

Data visualization in R Using ggplot2

Surya Lamichhane

01 March 2025

Contents

Part 2. (Data vizulization in R using ggplot)	1
2.1 Scatter plot	2
2.2. Line plots in R	7
2.3. Bar plot in R	10
2.4. Stacked bar plot	15
2.6. Histogram in R	19
2.7. Box plot	25
2.8. Facet Plots in ggplot2	33

By the end of this lesson, you will be able to:

- Visualize data using ggplot function from ggplot2 package
- Describe and Analyze date using ggplot functions

Part 2. (Data vizulization in R using ggplot)

ggplot2 is one of the most used packages for data visualization in R and it builds plots in layers.

ggplot2 builds graphs in layers. It divides the plot into three parts:

- Data: It is the dataframe that contains data to be plotted.
- Aesthetics: These are the variables mapping to the visual properties of the plot.
- Geometry: It refers to the type of graph that is used, such as bar graph or histogram
- Optional parts:
 - XY- axex label
 - theme

For more details see the **link** [<http://r-statistics.co/Top50-Ggplot2-Visualizations-MasterList-R-Code.html>]

- **syntax**

```
ggplot(data) +
aes(x, y) +
geom_..() + # geometry type such point, hist, bar
optional_layer # such as theme, xy-labels
```

-OR- (using dplyr piping)

```
data %>%
ggplot() %>%
aes(x, y) +
geom_..()
```

2.1 Scatter plot

```
# install.packages("ggplot2")
library(dplyr)
```

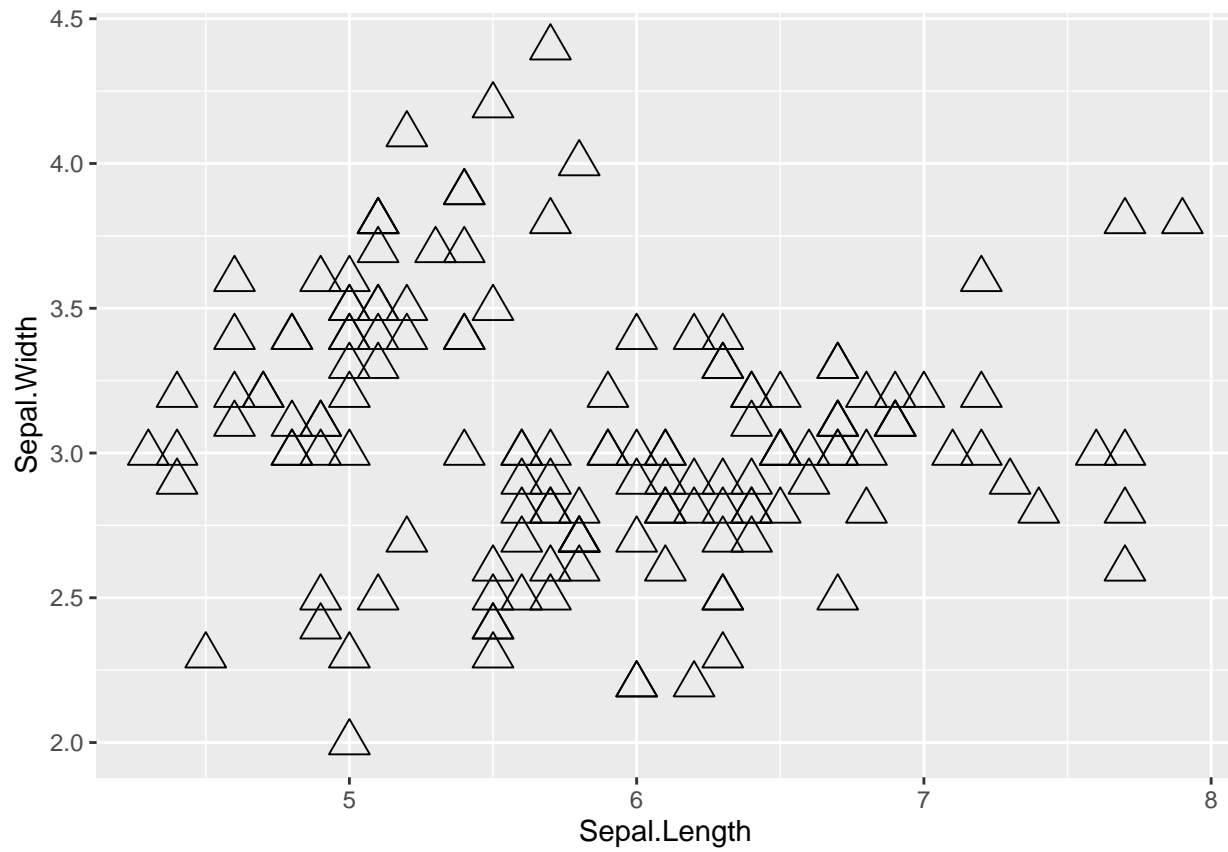
```
##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##
##   filter, lag

## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
```

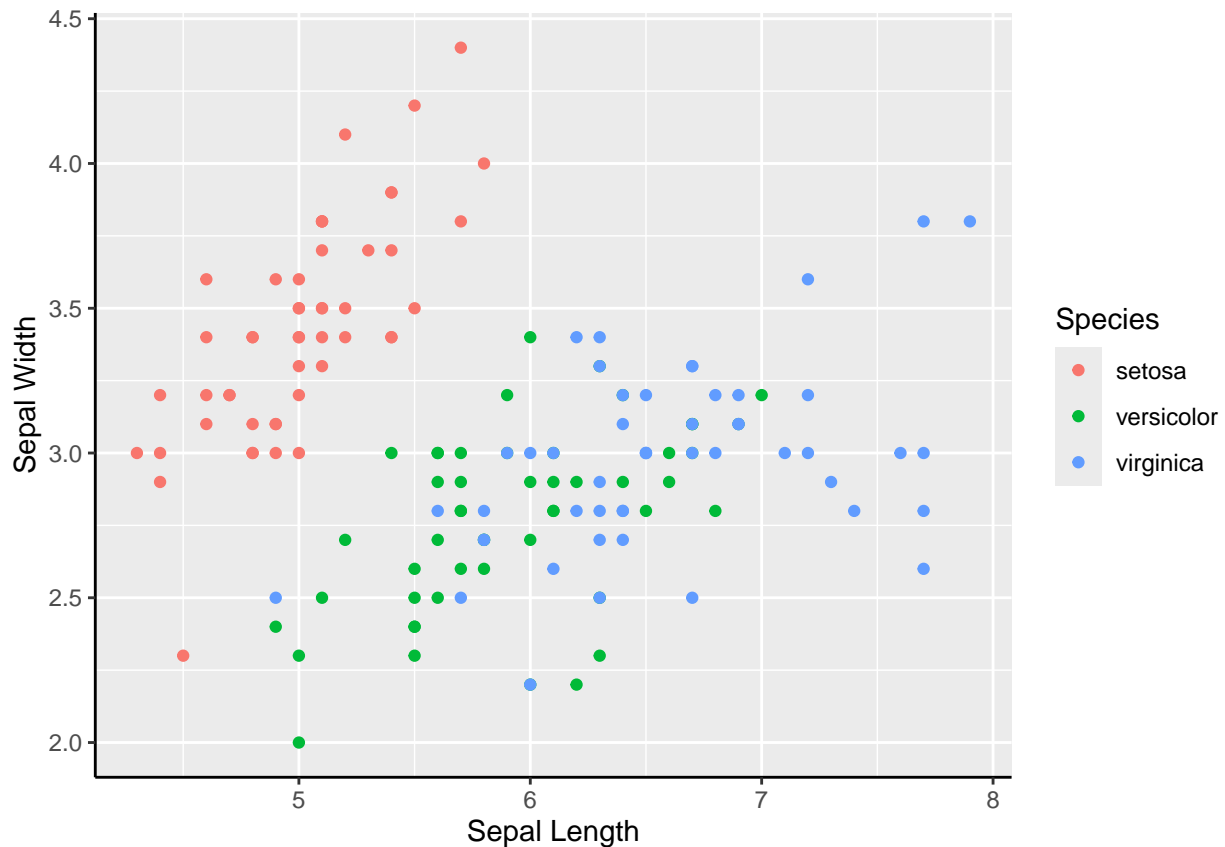
```
library(ggplot2)
```

```
iris %>%
ggplot() +
  aes(x = Sepal.Length, y = Sepal.Width) +
  #geom_point(aes(size=Sepal.Length)) # size of points varies as values
  geom_point(size=5, shape=2)
```



2.1.1. Label in the scatter plot

```
ggplot(iris, aes(x = Sepal.Length, y = Sepal.Width, color = Species)) + # Points color by Species
  geom_point() +
  xlab("Sepal Length") + # X-axis label
  ylab("Sepal Width") + # Y-axis label
  theme(axis.line = element_line(colour = "black", # Changes the default theme
                                linewidth = 0.5))
```



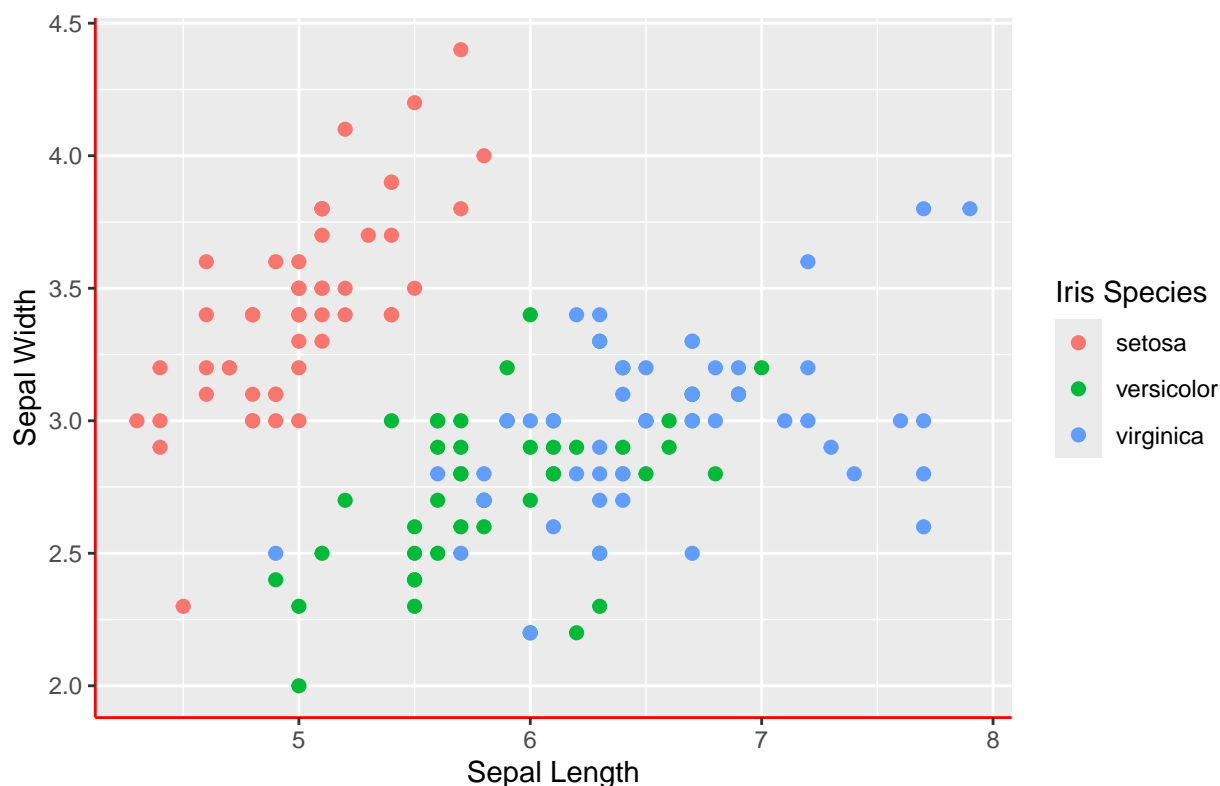
2.1.2. Plot Title and Change legend name

```
library(ggplot2)

ggplot(iris, aes(x = Sepal.Length, y = Sepal.Width, color = Species)) +
  geom_point(size = 2) + # Increased point size for better visibility
  ggtitle("Scatter Plot in R") +
  scale_color_discrete(name = "Iris Species") + # Correct way to rename legend
  xlab("Sepal Length") +
  ylab("Sepal Width") +
  theme(
    axis.line = element_line(colour = "red", size = 0.5), # Keep red xy-axes
    plot.title = element_text(hjust = 0.5, size = 20, face = "bold"), # Centered and bold title
  )
```

```
## Warning: The `size` argument of `element_line()` is deprecated as of ggplot2 3.4.0.
## i Please use the `linewidth` argument instead.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
```

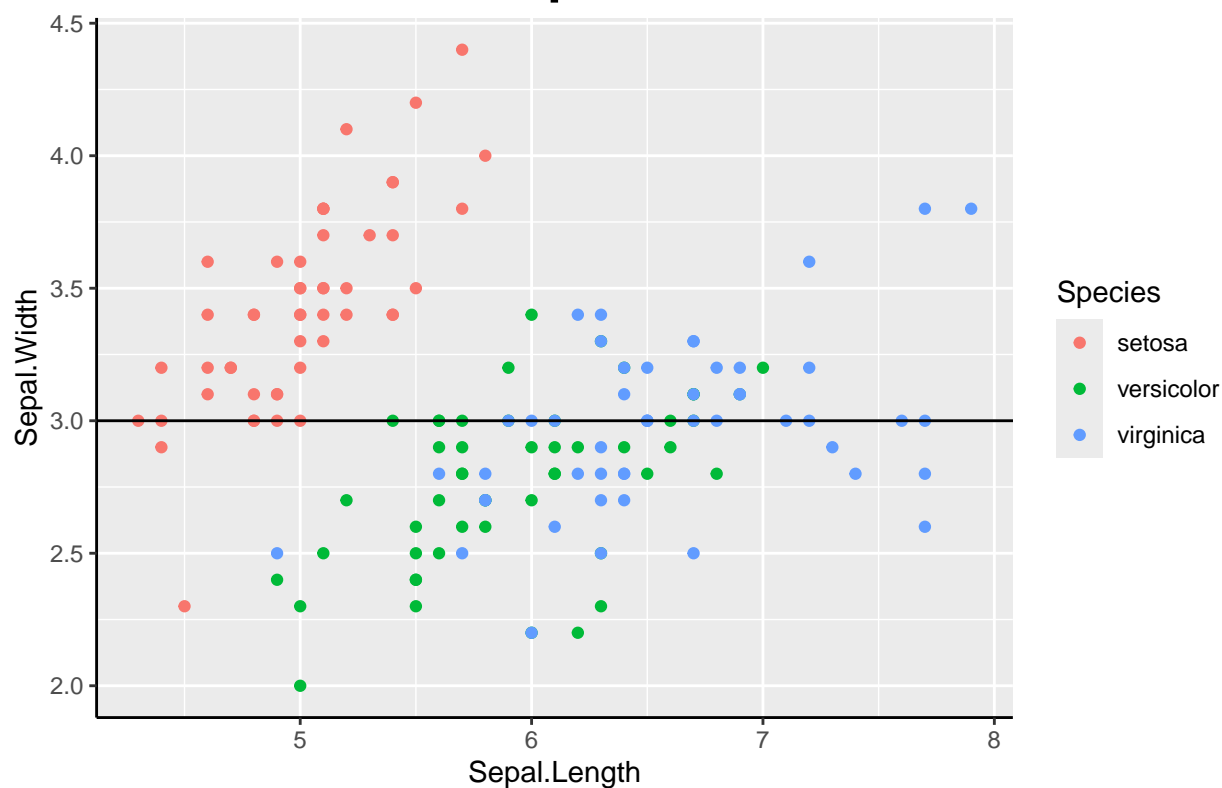
Scatter Plot in R



2.1.3. Straigh line using abline()

```
ggplot(iris, aes(x = Sepal.Length, y = Sepal.Width, color = Species)) +  
  geom_point() +  
  geom_abline(intercept = 3, slope = 0 ) + # draw a horizontal line  
  ggtitle("Scattor plot in R") +  
  scale_color_discrete(name = "Species") + # Legend title  
  xlab("Sepal.Length") + # X-axis label  
  ylab("Sepal.Width") + # Y-axis label  
  theme(axis.line = element_line(color = "black", size = 0.5), # Changes the default theme (xy-axes)  
        plot.title = element_text(hjust=0.5, size = 20, face = "bold")) # Assign title on center  
)
```

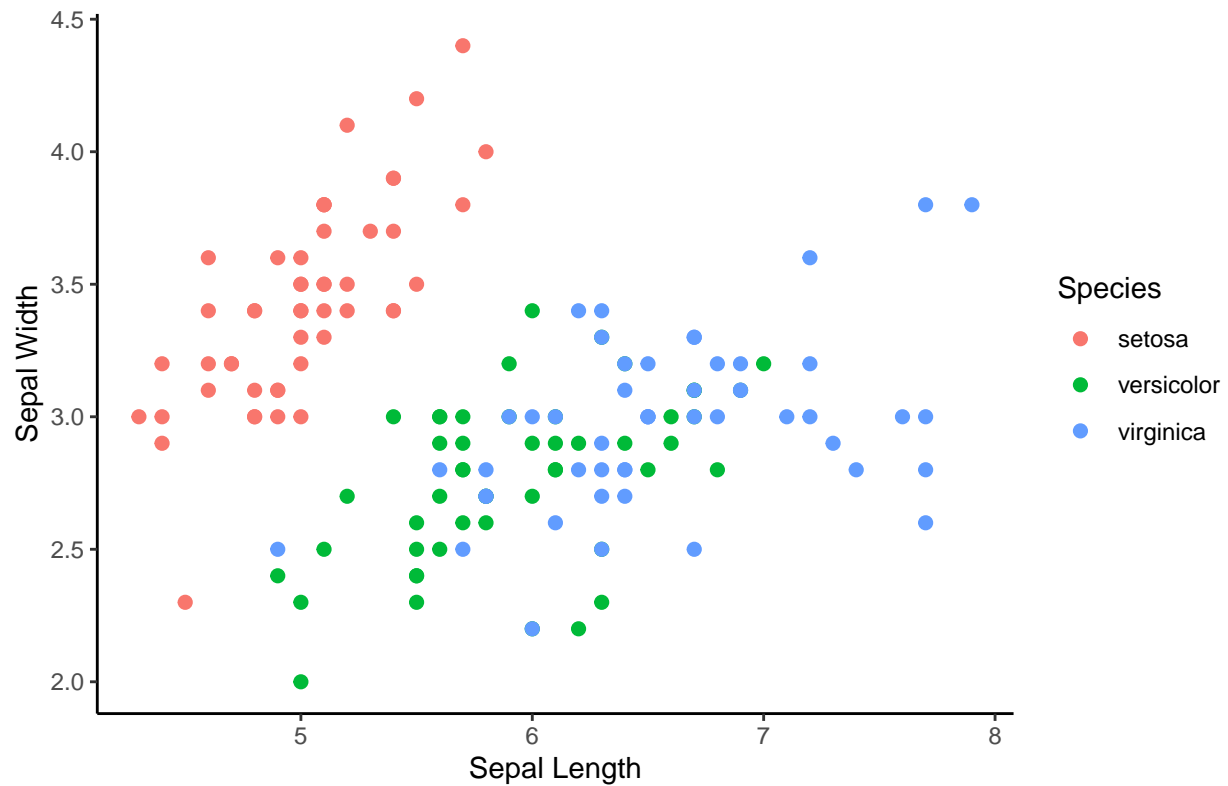
Scatter plot in R



2.1.4. Remove the grids and change background to white

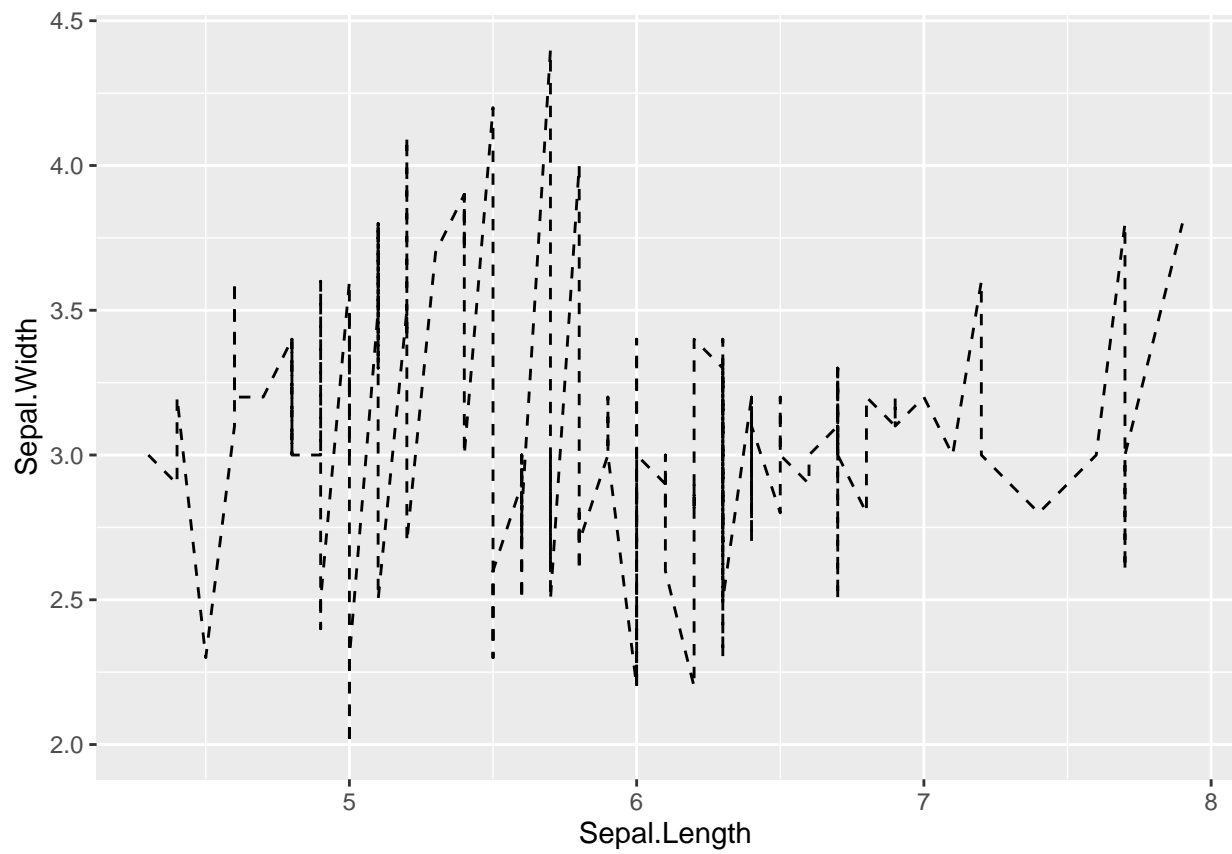
```
ggplot(iris, aes(x = Sepal.Length, y = Sepal.Width, color = Species)) +  
  geom_point(size = 2) + # Adds scatter points with color by Species  
  ggtitle("Scatter Plot in R") +  
  scale_color_discrete(name = "Species") + # Legend title  
  xlab("Sepal Length") +  
  ylab("Sepal Width") +  
  theme(  
    axis.line = element_line(colour = "black", size = 0.5), # Make xy-axes black  
    plot.title = element_text(hjust = 0.5, face = "bold", size = 20), # Center title, bold, increase s  
    panel.background = element_rect(fill = "white"), # White background  
    panel.grid.major = element_blank(), # Remove major grid lines  
    panel.grid.minor = element_blank() # Remove minor grid lines  
  )
```

Scatter Plot in R

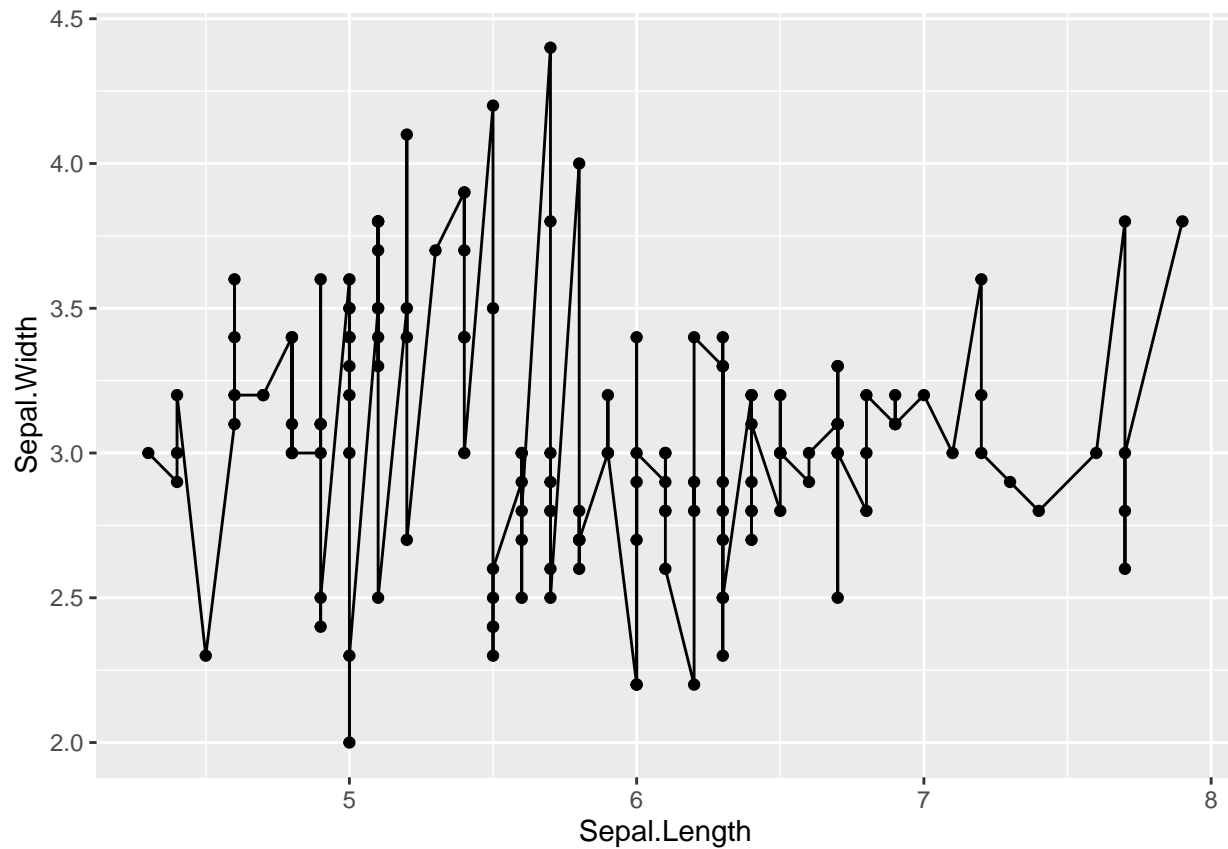


2.2. Line plots in R

```
# Change the line type
ggplot(data=iris, aes(x = Sepal.Length, y = Sepal.Width)) +
  geom_line(linetype = "dashed")
```



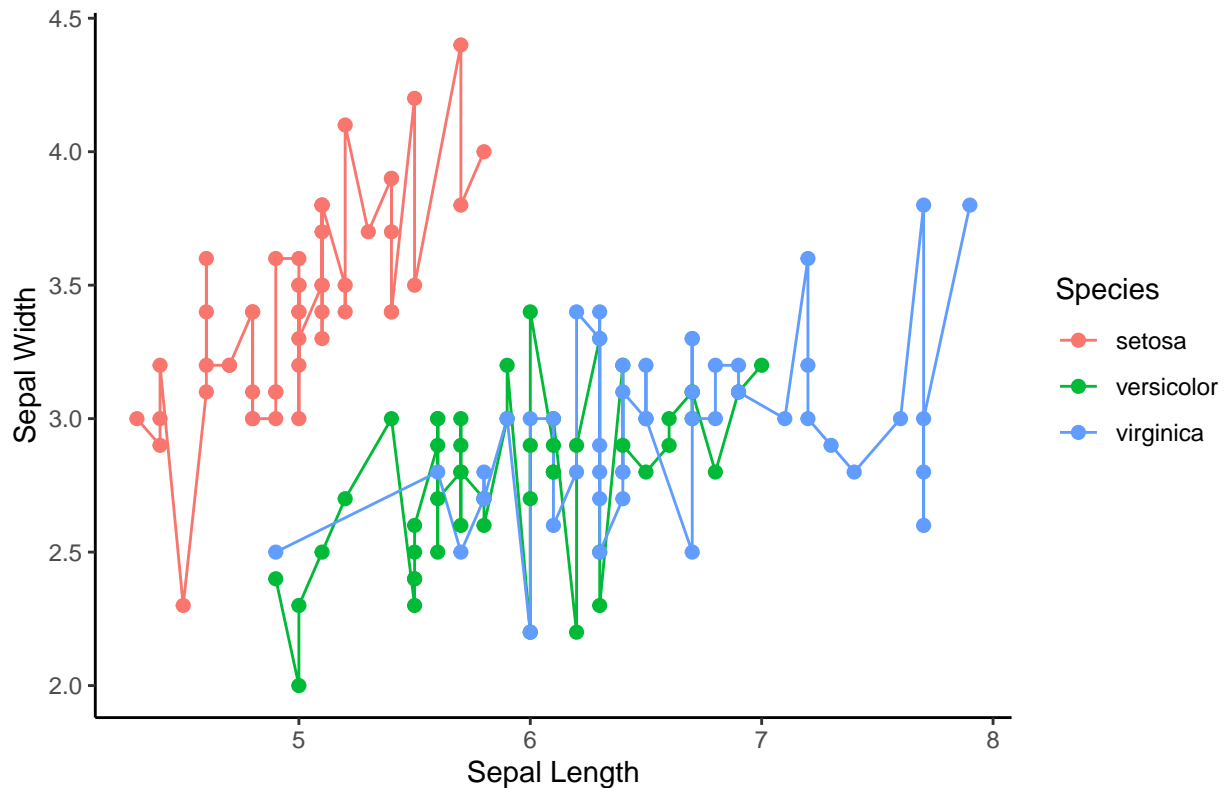
```
# add points
ggplot(data=iris, aes(x = Sepal.Length, y = Sepal.Width)) +
  geom_line(linetype = "solid")+
  geom_point()
```

```
# Line plot grouped by Species
```

```
ggplot(data = iris, aes(x = Sepal.Length, y = Sepal.Width, color = Species)) +
  geom_line() + # Line plot grouped by Species
  geom_point(size = 2) + # Scatter points with color by Species
  ggtitle("Sepal Length vs Sepal Width by Species") + # Title
  scale_color_discrete(name = "Species") + # Properly label legend
  xlab("Sepal Length") +
  ylab("Sepal Width") +
  theme(
    axis.line = element_line(colour = "black", size = 0.5), # Make xy-axes black, size is thickness
    plot.title = element_text(hjust = 0.5, size = 18, face = "bold"), # Centered bold title
    legend.position = "right", # Position legend to the right
    panel.background = element_rect(fill = "white"), # White background
    panel.grid.major = element_blank(), # Remove major grid lines
    panel.grid.minor = element_blank() # Remove minor grid lines
  )
```

Sepal Length vs Sepal Width by Species



2.3. Bar plot in R

```
str(mtcars)
```

```
## 'data.frame':  32 obs. of  11 variables:
##  $ mpg : num  21 21 22.8 21.4 18.7 18.1 14.3 24.4 22.8 19.2 ...
##  $ cyl : num   6  6  4  6  8  6  8  4  4  6 ...
##  $ disp: num  160 160 108 258 360 ...
##  $ hp  : num  110 110 93 110 175 105 245 62 95 123 ...
##  $ drat: num   3.9 3.9 3.85 3.08 3.15 2.76 3.21 3.69 3.92 3.92 ...
##  $ wt  : num   2.62 2.88 2.32 3.21 3.44 ...
##  $ qsec: num   16.5 17 18.6 19.4 17 ...
##  $ vs  : num    0  0  1  1  0  1  0  1  1  1 ...
##  $ am  : num    1  1  1  0  0  0  0  0  0  0 ...
##  $ gear: num    4  4  4  3  3  3  3  4  4  4 ...
##  $ carb: num    4  4  1  1  2  1  4  2  2  4 ...
```

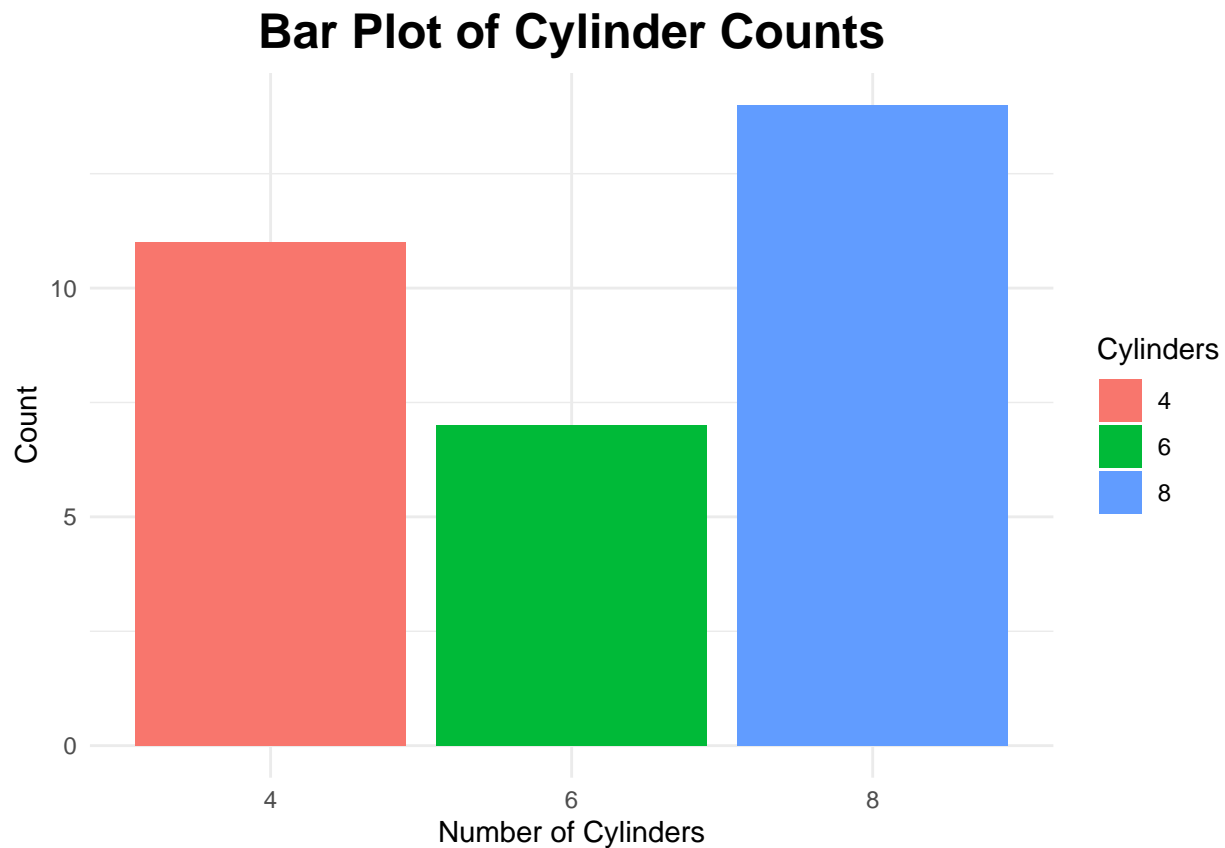
2.3.1. Creating a Bar Plot with Factor Variables in

```
mtcars %>%
  ggplot(aes(x = factor(cyl), fill = factor(cyl))) + # x variable must be factor, y is not required wi
  geom_bar() +
```

```

scale_fill_discrete(name = "Cylinders") + # Change Legend title
ggtitle("Bar Plot of Cylinder Counts") +
xlab("Number of Cylinders") +
ylab("Count") +
theme_minimal()+
theme(
  #axis.line = element_line(colour = "black", size = 0.5), # Make xy-axes black, size is thickness
  plot.title = element_text(hjust = 0.5, size = 18, face = "bold"), # Centered bold title
)

```



2.3.2. Creating a Bar Plot when we know the frequency (counts) of the labels

```

df = mtcars %>%
  group_by(cyl)%>%
  summarise(count = n())
df

```

```

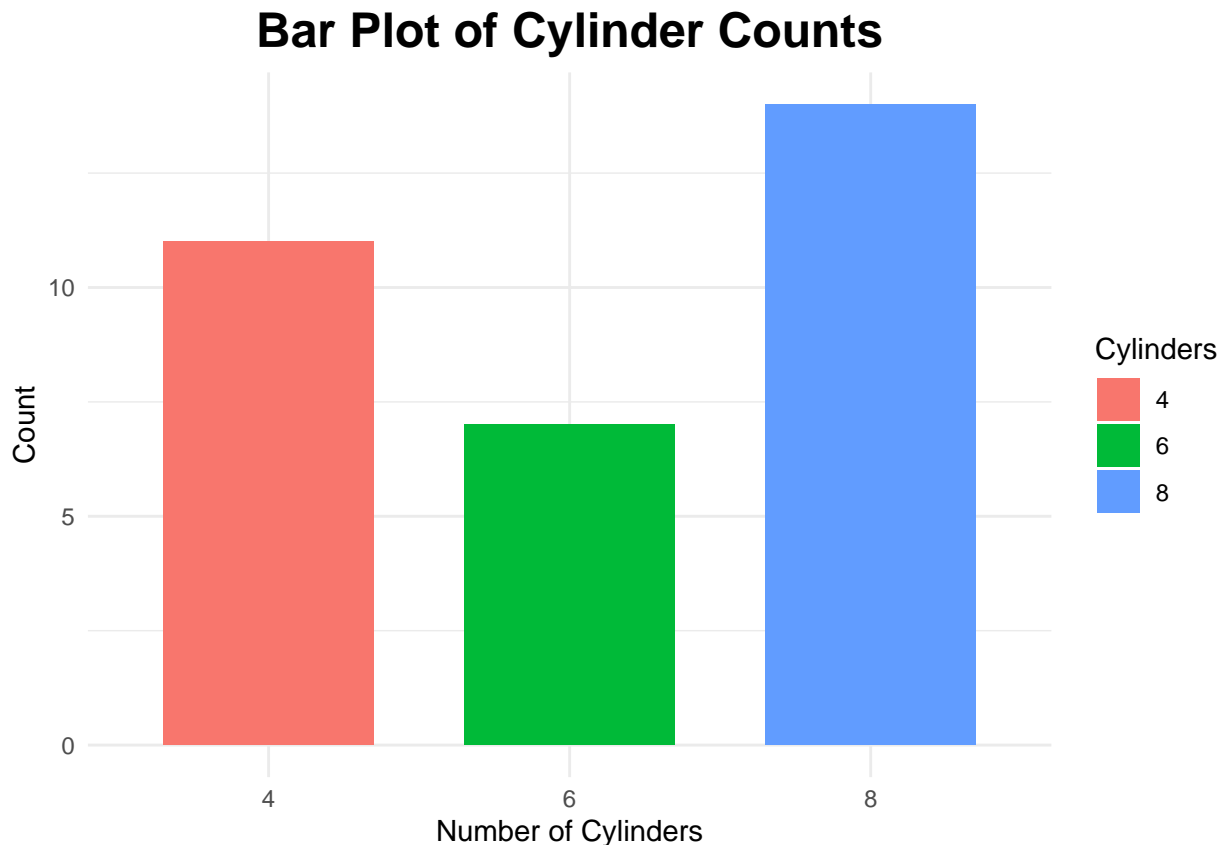
## # A tibble: 3 x 2
##   cyl count
##   <dbl> <int>
## 1     4     11
## 2     6      7
## 3     8     14

```

- In ggplot2, the `geom_bar()` function has a default statistical transformation (`stat = "count"`), which automatically counts occurrences of categories when no y variable is specified.
- However, if we already have pre-computed values (such as exact counts or aggregated data) and want to use them directly, we set `stat = "identity"` to ensure the bars represent the actual y values provided.

```
# Basic barplot
df$cyl = factor(df$cyl) # changing to factor is important here, otherwise R will consider the values as
# Create the bar plot
ggp <- ggplot(data = df, aes(x = cyl, y = count, fill = cyl)) +
  geom_bar(stat = "identity", width = 0.7) + # Identity means use precomputed counts
  scale_fill_discrete(name = "Cylinders") + # Update legend title
  ggtitle("Bar Plot of Cylinder Counts") +
  xlab("Number of Cylinders") +
  ylab("Count") +
  theme_minimal() + # White background
  theme(
    plot.title = element_text(hjust = 0.5, size = 18, face = "bold") # Centered bold title
  )

# Display the plot
print(ggp)
```



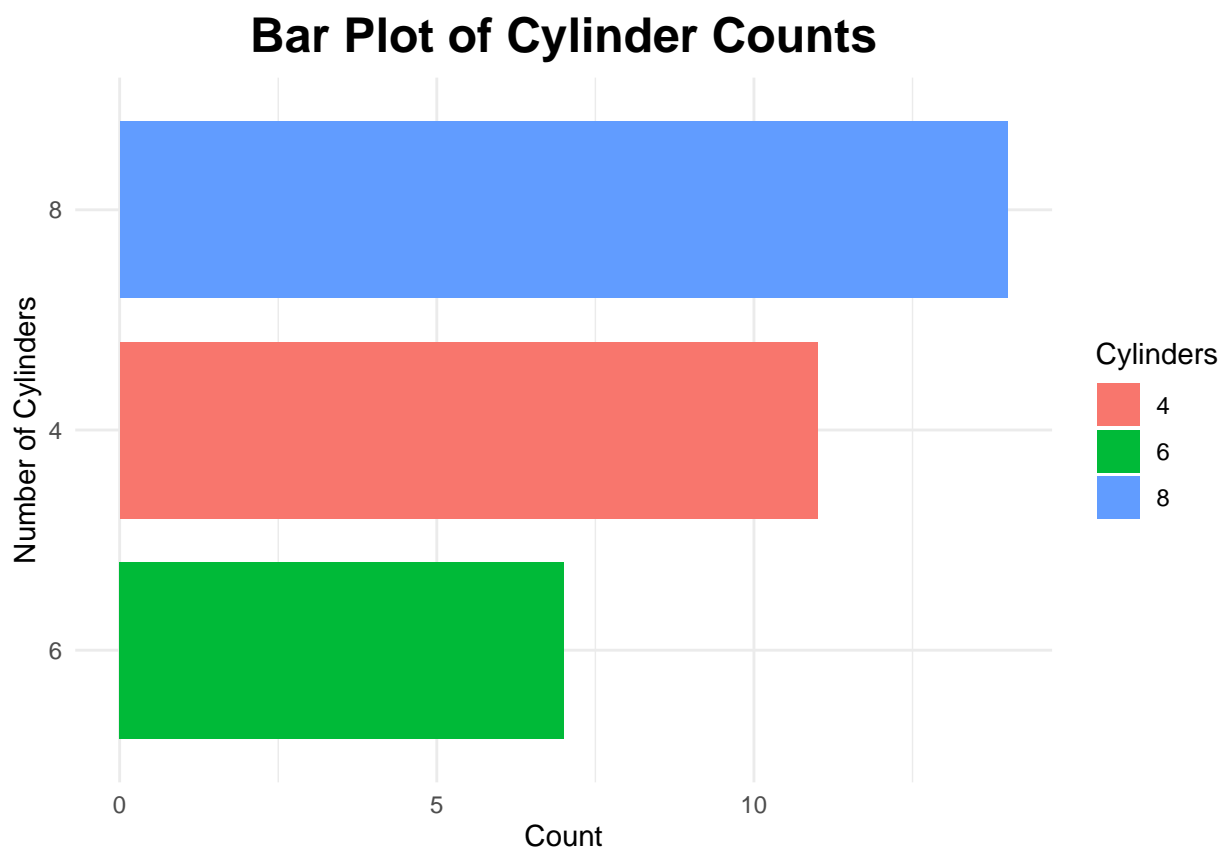
2.3.4 Horizontal Barplot

```

# for the summary x variable need to be a factor variable
ggp <- ggplot(data = df, aes(y = reorder(cyl,count), x = count, fill = cyl)) +
  geom_bar(stat="identity", width=0.8) +
  scale_fill_discrete(name = "Cylinders") + # Update legend title
  ggtitle("Bar Plot of Cylinder Counts") +
  ylab("Number of Cylinders") +
  xlab("Count") +
  theme_minimal() + # White background
  theme(
    plot.title = element_text(hjust = 0.5, size = 18, face = "bold") # Centered bold title
  )

ggp

```



2.3.5. Application to the Boston-311 Service Data

```

Boston311_2023_data =
read.csv("https://data.boston.gov/dataset/8048697b-ad64-4bfc-b090-ee00169f2323/resource/e6013a93-1321-4

library(stringr)
library(dplyr)
Boston311_2023_data$Parking_Enforcement_status <- str_detect(Boston311_2023_data$case_title,
  regex("\\bParking Enforcement\\b"))

```

```
Parking_Enforcement_by_nbd <- Boston311_2023_data %>%
  group_by(neighborhood) %>%
  summarise(nbd_count_Parking_Enforcement = n()) %>%
  arrange(desc(nbd_count_Parking_Enforcement))
head(Parking_Enforcement_by_nbd, 10)
```

```
## # A tibble: 10 x 2
##   neighborhood      nbd_count_Parking_Enforcement
##   <chr>              <int>
## 1 Dorchester        36272
## 2 Roxbury           21426
## 3 South Boston / South Boston Waterfront 18835
## 4 Allston / Brighton 18490
## 5 East Boston       17862
## 6 South End         15265
## 7 Jamaica Plain     13728
## 8 Downtown / Financial District 11526
## 9 Greater Mattapan  11191
## 10 Back Bay         10559
```

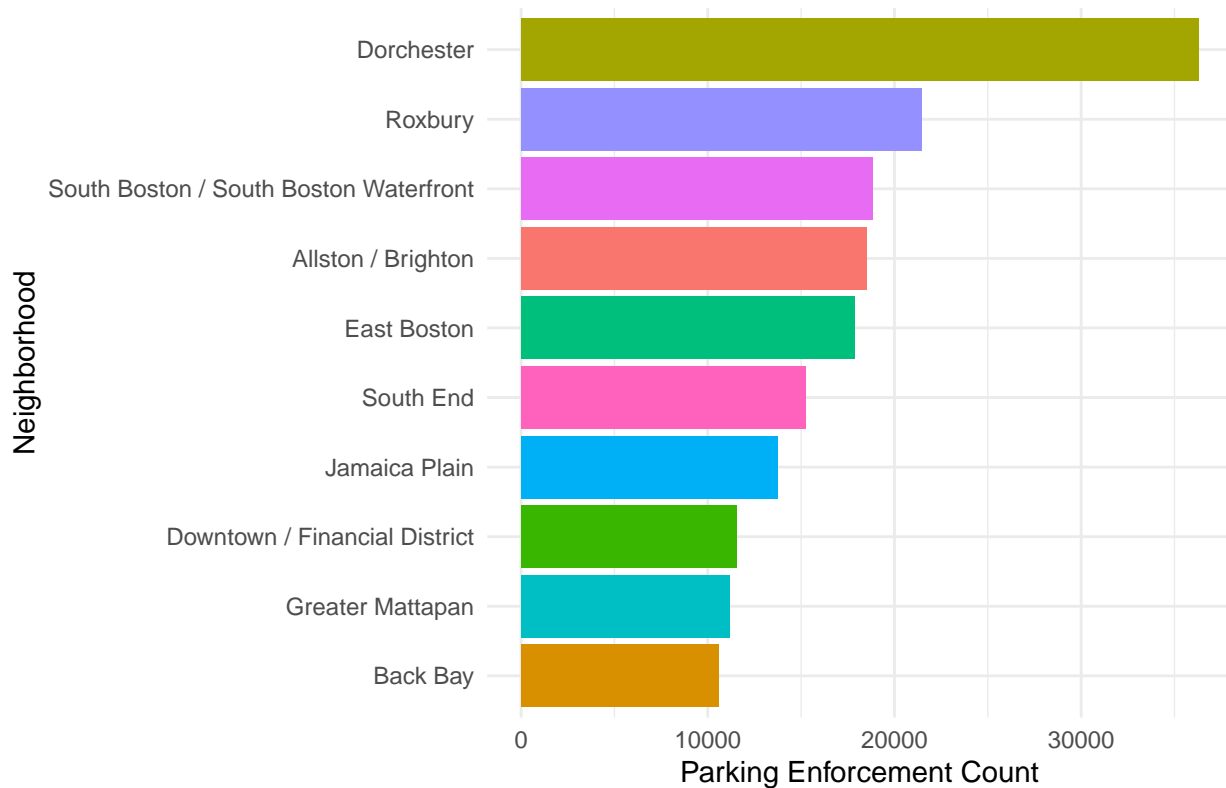
```
top_10_nbd = Parking_Enforcement_by_nbd[1:10, ]
top_10_nbd
```

```
## # A tibble: 10 x 2
##   neighborhood      nbd_count_Parking_Enforcement
##   <chr>              <int>
## 1 Dorchester        36272
## 2 Roxbury           21426
## 3 South Boston / South Boston Waterfront 18835
## 4 Allston / Brighton 18490
## 5 East Boston       17862
## 6 South End         15265
## 7 Jamaica Plain     13728
## 8 Downtown / Financial District 11526
## 9 Greater Mattapan  11191
## 10 Back Bay         10559
```

```
ggp_311 <- ggplot(top_10_nbd, aes(y = reorder(neighborhood, nbd_count_Parking_Enforcement),
                                   x = nbd_count_Parking_Enforcement,
                                   fill = neighborhood)) +
  geom_bar(stat = "identity") +
  scale_fill_discrete(name = "Neighborhood") + # Corrected legend title
  ggtitle("Top 10 Neighborhoods by Parking Enforcement Count") + # Added title
  xlab("Parking Enforcement Count") +
  ylab("Neighborhood") +
  theme_minimal() + # Clean background instead of theme_void()
  theme(
    plot.title = element_text(hjust = 0.5, size = 16, face = "bold"), # Centered, bold title
    legend.position = "none" # Removes the legend
  )

ggp_311
```

Top 10 Neighborhoods by Parking Enforcement



2.3.5. Saving Plots

```
ggsave("~/Desktop/NBD_complaint.png") # save in desktop with file name "NBD_complaint"
```

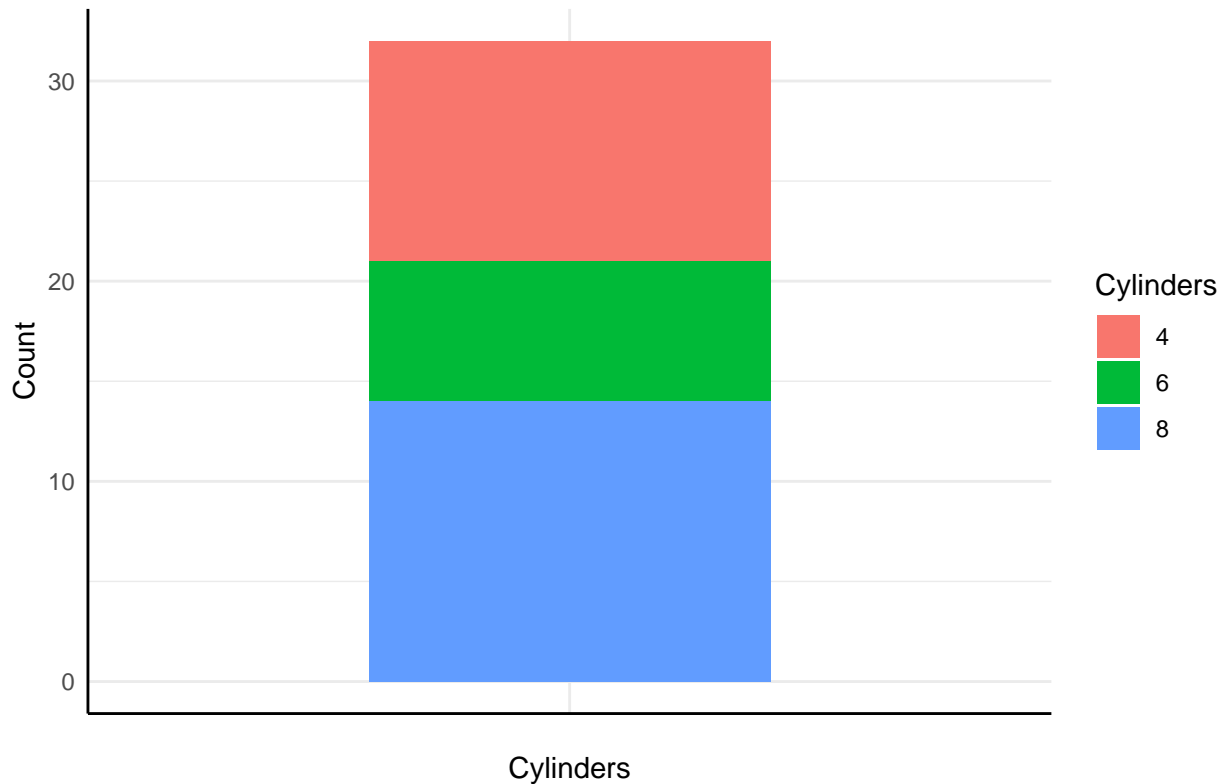
```
## Saving 6.5 x 4.5 in image
```

2.4. Stacked bar plot

a. Creating Single Stacked bar plot is exactly similar as bar-graph, but we leave x as x = " " (empty).

```
# Create bar plot
ggp <- ggplot(data = mtcars, aes(x = '', fill = factor(cyl))) +
  geom_bar(stat = "count", width = 0.5) +
  scale_fill_discrete(name = "Cylinders") + # Corrected legend title
  ggtitle("Cylinder Count in mtcars Dataset") +
  xlab("Cylinders") +
  ylab("Count") +
  theme_minimal() +
  theme(
    axis.line = element_line(colour = "black", size = 0.5), # Make xy-axes black, size is thickness
    plot.title = element_text(hjust = 0.5, size = 18, face = "bold") # Centered bold title
  )
ggp
```

Cylinder Count in mtcars Dataset

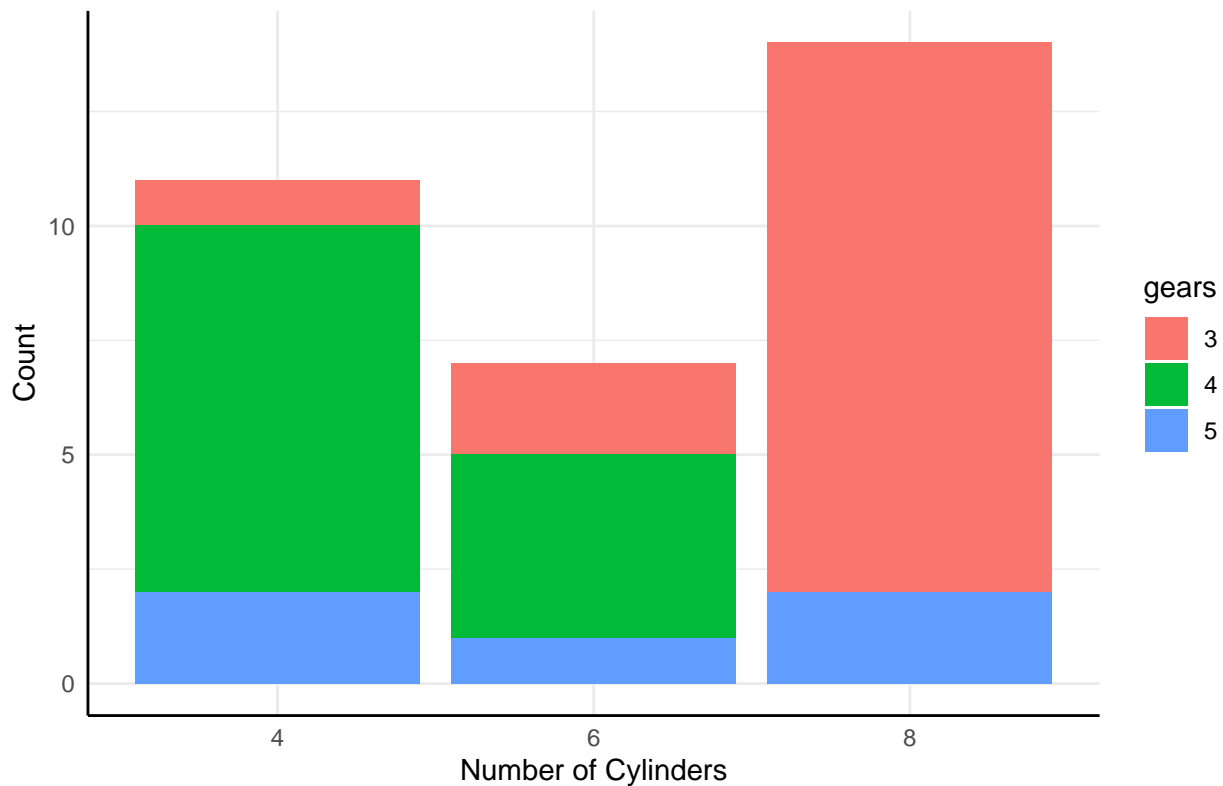


b. Multiple stacked bar plot based on cylinder and gears

```
library(ggplot2)

ggplot(mtcars, aes(x = factor(cyl), fill = factor(gear))) +
  geom_bar() +
  scale_fill_discrete(name = "gears") + # Corrected legend title
  ggtitle("Subdivided (Stacked) Bar Plot") +
  xlab("Number of Cylinders") +
  ylab("Count") +
  theme_minimal()+
  theme(
    axis.line = element_line(colour = "black", size = 0.5), # Make xy-axes black, size is thickness
    plot.title = element_text(hjust = 0.5, size = 18, face = "bold") # Centered bold title
  )
```


Subdivided (Stacked) Bar Plot



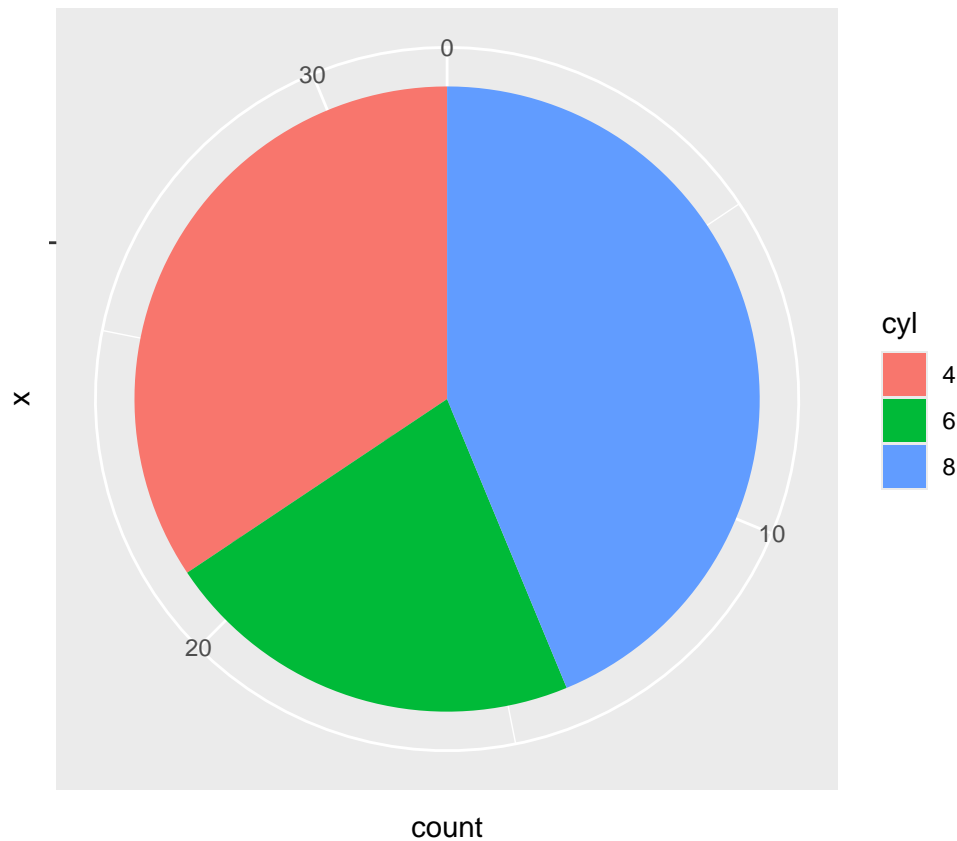
2.5. Pie chart

Creating a Pie Chart in ggplot2 is similar to a subdivided bar chart (stacked bar chart), but we need to add an additional layer using `coord_polar(theta = "y")` to transform it into a circular shape.

```
df <- mtcars %>%  
  group_by(cyl) %>%  
  summarize(count = n()) %>% # count the number of cars basedd on cylinders  
  mutate(cyl = factor(cyl)) # convert to factor variable  
df
```

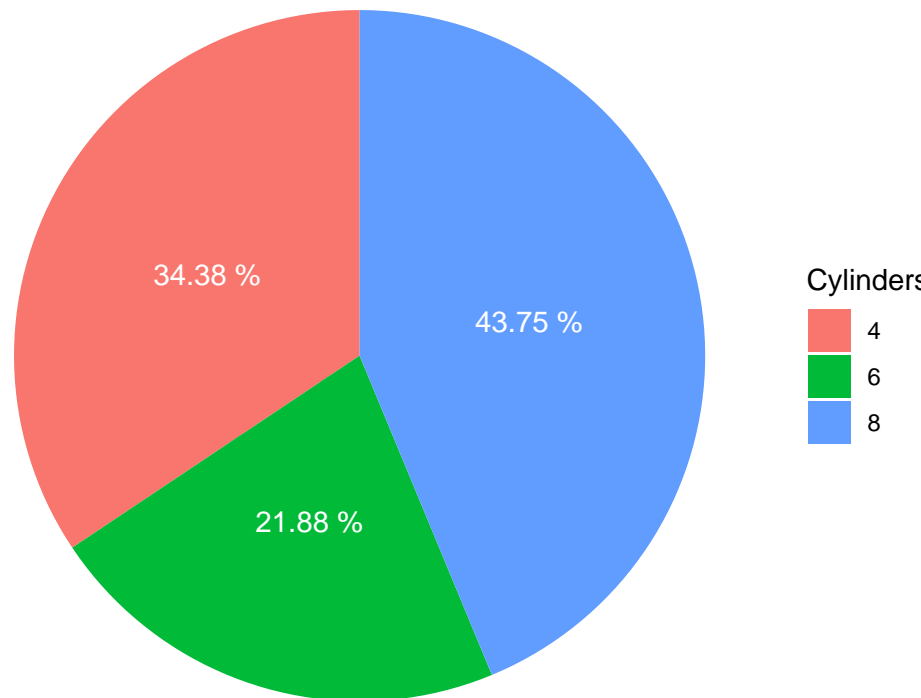
```
## # A tibble: 3 x 2  
##   cyl   count  
##   <fct> <int>  
## 1 4       11  
## 2 6        7  
## 3 8       14
```

```
ggp <- ggplot(data=df, aes(x = '', y = count, fill = cyl)) +  
  geom_bar(stat="identity", width=0.7) +  
  coord_polar("y", start=0)  
ggp
```



```
count_percent <- df$count/sum(df$count) *100
df$perc = round(count_percent, 2)

ggp <- ggplot(data=df, aes(x = '', y = perc, fill = cyl)) +
  geom_bar(stat="identity", width=0.7) +
  coord_polar("y", start=0)+
  scale_fill_discrete(name = "Cylinders") + # Corrected legend title
  geom_text(aes(label = paste(perc, '%')), color = rep("white", 3),
            position = position_stack(vjust = 0.5)) +
  theme_void()
ggp
```



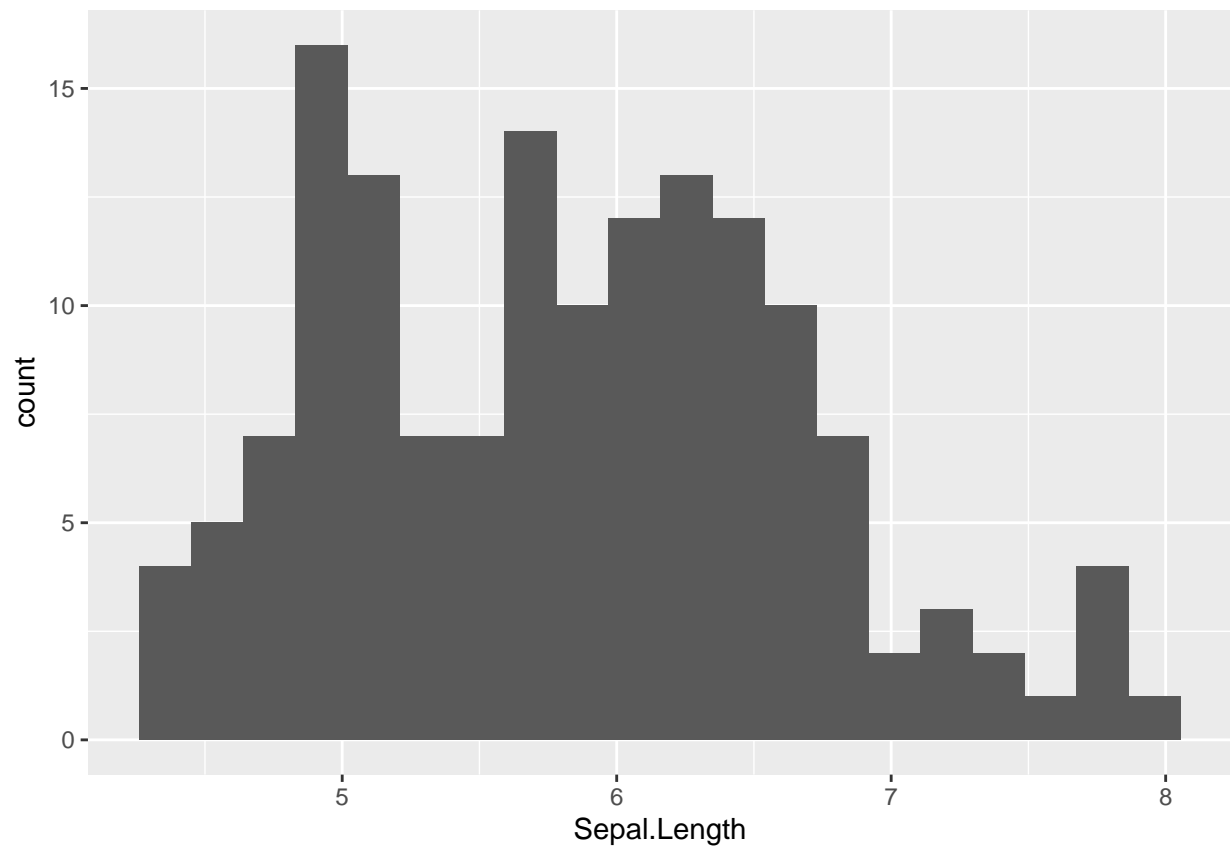
Let's add the percentage in each sector

2.6. Histogram in R

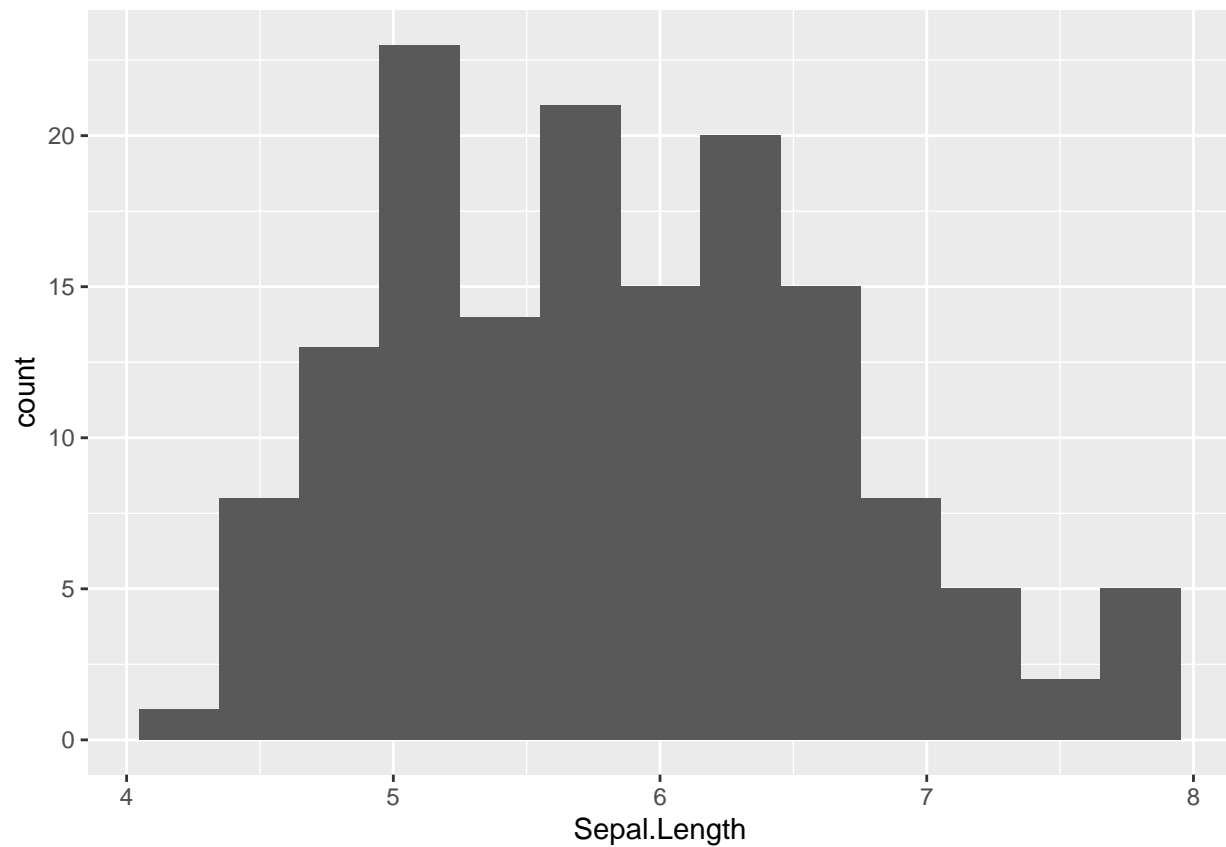
```
#?geom_histogram()
```

2.6.1. Basic histogram

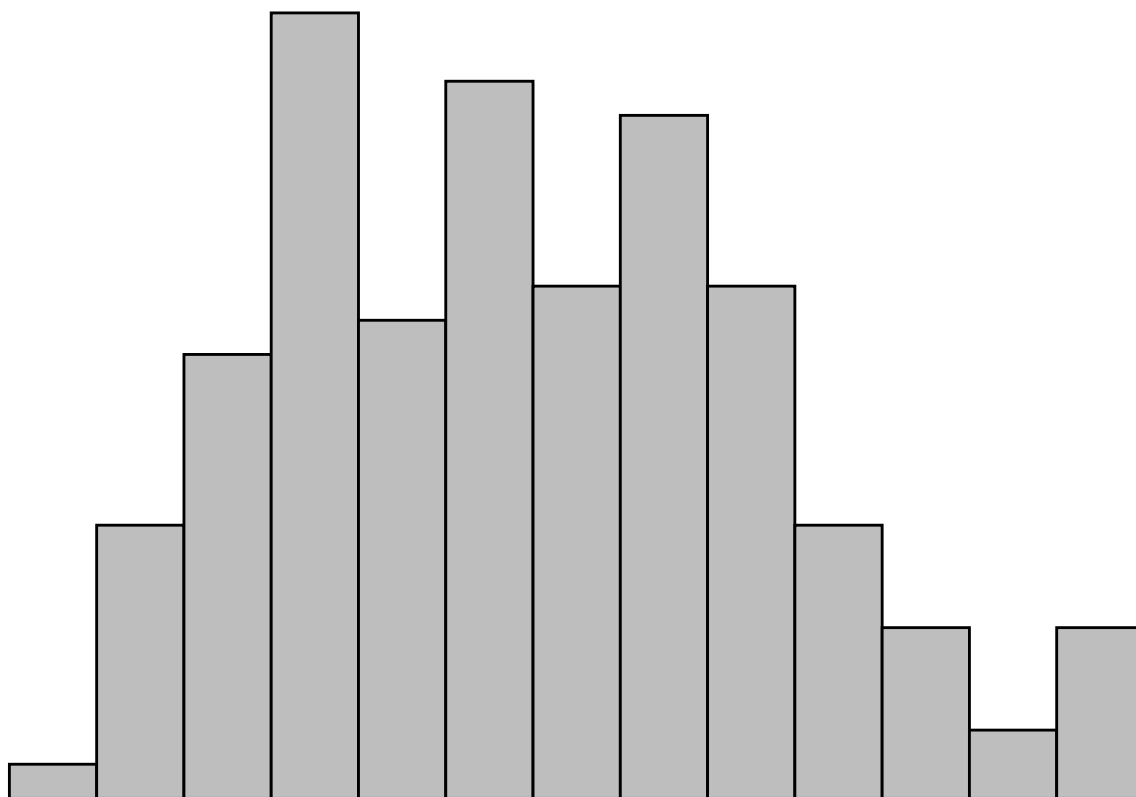
```
ggplot(iris, aes(x=Sepal.Length)) +  
  geom_histogram(bins = 20) # number of bins 20, you can change this
```



```
# Change the width of bins  
ggplot(iris, aes(x=Sepal.Length)) +  
  geom_histogram(binwidth=0.3, bins = 20)
```



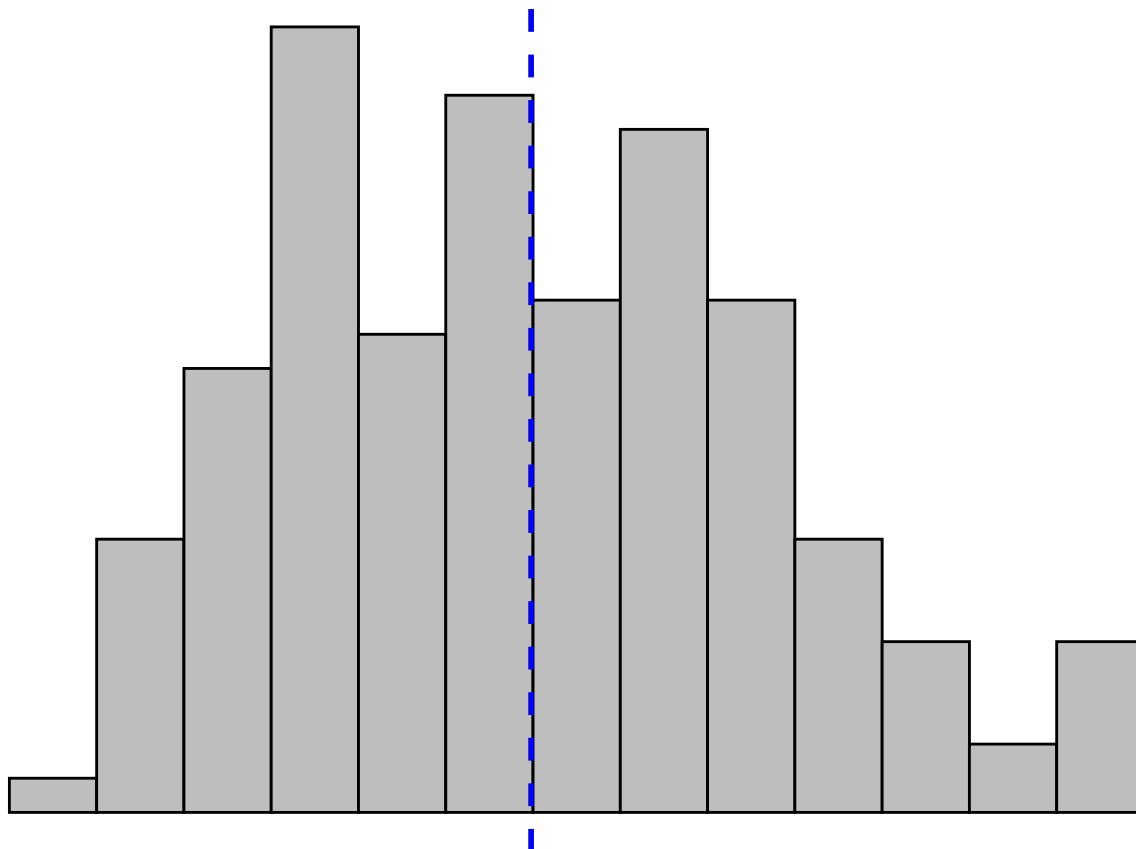
```
# Change colors
#` color means boundary color
#` fill means color fill inside the geometrical object
p <-ggplot(iris, aes(x=Sepal.Length)) +
  geom_histogram(binwidth=0.3,bins = 20, color="black", fill="gray")+
  theme_void()
p
```



Add mean line and density plot on the histogram

```
# Add mean line  
p + geom_vline(aes(xintercept=mean(Sepal.Length)),  
               color="blue", linetype="dashed", size=1)
```

```
## Warning: Using `size` aesthetic for lines was deprecated in ggplot2 3.4.0.  
## i Please use `linewidth` instead.  
## This warning is displayed once every 8 hours.  
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was  
## generated.
```

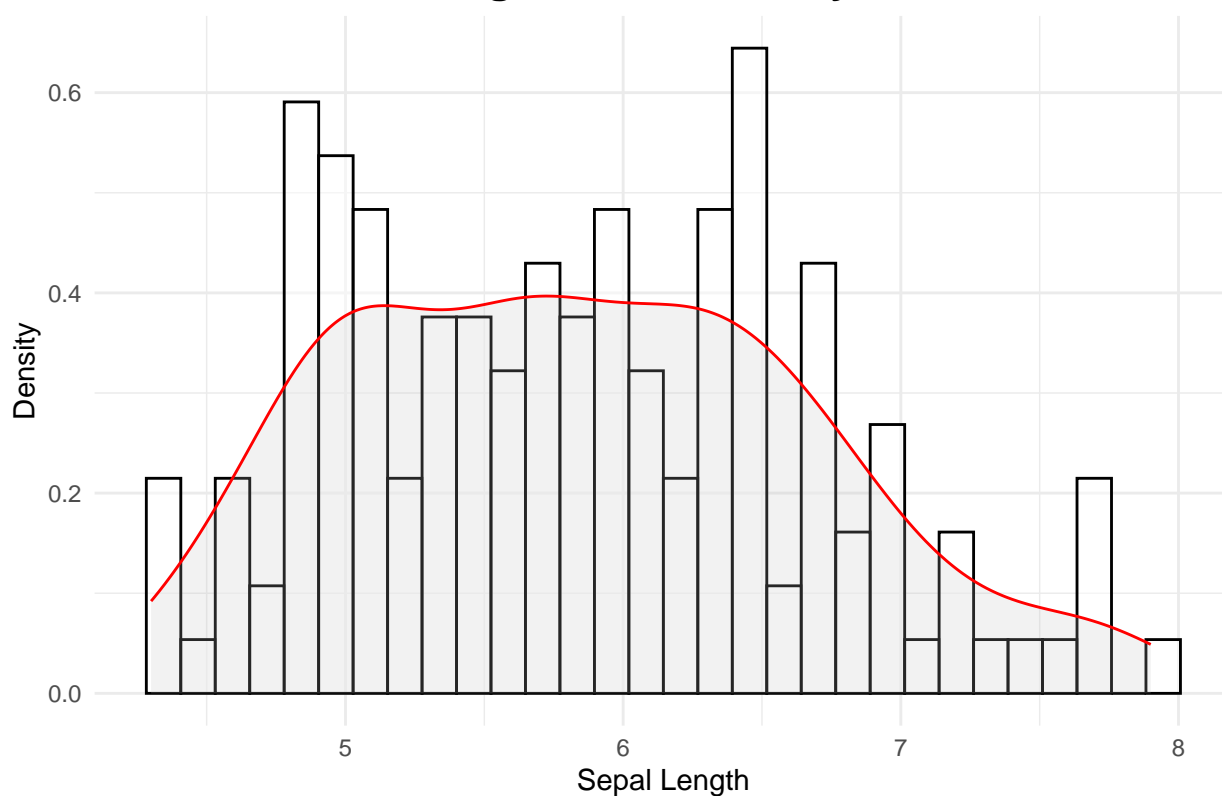


2.6.2. Histogram with density plot

```
ggplot(iris, aes(Sepal.Length)) +
  geom_histogram(aes(y = ..density..),
                 color = "black", fill = "white", alpha = 0.5, bins = 30) + # Adjust transparency and
  geom_density(color = "red", fill = "grey", alpha = 0.2) + # Density curve with transparency
  ggtitle("Histogram with Density Plot") +
  xlab("Sepal Length") +
  ylab("Density") +
  theme_minimal() +
  theme(
    plot.title = element_text(hjust = 0.5, size = 16, face = "bold") # Centered title
  )
```

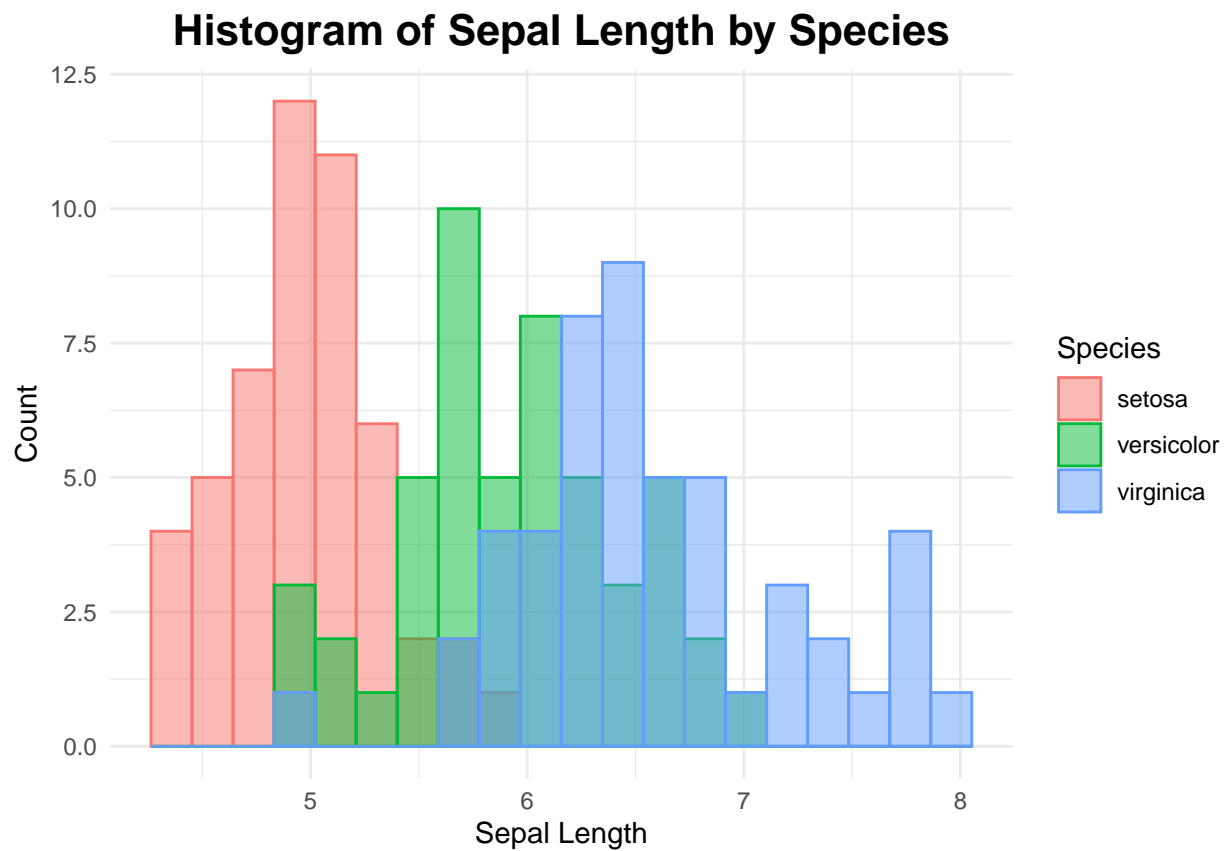
```
## Warning: The dot-dot notation (`..density..`) was deprecated in ggplot2 3.4.0.
## i Please use `after_stat(density)` instead.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
```

Histogram with Density Plot



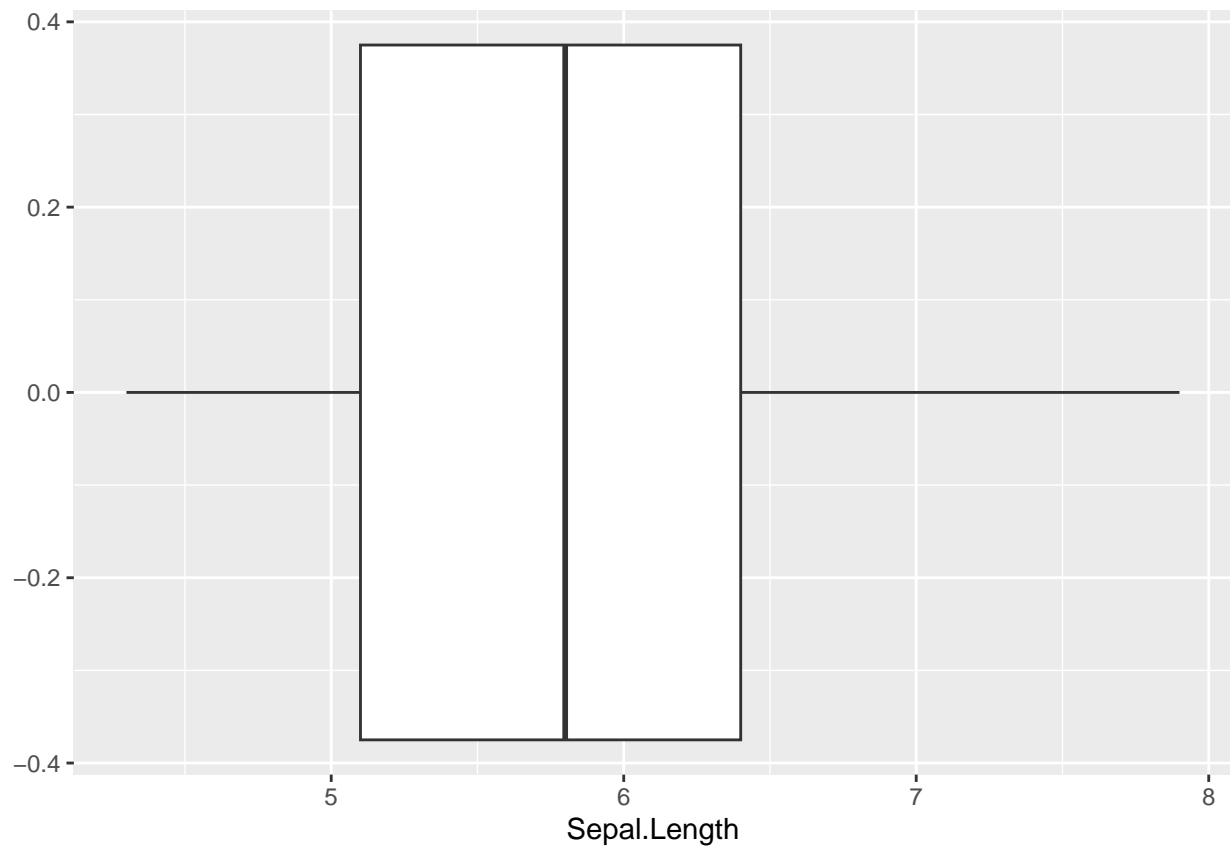
2.6.3. Histogram of a variable with multiple groups

```
# Change histogram plot line colors by groups
ggplot(iris, aes(x = Sepal.Length, color = Species, fill = Species)) +
  geom_histogram(alpha = 0.5, position = "identity", bins = 20) + # Adjust transparency & bins
  ggtitle("Histogram of Sepal Length by Species") +
  xlab("Sepal Length") +
  ylab("Count") +
  theme_minimal() +
  theme(
    plot.title = element_text(hjust = 0.5, size = 16, face = "bold") # Centered title
  )
```

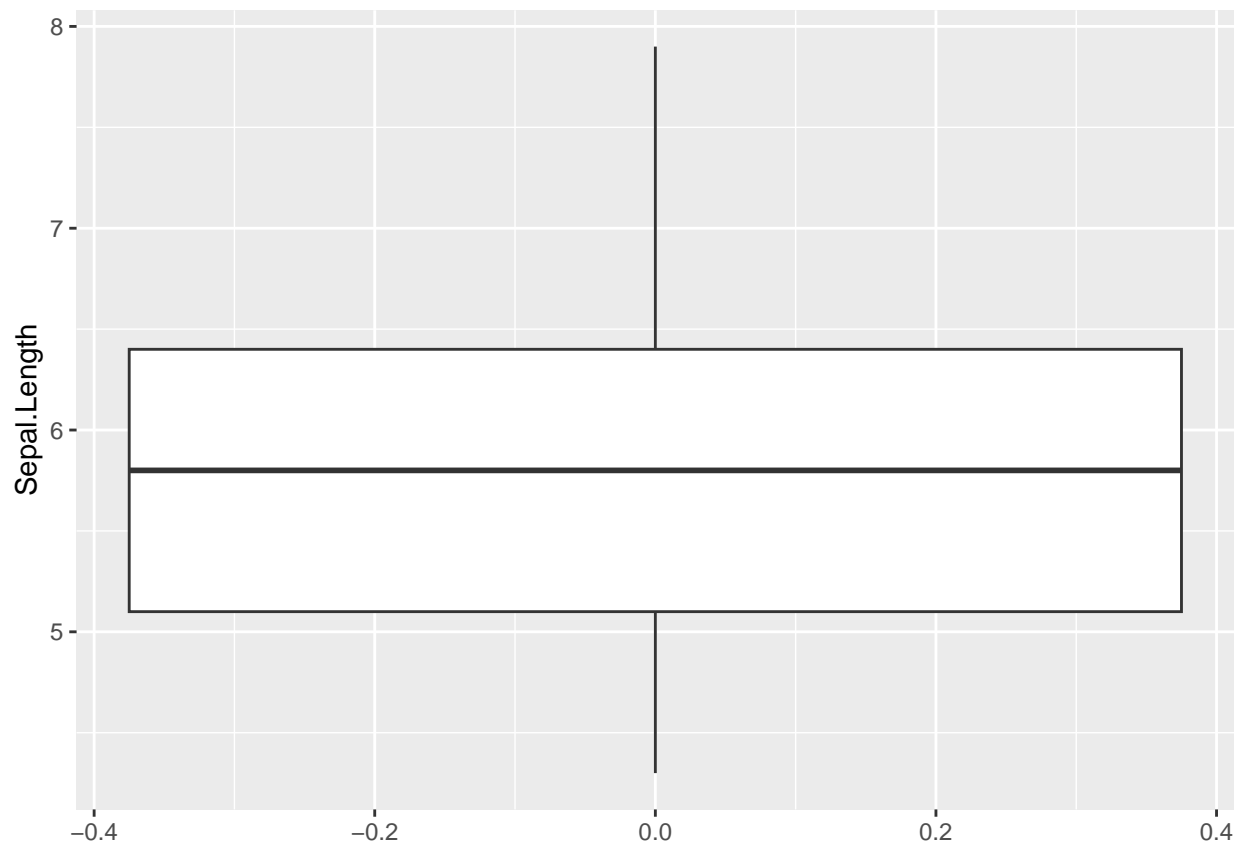



2.7. Box plot

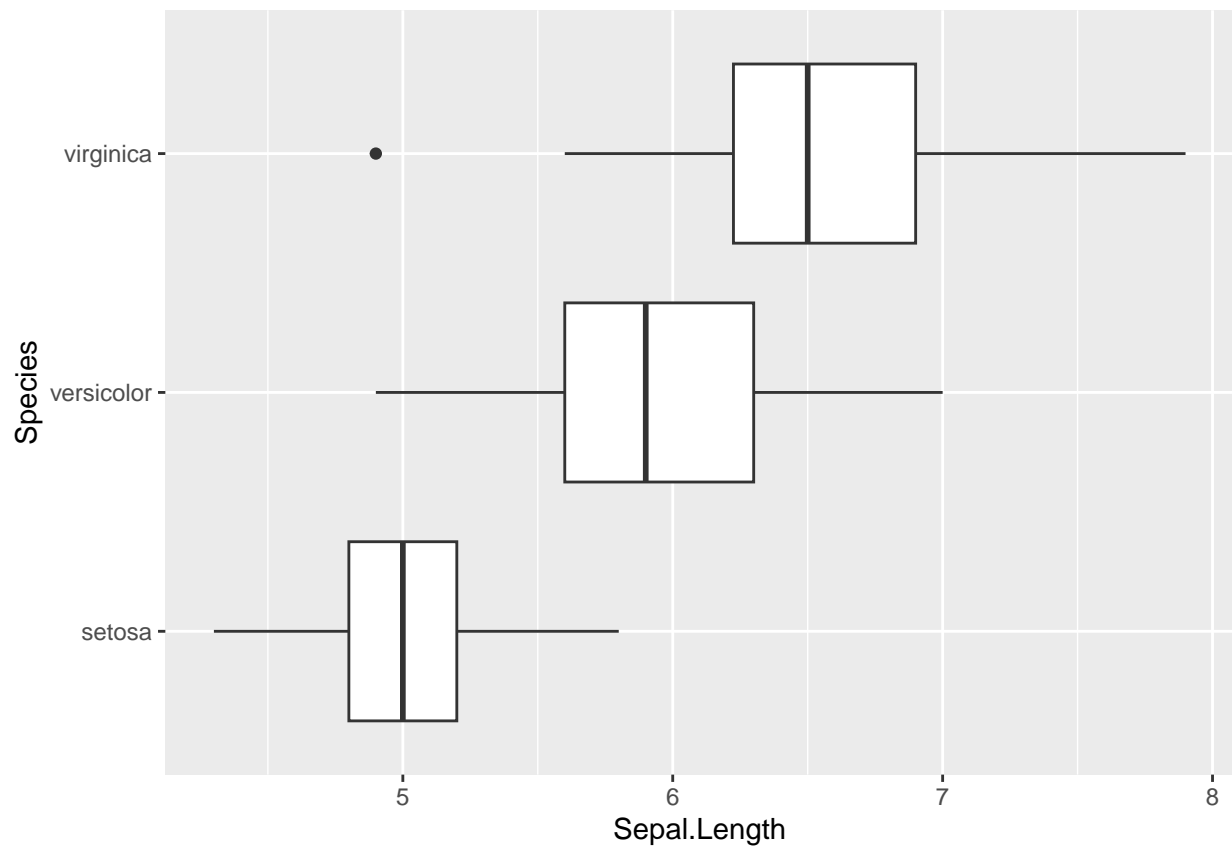
```
# Basic box plot
p <- ggplot(iris, aes(Sepal.Length)) +
  geom_boxplot()
p
```



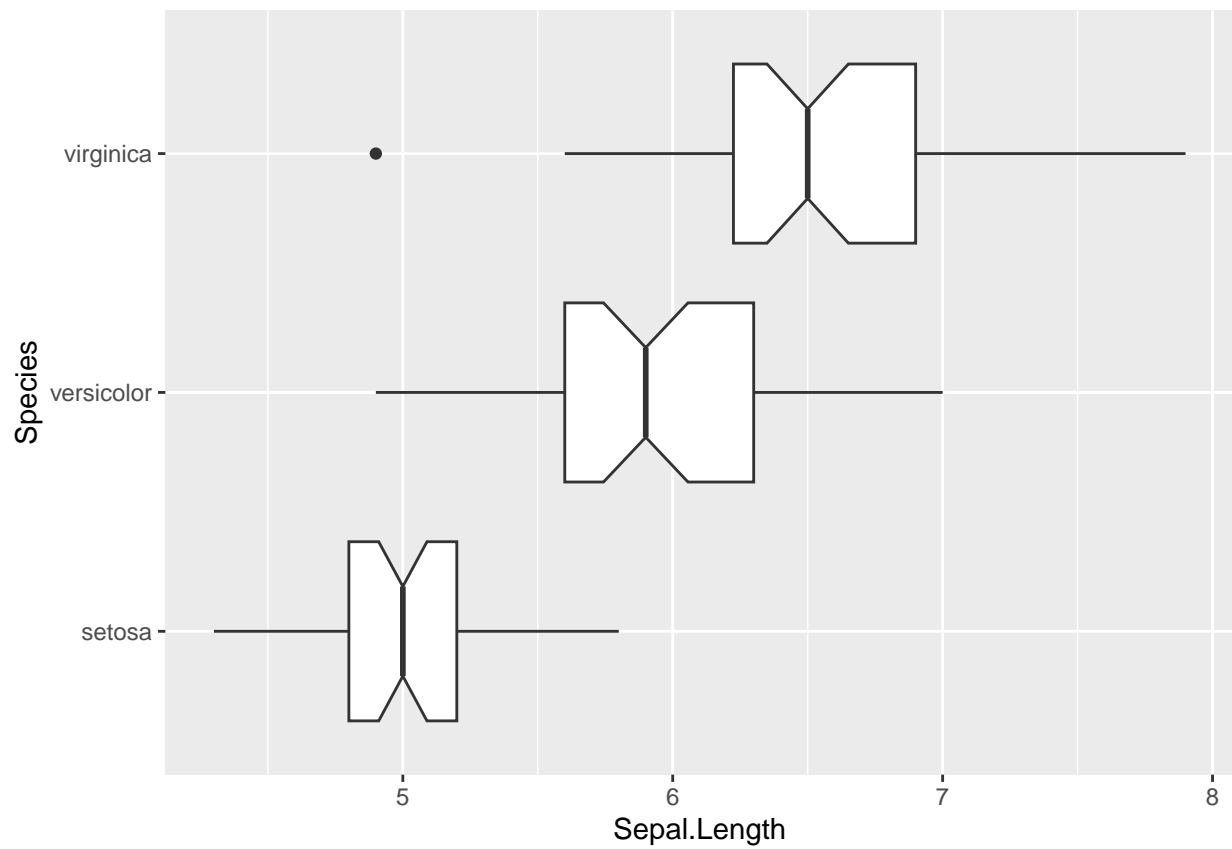
```
# Horizontal box plot  
p + coord_flip()
```



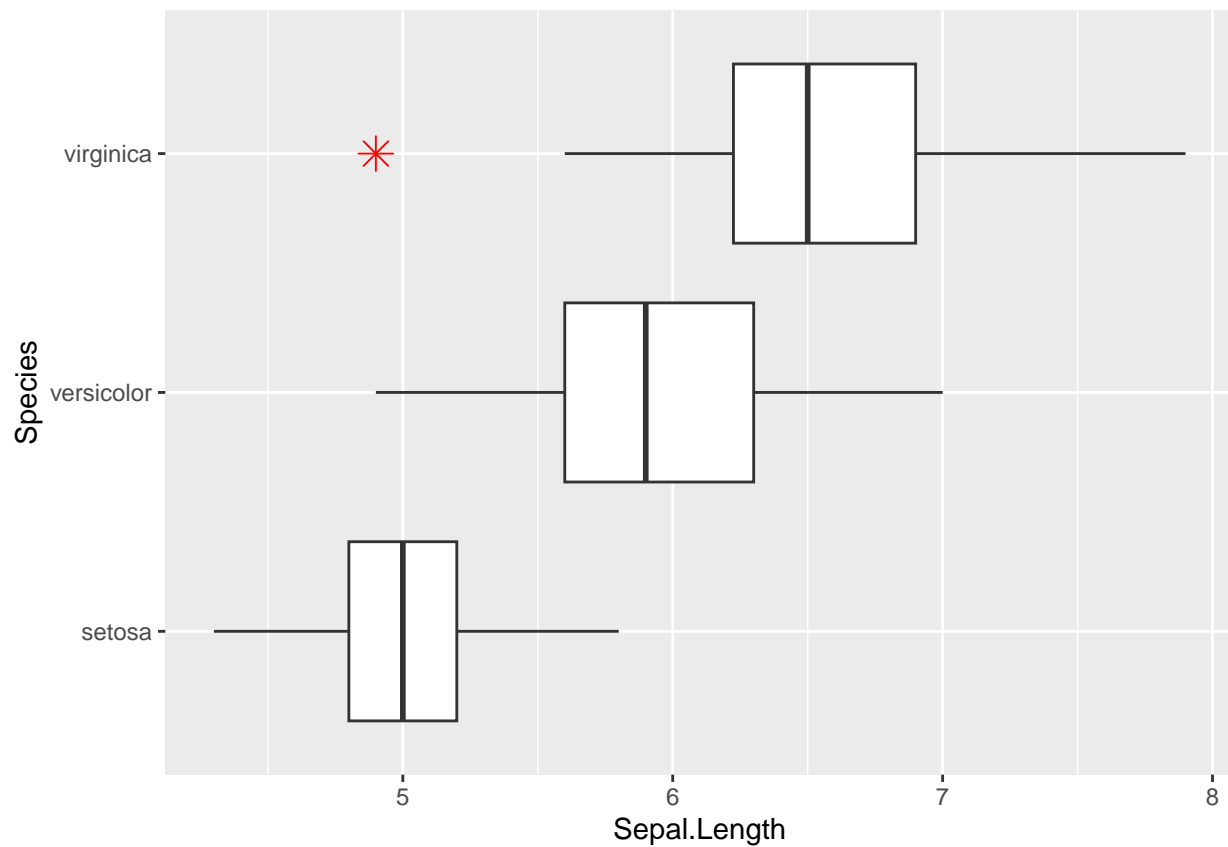
```
#box plot for multiple category  
ggplot(iris, aes(x=Sepal.Length, y=Species)) +  
geom_boxplot()
```



```
# Notched box plot
ggplot(iris, aes(x=Sepal.Length, y=Species)) +
  geom_boxplot(notch=TRUE)
```



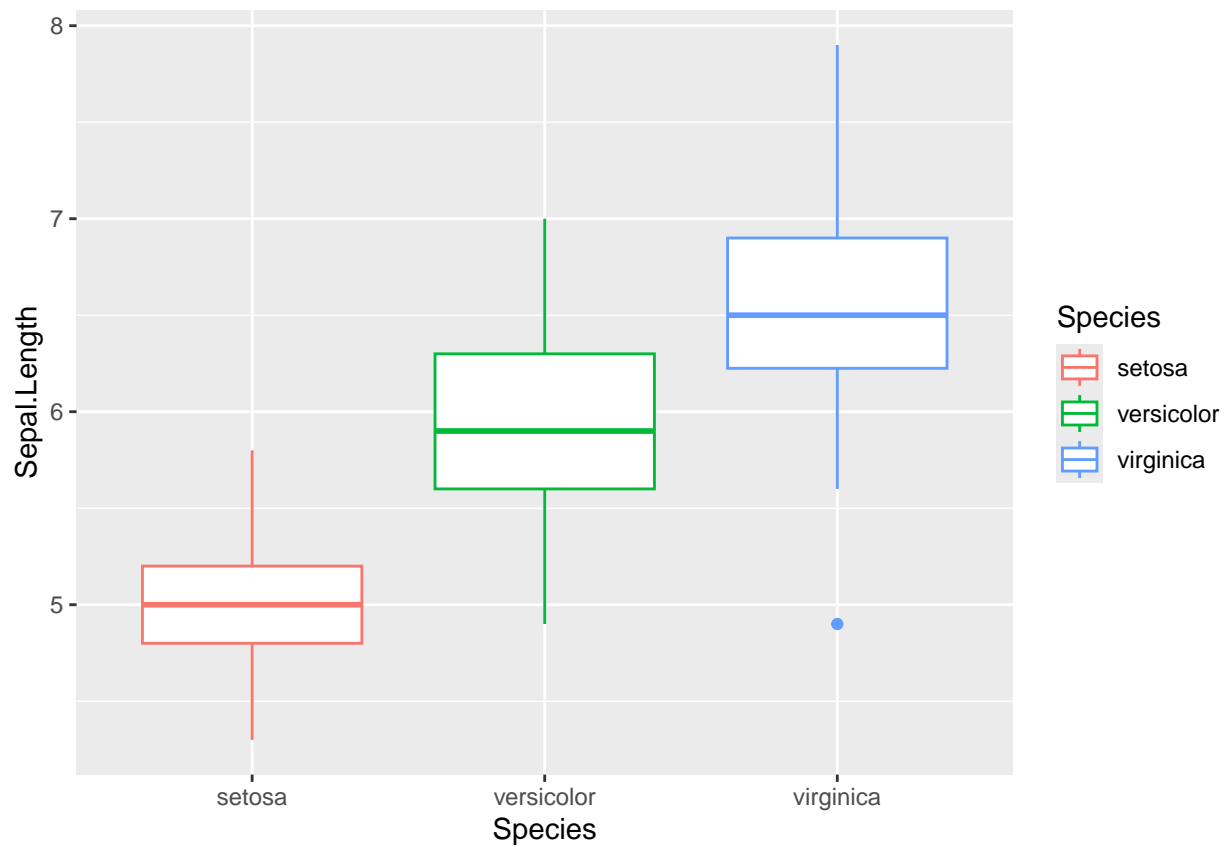
```
# Change outlier, color, shape and size
ggplot(iris, aes(x=Sepal.Length, y=Species)) +
  geom_boxplot(outlier.colour="red", outlier.shape=8,
    outlier.size=4)
```



Change box plot line colors

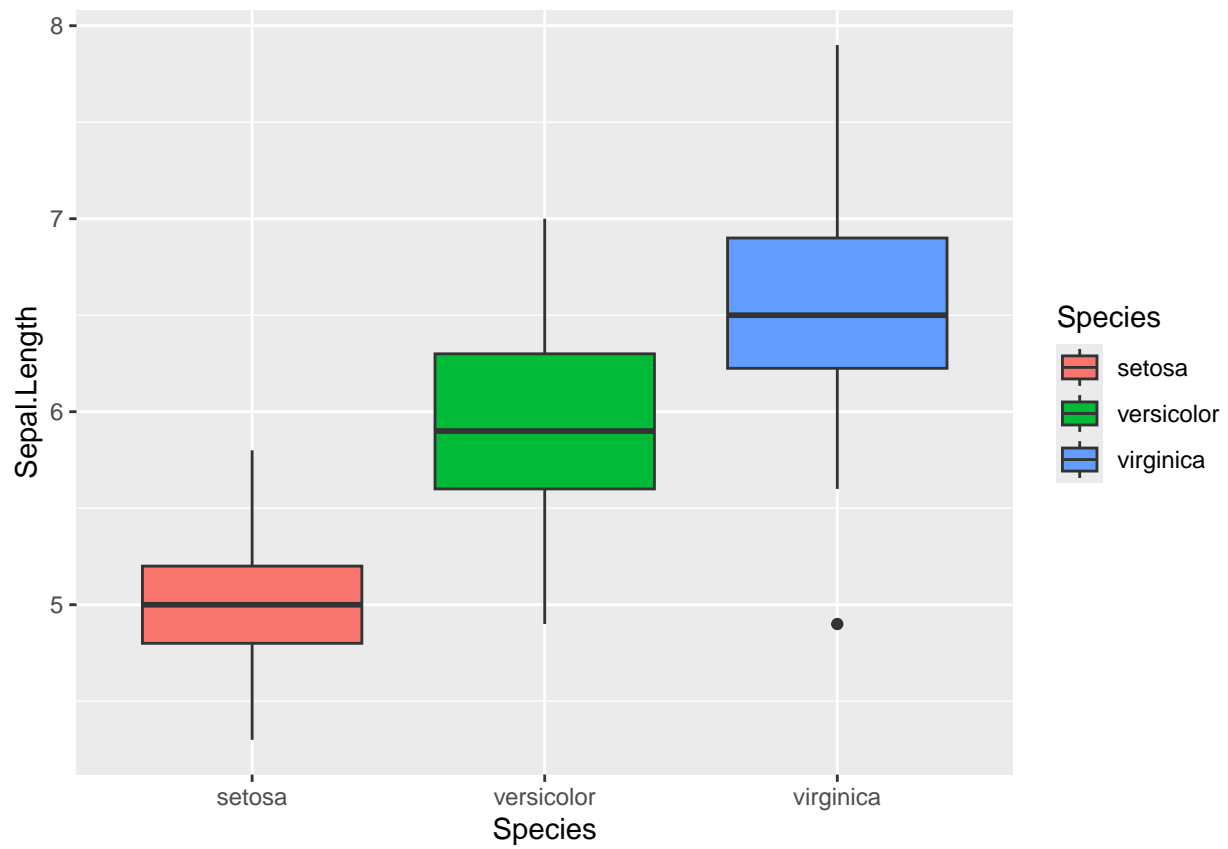
Box plot line colors can be automatically controlled by the level variable :

```
# Change box plot line colors by groups
p<-ggplot(iris, aes(y=Sepal.Length, x=Species, color = Species)) +
  geom_boxplot()
p
```

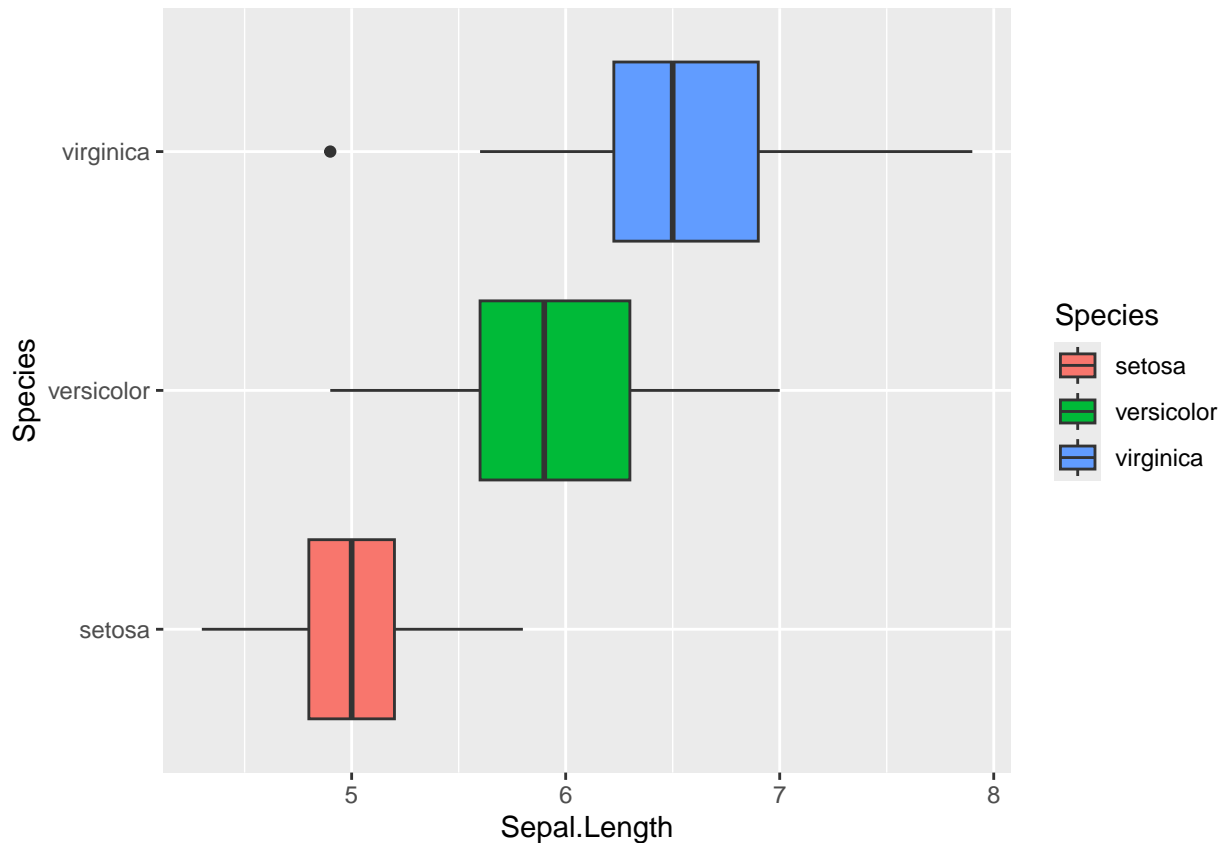


Change box plot fill colors

```
# Change box plot colors by groups
p<- ggplot(iris, aes(y=Sepal.Length, x=Species, fill= Species)) +
  geom_boxplot()
p
```



```
p + coord_flip() # horizontal box plot
```

2.8. Facet Plots in ggplot2

Faceting in ggplot2 allows for visual comparisons across categories. Below are several insightful facet grid examples, demonstrating different types of visualizations that can benefit from faceting.

- Best for understanding distribution differences across categories.
 - Faceted by Species for clearer separation of groups.
1. A Categorical Variable for Faceting Faceting is used to create multiple subplots based on levels of a categorical variable. The variable should be discrete (categorical) or grouped continuous data.
 2. Facet Functions:
 - `facet_wrap(~ variable)`: Creates small multiples in a wrapped format.
 - `facet_grid(rows ~ cols)`: Creates a grid of plots based on row and column factors.
 3. Data & Aesthetic Mappings (`aes()`) A dataset that contains the categorical variable for faceting. The x and y variables should be mapped inside `aes()`.

2.8.1. Faceted Histogram (Comparing Distributions Across Categories)

Question: How does the distribution of Sepal.Length vary across Species?

```
ggplot(iris, aes(x = Sepal.Length)) +
  geom_histogram(color = "black", fill = "steelblue", bins = 20, alpha = 0.7) +
  facet_grid(Species ~ .) + # Facet by Species
```

```
ggtitle("Faceted Histogram of Sepal Length by Species") +
xlab("Sepal Length") +
ylab("Count") +
theme_minimal()
```



Plot like this helps us to compare the distribution as follows:

- Shape: Iris setosa exhibits a symmetric distribution, while versicolor and virginica show slight right skewness.
- Measure of Center: Iris setosa has the shortest average sepal length, followed by versicolor, with virginica having the longest.
- Measure of Dispersion: Iris virginica displays the greatest variability in sepal length, whereas setosa shows the least.

2.8.2. Faceted Scatter Plot (Relationships Across Categories)

- Best for visualizing category-wise correlations.
- Helps detect trends, clusters, or outliers within groups.

Question: How does the relationship between Sepal.Length and Sepal.Width differ across Species?

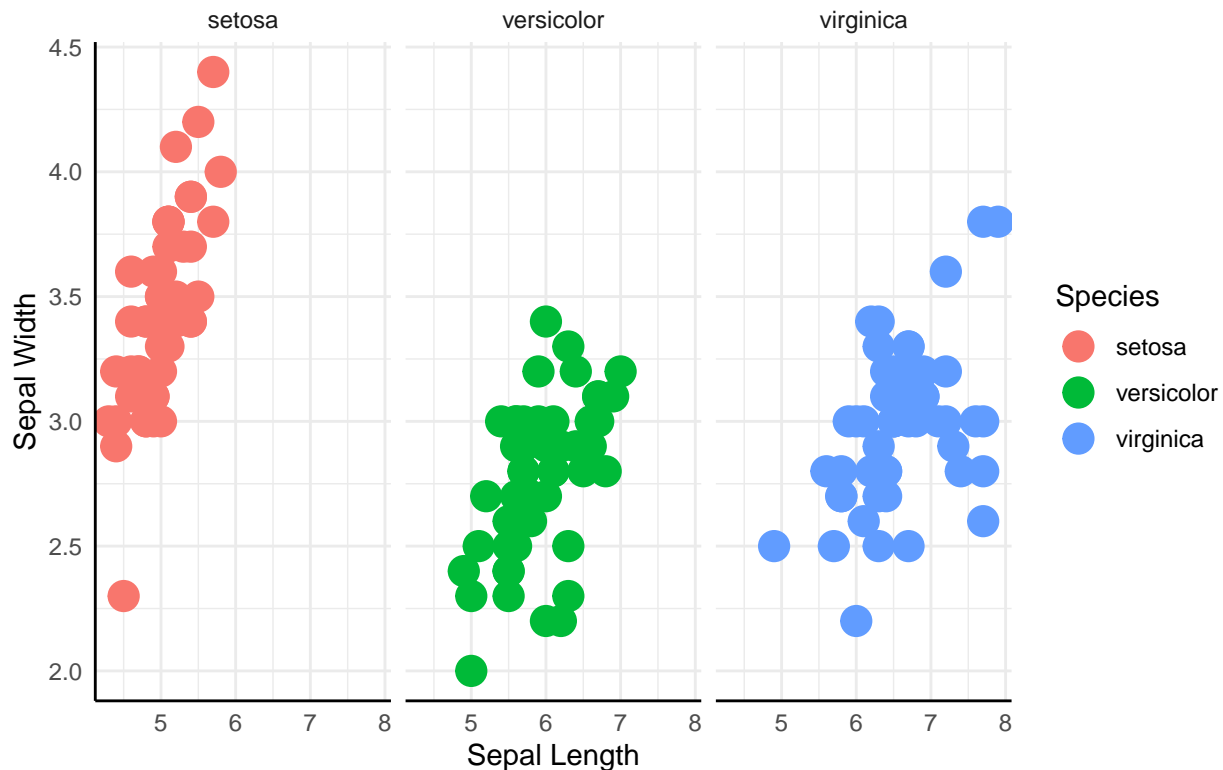
```
ggplot(iris, aes(x = Sepal.Length, y = Sepal.Width, color = Species)) +
geom_point(size = 5) +
facet_wrap(~ Species) + # Creates separate scatter plots per Species
```

```

ggtitle("Sepal Length vs Sepal Width Across Species") +
xlab("Sepal Length") +
ylab("Sepal Width") +
theme_minimal()+
  theme(
    axis.line = element_line(colour = "black", size = 0.5), # Make xy-axes black, size is thickness
    plot.title = element_text(hjust = 0.5, size = 18, face = "bold") # Centered bold title
  )

```

Sepal Length vs Sepal Width Across Species



2.8.3. Faceted Line Chart (Time Series Data Across Groups)

- Best for visualizing trends in different groups over a continuous variable.

Question: How does Sepal.Width trend with Sepal.Length across Iris-Species?

```

str(iris)

## 'data.frame':   150 obs. of  5 variables:
##  $ Sepal.Length: num  5.1 4.9 4.7 4.6 5 5.4 4.6 5 4.4 4.9 ...
##  $ Sepal.Width : num  3.5 3 3.2 3.1 3.6 3.9 3.4 3.4 2.9 3.1 ...
##  $ Petal.Length: num  1.4 1.4 1.3 1.5 1.4 1.7 1.4 1.5 1.4 1.5 ...
##  $ Petal.Width : num  0.2 0.2 0.2 0.2 0.2 0.4 0.3 0.2 0.2 0.1 ...
##  $ Species      : Factor w/ 3 levels "setosa","versicolor",...: 1 1 1 1 1 1 1 1 1 1 ...

```

```
ggplot(iris, aes(x = Sepal.Length, y = Sepal.Width, color = Species)) +
  geom_line() +
  facet_wrap(~ Species) + # One line chart per Species
  ggtitle("Sepal.Length vs Sepal.Width across Iris-Species") +
  xlab("Sepal.Length") +
  ylab("Sepal.Width") +
  theme_minimal()
```



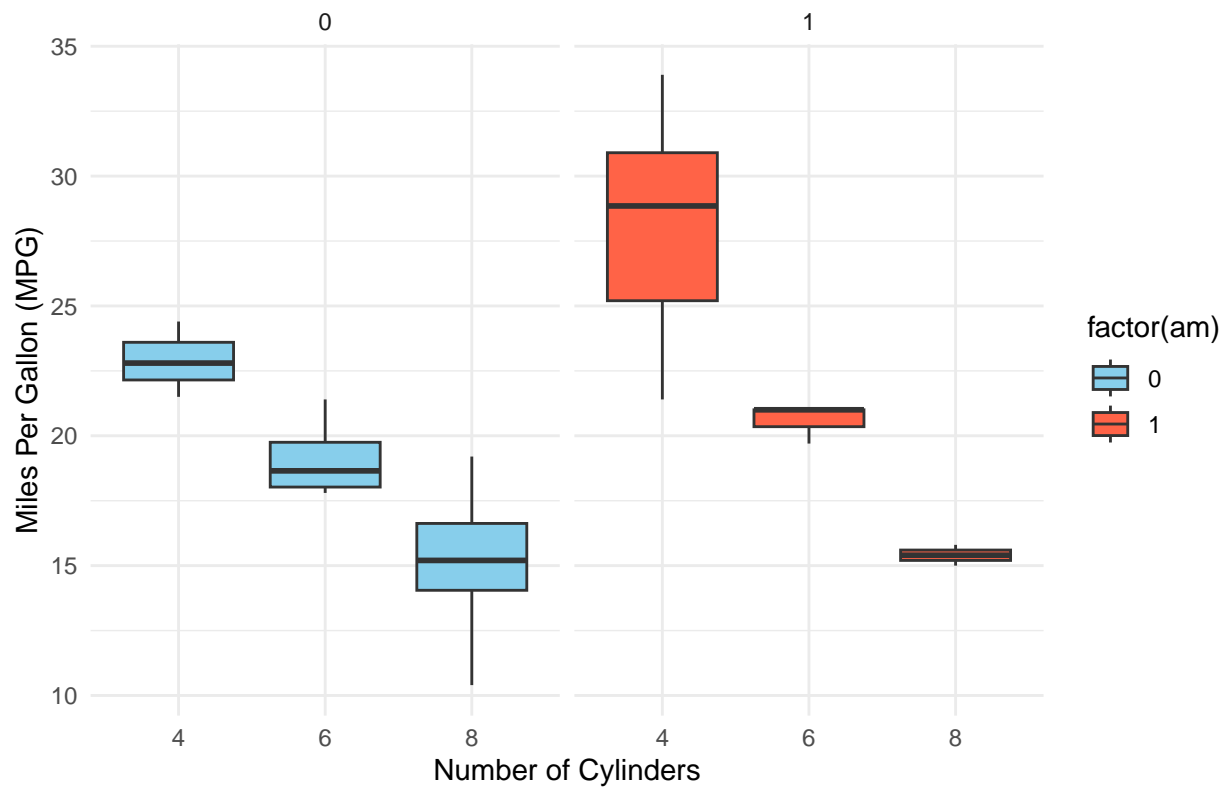
2.8.4. Faceted Box Plot (Comparing Distributions Side-by-Side)

- Best for comparing distributions across two categorical variables.

Question: How does mpg vary across transmission types and cylinders?

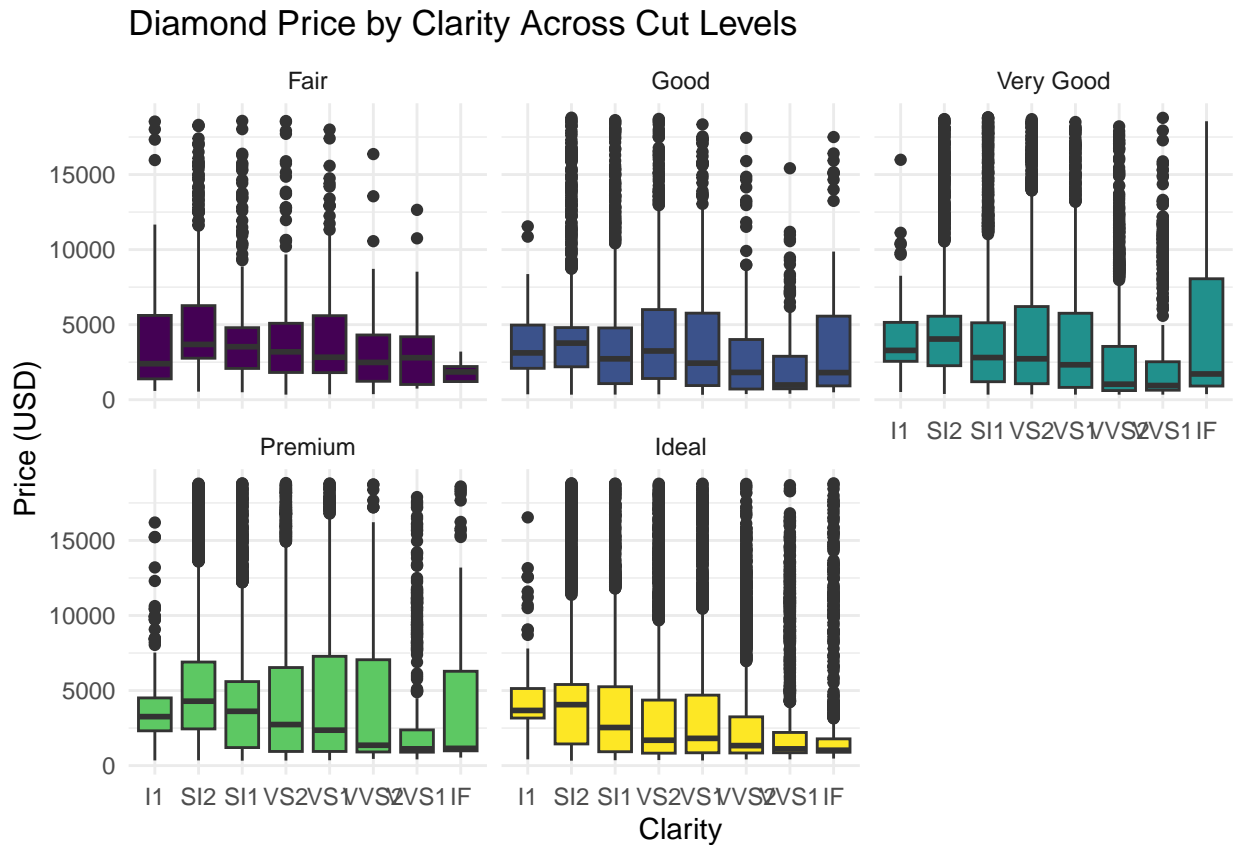
```
ggplot(mtcars, aes(x = factor(cyl), y = mpg, fill = factor(am))) +
  geom_boxplot() +
  facet_grid(. ~ am) + # Faceted by transmission (am: 0 = Automatic, 1 = Manual)
  scale_fill_manual(values = c("skyblue", "tomato")) + # Custom colors
  ggtitle("MPG Distribution Across Cylinders and Transmission Types") +
  xlab("Number of Cylinders") +
  ylab("Miles Per Gallon (MPG)") +
  theme_minimal()
```

MPG Distribution Across Cylinders and Transmission Types



```
data("diamonds")
```

```
ggplot(diamonds, aes(x = clarity, y = price, fill = cut)) +  
  geom_boxplot() +  
  facet_wrap(~ cut) +  
  ggtitle("Diamond Price by Clarity Across Cut Levels") +  
  xlab("Clarity") +  
  ylab("Price (USD)") +  
  theme_minimal() +  
  theme(legend.position = "none")
```

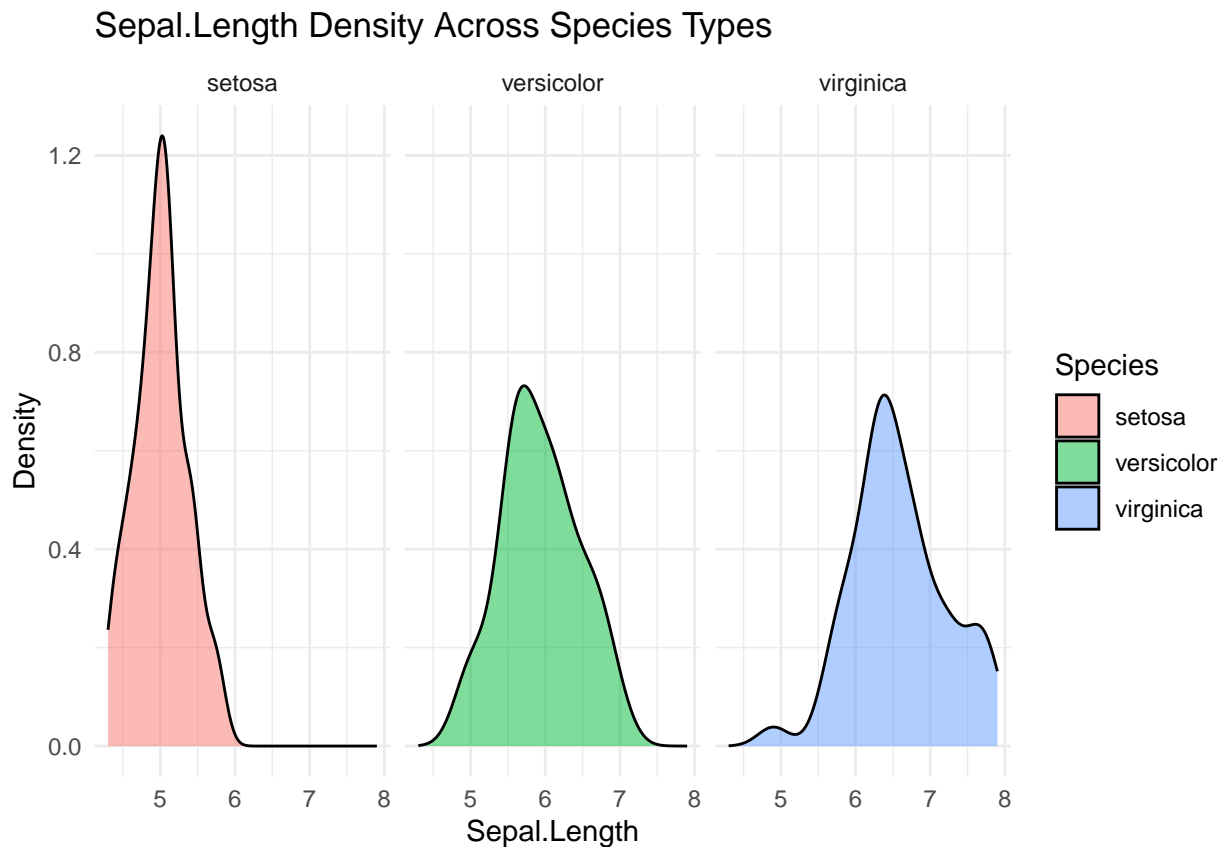


2.8.5. Faceted Density Plot (Comparing Density Distributions)

- Best for comparing smooth distributions instead of histograms.

Question: How does the density of Sepal.Length vary across different Species?

```
ggplot(iris, aes(x = Sepal.Length, fill = Species)) +
  geom_density(alpha = 0.5) +
  facet_wrap(~ Species) + # Separate density plots for each transmission type
  ggtitle("Sepal.Length Density Across Species Types") +
  xlab("Sepal.Length") +
  ylab("Density") +
  theme_minimal()
```



2.8.6. Faceted Bar Chart (Comparing Categorical Counts)

- Best for visualizing frequency differences across categories.
- Faceted by cyl to separate comparisons.

Question: How do gear counts vary across different cylinder types?

```
ggplot(mtcars, aes(x = factor(gear), fill = factor(cyl))) +
  geom_bar(position = "dodge") +
  facet_wrap(~ cyl) + # Separate bar charts per cylinder type
  ggtitle("Gear Count Across Cylinders") +
  xlab("Number of Gears") +
  ylab("Count") +
  theme_minimal()
```

