

# Tutorial-04

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## Visualisation in R

### Overview

Today you will be learning how to create visualisations in R. This will involve adding layers to a plot using the grammar of graphics.

### Learning Objectives

- Develop the skills to create plots in R
- Develop your understanding of the grammar of graphics
- Learn about different plot geometries
- Learn about different ways to encode colour
- Practice changing the plot appearance (e.g. theme)

## Preparation

- Ensure you installed R and RStudio
- Ensure you have installed the `ggplot2` package. You can install this package directly, but this package is also part of the `tidyverse` package.
- Set yourself an R project for this tutorial.

**If you are confused about any of the above, let us know! It's important you don't get left behind here and we want to help you.** You may also like to refer back to the Lecture 1 and Tutorial 1 material.

## Tasks

### Exercise 1

Today you will be creating visualisations in `ggplot2` to analyse sporting statistics from the Boston Celtics NBA basketball team.

1. Download the dataset from Moodle, `boston_celtics.csv`, and place it in a folder called `data` within the R Project you set up. Then read your data set into R.

We recommend installing the package [here](#) to help keep things organised when referencing files.

```
if(!require(here)){  
  install.packages("here")  
}  
library(here)
```

```
library(tidyverse)  
boston_celtics <- read_csv(here("data", "boston_celtics.csv"))
```

We also recommend following the Reading Data Checklist if you have any problems (see lecture).

```
# Check your working directory  
getwd()  
  
# Check your data file is where you think it is  
file.exists(here("data", "boston_celtics.csv"))
```

```
# Read in your data
library(tidyverse)
boston_celtics <- read_csv(here("data", "boston_celtics.csv"))

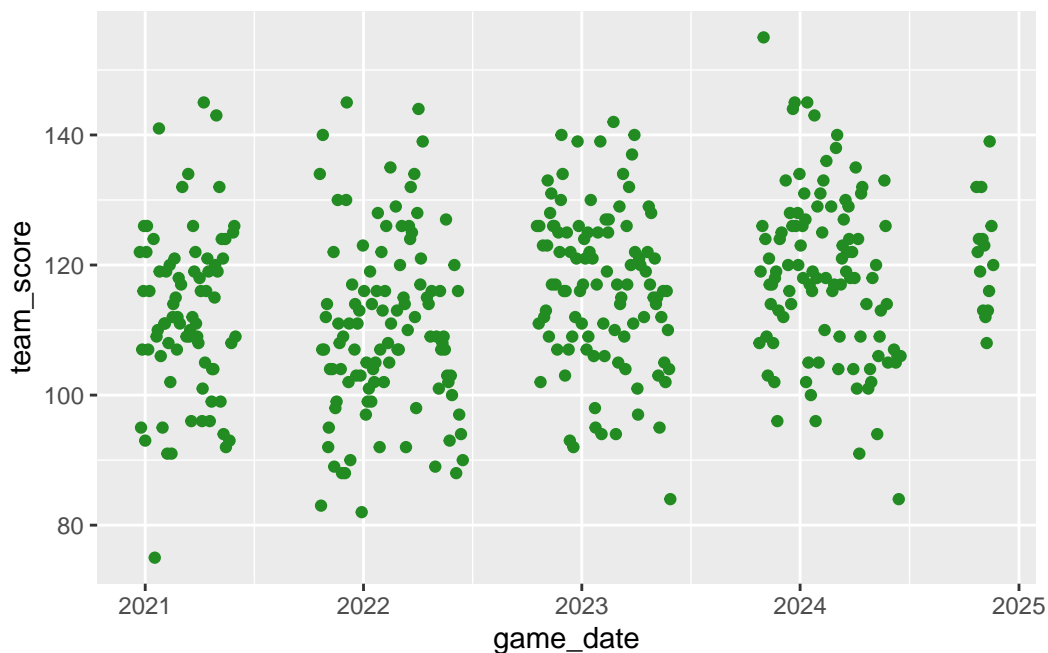
# Look at your data
View(boston_celtics)

# Look at a summary
summary(boston_celtics)
```

2. Create a scatter plot that shows the `game_date` and `team_scores`. Make the colour of the points `forestgreen`. Focus on the geometry and aesthetic layer for now.

```
ggplot(data = ???) +
  geom_???(aes(x = ???, y = ???), col = ???)
```

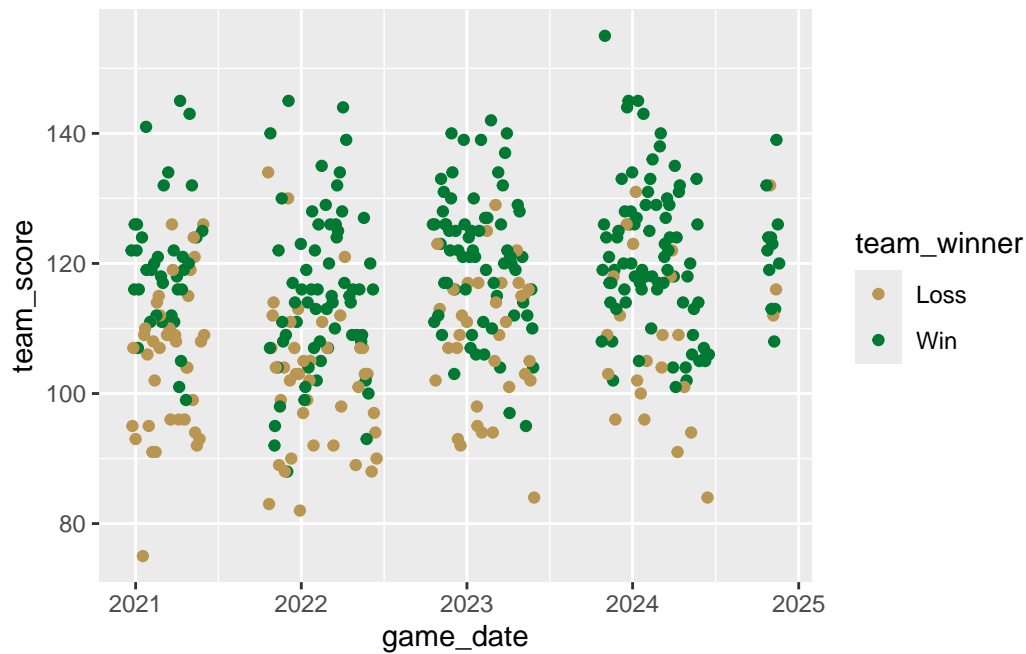
```
ggplot(data = boston_celtics) +
  geom_point(aes(x = game_date, y = team_score), col = "forestgreen")
```



3. Colour the points according to if the team won or lost. Use the hex colour codes from the [Boston Celtics team](#) to set the colour scheme. *Notice that the colour command now moves inside the bracket specify aesthetic mappings - Make sure you understand why this is.*

```
ggplot(data = ???) +
  geom_???(aes(x = ???, y = ???, col = ???))
```

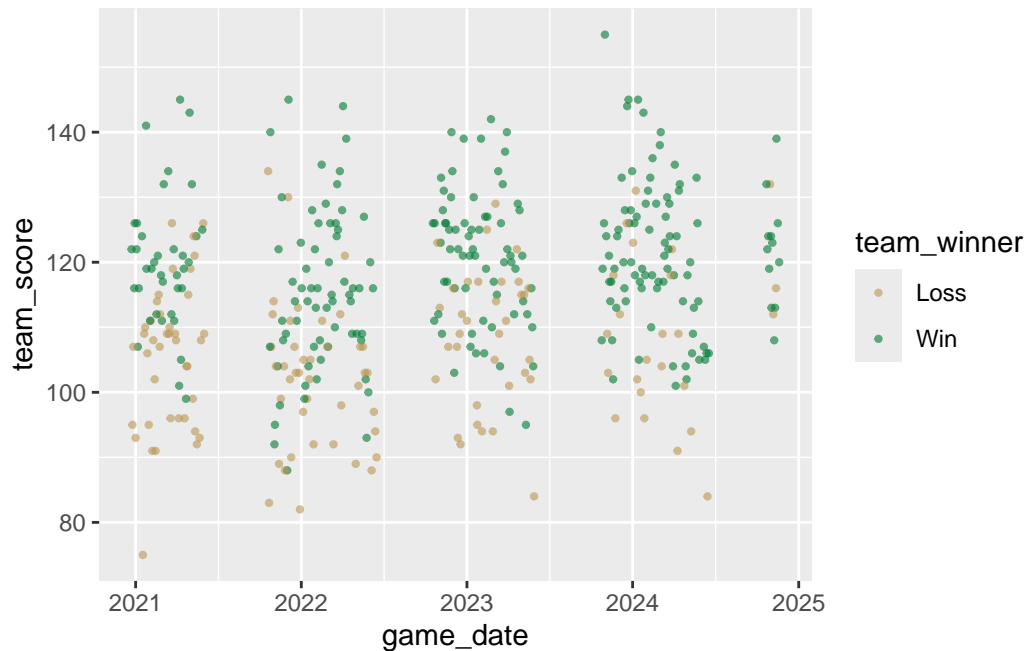
```
ggplot(data = boston_celtics) +
  geom_point(aes(x = game_date, y = team_score, col = team_winner)) +
  scale_colour_manual(label = c("Loss", "Win"), values = c("TRUE" = "#007A33", "FALSE" = "#B22222"))
```



4. As there are a lot of points, try changing the size and the transparency to see if that improves the plot.

```
ggplot(data = ???) +
  geom_???(
    aes(x = ???, y = ???, col = ???),
    size = ???, alpha = ???
  )
```

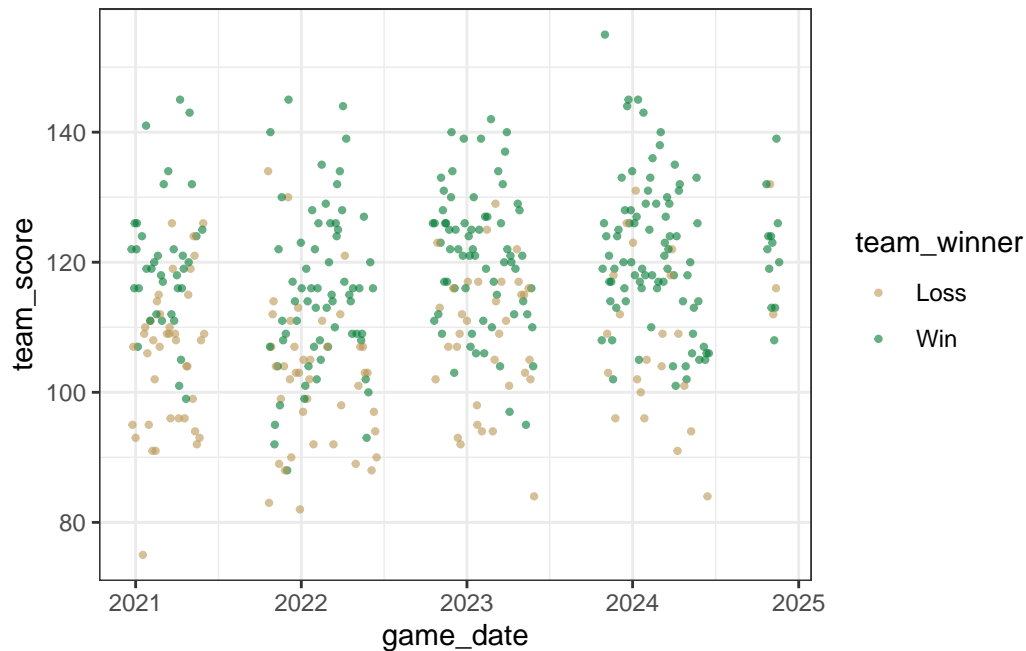
```
ggplot(data = boston_celtics) +
  geom_point(aes(x = game_date, y = team_score, col = team_winner), size = 0.8, alpha = 0.6)
  scale_colour_manual(label = c("Loss", "Win"), values = c("TRUE" = "#007A33", "FALSE" = "#B22222"))
```



5. Look at the [theme options](#) and try using a few different ones, before deciding on the theme for your plot.

```
ggplot(data = ???) +
  geom_???(
    aes(x = ???, y = ???, col = ???),
    size = ???, alpha = ???
  ) +
  theme_???
```

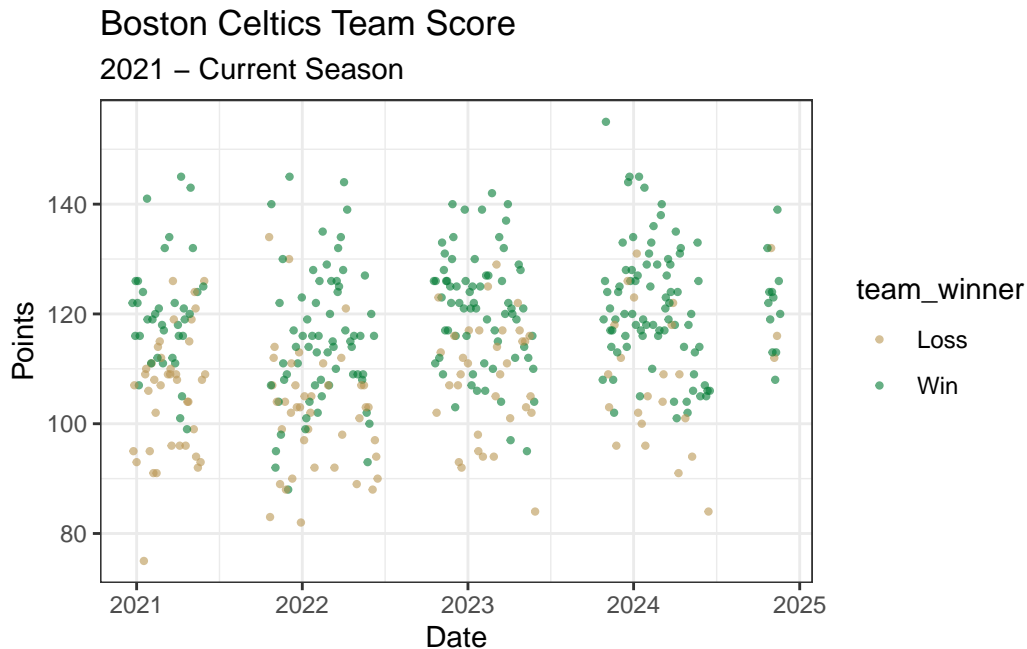
```
ggplot(data = boston_celtics) +
  geom_point(aes(x = game_date, y = team_score, col = team_winner), size = 0.8, alpha = 0.6)
  scale_colour_manual(label = c("Loss", "Win"), values = c("TRUE" = "#007A33", "FALSE" = "#B22222"))
  theme_bw()
```



5. Add appropriate labels to your plot.

```
ggplot(data = ???) +
  geom_???(
    aes(x = ???, y = ???, col = ???),
    size = ???, alpha = ???
  ) +
  theme_???() +
  labs(title = ???, subtitle = ???, x = ???, y = ???)
```

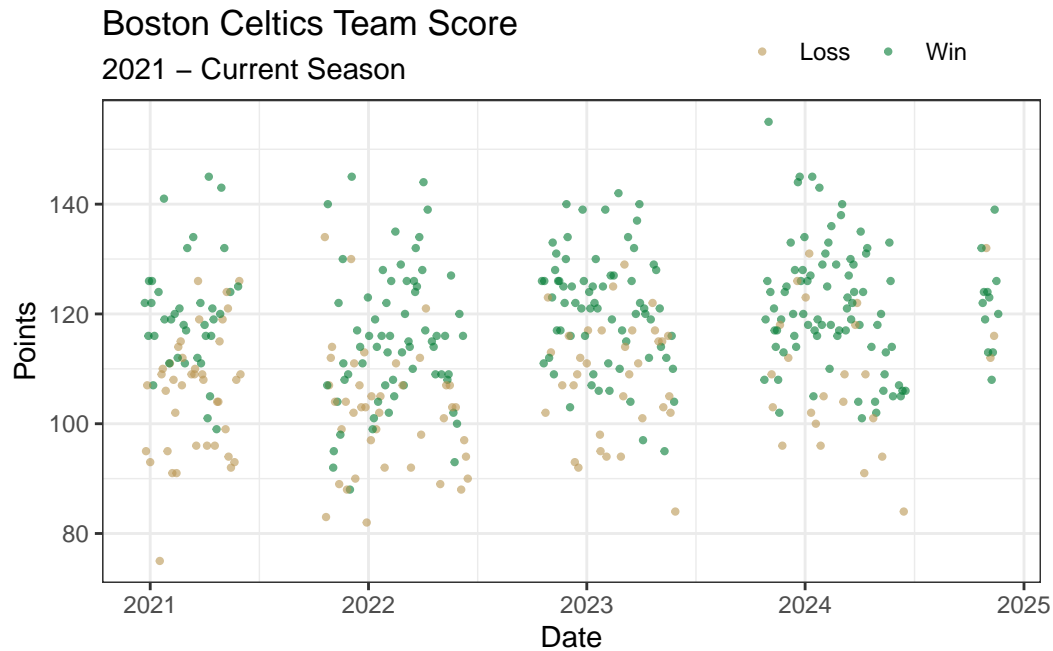
```
ggplot(data = boston_celtics) +
  geom_point(aes(x = game_date, y = team_score, col = team_winner), size = 0.8, alpha = 0.6)
  scale_colour_manual(label = c("Loss", "Win"), values = c("TRUE" = "#007A33", "FALSE" = "#B22222"))
  theme_bw() +
  labs(title = "Boston Celtics Team Score",
       subtitle = "2021 - Current Season",
       x = "Date",
       y = "Points")
```



6. Move the legend on the plot to increase the data-density. (Note the solutions show a more advanced way to move the legend than the lecture notes.)

```
ggplot(data = ???) +
  geom_???(
    aes(x = ???, y = ???, col = ???),
    size = ???, alpha = ???
  ) +
  theme_???() +
  labs(title = ???, subtitle = ???, x = ???, y = ???)
  theme(legend.position = ???)
```

```
ggplot(data = boston_celtics) +
  geom_point(aes(x = game_date, y = team_score, col = team_winner), size = 0.8, alpha = 0.6)
  scale_colour_manual(label = c("Loss", "Win"), values = c("TRUE" = "#007A33", "FALSE" = "#B22222"))
  theme_bw() +
  labs(title = "Boston Celtics Team Score",
       subtitle = "2021 – Current Season",
       x = "Date",
       y = "Points") +
  theme(legend.position = c(0.8, 1.1), legend.title = element_blank(), legend.direction = "h")
```



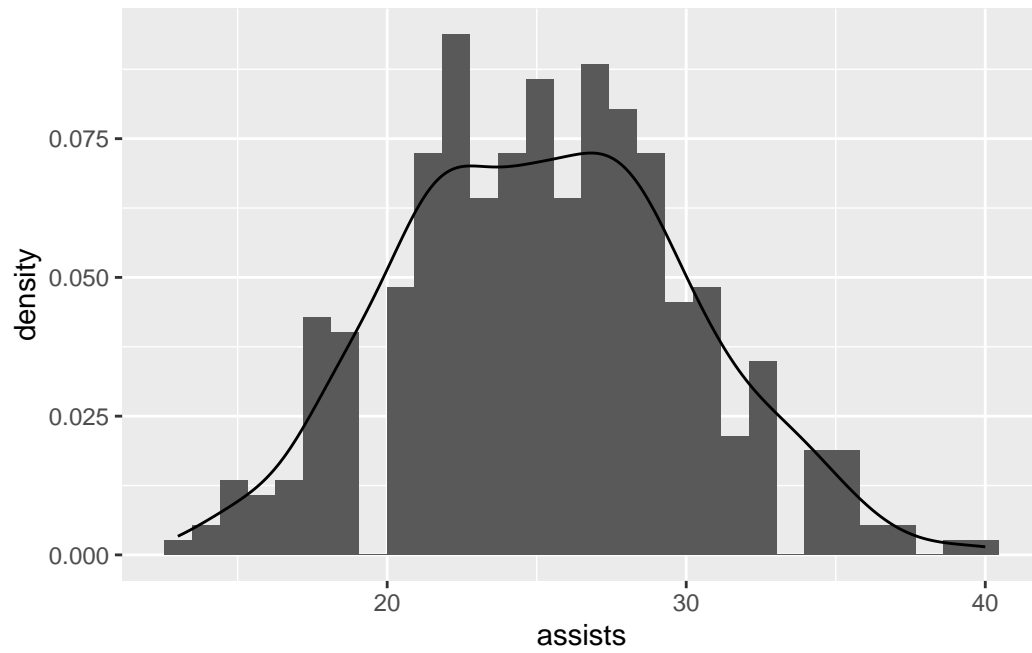
## Exercise 2

Now we've mastered the basic layers, take some time to creating some other visualisations that are interesting to you!

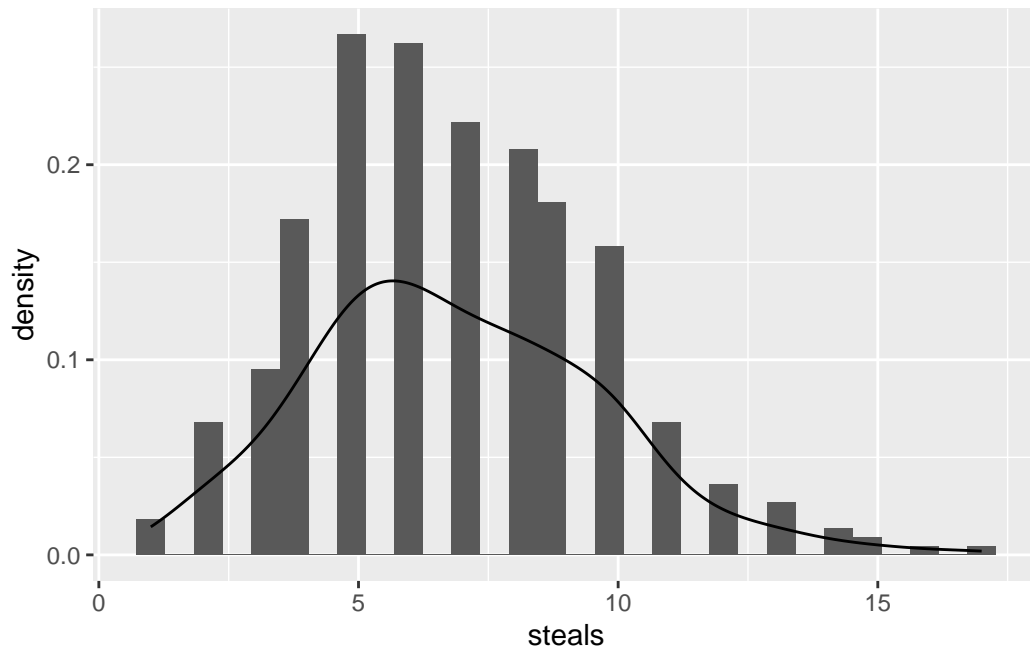
- You may like to produce plots looking at the distribution of some of the key variables. For example, **assists**, **rebounds**, **steals** etc.

```
boston_celtics |>
  ggplot(aes(x = assists)) +
  geom_histogram(aes(y = ..density..)) +
  geom_density()
```

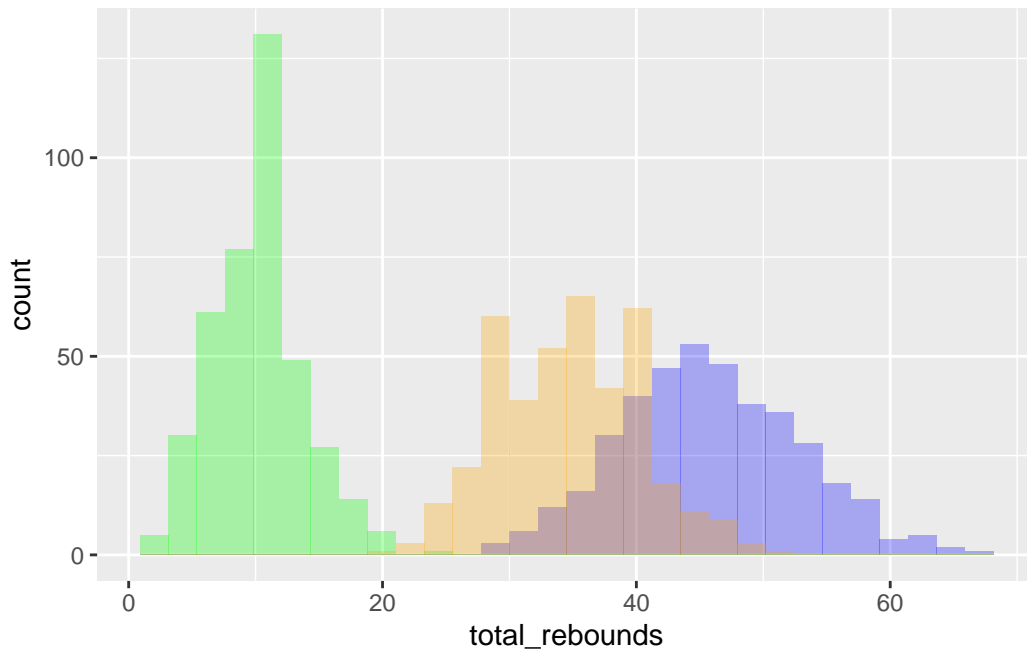




```
boston_celtics |>  
  ggplot(aes(x = steals)) +  
  geom_histogram(aes(y = ..density..)) +  
  geom_density()
```

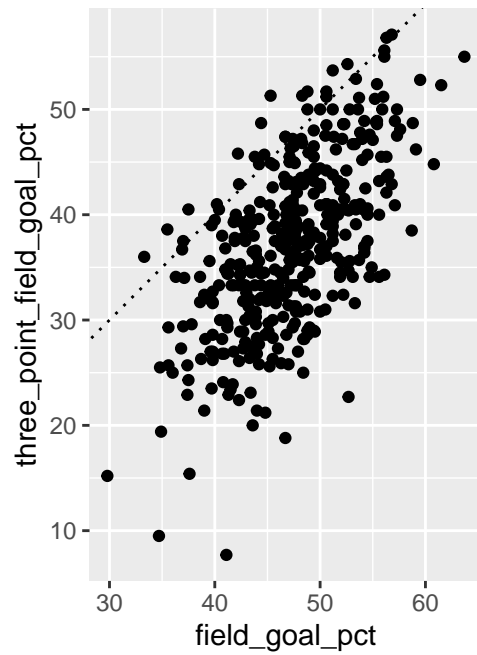


```
boston_celtics |>
  ggplot() +
  geom_histogram(aes(x = total_rebounds), fill = "blue", alpha = 0.3) +
  geom_histogram(aes(x = offensive_rebounds), fill = "green", alpha = 0.3) +
  geom_histogram(aes(x = defensive_rebounds), fill = "orange", alpha = 0.3)
```



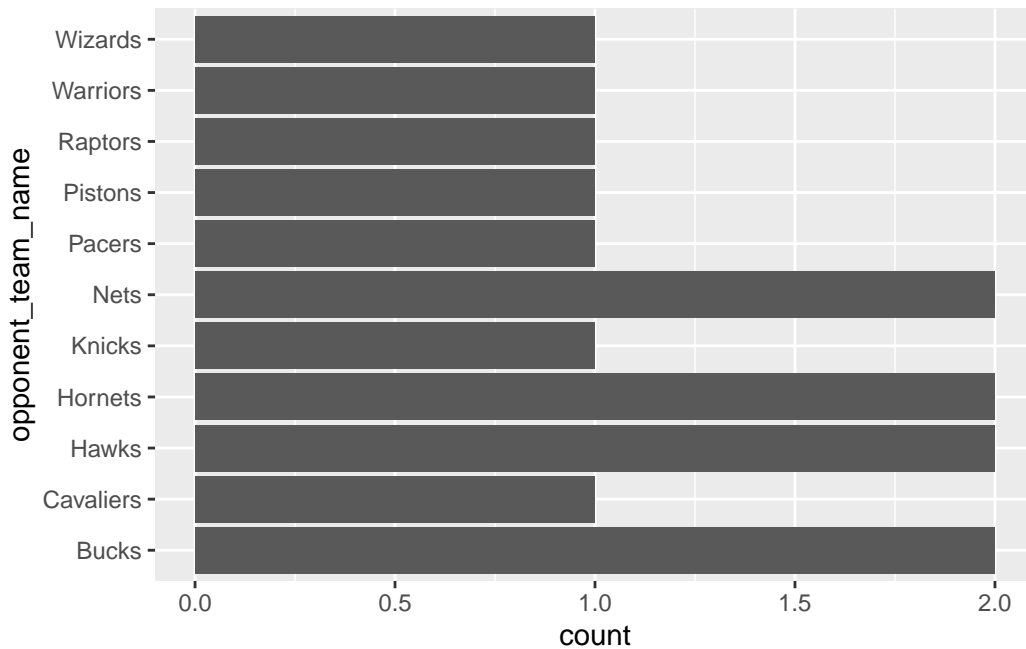
- Maybe you are interested in how the shot percentage varies between `field_goal_pct` (2 pts) and `three_point_field_goal_pct` (3 pts).

```
boston_celtics |>
  # filter(season == 2025) |>
  ggplot() +
  geom_point(aes(x = field_goal_pct, y = three_point_field_goal_pct)) +
  geom_abline(slope = 1, linetype = "dotted") +
  coord_fixed()
```



- Perhaps you want to create a bar chart (`geom_bar`) to see how many times they've played opposing teams this season.

```
boston_celtics |>  
  filter(season == 2025) |>  
  ggplot() +  
  geom_bar(aes(y = opponent_team_name))
```



## Wrap Up

By the end of this tutorial you should feel comfortable reading in a data set into R and creating a visualisation by adding layers in `ggplot2`.

**Important:** We will continue to build on this knowledge in coming weeks, so if you had any difficulties come along to consult to get some 1-on-1 help. We can slowly walk you through any parts you are struggling with so it makes sense.

## In Your Own Time

Go through the code in exercise 2 line by line to check your understanding. You may like to also add labels and themes to these plots. Have a think about what else you might do to improve them.

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