ODSC: ggplot2 Graph Gallery

Categories and distributions: distributions

by Martin Frigaard

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Resources:



The graphs

- The ggplot2 book by Hadley Wickham, Danielle Navarro, and Thomas Lin Pedersen
- Data Visualization: A Practical Introduction by Kieran Healy (2018)
- R Graphics Cookbook, 2nd edition by Winston Chang (2022)

Graph Categories

- Fundamentals of Data Visualization by Claus O. Wilke (2019)
- Data Visualisation: A Handbook for Data Driven Design by Andy Kirk (2019)
- Data Points by Nathan Yau (2013)

Graph Categories: The 'CHRTS' Families of Chart Types



From "Data Visualisation: A Handbook for Data Driven Design", Andy Kirk (2019)

Comparing categories and distributions

Hierarchies/part-to-whole relationships

Correlations and connections

Trends and intervals over time

Maps, overlays, and/or distortions

Graph Categories: Directory of Visualizations



From "Fundamentals of Data Visualization", Claus O. Wilke (2019)

Amounts

Distributions

Proportions

X-Y relationships

Geospatial Data

Uncertainty

Comprehensive Graph Gallery



Comparing categories and values

- Amounts
- Distributions

Hierarchies and proportions

• Part-to-whole relationships

Trends, correlations and connections

X–Y relationships

Maps, overlays, and distortions

Geospatial Data

Statistical measures

Uncertainty

Data



Data come from the following packages:

- -palmerpenguins
- fivethirtyeight
- -ggplot2movies

Or created using tribble()

variable 1	variable 2
<chr></chr>	<dbl></dbl>
a	1
b	2
С	3
3 rows	

Load data packages



```
library(palmerpenguins)
library(fivethirtyeight)
library(ggplot2movies)
```

palmerpenguins



palmerpenguins package website

palmerpenguins::penguins -> penguins

species	island	bill_length_mm	bill_depth_mm	flipper_length_mm	body_mass_g sex	year
<fct></fct>	<fct></fct>	<dbl></dbl>	<dbl></dbl>	<int></int>	<int> <fct></fct></int>	<int></int>
Adelie	Torgersen	39.1	18.7	181	3750 male	2007
Adelie	Torgersen	39.5	17.4	186	3800 female	2007
Adelie	Torgersen	40.3	18.0	195	3250 female	2007
Adelie	Torgersen	NA	NA	NA	NA NA	2007
Adelie	Torgersen	36.7	19.3	193	3450 female	2007
Adelie	Torgersen	39.3	20.6	190	3650 male	2007
Adelie	Torgersen	38.9	17.8	181	3625 female	2007
Adelie	Torgersen	39.2	19.6	195	4675 male	2007
Adelie	Torgersen	34.1	18.1	193	3475 <i>NA</i>	2007
Adelie	Torgersen	42.0	20.2	190	4250 NA	2007
1-10 of 34	4 rows			Previous	s 1 2 3 4 5 6 3	35 Next

fivethirtyeight



fivethirtyeight package website

All datasets are listed below with descriptions

datasets("fivethirtyeight")

dataset								
<chr></chr>								
US_births_1994_2003								
US_births_2000_2014								
ahca_polls								
airline_safety								
antiquities_act								
august_senate_polls								
avengers								
bachelorette								
bad_drivers								
bechdel								
1-10 of 129 rows 1-1 of 2 columns	Previous	1	2 3	3 4	5	6	13 1	Vext

ggplot2movies



ggplot2movies package website

We're using movies_data (derived version of the ggplot2movies::movies)

movies_data

title	year	length	budget	rating mpaa
<chr></chr>	<int></int>	<int></int>	<int></int>	<dbl> <fct></fct></dbl>
100 Mile Rule	2002	98	1100000	5.6 R
13 Going On 30	2004	98	37000000	6.4 PG-13
15 Minutes	2001	120	42000000	6.1 R
2 Fast 2 Furious	2003	107	76000000	5.1 PG-13
2046	2004	129	12000000	7.6 R
21 Grams	2003	124	20000000	8.0 R
25th Hour	2002	135	15000000	7.8 R
3000 Miles to Graceland	2001	125	62000000	5.4 R
40 Days and 40 Nights	2002	96	17000000	5.4 R
50 First Dates	2004	99	75000000	6.8 PG-13
1-10 of 751 rows 1-6 of 7 columns		Previ	ous 1 2 3	4 5 6 76 Next

Comparing Categories and Distributions



Distributions

Distributions: Histograms



Histograms use bars, but the x axis is divided into 'bins' that cover the range of the variable. The standard number of bins is 30 (but you should experiment to see how many bins fit your variable's distribution). In ggplot2, the geom for creating histograms is geom_histogram()

Distributions: Histograms



penguins

species	island	bill_length_mm	bill_depth_mm	flipper_length_mm	body_mass_g sex	year
<fct></fct>	<fct></fct>	<dbl></dbl>	<dbl></dbl>	<int></int>	<int> <fct></fct></int>	<int></int>
Adelie	Torgersen	39.1	18.7	181	3750 male	2007
Adelie	Torgersen	39.5	17.4	186	3800 female	2007
Adelie	Torgersen	40.3	18.0	195	3250 female	2007
Adelie	Torgersen	NA	NA	NA	NA NA	2007
Adelie	Torgersen	36.7	19.3	193	3450 female	2007
Adelie	Torgersen	39.3	20.6	190	3650 male	2007
Adelie	Torgersen	38.9	17.8	181	3625 female	2007
Adelie	Torgersen	39.2	19.6	195	4675 male	2007
Adelie	Torgersen	34.1	18.1	193	3475 <i>NA</i>	2007
Adelie	Torgersen	42.0	20.2	190	4250 NA	2007
1-10 of 34	4 rows			Previous	s 1 2 3 4 5 6 3	35 Next

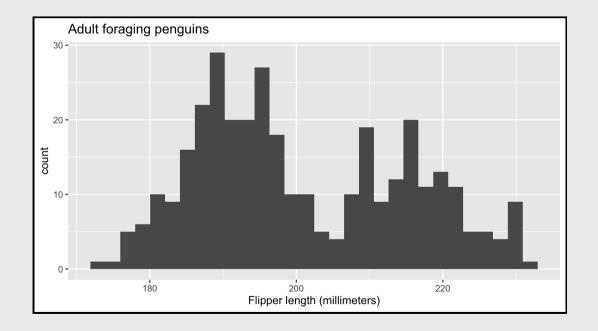
Distributions: Histograms



Map flipper_length_mm to the x axis, add the geom_histogram() layer and the labels

```
labs_histogram <- labs(
  x = "Flipper length (millimeters)",
  title = "Adult foraging penguins")</pre>
```

```
ggplot(data = penguins,
    aes(x = flipper_length_mm)) +
    geom_histogram() +
    labs_histogram
```





Frequency polygons (geom_freqpoly()) are similar to histograms, but use lines instead of bars to represent the variable distribution.



penguins

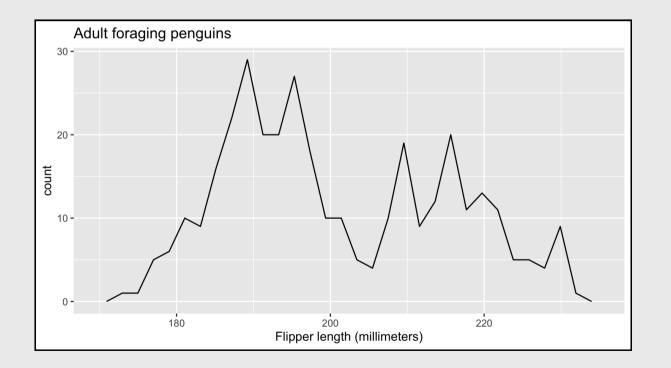
species	island	bill_length_mm	bill_depth_mm	flipper_length_mm	body_mass_g sex	year
<fct></fct>	<fct></fct>	<dbl></dbl>	<dbl></dbl>	<int></int>	<int> <fct></fct></int>	<int></int>
Adelie	Torgersen	39.1	18.7	181	3750 male	2007
Adelie	Torgersen	39.5	17.4	186	3800 female	2007
Adelie	Torgersen	40.3	18.0	195	3250 female	2007
Adelie	Torgersen	NA	NA	NA	NA NA	2007
Adelie	Torgersen	36.7	19.3	193	3450 female	2007
Adelie	Torgersen	39.3	20.6	190	3650 male	2007
Adelie	Torgersen	38.9	17.8	181	3625 female	2007
Adelie	Torgersen	39.2	19.6	195	4675 male	2007
Adelie	Torgersen	34.1	18.1	193	3475 <i>NA</i>	2007
Adelie	Torgersen	42.0	20.2	190	4250 NA	2007
1-10 of 34	4 rows			Previous	s 1 2 3 4 5 6 3	35 Next



Map flipper_length_mm to the x axis, add the geom_freqpoly() layer and the labels

```
labs_freqpoly <- labs(
  x = "Flipper length (millimeters)",
  title = "Adult foraging penguins")</pre>
```

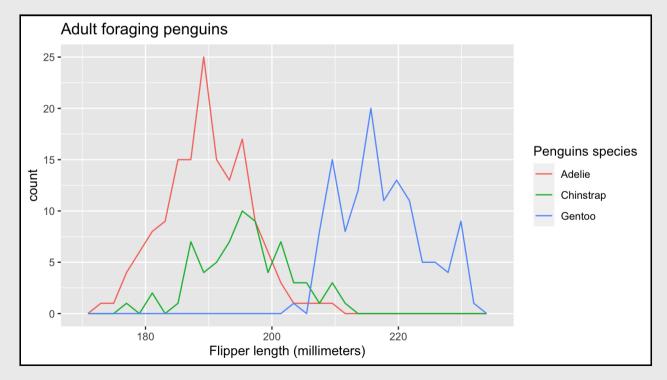
```
ggplot(data = penguins,
    aes(x = flipper_length_mm)) +
    geom_freqpoly() +
    labs_freqpoly
```





Frequency polygons are helpful when we want to look at a continuous variable across the levels of a categorical variable

```
labs_freqpoly_2 <- labs(
  x = "Flipper length (millimeters)",
  color = "Penguins species",
  title = "Adult foraging penguins")</pre>
```





Dot-plots (geom_dotplot()) are similar to histograms and frequency polygons, except instead of using bars or lines, they use dots to represent the values of a given variable.



penguins

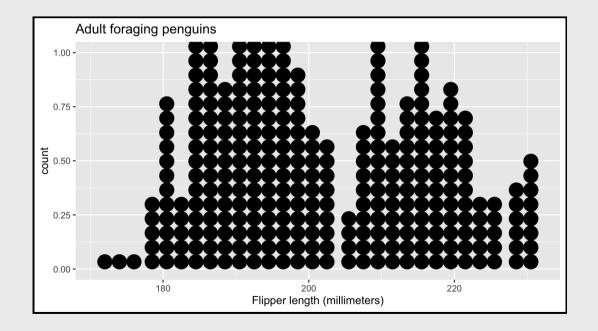
species	island	bill_length_mm	bill_depth_mm	flipper_length_mm	body_mass_g sex	year
<fct></fct>	<fct></fct>	<dbl></dbl>	<dbl></dbl>	<int></int>	<int> <fct></fct></int>	<int></int>
Adelie	Torgersen	39.1	18.7	181	3750 male	2007
Adelie	Torgersen	39.5	17.4	186	3800 female	2007
Adelie	Torgersen	40.3	18.0	195	3250 female	2007
Adelie	Torgersen	NA	NA	NA	NA NA	2007
Adelie	Torgersen	36.7	19.3	193	3450 female	2007
Adelie	Torgersen	39.3	20.6	190	3650 male	2007
Adelie	Torgersen	38.9	17.8	181	3625 female	2007
Adelie	Torgersen	39.2	19.6	195	4675 male	2007
Adelie	Torgersen	34.1	18.1	193	3475 <i>NA</i>	2007
Adelie	Torgersen	42.0	20.2	190	4250 NA	2007
1-10 of 34	4 rows			Previous	s 1 2 3 4 5 6 3	35 Next



Map flipper_length_mm to the x axis, add the geom_dotplot() layer and the labels

```
labs_dotplot <- labs(
  x = "Flipper length (millimeters)",
  title = "Adult foraging penguins")</pre>
```

```
ggplot(data = penguins,
   aes(x = flipper_length_mm)) +
   geom_dotplot() +
   labs_dotplot
```





penguins_histodot <- filter(penguins, !is.na(sex))</pre>

species	island	bill_length_mm	bill_depth_mm	flipper_length_mm	body_mass_g sex	year
<fct></fct>	<fct></fct>	<dbl></dbl>	<dbl></dbl>	<int></int>	<int> <fct></fct></int>	<int></int>
Adelie	Torgersen	39.1	18.7	181	3750 male	2007
Adelie	Torgersen	39.5	17.4	186	3800 female	2007
Adelie	Torgersen	40.3	18.0	195	3250 female	2007
Adelie	Torgersen	36.7	19.3	193	3450 female	2007
Adelie	Torgersen	39.3	20.6	190	3650 male	2007
Adelie	Torgersen	38.9	17.8	181	3625 female	2007
Adelie	Torgersen	39.2	19.6	195	4675 male	2007
Adelie	Torgersen	41.1	17.6	182	3200 female	2007
Adelie	Torgersen	38.6	21.2	191	3800 male	2007
Adelie	Torgersen	34.6	21.1	198	4400 male	2007
1-10 of 33	3 rows			Previous	s 1 2 3 4 5 6 3	34 Next

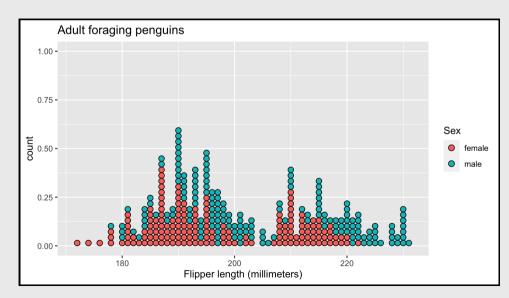


We can also use dot-plots to look at the range of a continuous (numerical) variable across the levels of a categorical (character) variable (like sex below).

The default setting for the size of the dots is '1/30 of the range of the data.' We can adjust the size with binwidth (and method = "histodot")

```
labs_histodot <- labs(
  x = "Flipper length (millimeters)",
  fill = "Sex",
  title = "Adult foraging penguins")</pre>
```

```
ggplot(data = penguins_histodot,
   aes(x = flipper_length_mm,
      fill = factor(sex))) +
geom_dotplot(
   stackgroups = TRUE,
   binwidth = 1,
   method = "histodot") +
labs_histodot
```



Distributions: Bee-swarm Plots



We can also use dots to show the spread of values for a particular variable with beeswarm plots. These display the distribution of numeric values across the levels of a categorical variable.

Distributions: Bee-swarm Plots



penguins

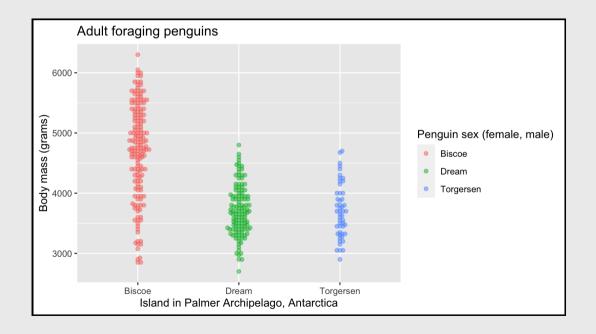
title	year	length	budget	rating mpaa	
<chr></chr>	<int></int>	<int></int>	<int></int>	<dbl> <fct></fct></dbl>	
100 Mile Rule	2002	98	1100000	5.6 R	
13 Going On 30	2004	98	37000000	6.4 PG-13	
15 Minutes	2001	120	42000000	6.1 R	
2 Fast 2 Furious	2003	107	76000000	5.1 PG-13	
2046	2004	129	12000000	7.6 R	
21 Grams	2003	124	20000000	8.0 R	
25th Hour	2002	135	15000000	7.8 R	
3000 Miles to Graceland	2001	125	62000000	5.4 R	
40 Days and 40 Nights	2002	96	17000000	5.4 R	
50 First Dates	2004	99	75000000	6.8 PG-13	
1-10 of 751 rows 1-6 of 7 columns		Previ	ous 1 2 3	4 5 6 76 Nex	it

Distributions: Bee-swarm Plots



Map island to the x axis and color, $body_mass_g$ to the y axis, the $geom_beeswarm()$ layer (with alpha), and the labels

```
labs_beeswarm <- labs(
  x = "Island in Palmer Archipelago, Antarctica",
  y = "Body mass (grams)",
  color = "Penguin sex (female, male)",
  title = "Adult foraging penguins")</pre>
```





Density plots are similar to frequency polygons and histograms, except the line has been 'smoothed.' Instead of dividing the x axis into discrete quantitative 'bins' to create groups for the variable values, density plots transform the distribution according to a 'bandwidth' parameter.



penguins

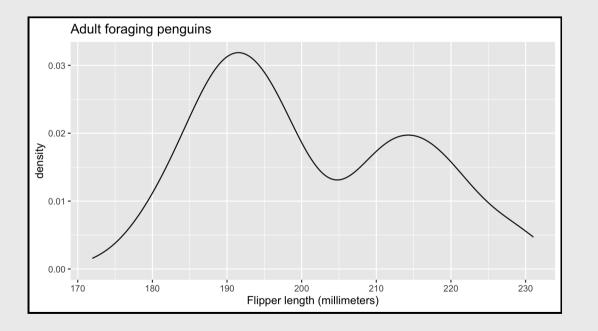
species	island	bill_length_mm	bill_depth_mm	flipper_length_mm	body_mass_g sex	year
<fct></fct>	<fct></fct>	<dbl></dbl>	<dbl></dbl>	<int></int>	<int> <fct></fct></int>	<int></int>
Adelie	Torgersen	39.1	18.7	181	3750 male	2007
Adelie	Torgersen	39.5	17.4	186	3800 female	2007
Adelie	Torgersen	40.3	18.0	195	3250 female	2007
Adelie	Torgersen	NA	NA	NA	NA NA	2007
Adelie	Torgersen	36.7	19.3	193	3450 female	2007
Adelie	Torgersen	39.3	20.6	190	3650 male	2007
Adelie	Torgersen	38.9	17.8	181	3625 female	2007
Adelie	Torgersen	39.2	19.6	195	4675 male	2007
Adelie	Torgersen	34.1	18.1	193	3475 <i>NA</i>	2007
Adelie	Torgersen	42.0	20.2	190	4250 NA	2007
1-10 of 34	4 rows			Previous	s 1 2 3 4 5 6 3	35 Next



Map flipper_length_mm to the x axis, add the geom_density() layer and the labels

```
labs_density <- labs(
  x = "Flipper length (millimeters)",
  title = "Adult foraging penguins")</pre>
```

```
ggplot(data = penguins,
    aes(x = flipper_length_mm)) +
    geom_density() +
    labs_density
```

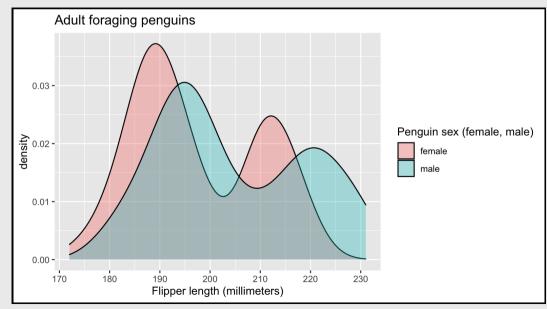




Similar to frequency polygons, geom_density() is useful when we want to look at the distribution of a continuous variable across the levels of a categorical variable

We can set the fill argument to a categorical variable, and use the alpha to handle the overlapping areas.

```
labs_density_alpha <- labs(
  x = "Flipper length (millimeters)",
  fill = "Penguin sex (female, male)",
  title = "Adult foraging penguins")</pre>
```



Distributions: Ridgeline Plots



If we want to plot density curves but retain the interpretability of the axes, consider comparing multiple distributions using ridgeline plots (geom_density_ridges())

Distributions: Ridgeline Plots



penguins

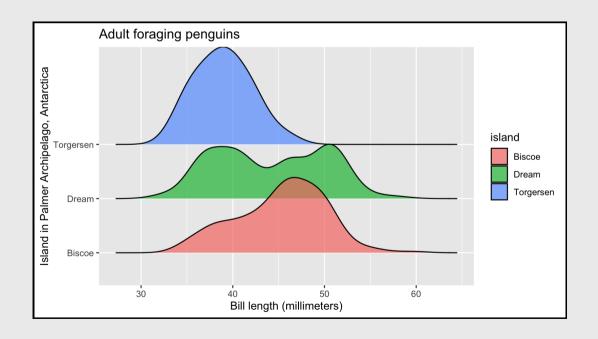
species	island	bill_length_mm	bill_depth_mm	flipper_length_mm	body_mass_g sex	year
<fct></fct>	<fct></fct>	<dbl></dbl>	<dbl></dbl>	<int></int>	<int> <fct></fct></int>	<int></int>
Adelie	Torgersen	39.1	18.7	181	3750 male	2007
Adelie	Torgersen	39.5	17.4	186	3800 female	2007
Adelie	Torgersen	40.3	18.0	195	3250 female	2007
Adelie	Torgersen	NA	NA	NA	NA NA	2007
Adelie	Torgersen	36.7	19.3	193	3450 female	2007
Adelie	Torgersen	39.3	20.6	190	3650 male	2007
Adelie	Torgersen	38.9	17.8	181	3625 female	2007
Adelie	Torgersen	39.2	19.6	195	4675 male	2007
Adelie	Torgersen	34.1	18.1	193	3475 NA	2007
Adelie	Torgersen	42.0	20.2	190	4250 NA	2007
1-10 of 34	4 rows			Previous	s 1 2 3 4 5 6 3	35 Next

Distributions: Ridgeline Plots



Map bill_length_mm to the x axis, island to the y axis and fill, the geom_density_ridges() layer (with alpha) and the labels

```
labs_density_ridges <- labs(
  x = "Bill length (millimeters)",
  y = "Island in Palmer Archipelago, Antarctica",
  title = "Adult foraging penguins")</pre>
```





Box-plots (sometimes called box-and-whisker plots) are great because they display a collection of statistics in a single graph. We're going to build a box-plot of a single numeric variable and review it's contents.



movies_data

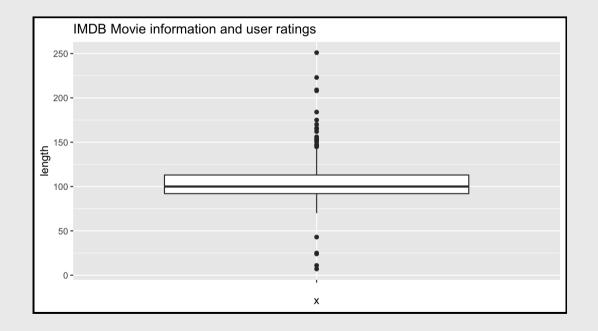
title	year	length	budget	rating mpaa
<chr></chr>	<int></int>	<int></int>	<int></int>	<dbl> <fct></fct></dbl>
100 Mile Rule	2002	98	1100000	5.6 R
13 Going On 30	2004	98	37000000	6.4 PG-13
15 Minutes	2001	120	42000000	6.1 R
2 Fast 2 Furious	2003	107	76000000	5.1 PG-13
2046	2004	129	12000000	7.6 R
21 Grams	2003	124	20000000	8.0 R
25th Hour	2002	135	15000000	7.8 R
3000 Miles to Graceland	2001	125	62000000	5.4 R
40 Days and 40 Nights	2002	96	17000000	5.4 R
50 First Dates	2004	99	75000000	6.8 PG-13
1-10 of 751 rows 1-6 of 7 columns		Previ	ous 1 2 3	4 5 6 76 Next



Map a blank character string (" ") to the x axis, length to the y axis, the geom_boxplot() layer, and the labels

```
labs_boxplot <- labs(
   y = "length",
   title = "IMDB Movie information and user ratings")</pre>
```

```
ggplot(data = movies_data,
    # place an empty string in the
    # x axis
        aes(x = " ",
    # place the length on the y
        y = length)) +
    geom_boxplot() +
    labs_boxplot
```



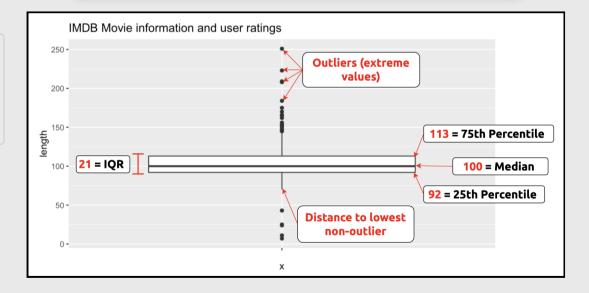


The table below shows the 25th percentile, the median, the 75th percentile, the IQR, and a histogram of the length column from the movies_data dataset.

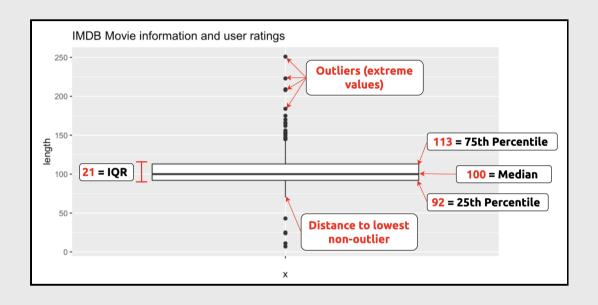
25th	Median	75th	IQR	Histogram
<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dpl></dpl>	<chr></chr>
92	100	113	21	
1 row				

These three horizontal lines give us a picture of the 'spread' of the data. If there is equal distance on either side of the middle (Median) line, this tells us the distribution is symmetrical.

Use these numbers to help you interpret the structure of the box-plot.







As we can see, the box-plot combines multiple summary statistics.

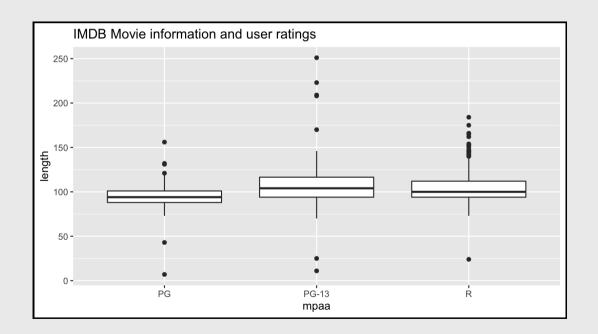
The 25th percentile (first quartile), the median (50th percentile or second quartile), and the 75th percentile (third quartile) values are common to all box-plots.

In ggplot2, values that fall more than 1.5 times the IQR are displayed as individual points (aka outliers). The lines extending from the bottom and top of the main box represent the last non-outlier value in the distribution.



Because box-plot provides so many helpful statistical measures, they are also helpful for viewing how a continuous variable varies across a categorical variable

```
labs_boxplots <- labs(
    x = "mpaa",
    y = "length",
    title = "IMDB Movie information and user ratings")</pre>
```





Compare the four graphs of length from movie_data below to the box-plot:

