# Exploring data #1

Data types and vector classes

# Data types and vector classes

Here are a few common vector classes in R:

Class	Example
character	"Chemistry", "Physics", "Mathematics"
numeric	10, 20, 30, 40
factor	Male [underlying number: 1], Female [2]
Date	"2010-01-01" [underlying number: 14,610]
logical	TRUE, FALSE [underlying numbers: 1, 0]

### **Numeric vectors**

To explore numeric vectors, there are a few base R functions that are very helpful. For example:

Description
Minimum of values in the vector
Maximum of values in the vector
Mean of values in the vector
Median of values in the vector

3

## Simple statistic examples

## [1] -999

All of these take, as the main argument, the vector(s) for which you want the statistic.

```
mean(x = beijing_pm$value)
## [1] 63.18646
min(x = beijing_pm$value)
```

If there are missing values in the vector, you'll need to add an option to say what to do when them (e.g., na.rm or use="complete.obs"—see help files).

# Simple statistic examples

## [1] 63.18646

These functions require a **numeric vector** as input.

Remember that you can pull a column from a dataframe as a vector using either \$ or the pluck function from purrr. Therefore, you can use either of these calls to get the mean weight of the children in the dataset:

```
mean(beijing_pm$value)
## [1] 63.18646
library("purrr")
## Warning: package 'purrr' was built under R version
## 3.5.2
beijing pm %>%
  pluck("value") %>%
  mean()
```

Within a "tidy" workflow, you can use the summarize function from the dplyr package to create summary statistics for a dataframe. This function inputs a dataframe and outputs a dataframe with the specified summary measures.

The basic format for using summarize is:

As an example, to summarize the beijing\_pm dataset to get the minimum, mean, and maximum  $PM_{2.5}$  concentrations, you could run:

```
summarize(beijing_pm,
    min_pm = min(value),
    mean_pm = mean(value),
    max_pm = max(value))
```

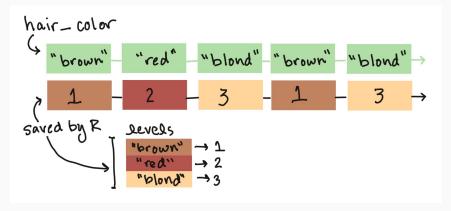
```
## # A tibble: 1 x 3
## min_pm mean_pm max_pm
## <dbl> <dbl> <dbl>
## 1 -999 63.2 684
```

Notice that the output is one row (since the summary was on ungrouped data), with three columns (since we defined three summaries in the summarize function).

Because the first input to the summarize function is a dataframe, you can "pipe into" a summarize call. For example, we could have written the code on the previous slide as:

As another note, because the output from summarize is also a dataframe, we could also "pipe into" another tidyverse function after running summarize.

Factor vectors are used in R for **categorical variables**, where more than one observation can have the same category.



Factor variables have one or more **levels**. While you will always see a factor printed with its factor level labels, R "remembers" the variable with each level assigned a number.

In tibbles, factors will be noted with "fctr" under the column name. For example, look at the aqi column in the beijing\_pm data:

```
head(beijing_pm, n = 3)
```

You can use the levels function to see the levels of a factor vector, as well as the order those levels are recorded in R.

```
levels(beijing_pm$aqi)
```

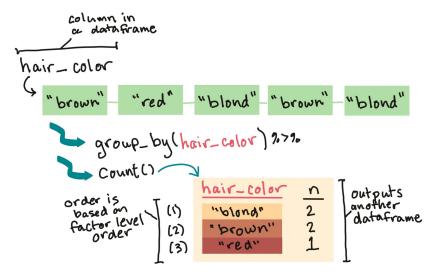
```
## [1] "Good"
## [2] "Moderate"
## [3] "Unhealthy for Sensitive Groups"
## [4] "Unhealthy"
## [5] "Very Unhealthy"
## [6] "Hazardous"
## [7] "Beyond Index"
```

To explore a factor vector, you'll often want to **count** the number of observations in each category. You can do that with two functions in the dplyr package, group\_by and count.

Start with a dataframe that includes the factor variable as a column. First group\_by the factor, then pipe the output of that into the count function.

This will create a new summary dataframe, with a row for each level of the factor. A column called n will give the number of observations in the original data that had that level of the factor.

You can **count** how many observations have each level of a factor.



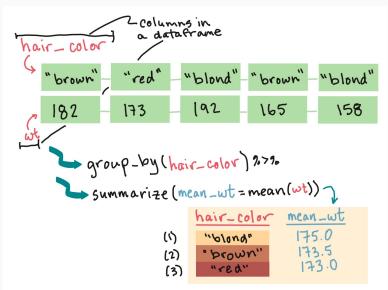
```
beijing_pm %>%
  group_by(aqi) %>%
  count()
## # A tibble: 8 x 2
## # Groups: aqi [8]
##
   aqi
                                         n
## <fct>
                                     <int>
## 1 Good
                                      2438
## 2 Moderate
                                      1021
                                       374
## 3 Unhealthy for Sensitive Groups
## 4 Unhealthy
                                       167
## 5 Very Unhealthy
                                       179
## 6 Hazardous
                                       107
## 7 Beyond Index
                                        27
## 8 <NA>
                                        31
```

You can jointly explore multiple columns in a dataframe.

For example, if one column is a factor and one is numeric, it can be useful to explore values of the numeric column within each level of the factor column.

For the Beijing data, you may want to find out the mean comcentration of  $\mathsf{PM}_{2.5}$  within each AQI level.

You can summarize a numeric column within levels of a factor column:



To do this, pipe the dataframe into group\_by (where you can group by the factor column) and then into summarize, where you can calculate summaries.

```
beijing_pm %>%
  group by(aqi) %>%
  summarize(mean_pm = mean(value))
## # A tibble: 8 x 2
##
    aqi
                                     mean pm
## <fct>
                                        <dbl>
## 1 Good
                                        23.5
## 2 Moderate
                                        70.7
## 3 Unhealthy for Sensitive Groups
                                        122.
   4 Unhealthy
                                        172.
## 5 Very Unhealthy
                                       243.
## 6 Hazardous
                                       378.
## 7 Beyond Index
                                       554.
```

You can create several summaries at once:

```
## # A tibble: 8 x 3
                                  min_pm max_pm
## aqi
## <fct>
                                   <dbl> <dbl>
## 1 Good
                                             50
## 2 Moderate
                                      51
                                            100
## 3 Unhealthy for Sensitive Groups
                                     101
                                            150
                                     151
                                            200
## 4 Unhealthy
## 5 Very Unhealthy
                                     202
                                            300
## 6 Hazardous
                                     301 500
                                     505
                                            684
## 7 Beyond Index
## 8 <NA>
                                    -999
                                            -2
```

As a note, there's a function called n() that you can use inside summarize to replace count. For example, these two expressions give the same output:

```
beijing_pm %>%
  group_by(aqi) %>%
  count()

beijing_pm %>%
  group_by(aqi) %>%
  summarize(n = n())
```

If a column is in a character class, but you'd like it to be a factor, you can use as.factor:

```
beijing_pm %>%
 mutate(qc = as.factor(qc))
## # A tibble: 4,344 x 4
##
     sample_time value qc aqi
##
     <chr> <dbl> <fct> <fct>
   1 1/1/2017 0:00 505 Valid Beyond Index
##
   2 1/1/2017 1:00 485 Valid Hazardous
##
   3 1/1/2017 2:00 466 Valid Hazardous
##
   4 1/1/2017 3:00 435 Valid Hazardous
##
   5 1/1/2017 4:00 405 Valid Hazardous
##
   6 1/1/2017 5:00 402 Valid Hazardous
##
##
   7 1/1/2017 6:00 407 Valid Hazardous
   8 1/1/2017 7:00 435 Valid Hazardous
##
##
   9 1/1/2017 8:00 472 Valid Hazardous
```