

Information about the tidyverse

Write your name here

Contents

1	Code to Create this Document	1
2	R packages for data science	1
3	Some Core Packages	1
3.1	dplyr	1
3.2	ggplot2	2
3.3	readr	3
3.4	tidyr	3

1 Code to Create this Document

*#This is where your code with the render function should appear.
#Remember that you don't want this code chunk to evaluate but
#you do want it to show up in your document.*

2 R packages for data science

The tidyverse is an opinionated collection of R packages designed for data science. All packages share an underlying design philosophy, grammar, and data structures.

Install the complete tidyverse with:

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

3 Some Core Packages

The four *core* packages that we'll use the most are given below along with their purpose and a quick example of some functionality.

3.1 dplyr



[dplyr is a grammar of data manipulation](#), providing a consistent set of verbs that help you solve the most common data manipulation challenges:

- `mutate()` adds new variables that are functions of existing variables
- `select()` picks variables based on their names.
- `filter()` picks cases based on their values.
- `summarise()` reduces multiple values down to a single summary.
- `arrange()` changes the ordering of the rows.

These all combine naturally with `group_by()` which allows you to perform any operation “by group”. You can learn more about them in `vignette(“dplyr”)`. As well as these single-table verbs, dplyr also provides a variety of two-table verbs, which you can learn about in `vignette(“two-table”)`.

If you are new to dplyr, the best place to start is the data transformation chapter in *R for data science*.

```
library(dplyr)

starwars %>%
  filter(species == "Droid")

## # A tibble: 5 x 13
##   name height mass hair_color skin_color eye_color
##   <chr>   <int> <dbl> <chr>      <chr>      <chr>
## 1 C-3P0    167    75 <NA>       gold        yellow
## 2 R2-D2     96    32 <NA>       white, bl~ red
## 3 R5-D4     97    32 <NA>       white, red red
## 4 IG-88    200   140 none       metal       red
## 5 BB8       NA     NA none       none       black
## # ... with 7 more variables: birth_year <dbl>,
## #   gender <chr>, homeworld <chr>, species <chr>,
## #   films <list>, vehicles <list>, starships <list>
```

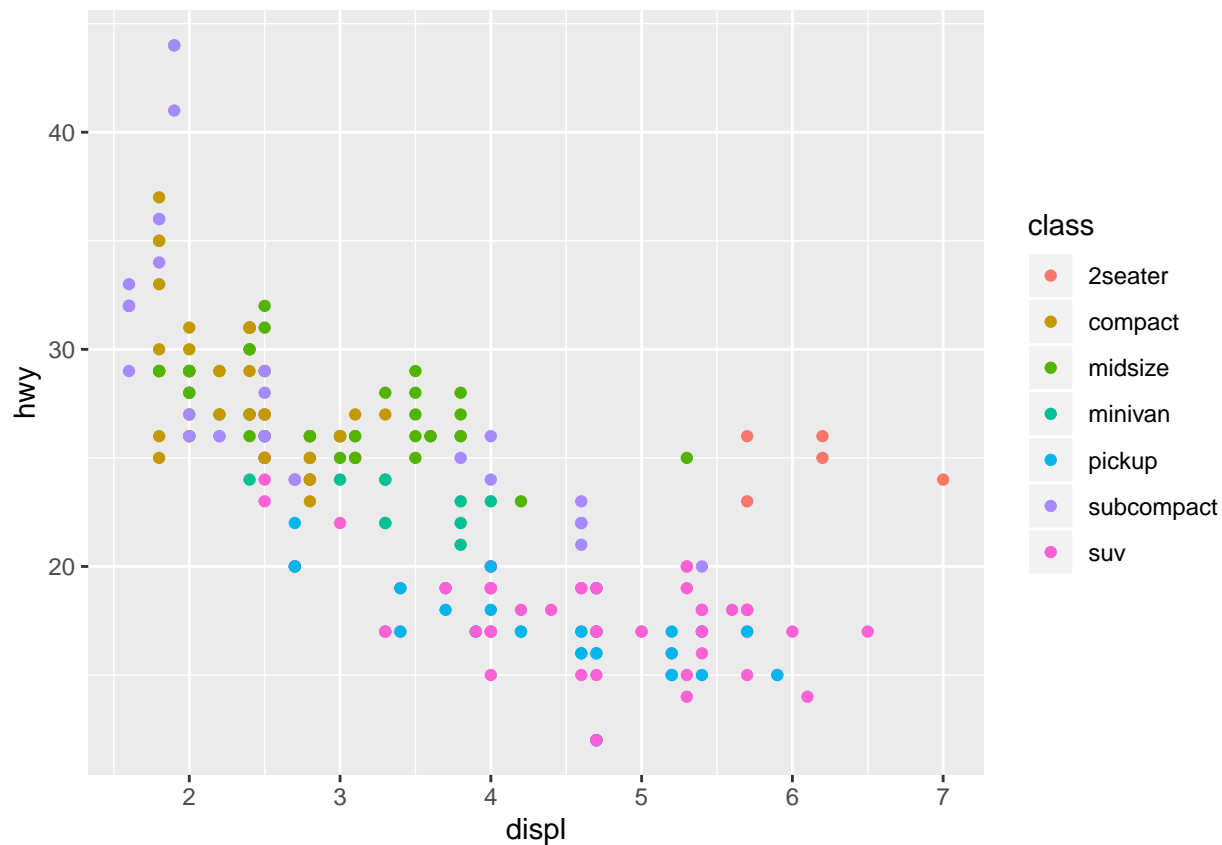
3.2 ggplot2



[ggplot2 is a system for declaratively creating graphics, based on The Grammar of Graphics](#). You provide the data, tell ggplot2 how to map variables to aesthetics, what graphical primitives to use, and it takes care of the details.

```
library(ggplot2)

ggplot(mpg, aes(displ, hwy, colour = class)) +
  geom_point()
```

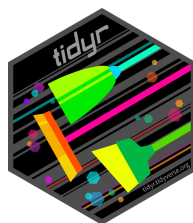


3.3 readr



The goal of [readr](#) is to provide a fast and friendly way to read rectangular data (like csv, tsv, and fwf). It is designed to flexibly parse many types of data found in the wild, while still cleanly failing when data unexpectedly changes. If you are new to readr, the best place to start is the data import chapter in R for data science.

3.4 tidyr



The goal of [tidyr](#) is to help you create tidy data. Tidy data is data where:

1. Every column is variable.

2. Every row is an observation.
3. Every cell is a single value.

Tidy data describes a standard way of storing data that is used wherever possible throughout the tidyverse. If you ensure that your data is tidy, you'll spend less time fighting with the tools and more time working on your analysis. Learn more about tidy data in `vignette("tidy-data")`.

```
library(tidyr)
```

```
relig_income
```

```
## # A tibble: 18 x 11
##   religion `<$10k` ` $10-20k` ` $20-30k` ` $30-40k`
##   <chr>      <dbl>      <dbl>      <dbl>      <dbl>
## 1 Agnostic      27        34        60        81
## 2 Atheist       12        27        37        52
## 3 Buddhist      27        21        30        34
## 4 Catholic    418       617       732       670
## 5 Don't k~      15        14        15        11
## 6 Evangel~    575       869      1064      982
## 7 Hindu         1         9         7         9
## 8 Histori~    228       244       236       238
## 9 Jehovah~     20        27        24        24
## 10 Jewish      19        19        25        25
## 11 Mainlin~    289       495       619       655
## 12 Mormon      29        40        48        51
## 13 Muslim       6         7         9        10
## 14 Orthodox    13        17        23        32
## 15 Other C~     9         7        11        13
## 16 Other F~    20        33        40        46
## 17 Other W~     5         2         3         4
## 18 Unaffil~   217       299       374       365
## # ... with 6 more variables: ` $40-50k` <dbl>,
## #   ` $50-75k` <dbl>, ` $75-100k` <dbl>,
## #   ` $100-150k` <dbl>, ` >150k` <dbl>, `Don't
## #   know/refused` <dbl>
```

```
relig_income %>%
```

```
  pivot_longer(-religion, names_to = "income", values_to = "frequency")
```

```
## # A tibble: 180 x 3
##   religion income      frequency
##   <chr>    <chr>      <dbl>
## 1 Agnostic <$10k          27
## 2 Agnostic $10-20k          34
## 3 Agnostic $20-30k          60
## 4 Agnostic $30-40k          81
## 5 Agnostic $40-50k          76
## 6 Agnostic $50-75k         137
## 7 Agnostic $75-100k        122
## 8 Agnostic $100-150k       109
## 9 Agnostic >150k           84
## 10 Agnostic Don't know/refused 96
## # ... with 170 more rows
```