectives

- Manipulating data frames.
- Calculating summary statistics.
- Using the basic functions of dplyr
- Chapter 5 of RDS

#### Background

- A data frame consists of variables along the columns and observations along the rows.
- For example, in the msleep data frame, the observations are animals and the the variables are properies of those animals (body weight, total sleep time, etc).
- Data frames are the fundamental data type in most analyses.
- Common operations on a data frame during an analysis:
  - Select specific variables (select()).
  - Select observational units by the values of some variables (filter()).
  - Create new variables from old variables (mutate())
  - Reorder the observional units (arrange())
  - Create summary statistics from many observational units (summarize())
  - Group the observational units by the values of some variables (group\_by()).
- As a taste, let's look at an example from the flights data frame from the nycflights13 package:

```
library(nycflights13)
data("flights")
```

- Suppose we want calculate the average departure delay for the flights from carrier in the second half of the year. The steps would be
  - 1. Select only flights from the second half of the year.
  - 2. Group the flights by the carrier.
  - 3. Calculate the average departure delay time within each carrier.
- In base R, this operation would look like:

```
flights2 <- flights[flights$month >= 7, ]
flights3 <- aggregate(dep_time ~ carrier, FUN = mean, data = flights2)
flights3</pre>
```

• In tidyverse, this looks like

```
suppressPackageStartupMessages(library(tidyverse))
```

```
flights %>%
  filter(month >= 7) %>%
  group_by(carrier) %>%
  summarize(mean_dep = mean(dep_time, na.rm = TRUE))
```

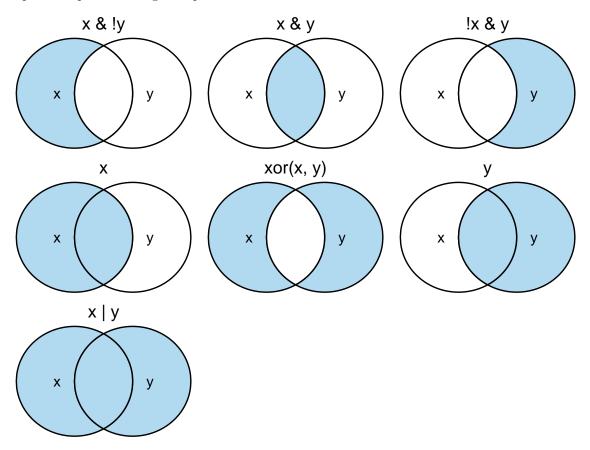
- In the tidyverse:
  - We get to use piping.
  - It's more expressive/clear.

### Filter Rows Based on Variable Values

- In the tidyverse, we use the filter() function to select rows (observations) based on the values of some variables.
- You create <u>logical conditions</u> and the rows that satisfy these <u>logical conditions</u> (return TRUE) are selected.
- Let's extract all flights from new york that occurred in January.

```
flights %>%
  filter(month == 1)
```

- You can filter based on more than two variables using logical operators.
- Graphical Depiction of Logical Operations:



• Let's get all flights that were both in January and from JFK.

```
flights %>%
filter(month == 1 & origin == "JFK")
```

• If you don't know what variable values are possible in a categorical variable, then you can try two things:

- 1. levels() if the variable is a factor.
- 2. unique() otherwise.

```
unique(flights$origin)
```

```
## [1] "EWR" "LGA" "JFK"
```

• Because the and operator is the most used, filter() will also perform the and operation if you separate logical conditions by a comma.

```
flights %>%
filter(month == 1, origin == "JFK")
```

- You should still know the logical operators in case the filtering gets super complicated.
- Let's extract the January LGA flights and the December JFK flights.

```
flights %>%
filter((month == 1 & origin == "LGA") | (month == 12 & origin == "JFK"))
```

- Exercise: Extract all flights that either occur on odd months, or on odd days of even months.
- Exercise (RDS 5.2.4.1) Find all flights that satisfy the following conditions
  - 1. Had an arrival delay of two or more hours
  - 2. Flew to Houston (IAH or HOU)
  - 3. Were operated by United, American, or Delta
  - 4. Departed in summer (July, August, and September)
  - 5. Arrived more than two hours late, but didn't leave late

#### Missing Values

- filter() will exclude observations with missing values.
- If you want to extract those rows as well, you have to ask for them explicitly using is.na().

• You cannot use NA == NA. If two observations are missing, then you don't know if they are equal, so R will return NA to this:

```
NA == NA
## [1] NA
```

#### near()

- Unless you explicitely tell it, R treats all numerics as floats.
- It's thus dangerous to use == for numerics.

• Instead, use the near() function.

```
sqrt(2) ^ 2
  ## [1] 2
  sqrt(2) ^2 == 2
  ## [1] FALSE
  near(sqrt(2) ^ 2, 2)
  ## [1] TRUE
• If a variable is an integer <int>, then it's OK to use ==
  twoint <- as.integer(sqrt(2) ^ 2)</pre>
  twoint == 2
  ## [1] TRUE
```

#### Arrange order of rows

## 3 2 1

• Use arrange() to order the rows by the value of a variable.

```
flights %>%
  arrange(dep_delay)
```

• The default is the arrange in ascending order. To arrange in descending order, use the desc() function.

```
flights %>%
  arrange(desc(dep_delay))
```

```
• If there are ties, then you can break the ties by arranging by another variable.
  dfdat \leftarrow data.frame(x = c(1, 2, 1, 2),
                        y = c(2, 2, 1, 1))
  dfdat
  ##
     х у
  ## 1 1 2
  ## 2 2 2
  ## 3 1 1
  ## 4 2 1
  dfdat %>%
    arrange(x)
      х у
  ## 1 1 2
  ## 2 1 1
  ## 3 2 2
  ## 4 2 1
  dfdat %>%
    arrange(x, y)
  ##
      х у
  ## 1 1 1
  ## 2 1 2
```

```
## 4 2 2
```

• Observations with missing values are always placed at the end (even when using the desc() function)

### Select Specific Columns

- The select() function will extract variables and place them in a smaller data frame.
- Select specific variables

```
flights %>%
  select(dep_delay, arr_delay)
```

• Select a range of variables with ::

```
flights %>%
select(year:day)
```

• Select all variables except certain ones with -:

```
flights %>%
select(-dep_delay, -arr_delay)
```

• Select all variables except within a range of columns.

```
flights %>%
select(-(year:day))
```

- Useful helper functions for select():
  - starts\_with("abc"): matches names that begin with "abc".
  - ends\_with("xyz"): matches names that end with "xyz".
  - contains("ijk"): matches names that contain "ijk".
  - matches("(.)\\1"): selects variables that match a regular expression. This one matches any
    variables that contain repeated characters. You'll learn more about regular expressions in strings.
  - num\_range("x", 1:3): matches x1, x2, and x3.

## Select Specific Rows

#### Create New Variables

# Group Observations by Variables

#### Calculate Summaries