# Lecture 06: Variable Types in R

Taylor Arnold

### Data Types

We have already discussed that variables in R have different *data types*, most likely one of the following:

- ▶ int stands for integers.
- ▶ dbl stands for doubles, or real numbers.
- chr stands for character vectors, or strings.

#### Example

## 6

An example with the mammals sleep data:

```
data(msleep)
msleep
## # A tibble: 83 x 11
##
                                                order
                       name genus vore
##
                      <chr> <chr> <chr>
                                                <chr>
## 1
                    Cheetah Acinonyx carni Carnivora
## 2
                  Owl monkey Aotus omni Primates
## 3
             Mountain beaver Aplodontia herbi Rodentia
## 4 Greater short-tailed shrew Blarina omni Soricomorpha
## 5
                                 Bos herbi Artiodactyla
                        Cow
```

sleep total <dbl>, sleep rem <dbl>, sleep cycle <dbl>, ## # awake <dbl>, brainwt <dbl>, bodywt <dbl>

Three-toed sloth Bradypus herbi ## # ... with 77 more rows, and 7 more variables: conservation <chr>

3

#### Example

The names and genus are characters whereas numbers such as awake are doubles. These slides shows how to convert numeric variables into categorical ones and way to manipulate categorical variables. There is generally no straightforward way to convert a categorical variable into a number, so we will not discuss that here.

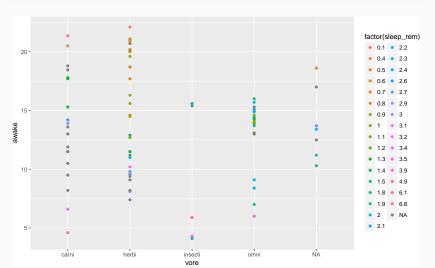
# Numeric to Categorical

# With factor()

To convert every unique values of a numeric variable to a category in a new categorical variable, we simply use the function factor. For example, we can convert sleep\_rem into categories. This changes, amongst other things, the way color is used in a plot:

Notice that missing values became there own category. This is often very useful.

# With factor()



# With cut() and bin()

Often we do not want each numeric variable to be its own category but instead wish to group values that are close into a single category. There are two related functions for doing this: cut and bin. Both take the variable followed by the number of bins to use. The function cut splits the range of the variable into equally spaced buckets where as bin breaks the range up so that each bin has (roughly) the same proportion of data points.

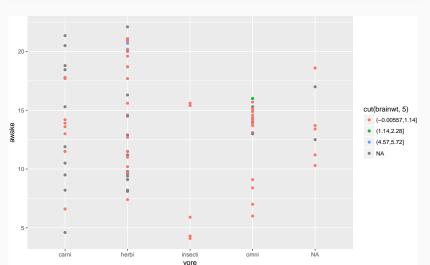
For most variables the difference between these two functions is small; I usually use cut because the interval cutoffs print out nicer. However, when a variable is fairly skewed, I find that bin works better.

# With cut() and bin()

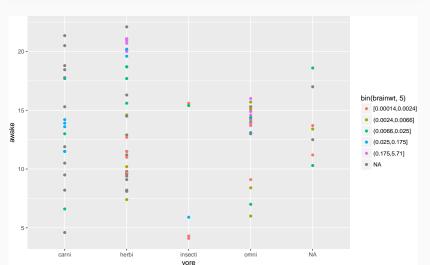
For example, notice how different cutting and binning the brainwt variable is:

This variable is very skewed and the cutting algorithm puts almost all of the mammals in the same category.

## brainwt with cut()



# brainwt with bin()



# **Manipulating Categorical Data**

#### concept

Likewise, it is sometimes useful to convert one categorical variable into another categorical variable by grouping or ordering the categories in a different way. To work with these, we will use the **forcats** package:

#### library(forcats)

By default categories are order alphabetically.

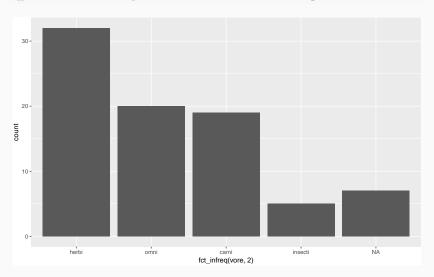
## fct\_infreq

The function fct\_infreq order the categories from largest to smallest.

```
qplot(fct_infreq(vore), data = msleep)
```

### fct\_infreq

### qplot(fct\_infreq(vore, 2), data = msleep)



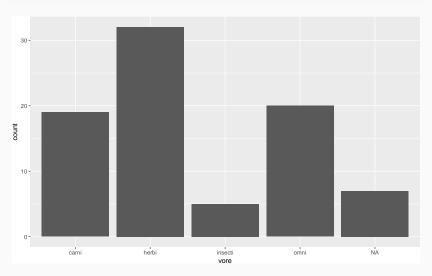
### fct\_lump

The function fct\_lump take a second argument that gives the maximum number of categories allowed. The most frequent categories are included and all other categories are lumped into the other category.

```
qplot(vore, data = msleep)
qplot(fct_lump(vore, 2), data = msleep)
```

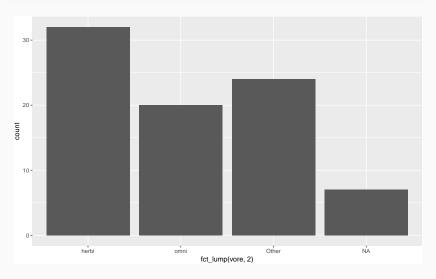
## fct\_lump

## qplot(vore, data = msleep)



### fct\_lump

## qplot(fct\_lump(vore, 2), data = msleep)

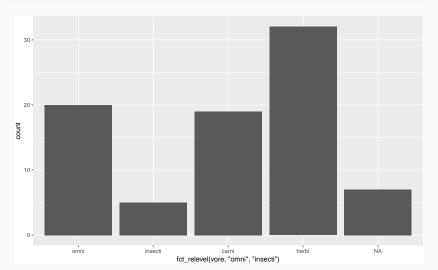


#### fct\_relevel

Finally, the function fct\_relevel takes multiple arguments that gives the categories that should be first, second, and so on.

Unmentioned categories are left in their original order:

#### fct\_relevel

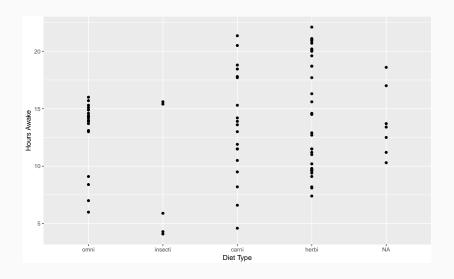


# **Plot Labels**

#### **Axes**

All this messing with the input variables makes the default axis labels become quite messy. To modify these, simply add the functions xlab() and/or ylab() to the plot:

### Axes



#### **Titles**

Similarly, titles are added with ggtitle() and labels for legends with the labs() function:

```
qplot(sleep_rem, awake, color = vore, data = msleep) +
    ggtitle("An Awesome Plot") +
    labs(color = "Diet Type") +
    xlab("Hours REM Sleep") +
    ylab("Hours Awake")
```

#### **Titles**

