

Lecture 6: Augmented vectors, Factor + Labelled Variables

1 Introduction

Reading to do before next class:

GW 15.1 - 15.2 (factors) [this is like 2-3 pages]

[OPTIONAL] GW 15.3 - 15.5 (remainder of “factors” chapter)

[OPTIONAL] GW 20.6 - 20.7 (attributes and augmented vectors)

[OPTIONAL] GW 10 (tibbles)

Explanation about `beamer_header.tex` in YAML header:

We are calling the `beamer_header.tex` file in the background to customize our slides. Without this LaTeX file, our slides would compile according to the default beamer presentation (PDF).

Why would we want to do this?

We can customize our slides with the `beamer_header.tex` LaTeX file to include page numbers, change heading options, or change slide colors (in addition to other things).

`includes` option in the YAML header customizes the beamer presentation slides

Here is a [link](#) to a short description of the `includes` option in the YAML header.

What we will do today

1. Introduction

2. Augmented vectors

- 2.1 Review data types and structures
- 2.2 Attributes and augmented vectors
- 2.3 Object class
- 2.4 Class == factor
- 2.5 Class == labelled
- 2.6 Comparing labelled class to factor class

3. Appendix. Creating factor variables

Libraries we will use today

“Load” the package we will use today (output omitted)

you must run this code chunk after installing these packages

```
library(tidyverse)
library(haven)
library(labelled)
library(lubridate)
```

If package not yet installed, then must install before you load. Install in “console” rather than .Rmd file

Generic syntax: `install.packages("package_name")`

Install “tidyverse”: `install.packages("tidyverse")`

Note: when we load package, name of package is not in quotes; but when we install package, name of package is in quotes:

```
install.packages("tidyverse")
```

```
library(tidyverse)
```

2 Augmented vectors

Data we will use to introduce augmented vectors

```
rm(list = ls()) # remove all objects

#load("../..data/prospect_list/western_washington_college_board_list.RData")
load(url("https://github.com/ozanj/rclass/raw/master/data/prospect_list/wwlist_m
```

2.1 Review data types and structures

Vectors are the primary data structures in R

Two types of vectors:

1. **atomic vectors**
2. **lists**

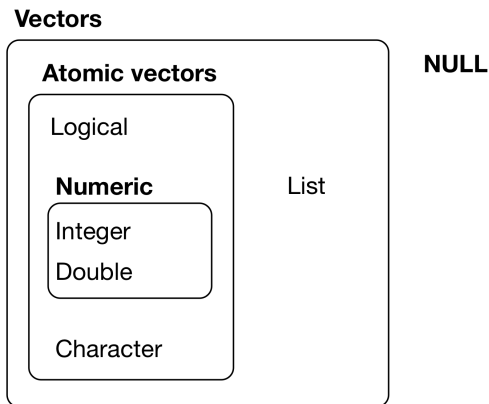


Figure 1: Overview of data structures (Grolemund and Wickham, 2018)

Review data structures: atomic vectors

An **atomic vector** is a collection of values

each value in an atomic vector is an **element**

all elements within vector must have same **data type**

```
(a <- c(1,2,3)) # parentheses () assign and print object in one step
#> [1] 1 2 3
length(a)
#> [1] 3
typeof(a)
#> [1] "double"
str(a)
#>  num [1:3] 1 2 3
```

Can assign **names** to vector elements, creating a **named atomic vector**

```
(b <- c(v1=1,v2=2,v3=3))
#> v1 v2 v3
#> 1 2 3
length(b)
#> [1] 3
typeof(b)
#> [1] "double"
str(b)
#> Named num [1:3] 1 2 3
#> - attr(*, "names")= chr [1:3] "v1" "v2" "v3"
```

Review data structures: lists

Like atomic vectors, **lists** are objects that contain **elements**

However, **data type** can differ across elements within a list

an element of a list can be another list

```
list_a <- list(1,2,"apple")
```

```
typeof(list_a)
```

```
#> [1] "list"
```

```
length(list_a)
```

```
#> [1] 3
```

```
str(list_a)
```

```
#> List of 3
```

```
#> $ : num 1
```

```
#> $ : num 2
```

```
#> $ : chr "apple"
```

```
list_b <- list(1, c("apple", "orange"), list(1, 2))
```

```
length(list_b)
```

```
#> [1] 3
```

```
str(list_b)
```

```
#> List of 3
```

```
#> $ : num 1
```

```
#> $ : chr [1:2] "apple" "orange"
```

```
#> $ :List of 2
```

```
#> ..$ : num 1
```

```
#> ..$ : num 2
```

Review data structures: lists

Like atomic vectors, elements within a list can be named, thereby creating a **named list**

```
# not named
str(list_b)
#> List of 3
#> $ : num 1
#> $ : chr [1:2] "apple" "orange"
#> $ :List of 2
#> ..$ : num 1
#> ..$ : num 2

# named
list_c <- list(v1=1, v2=c("apple", "orange"), v3=list(1, 2, 3))
str(list_c)
#> List of 3
#> $ v1: num 1
#> $ v2: chr [1:2] "apple" "orange"
#> $ v3:List of 3
#> ..$ : num 1
#> ..$ : num 2
#> ..$ : num 3
```

Review data structures: a data frame is a list

A **data frame** is a list with the following characteristics:

All the elements must be **vectors** with the same **length**

Data frames are **augmented lists** because they have additional **attributes** [described later]

```
#a regular list
list_d <- list(col_a = c(1,2,3), col_b = c(4,5,6), col_c = c(7,8,9))
typeof(list_d)
#> [1] "list"
str(list_d)
#> List of 3
#> $ col_a: num [1:3] 1 2 3
#> $ col_b: num [1:3] 4 5 6
#> $ col_c: num [1:3] 7 8 9

#a data frame
df_a <- data.frame(col_a = c(1,2,3), col_b = c(4,5,6), col_c = c(7,8,9))
typeof(df_a)
#> [1] "list"
str(df_a)
#> 'data.frame':    3 obs. of  3 variables:
#> $ col_a: num  1 2 3
#> $ col_b: num  4 5 6
#> $ col_c: num  7 8 9
```

2.2 Attributes and augmented vectors

Atomic vectors versus augmented vectors

Atomic vectors [our focus so far]

I think of atomic vectors as “just the data”

Atomic vectors are the building blocks for augmented vectors

Augmented vectors

Augmented vectors are atomic vectors with additional **attributes** attached

Attributes

Attributes are additional “metadata” that can be attached to any object (e.g., vector or list)

Examples of some important attributes in R:

Names: name the elements of a vector (e.g., variable names)

value labels: character labels (e.g., “Charter School”) attached to numeric values

Object class: How object should be treated by object oriented programming language [discussed below]

Main takaway:

Augmented vectors are atomic vectors (just the data) with additional attributes attached

Attributes in vectors

Identify attributes in any object using the `attributes()` function

```
#vector with no attributes
```

```
vector1 <- c(1,2,3,4)
```

```
vector1
```

```
#> [1] 1 2 3 4
```

```
attributes(vector1)
```

```
#> NULL
```

```
#vector with name attributes
```

```
vector2 <- c(a = 1, b = 2, c = 3, d = 4)
```

```
vector2
```

```
#> a b c d
```

```
#> 1 2 3 4
```

```
attributes(vector2)
```

```
#> $names
```

```
#> [1] "a" "b" "c" "d"
```


Attributes in lists

```
#no attributes
list1 <- list(c(1,2,3), c(4,5,6))
attributes(list1)
#> NULL

#list with attributes
list2 <- list(col_a = c(1,2,3), col_b = c(4,5,6))
str(list2)
#> List of 2
#> $ col_a: num [1:3] 1 2 3
#> $ col_b: num [1:3] 4 5 6
attributes(list2)
#> $names
#> [1] "col_a" "col_b"

#data frame with attributes
list3 <- data.frame(col_a = c(1,2,3), col_b = c(4,5,6))
str(list3)
#> 'data.frame':    3 obs. of  2 variables:
#> $ col_a: num  1 2 3
#> $ col_b: num  4 5 6
attributes(list3)
#> $names
#> [1] "col_a" "col_b"
#>
#> $class
#> [1] "data.frame"
```

2.3 Object class

Object class

Every object in R has a **class**

Object class defines rules for how object can be treated by object oriented programming language (e.g., which functions you can apply to object)

class is an **attribute** of an object

Identify the class of an object using the `class()` function

```
(vector2 <- c(a = 1, b = 2, c = 3, d = 4))
```

```
#> a b c d
```

```
#> 1 2 3 4
```

```
class(vector2)
```

```
#> [1] "numeric"
```

When I encounter a new object I often investigate object by applying

`typeof()`, `class()`, and `attributes()` functions to that object

```
vector2
```

```
#> a b c d
```

```
#> 1 2 3 4
```

```
typeof(vector2)
```

```
#> [1] "double"
```

```
class(vector2)
```

```
#> [1] "numeric"
```

```
attributes(vector2)
```

```
#> $names
```

```
#> [1] "a" "b" "c" "d"
```

Object class

Why is **class** important?

Specific functions usually work with only particular **classes** of objects

e.g., "date" functions usually only work on objects with a date class

"string" functions usually only work with on objects with a character class

Functions that do mathematical computation usually work on objects with a numeric class

Note: functions care about object **class**, not object **type**

object with `numeric` class (output omitted)

```
str(wwlist)
```

```
typeof(wwlist$med_inc_zip)
```

```
class(wwlist$med_inc_zip)
```

```
sum(wwlist$med_inc_zip[1:10], na.rm = TRUE) # numeric function
```

```
# load library with date functions
```

```
library(lubridate)
```

```
#Sys.setenv(TZ="America/Los_Angeles") #setting time zone to Los Angeles time
```

```
year(wwlist$med_inc_zip[1:10]) # date function
```

Object class

Why is **class** important?

Specific functions usually work with only particular **classes** of objects

Note: functions care about object **class**, not object **type**

Object with `character` class

```
str(wwlist$hs_city)
typeof(wwlist$hs_city)
class(wwlist$hs_city)

tolower(wwlist$hs_city[1:10]) # string function
sum(wwlist$hs_city, na.rm = TRUE) # numeric function
```

Object with a date class

```
typeof(wwlist$receive_date)
class(wwlist$receive_date)

year(wwlist$receive_date[1:10]) # date function
sum(wwlist$receive_date) # numeric function
```

Class and object oriented programming

Definition of object oriented programming from this [LINK](#)

“Object-oriented programming (OOP) refers to a type of computer programming in which programmers define not only the data type of a data structure, but also the types of operations (functions) that can be applied to the data structure.”

Object **class** is fundamental to object oriented programming because:

- object class determines which functions can be applied to the object

- object class also determines what those functions do to the object

Many different object classes exist in R

- we can also create our own classes

- but in this course we will work with classes that have been created by others

2.4 Class == factor

Factors

Factors are an object *class* used to display categorical data (e.g., marital status)

A factor is an **augmented vector** built by attaching a “levels” attribute to an (atomic) integer vectors

Usually, we would prefer a categorical variable (e.g., race, school type) to be a factor variable rather than a character variable

So far in the course I have made all categorical variables character variables because we had not introduced factors yet

Below, I'll create a factor version of the character variable `ethn_code`

(don't worry about understanding this code; I'll explain it later)

```
str(wwlist$ethn_code)
#> chr [1:268396] "other-2 or more" "white" "white" "other-2 or more" ...
class(wwlist$ethn_code)
#> [1] "character"
# create factor var; tidyverse approach
wwlist <- wwlist %>% mutate(ethn_code_fac = factor(ethn_code))
#wwlist$ethn_code_fac <- factor(wwlist$ethn_code) # base r approach

str(wwlist$ethn_code_fac)
#> Factor w/ 10 levels "american indian or alaska native",...: 8 10 10 8 10 8 8
```


Factors

A factor is an **augmented vector** built by attaching a “levels” attribute to an (atomic) integer vector

Compare (character) `ethn_code` to (factor) `ethn_code_fac` (output omitted)

```
#character var
typeof(wwlist$ethn_code)
class(wwlist$ethn_code)
str(wwlist$ethn_code)
attributes(wwlist$ethn_code)

#factor var
typeof(wwlist$ethn_code_fac)
class(wwlist$ethn_code_fac)
str(wwlist$ethn_code_fac)
attributes(wwlist$ethn_code_fac)
```

Main takeaway

`ethn_code_fac` has `type=integer` and `class=factor` because the variable has a “levels” attribute

Underlying data are integers but levels attribute is used to display the data.

```
wwlist$ethn_code_fac[1:4] # print first few obs of ethn_code_fac
#> [1] other-2 or more white          white          other-2 or more
#> 10 Levels: american indian or alaska native ...
```

Working with factor variables

```
attributes(wwlist$ethn_code_fac)
```

Refer to categories of a factor by the values of the **level attribute** rather than the underlying values of the variable

Task

count the number of prospects in object `wwlist` who identify as “white”

```
# referring to variable value; this doesn't work
```

```
wwlist %>% filter(ethn_code_fac==10) %>% count
```

```
#> # A tibble: 1 x 1
```

```
#>       n
```

```
#>   <int>
```

```
#> 1     0
```

```
#referring to value of level attribute; this works
```

```
wwlist %>% filter(ethn_code_fac=="white") %>% count
```

```
#> # A tibble: 1 x 1
```

```
#>       n
```

```
#>   <int>
```

```
#> 1 159680
```

Working with factor variables

Task

count the number of prospects in object `wwlist` who identify as “white”

If you want to refer to underlying values, then apply `as.integer()` function to the factor variable

```
attributes(wwlist$ethn_code_fac)
#> $levels
#> [1] "american indian or alaska native"
#> [2] "asian or native hawaiian or other pacific islander"
#> [3] "black or african american"
#> [4] "cuban"
#> [5] "mexican/mexican american"
#> [6] "not reported"
#> [7] "other spanish/hispanic"
#> [8] "other-2 or more"
#> [9] "puerto rican"
#> [10] "white"
#>
#> $class
#> [1] "factor"
wwlist %>% filter(as.integer(ethn_code_fac)==10) %>% count
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1 159680
```

How to identify the variable values associated with factor levels

Let's create a factor version of the character variable `psat_range`

```
wwlist <- wwlist %>% mutate(psat_range_fac = factor(psat_range)) # create factor
```

Run below code in console rather than code chunk to see values associated with each factor

```
wwlist %>% count(psat_range_fac)
#> Warning: Factor `psat_range_fac` contains implicit NA, consider using
#> `forcats::fct_explicit_na`
attributes(wwlist$psat_range_fac)
levels(wwlist$psat_range_fac) #starts at 1
nlevels(wwlist$psat_range_fac) #7 levels total
levels(wwlist$psat_range_fac)[1:3] #prints levels 1-3
```

Once you know values associated with factor, you can filter based on values

```
wwlist %>% filter(as.integer(psat_range_fac)==4) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1  8348
```

Or you can just filter based on value of **factor levels**

```
wwlist %>% filter(psat_range=="1270-1520") %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1  8348
```

Creating factor variables from character variables or from integer variables

See Appendix

Factor student exercise

1. After running the code below, use `typeof`, `class`, `str`, and `attributes` functions to check the new variable `receive_year`
2. Create a factor variable from the input variable `receive_year` and name it `receive_year_fac`
3. Run the same functions (`typeof`, `class`, etc.) from the first question using the new variable you created
4. Get a count of `receive_year_fac`. **hint:** you could also run this in the console to see values associated with each factor

Run this code to create a year variable from the input variable "receive_date"

```
#wwlist %>% glimpse()
```

```
library(lubridate) #load library if you haven't already
```

```
wwlist <- wwlist %>%
```

```
  mutate(receive_year = year(receive_date)) #creating year variable with the lub
```

```
#Check variable
```

```
wwlist %>%
```

```
  count(receive_year)
```

```
wwlist %>%
```

```
  group_by(receive_year) %>%
```

```
  count(receive_date)
```

Factor student exercise solutions

1. Use `typeof`, `class`, `str`, and `attributes` functions to check the new variable `receive_year`

```
typeof(wwlist$receive_year)
#> [1] "double"
class(wwlist$receive_year)
#> [1] "numeric"
str(wwlist$receive_year)
#>  num [1:268396] 2016 2016 2016 2016 2016 2016 ...
attributes(wwlist$receive_year)
#> NULL
```

Factor student exercise solutions

2. Now create a factor variable from the input variable `receive_year` and name it `receive_year_fac`

```
# create factor var; tidyverse approach
```

```
wwlist <- wwlist %>%  
  mutate(receive_year_fac = factor(receive_year))
```


Factor student exercise solutions

3. Run the same functions (`typeof` , `class` , etc.) from the first question using the new variable you created

```
typeof(wwlist$receive_year_fac)
#> [1] "integer"
class(wwlist$receive_year_fac)
#> [1] "factor"
str(wwlist$receive_year_fac)
#> Factor w/ 3 levels "2016","2017",...: 1 1 1 1 1 1 1 1 1 1 ...
attributes(wwlist$receive_year_fac)
#> $levels
#> [1] "2016" "2017" "2018"
#>
#> $class
#> [1] "factor"
```

Factor student exercise solutions

4. Get a count of `receive_year_fac`. **hint:** you could also run this in the console to see values associated with each factor

```
wwlist %>%  
  count(receive_year_fac)  
#> # A tibble: 3 x 2  
#>   receive_year_fac      n  
#>   <fct>          <int>  
#> 1 2016          89637  
#> 2 2017          89816  
#> 3 2018          88943
```

2.5 Class == labelled

Data we will use to introduce labelled class

High school longitudinal surveys from National Center for Education Statistics (NCES)

Follow U.S. students from high school through college, labor market

We will be working with [High School Longitudinal Study of 2009 \(HSL:09\)](#)

Follows 9th graders from 2009

Data collection waves

- Base Year (2009)
- First Follow-up (2012)
- 2013 Update (2013)
- High School Transcripts (2013-2014)
- Second Follow-up (2016)

haven package

`haven`, which is part of **tidyverse**, “enables R to read and write various data formats” from the following statistical packages:

SAS

SPSS

Stata

When using `haven` to read data, resulting R objects have these characteristics:

Are **tibbles**, a particular type of data frame we discuss in future weeks

Transform variables with “value labels” into the `labelled()` class [our focus today]

`labelled` is an object **class** created by folks who created `haven` package

`labelled` is an object class, just like `factor` is an object class

`labelled` and `factor` classes are both viable alternatives for categorical variables

Helpful description of `labelled` class [HERE](#)

Dates and times converted to R date/time classes

Character vectors not converted to factors

haven package

Use `read_dta()` function from `haven` to import Stata dataset into R

```
hsls <- read_dta(file="https://github.com/ozanj/rclass/raw/master/data/hsls/hsls")
```

Let's examine the data [you **must** run this code chunk]

```
names(hsls)
names(hsls) <- tolower(names(hsls)) # convert names to lowercase
names(hsls)

str(hsls) # ugh

str(hsls$s3classes)
attributes(hsls$s3classes)
typeof(hsls$s3classes)
class(hsls$s3classes)
```

labelled package

Purpose of the `labelled` package is to work with data imported from SPSS/Stata/SAS using the `haven` package.

In particular, `labelled` package creates functions to work with objects that have `labelled` class

From package documentation: "purpose of the `labelled` package is to provide functions to manipulate *metadata* as variable labels, value labels and defined missing values using the `labelled` class and the `label` attribute introduced in `haven` package.

More info on the `labelled` package: [LINK](#)

Functions in `labelled` package

[Full list](#)

A couple relevant functions

`val_labels` : get or set variable *value labels*

`var_label` : get or set a *variable label*

```
attributes(hs1s$s3classes)
```

```
hs1s %>% select(s3classes) %>% var_label()
```

```
hs1s %>% select(s3classes) %>% val_labels()
```

Core concepts for understanding labelled class [SKIP]

atomic vectors (and lists) the underlying data

data structures: vector or list

data type: numeric (integer or double); character; logical

```
typeof(hs1s$s3classes)
```

```
#> [1] "double"
```

augmented vectors are atomic vectors with **attributes** attached

attributes are “metadata” attached to an object. Examples

names: names of elements of a vector or list (e.g., variable names)

levels: display output associated with values of a factor variable

class: e.g., factor, labelled

```
attributes(hs1s$s3classes)
```

class is an object oriented programming concept. The `class` of an object determines which functions can be applied to the object and what those functions do

e.g., can't apply `sum()` to an object where `class=character`

What is labelled class?

`labelled` is an object class created by the `haven` package for importing variables from SAS/SPSS/Stata that have **value labels**

value labels [in Stata] are labels attached to specific values of a variable:

e.g., variable value `1` attached to value label "married", `2`="single", `3`="divorced"

Variables in an R data frame with `class==labelled` :

data `type` can be numeric(double) or character

To see `value labels` associated with each value:

```
attr(data_frame_name$variable_name,"labels")
```

e.g., `attr(hs1s$s3classes,"labels")`

Let's investigate the attributes of `hs1s$s3classes`

```
typeof(hs1s$s3classes)
class(hs1s$s3classes)
str(hs1s$s3classes)
attributes(hs1s$s3classes)
```

use `attr(object_name,"attribute_name")` to refer to each attribute

```
attr(hs1s$s3classes,"label")
attr(hs1s$s3classes,"labels")
attr(hs1s$s3classes,"class")
attr(hs1s$s3classes,"format.stata")
```

Working with labelled class data

Show variable labels (`var_label`); and show value labels (`val_labels`)

```
hsls %>% select(s3classes,s3clglvl) %>% var_label #show variable label
hsls %>% select(s3classes,s3clglvl) %>% val_labels #show value labels
```

Create frequency tables with `labelled` class variables using `count()`

Default setting is to show variable **values** not **value labels**

```
hsls %>% count(s3classes)
#investigate the object created
hsls_freq_temp <- hsls %>% count(s3classes)
hsls_freq_temp
rm(hsls_freq_temp)
```

To make frequency table show **value labels** add `%>% as_factor()` to pipe

`as_factor()` is function from `haven` that converts an object to a factor

```
hsls %>% count(s3classes) %>% as_factor()
#investigate the object created
hsls_freq_temp <- hsls %>% count(s3classes) %>% as_factor()
hsls_freq_temp
rm(hsls_freq_temp)
```

Working with labelled class data

To isolate values of labelled class variables in `filter()` function:

refer to variable **value**, not the **value label**

Task

how many observations in var `s3classes` associated with “Unit non-response”

how many observations in var `s3classes` associated with “Yes”

General steps to follow:

1. investigate object
2. use filter to isolate desired observations

Investigate object

```
class(hs1s$s3classes)
hs1s %>% select(s3classes,s3clglv1) %>% var_label #show variable label
hs1s %>% select(s3classes,s3clglv1) %>% val_labels #show value label
hs1s %>% count(s3classes) # freq table, values
hs1s %>% count(s3classes) %>% as_factor() # freq table, value labels
```

filter specific values

```
hs1s %>% filter(s3classes== -8) %>% count() # -8 = unit non-response
hs1s %>% filter(s3classes== 1) %>% count() # 1 = yes
```

Labelled student exercise

1. Get variable and value labels of `s3hs`
2. Get a count of the variable showing the values and the value labels. **hint** use `factor()`
3. Filter if value is associated with “Missing”
4. Filter if value is associated with “Missing” or “Unit non-response”

Labelled student exercise solutions

1. Get variable and value labels of `s3hs`

```
hsls %>%  
  select(s3hs) %>%  
  var_label()  
#> $s3hs  
#> [1] "S3 B01F Attending high school or homeschool as of Nov 1 2013"
```

```
hsls %>%  
  select(s3hs) %>%  
  val_labels()  
#> $s3hs  
#>                                     Missing  
#>                                     -9  
#>                                     Unit non-response  
#>                                     -8  
#>                                     Item legitimate skip/NA  
#>                                     -7  
#>                                     Component not applicable  
#>                                     -6  
#> Item not administered: abbreviated interview  
#>                                     -4  
#>                                     Yes  
#>                                     1  
#>                                     No  
#>                                     2  
#>                                     Don't know
```

Labelled student exercise solutions

2. Get a count of the variable `s3hs` showing the value labels. **hint** use `factor()`

```
hsls %>%  
  count(s3hs)  
#> # A tibble: 6 x 2  
#>       s3hs      n  
#>   <dbl+lbl> <int>  
#> 1 -9 [Missing]      22  
#> 2 -8 [Unit non-response] 4945  
#> 3 -7 [Item legitimate skip/NA] 16770  
#> 4  1 [Yes]      624  
#> 5  2 [No]      985  
#> 6  3 [Don't know] 157
```

```
hsls %>%  
  count(s3hs) %>%  
  as_factor()  
#> # A tibble: 6 x 2  
#>   s3hs      n  
#>   <fct>   <int>  
#> 1 Missing      22  
#> 2 Unit non-response 4945  
#> 3 Item legitimate skip/NA 16770  
#> 4 Yes      624  
#> 5 No      985  
#> 6 Don't know 157
```

Labelled student exercise solutions

3. Filter if value is associated with "Missing"

```
hsls %>%  
  filter(s3hs== -9) %>%  
  count()  
#> # A tibble: 1 x 1  
#>       n  
#>   <int>  
#> 1    22
```

Labelled student exercise solutions

4. Filter if value is associated with “Missing” or “Unit non-response”

```
hsls %>%  
  filter(s3hs== -9 | s3hs== -8) %>%  
  count()  
#> # A tibble: 1 x 1  
#>       n  
#>   <int>  
#> 1  4967
```


2.6 Comparing labelled class to factor class

Comparing `class==labelled` to `class==factor`

	<code>class==labelled</code>	<code>class==factor</code>
data type	numeric or character	integer
name of value label attribute	labels	levels
refer to data using	variable values	levels attribute

Converting `class==labelled` to `class==factor`

The `as_factor()` function from `haven` package converts variables with `class==labelled` to `class==factor`

Can be used for descriptive statistics

```
hsls %>% select(s3classes) %>% count(s3classes) %>% as_factor()
```

Can create object with some or all `labelled` vars converted to `factor`

```
hsls_f <- as_factor(hsls, only_labelled = TRUE)
```

Let's examine this object

```
glimpse(hsls_f)
hsls_f %>% select(s3classes, s3c1g1v1) %>% str()
typeof(hsls_f$s3classes)
class(hsls_f$s3classes)
attributes(hsls_f$s3classes)

hsls_f %>% select(s3classes) %>% var_label()
hsls_f %>% select(s3classes) %>% val_labels()
```

Working with `class==factor` data

Showing values associated with factor levels

```
hsls_f %>% count(s3classes)
#> # A tibble: 5 x 2
#>   s3classes      n
#>   <fct>      <int>
#> 1 Missing      59
#> 2 Unit non-response 4945
#> 3 Yes      13477
#> 4 No       3401
#> 5 Don't know  1621
```

In code, refer `level` attribute not variable value

```
hsls_f %>% filter(s3classes=="Yes") %>% count(s3classes)
#> # A tibble: 1 x 2
#>   s3classes      n
#>   <fct>      <int>
#> 1 Yes      13477
```

3 Appendix. Creating factor variables

Create factors [from string variables]

To create a factor variable from string variable

1. create a character vector containing underlying data
2. create a vector containing valid levels
3. Attach levels to the data using the `factor()` function

```
#underlying data: months my fam is born
x1 <- c("Jan", "Aug", "Apr", "Mar")
#create vector with valid levels
month_levels <- c("Jan", "Feb", "Mar", "Apr", "May", "Jun",
  "Jul", "Aug", "Sep", "Oct", "Nov", "Dec")
#attach levels to data
x2 <- factor(x1, levels = month_levels)
```

Note how attributes differ

```
str(x1)
#> chr [1:4] "Jan" "Aug" "Apr" "Mar"
str(x2)
#> Factor w/ 12 levels "Jan","Feb","Mar",...: 1 8 4 3
```

Sorting differs

```
sort(x1)
#> [1] "Apr" "Aug" "Jan" "Mar"
sort(x2)
#> [1] Jan Mar Apr Aug
#> Levels: Jan Feb Mar Apr May Jun Jul Aug Sep Oct Nov Dec
```

Create factors [from string variables]

Let's create a character version of variable `hs_state` and then turn it into a factor

```
#wwlist %>%  
#  count(hs_state)  
#Subset obs to West Coast states  
wwlist_temp <- wwlist %>%  
  filter(hs_state %in% c("CA", "OR", "WA"))  
  
#Create character version of high school state for West Coast states only  
wwlist_temp$hs_state_char <- as.character(wwlist_temp$hs_state)  
  
#investigate character variable  
str(wwlist_temp$hs_state_char)  
table(wwlist_temp$hs_state_char)  
  
#create new variable that assigns levels  
wwlist_temp$hs_state_fac <- factor(wwlist_temp$hs_state_char, levels = c("CA", "OR", "WA"))  
str(wwlist_temp$hs_state_fac)  
attributes(wwlist_temp$hs_state_fac)  
  
#wwlist_temp %>%  
#  count(hs_state_fac)  
rm(wwlist_temp)
```

Create factors [from string variables]

How the `levels` argument works when underlying data is character

- Matches value of underlying data to value of the level attribute

- Converts underlying data to integer, with level attribute attached

See chapter 15 of Wickham for more on factors (e.g., modifying factor order, modifying factor levels)

Creating factors [from integer vectors]

Factors are just integer vectors with level attributes attached to them. So, to create a factor:

1. create a vector for the underlying data
2. create a vector that has level attributes
3. Attach levels to the data using the `factor()` function

```
a1 <- c(1,1,1,0,1,1,0) #a vector of data
a2 <- c("zero","one") #a vector of labels

#attach labels to values
a3 <- factor(a1, labels = a2)
a3
#> [1] one one one zero one one zero
#> Levels: zero one
str(a3)
#> Factor w/ 2 levels "zero","one": 2 2 2 1 2 2 1
```

Note: By default, `factor()` function attached “zero” to the lowest value of vector `a1` because “zero” was the first element of vector `a2`

Creating factors [from integer vectors]

Let's turn an integer variable into a factor variable in the `wwlist` data frame

Create integer version of `receive_year`

```
#typeof(wwlist_temp$receive_year)
wwlist$receive_year_int <- as.integer(wwlist$receive_year)
str(wwlist$receive_year_int)
#> int [1:268396] 2016 2016 2016 2016 2016 2016 2016 2016 2016 2016 ...
typeof(wwlist$receive_year_int)
#> [1] "integer"
```

Assign levels to values of integer variable

```
wwlist$receive_year_fac <- factor(wwlist$receive_year_int,
  labels=c("Twenty-sixteen", "Twenty-seventeen", "Twenty-eighteen"))
str(wwlist$receive_year_fac)
str(wwlist$receive_year)

#Check variable
wwlist %>%
  count(receive_year_fac)

wwlist %>%
  count(receive_year)
```