- This chapter discusses an interface for data visualization based on the *Grammar of Graphics* that was proposed in the 1990s by Leland Wilkinson (worked for SPSS and Tableau Software).
- A renowned member of the R development team, Hadley Wickham, built the R package ggplot2, which is based on Wilkinson's philosophy.
- Wickham also developed the tidyverse, which is a collection of R packages including the important packages ggplot2 and dplyr.

- There are other traditional ways to create graphs in R, using base R and the lattice package, with are often fine for basic plots and graphs.
- But ggplot2 uses certain terms and tools that are related to the taxonomy of graphics discussed in Chapter 2.
- It also allows you to build graphs incrementally in a sensible way by adding bits of code.

- A *aesthetic* connects a variable to a visual cue.
- A glyph (also called a mark or a symbol) is a plotting symbol that represents one observation.
- A graph can have several aesthetics that can each correspond to different visual cues.
- The foundation for the graphic is a data table:
- Each row in the data table is an observation (also called a case), and each case is represented by a glyph (mark) on the graph.

Adding Aesthetics and Changing Scales

- In ggplot2, we can create (and assign a name to) a basic plot object with the ggplot function.
- Then we can incrementally add aesthetics to the basic plot with other functions, such as geom_point and geom_text.
- See Figures 3.1, 3.2, 3.3, 3.4 as we gradually add aesthetics (x-y position, color, area).
- The coord_trans directive can change the scale of the axes of a graph, such as from linear to logarithmic.
- The scale-type functions (such as scale_x_continuous, scale_x_discrete, scale_x_log10, etc.) will do similar operations.
- scale_color-type functions can alter the scale of the color palette for the color aesthetic.
- Guides such as legends or axis labels provide written context for visual cues to aid interpretability.

Facets and Layers

- Facets are multiple graphs (side-by-side or in a grid) showing the same type of plot for several levels of a categorical variable.
- The R function facet_wrap creates side-by-side plots based on a single categorical variable; facet_grid arranges the plots in a grid.
- See Figure 3.7 and an extension of this graphic.
- Using *layers*, you can place data from two separate data tables onto the same graph (one on top of the other).
- Typically the information from the two data tables would be plotted with different glyphs — you should take care to choose the glyphs wisely to preserve readability.
- See Figures 3.8 and 3.9 for examples.

Standard Graphical Displays for Univariate Numerical Data

- To display the distribution of one numerical variable, the histogram or the density plot are popular choices.
- In ggplot2, the geom_histogram function produces a histogram.
- The binwidth argument guides the number of bins in the histogram, which is a user-specified tuning parameter.
- The bins argument may alternatively be used to directly specify the number of bins.
- Often between 5 to 20 bins is sensible; typically larger data sets should have a larger number of bins.
- Some people recommend the number of bins should be around the square root of the number of observations.

- The geom_density function gives a kernel density estimate, which is a smooth estimate of the underlying density function of a random variable.
- The adjust argument changes the *bandwidth* of the kernel

 a larger bandwidth produces a smoother density plot and a
 smaller bandwidth a more wiggly density plot.
- See Figures 3.10 and 3.11 and the related code.

Standard Graphical Displays for Univariate Categorical Data

- To display the distribution of one categorical variable, the bar graph or the pie chart are popular choices – the bar graph usually considered preferable (Length easier to see than Area).
- The geom_col and geom_bar functions are useful for creating bar graphs.
- See Figures 3.12 and 3.13 and related code.

Standard Graphical Displays for Multivariate Data

- When displaying data on multiple variables, we are often interested in the relationships between variables.
- The scatterplot is a classical graph to see the association between two numerical variables.
- In ggplot2, the geom_point function produces a scatterplot.
- The geom_smooth function can add a trend line or a smooth trend curve on top of the scatterplot.
- We can add more information on the plot by faceting, layering, and/or using Color and Glyph aesthetics to distinguish cases by category.
- The facet_wrap function helps create a symbolic scatterplot showing levels of a categorical variable.
- Sometimes reordering levels of a categorical variable can produce a better display: The functions reorder and fct_relevel can be helpful.

- A time series plot is a way to show how a numerical variable changes over time, with the variable on the y-axis and time on the x-axis.
- Smooth trend curves are often added to time series plots with geom_smooth.
- A single boxplot is a simple way to show the distribution of a numerical variable (not as detailed as a histogram or density plot).
- To display one numerical variable against one categorical variable, side-by-side boxplots are common.
- The geom_boxplot function is helpful.

- A mosiac plot can show associations between two (or three) categorical variables.
- The areas in cells of a diagram correspond the relative frequencies of combinations of categories.
- The geom_mosaic function in the ggmosaic package produces mosaic plots.
- See Figure 3.19 for an example.

- For geographic spatial data, a choropleth map displays the magnitude of a variable in various regions of the map.
- A network diagram shows connections between entities, which may be cases or categories.
- Sections 3.2.3 and 3.2.4 give examples of such maps.

- Section 3.3 has an extended example about historical baby names in the U.S.
- Let's look at some example code and some variations on that example.