

Datacamp__Data Visualization with ggplot2 (Part 2)____Coordinates and Facets

dizhen

2019/4/11

Coordinates

1. Coordinates Layer

- Controls plot dimensions
- `coord__`
- `coord_cartesian()`

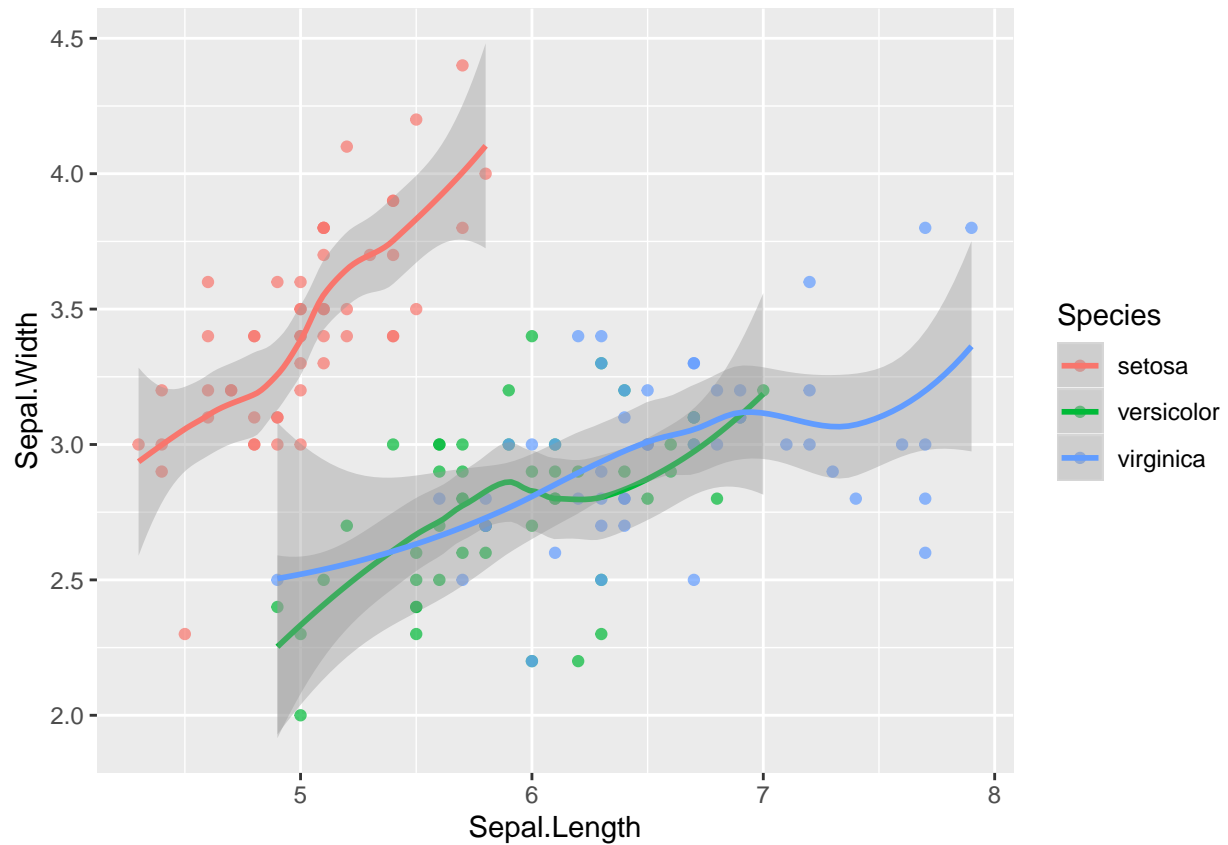
2. Zooming in

- `scale_x_continuous(limits = ...)`
- `xlim()`
- `coord_cartesian(xlim = ...)`

```
library("ggplot2")

# Original Plot
iris.smooth <- ggplot(iris, aes(x = Sepal.Length, y = Sepal.Width, col = Species)) +
  geom_point(alpha = 0.7) +
  geom_smooth()
iris.smooth
```

```
## `geom_smooth()` using method = 'loess' and formula 'y ~ x'
```

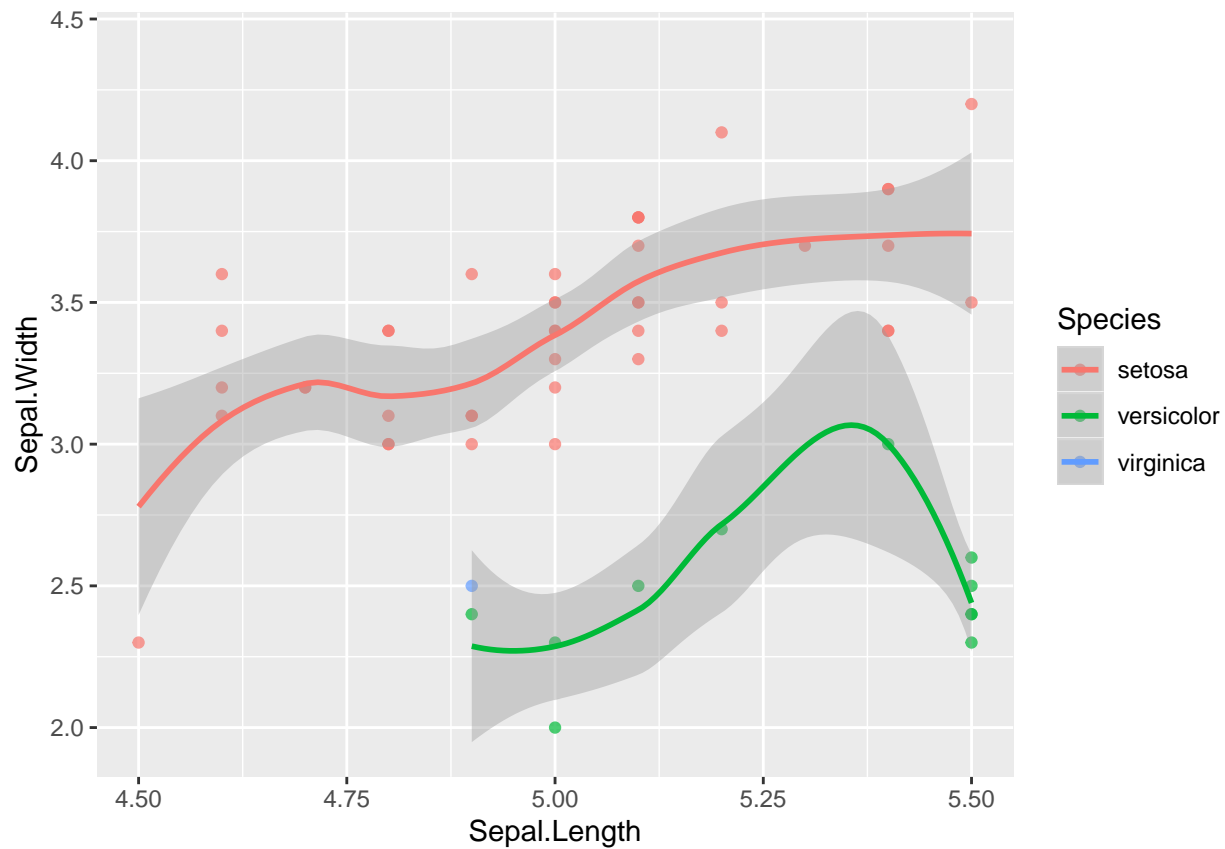


```
# scale_x_continuous
iris.smooth + scale_x_continuous(limits = c(4.5, 5.5))
```

```
## `geom_smooth()` using method = 'loess' and formula 'y ~ x'
```

```
## Warning: Removed 95 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 95 rows containing missing values (geom_point).
```

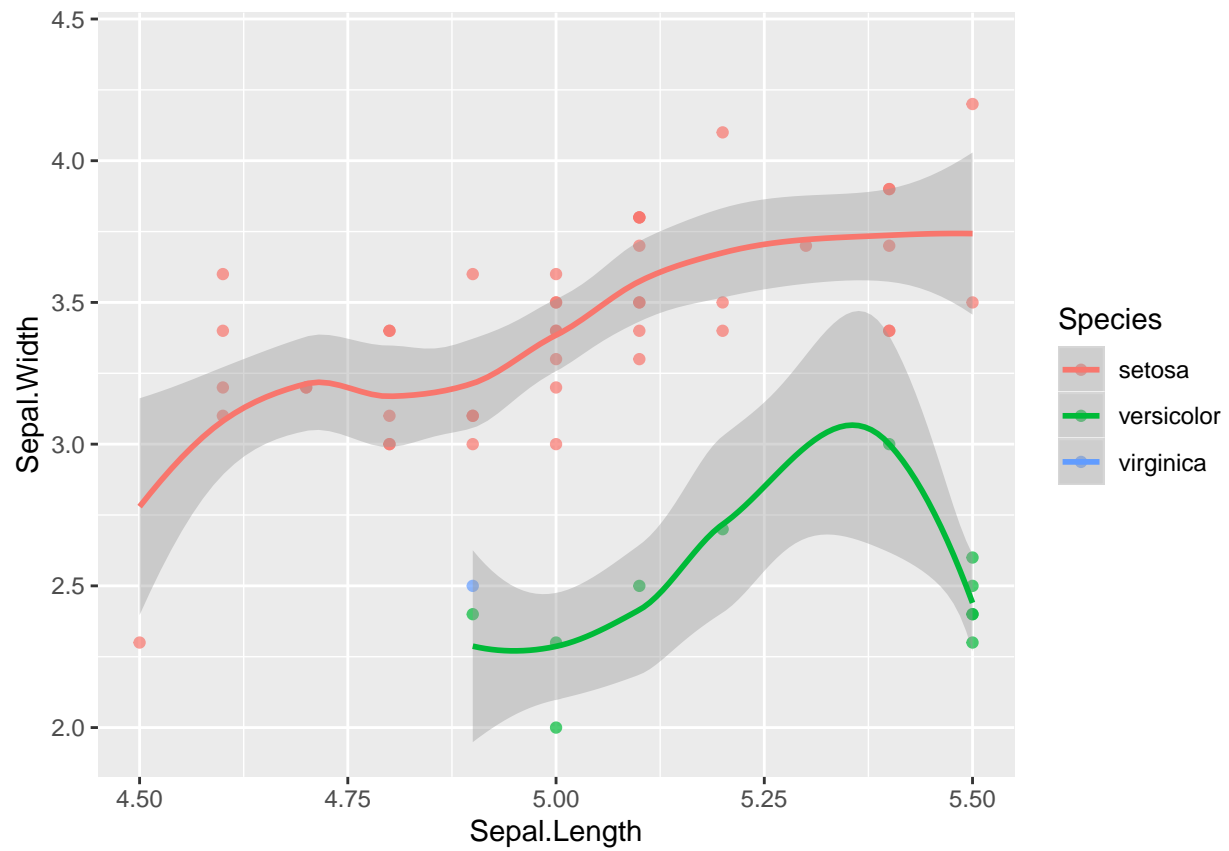


```
# xlim
iris.smooth + xlim(c(4.5, 5.5))
```

```
## `geom_smooth()` using method = 'loess' and formula 'y ~ x'
```

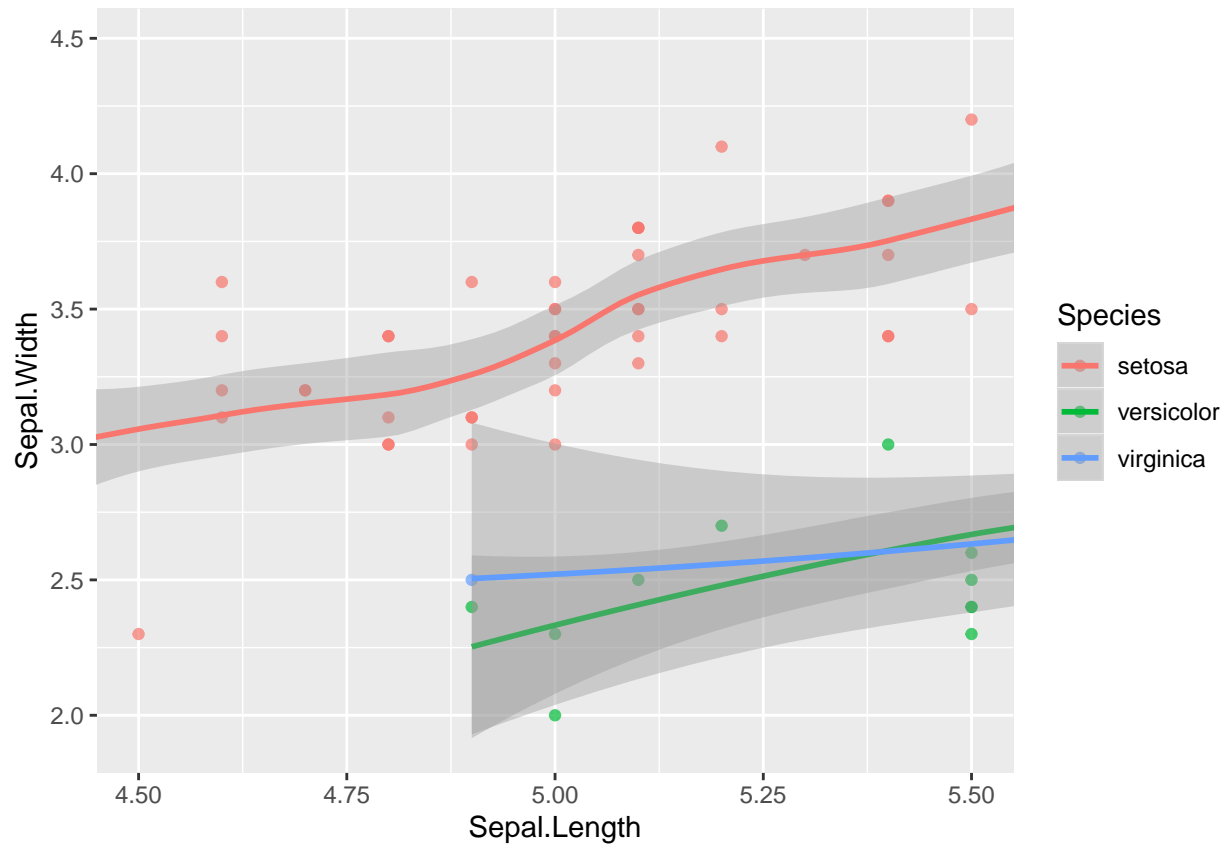
```
## Warning: Removed 95 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 95 rows containing missing values (geom_point).
```



```
# coord_cartesian
iris.smooth + coord_cartesian(xlim = c(4.5, 5.5))
```

```
## `geom_smooth()` using method = 'loess' and formula 'y ~ x'
```



3. Aspect Ratio

- Height-to-width ratio
- Deception
- Standardization attempts
- Typically 1:1

```
# Sunspots
library(reshape2); library(zoo)

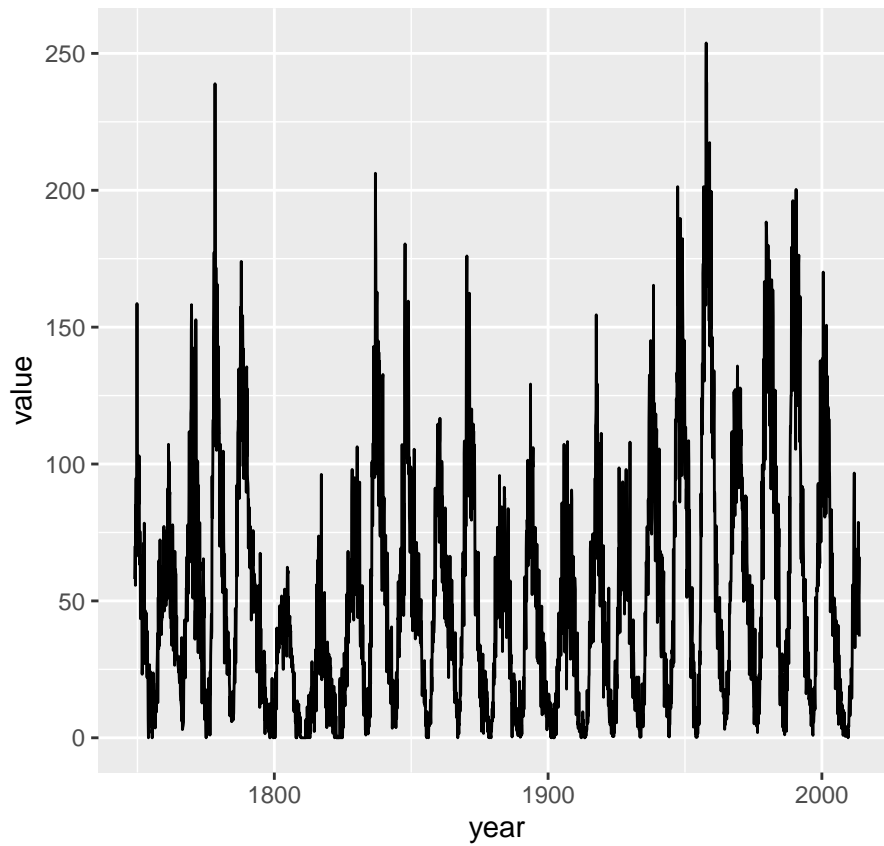
##
## Attaching package: 'zoo'

## The following objects are masked from 'package:base':
##
##   as.Date, as.Date.numeric

sunspots.m <- data.frame(year = index(sunspot.month), value = melt(sunspot.month)$value)

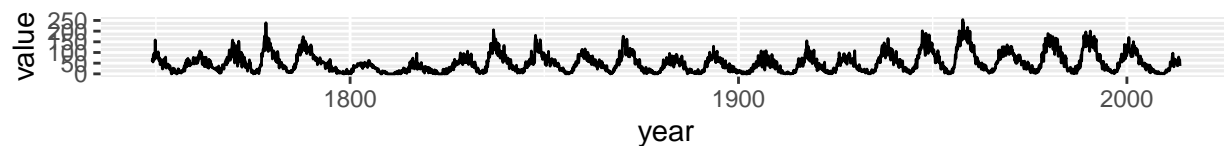
ggplot(sunspots.m, aes(x = year, y = value)) +
  geom_line() +
  coord_equal() # a 1:1 aspect ratio
```

```
## Don't know how to automatically pick scale for object of type ts. Defaulting to continuous.
```



```
ggplot(sunspots.m, aes(x = year, y = value)) +  
  geom_line() +  
  coord_fixed(0.055)
```

```
## Don't know how to automatically pick scale for object of type ts. Defaulting to continuous.
```



Practice

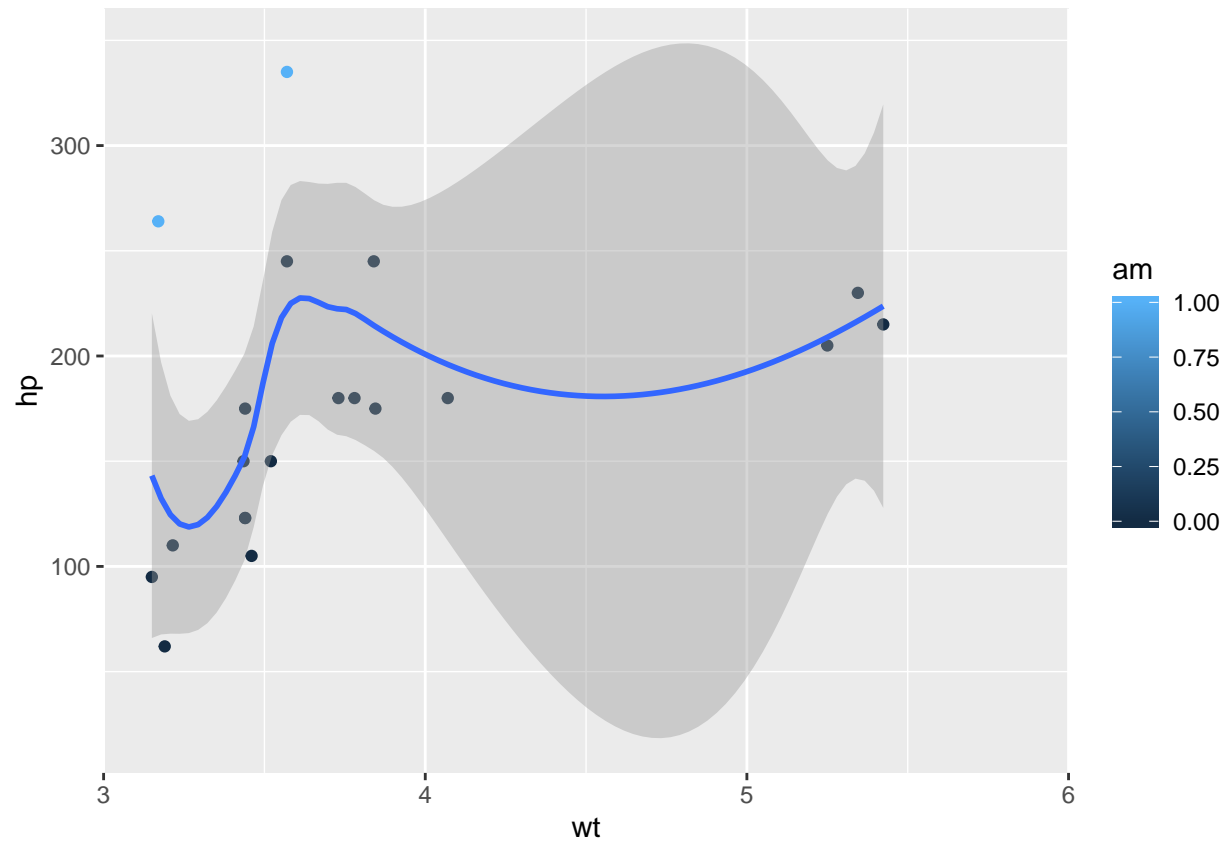
```
# Basic ggplot() command, coded for you
p <- ggplot(mtcars, aes(x = wt, y = hp, col = am)) + geom_point() + geom_smooth()

# Add scale_x_continuous()
p + scale_x_continuous(limit = c(3,6),expand = c(0,0))
```

```
## `geom_smooth()` using method = 'loess' and formula 'y ~ x'
```

```
## Warning: Removed 12 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 12 rows containing missing values (geom_point).
```

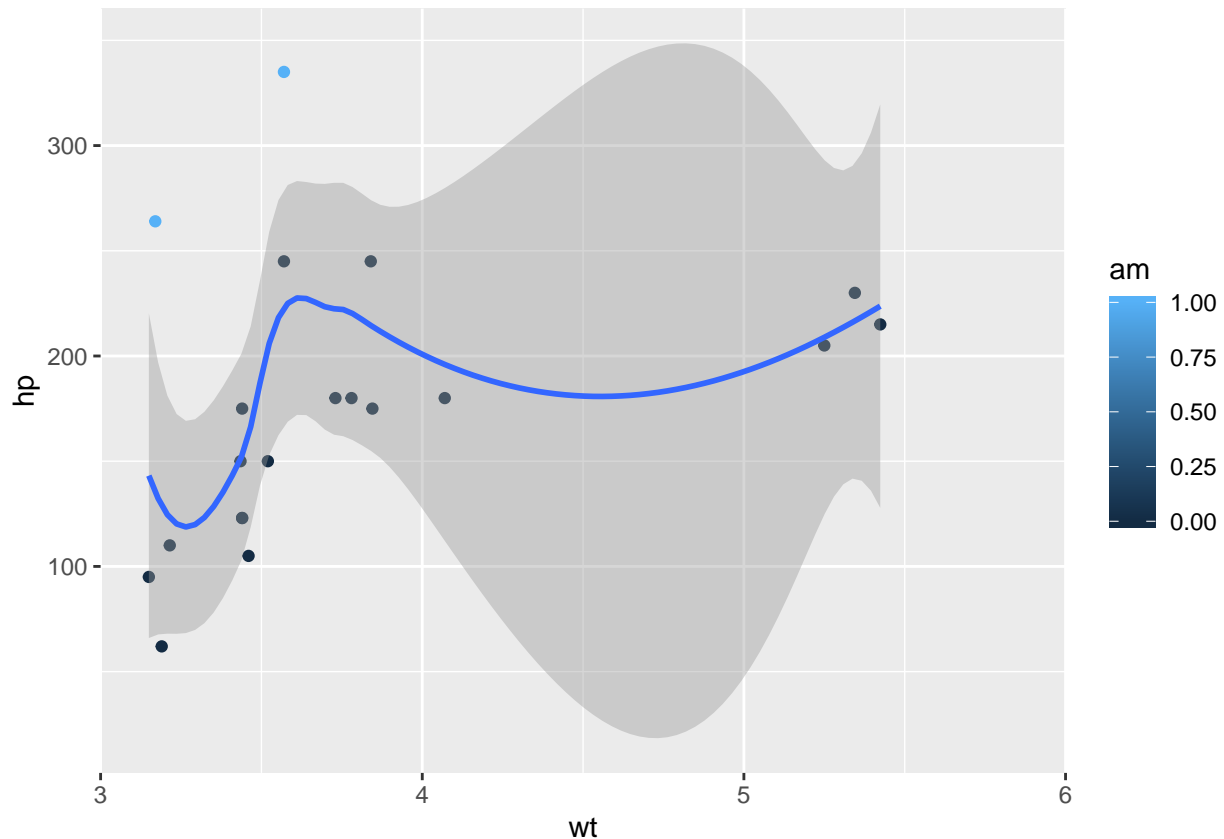


```
# Add coord_cartesian(): the proper way to zoom in
p + scale_x_continuous(limit = c(3,6),expand = c(0,0)) + coord_cartesian(xlim = c(3,6))
```

```
## `geom_smooth()` using method = 'loess' and formula 'y ~ x'
```

```
## Warning: Removed 12 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 12 rows containing missing values (geom_point).
```

Aspect Ratio

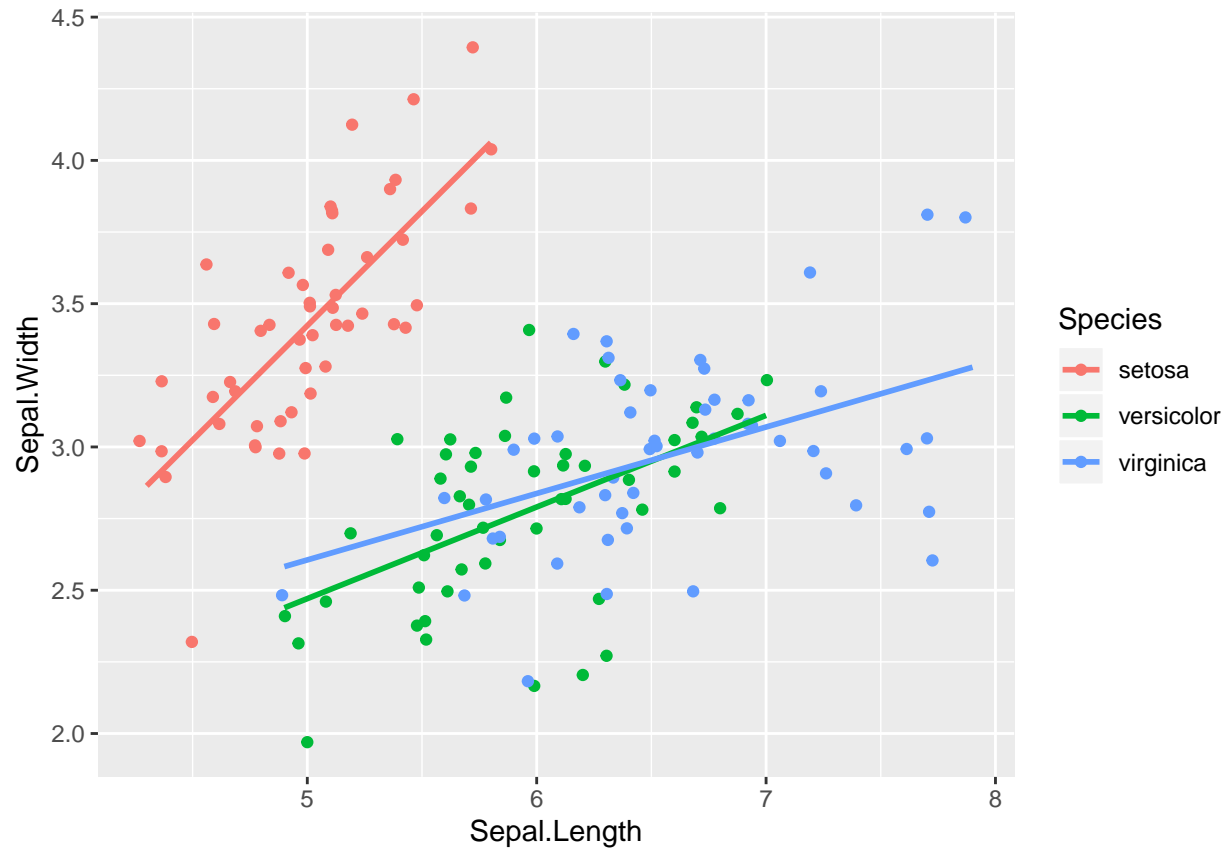
We can set the aspect ratio of a plot with `coord_fixed()` or `coord_equal()`. Both use `ratio = 1` as a default. A 1:1 aspect ratio is most appropriate when two continuous variables are on the same scale, as with the iris dataset.

All variables are measured in centimeters, so it only makes sense that one unit on the plot should be the same physical distance on each axis. This gives a more truthful depiction of the relationship between the two variables since the aspect ratio can change the angle of our smoothing line. This would give an erroneous impression of the data.

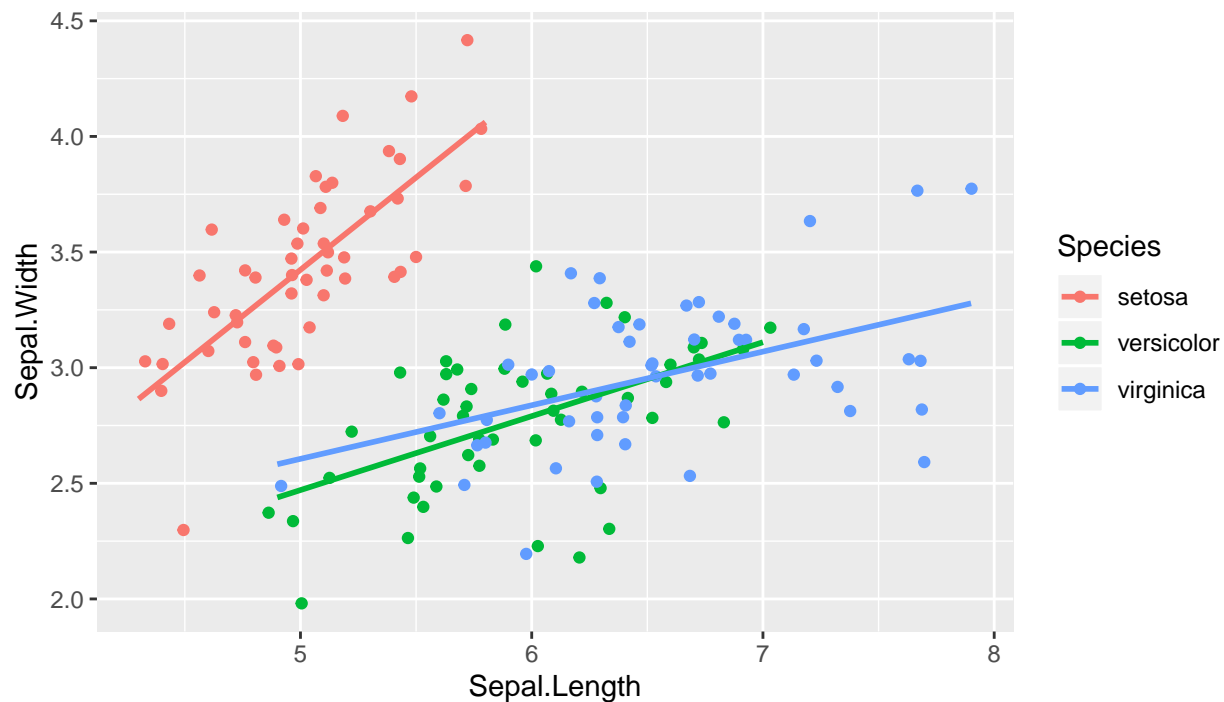
Of course the underlying linear models don't change, but our perception can be influenced by the angle drawn.

```
# Complete basic scatter plot function
base.plot <- ggplot(iris, aes(x = Sepal.Length, y = Sepal.Width, col = Species)) +
  geom_jitter() +
  geom_smooth(method = "lm", se = FALSE)

# Plot base.plot: default aspect ratio
base.plot
```



```
# Fix aspect ratio (1:1) of base.plot  
base.plot + coord_equal()
```



Pie Charts

The `coord_polar()` function converts a planar x-y Cartesian plot to polar coordinates. This can be useful if you are producing pie charts.

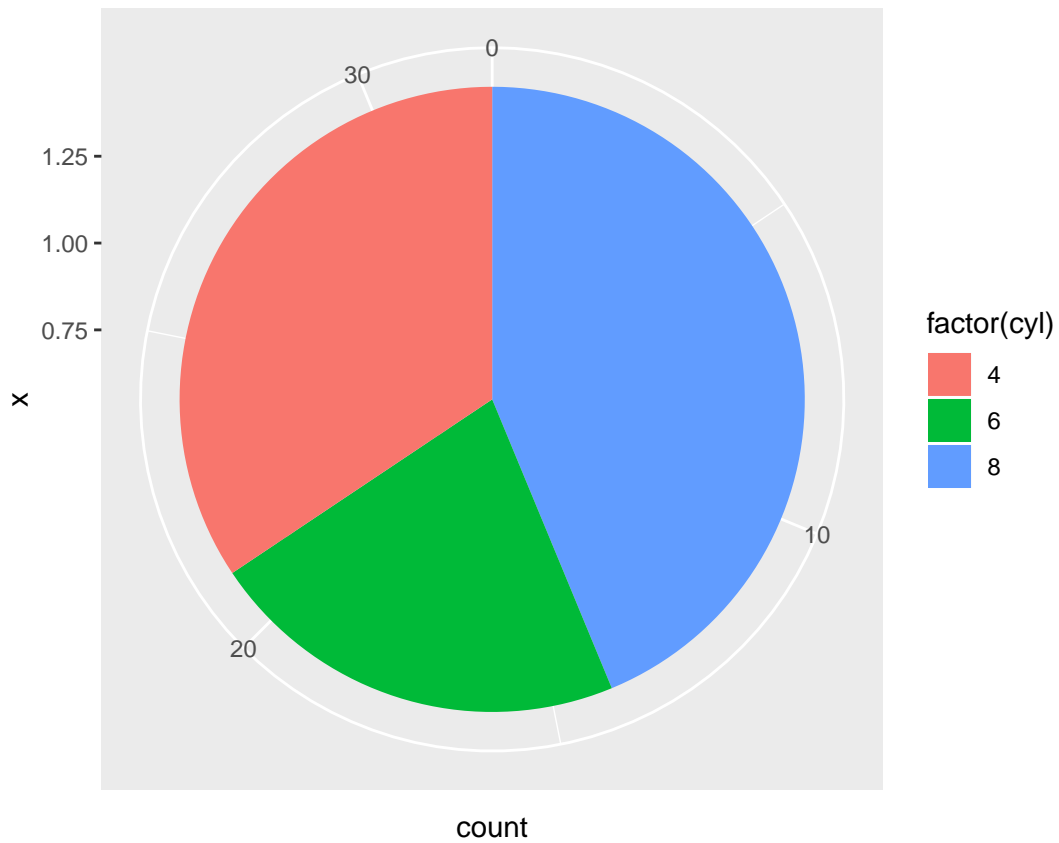
We can imagine two forms for pie charts - the typical filled circle, or a colored ring.

As an example, consider the stacked bar chart shown in the viewer. Imagine that we just take the y axis on the left and bend it until it loops back on itself, while expanding the right side as we go along. We'd end up with a pie chart - it's simply a bar chart transformed onto a polar coordinate system.

Typical pie charts omit all of the non-data ink.

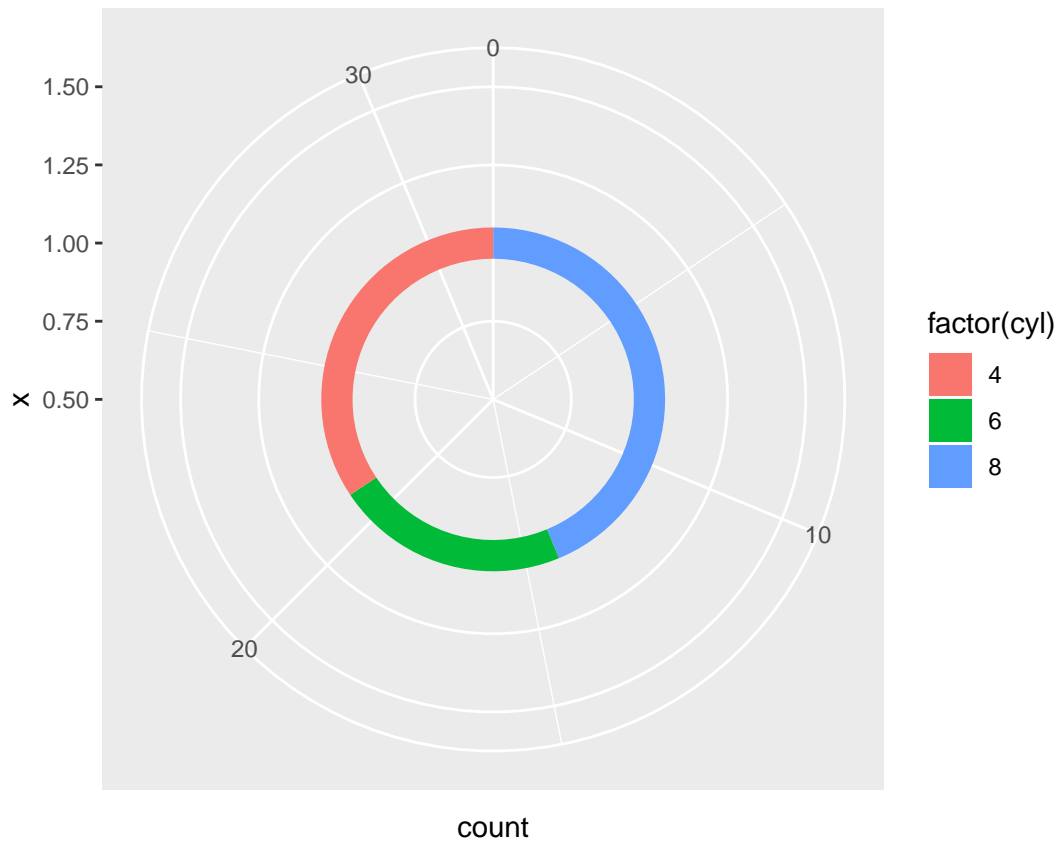
```
# Create a stacked bar plot: wide.bar
wide.bar <- ggplot(mtcars, aes(x = 1, fill = factor(cyl))) +
  geom_bar()

# Convert wide.bar to pie chart
wide.bar +
  coord_polar(theta = "y")
```



```
# Create stacked bar plot: thin.bar
thin.bar <- ggplot(mtcars, aes(x = 1, fill = factor(cyl))) +
  geom_bar(width = 0.1) +
  scale_x_continuous(limits = c(0.5, 1.5))

# Convert thin.bar to "ring" type pie chart
thin.bar +
  coord_polar(theta = "y")
```



Facets

1. Facets

- Straight-forward yet useful
- Concept of Small Multiples

Edward Tufte -Visualization of Quantitative Information, 1983

```
library(tidyr)
```

```
##
## Attaching package: 'tidyr'

## The following object is masked from 'package:reshape2':
##
## smiths
```

```
library(tidyverse)
```

```
## -- Attaching packages ----- tidyverse 1.2.1 --
```

```
## v tibble 2.1.3      v dplyr 0.8.3
## v readr  1.3.1      v stringr 1.4.0
## v purrr  0.3.2      v forcats 0.4.0

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
```

```
iris$Flower <- 1:nrow(iris)

iris.wide <- iris %>%
  gather(key, value, -Flower, -Species) %>%
  separate(key, c("Part", "Measure"), "\\.") %>%
  spread(Measure, value)

iris.tidy <- iris %>%
  gather(key, Value, -Flower, -Species) %>%
  separate(key, c("Part", "Measure"), "\\.")

iris.wide2 <- rbind(
  data.frame(Measure = "Length", Part = "Petal", Setosa = iris[iris$Species == "setosa", "Petal.Length"],
  data.frame(Measure = "Width", Part = "Petal", Setosa = iris[iris$Species == "setosa", "Petal.Width"],
  data.frame(Measure = "Length", Part = "Sepal", Setosa = iris[iris$Species == "setosa", "Sepal.Length"],
  data.frame(Measure = "Width", Part = "Sepal", Setosa = iris[iris$Species == "setosa", "Sepal.Width"],
```

For iris.wide2, each plot has separate y axis. Three different plot functions

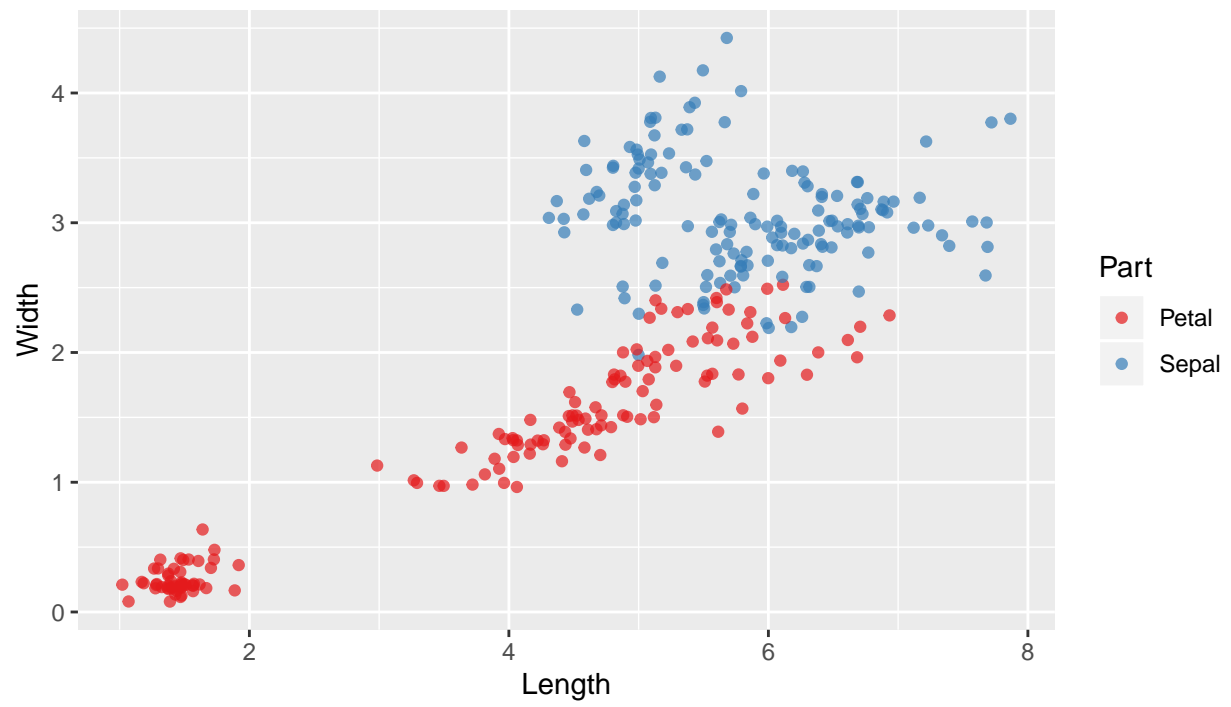
```
head(iris.wide2)
```

```
## Measure Part Setosa Versicolor Virginica
## 1 Length Petal 1.4 4.7 6.0
## 2 Length Petal 1.4 4.5 5.1
## 3 Length Petal 1.3 4.9 5.9
## 4 Length Petal 1.5 4.0 5.6
## 5 Length Petal 1.4 4.6 5.8
## 6 Length Petal 1.7 4.5 6.6
```

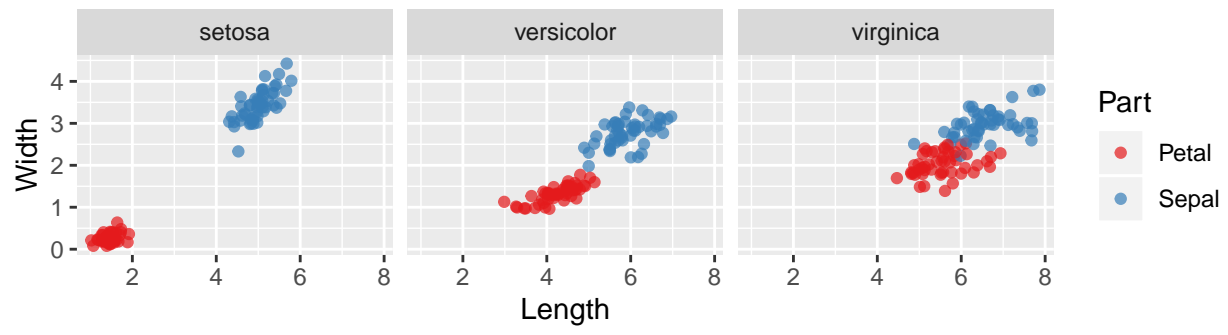
```
str(iris.wide2)
```

```
## 'data.frame': 200 obs. of 5 variables:
## $ Measure : Factor w/ 2 levels "Length","Width": 1 1 1 1 1 1 1 1 1 1 ...
## $ Part : Factor w/ 2 levels "Petal","Sepal": 1 1 1 1 1 1 1 1 1 1 ...
## $ Setosa : num 1.4 1.4 1.3 1.5 1.4 1.7 1.4 1.5 1.4 1.5 ...
## $ Versicolor: num 4.7 4.5 4.9 4.4 4.6 4.5 4.7 3.3 4.6 3.9 ...
## $ Virginica : num 6 5.1 5.9 5.6 5.8 6.6 4.5 6.3 5.8 6.1 ...
```

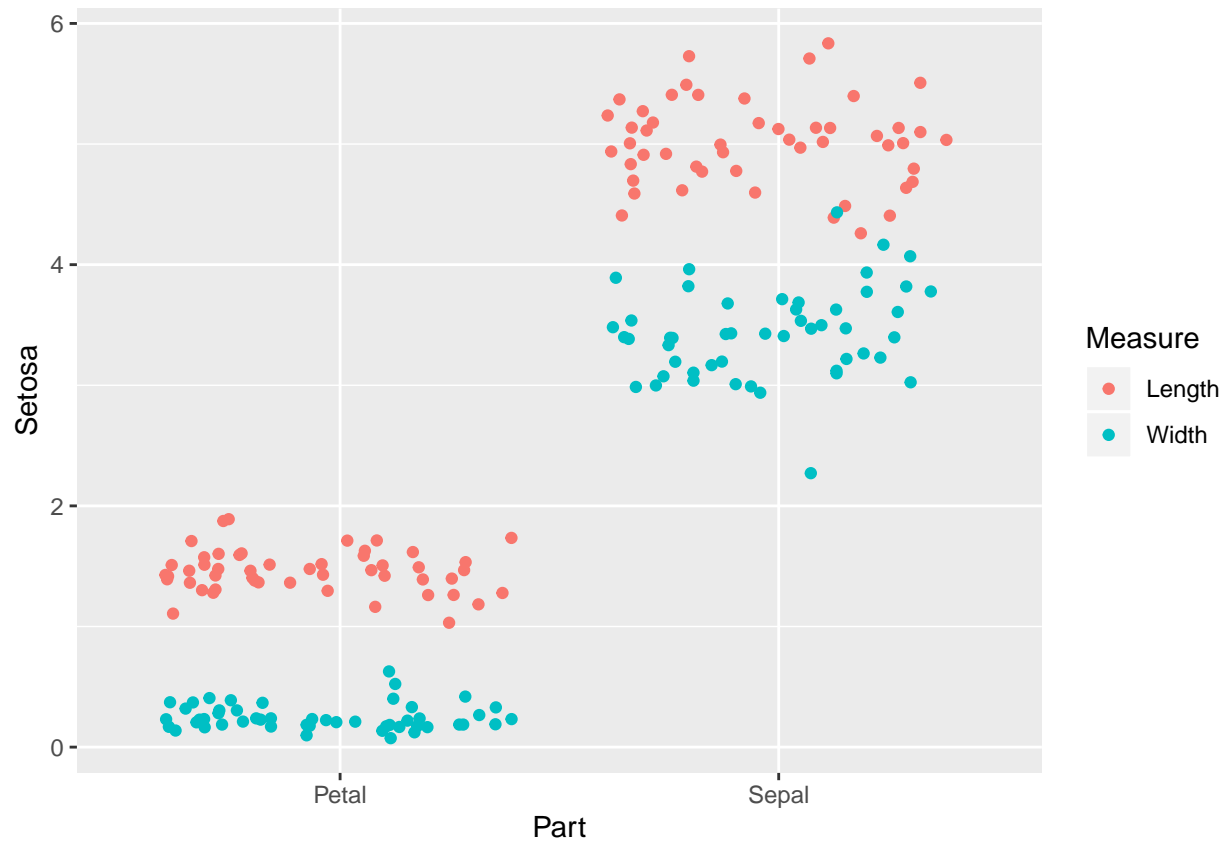
```
# iris.wide
p <- ggplot(iris.wide, aes(x = Length, y = Width, col = Part)) +
  geom_point(position = position_jitter(), alpha = 0.7) +
  scale_color_brewer(palette = "Set1") +
  coord_fixed()
p
```



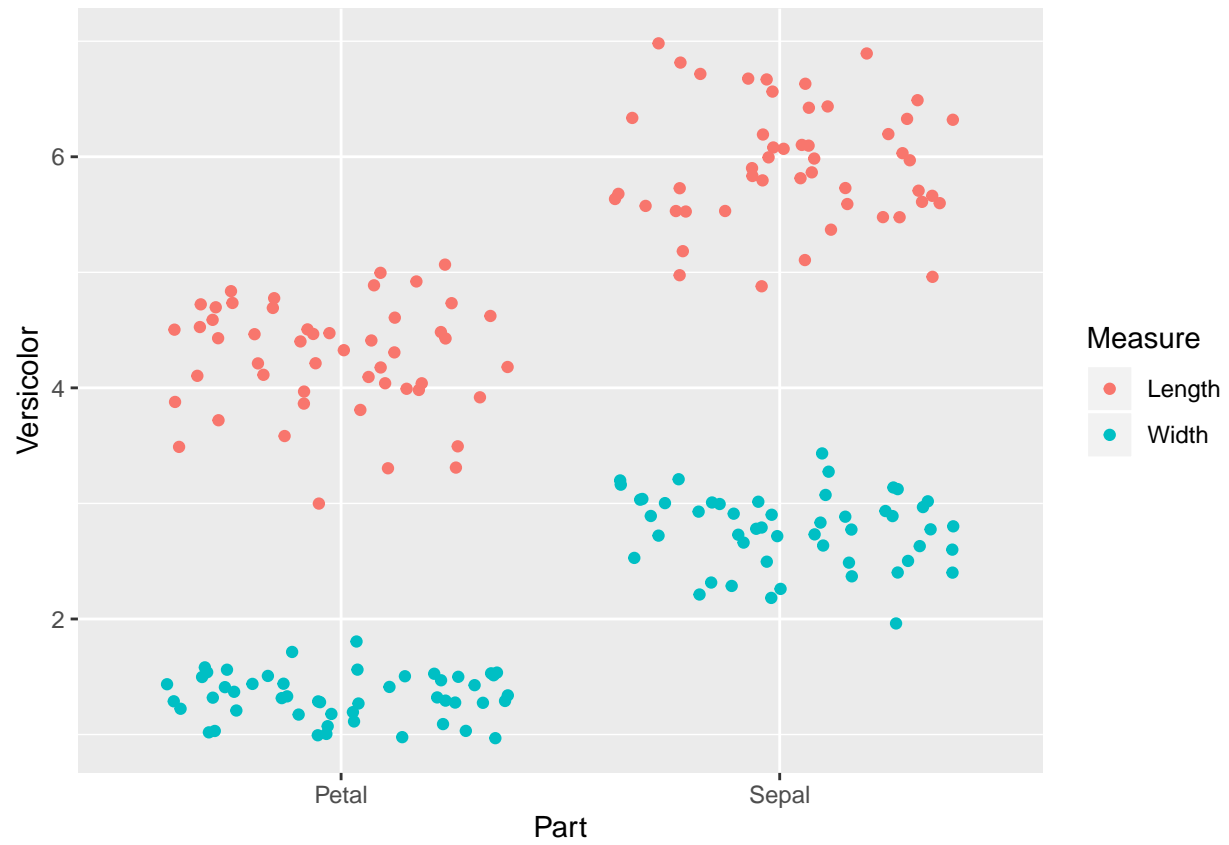
```
p + facet_grid(. ~ Species) # row ~ column
```



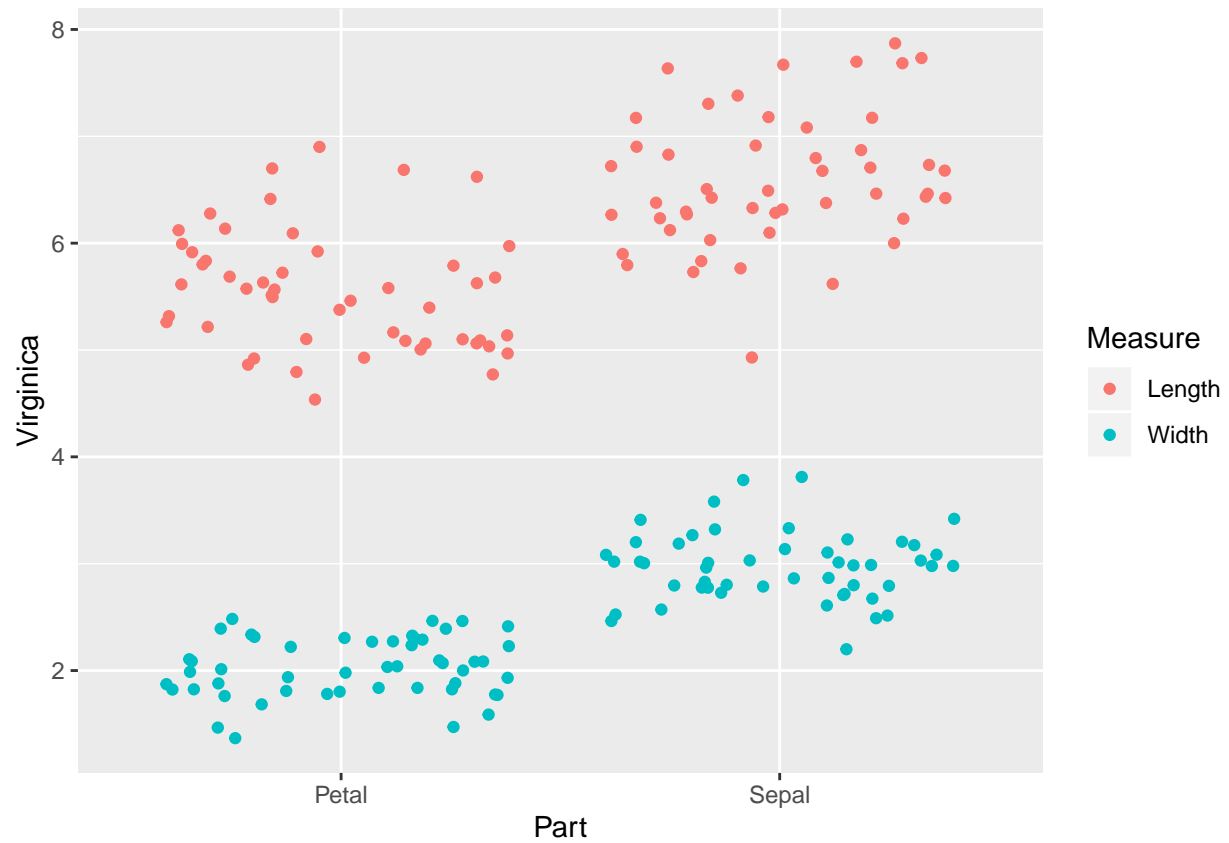
```
# iris.wide2  
ggplot(iris.wide2, aes(x = Part, y = Setosa, col = Measure)) +  
  geom_jitter()
```

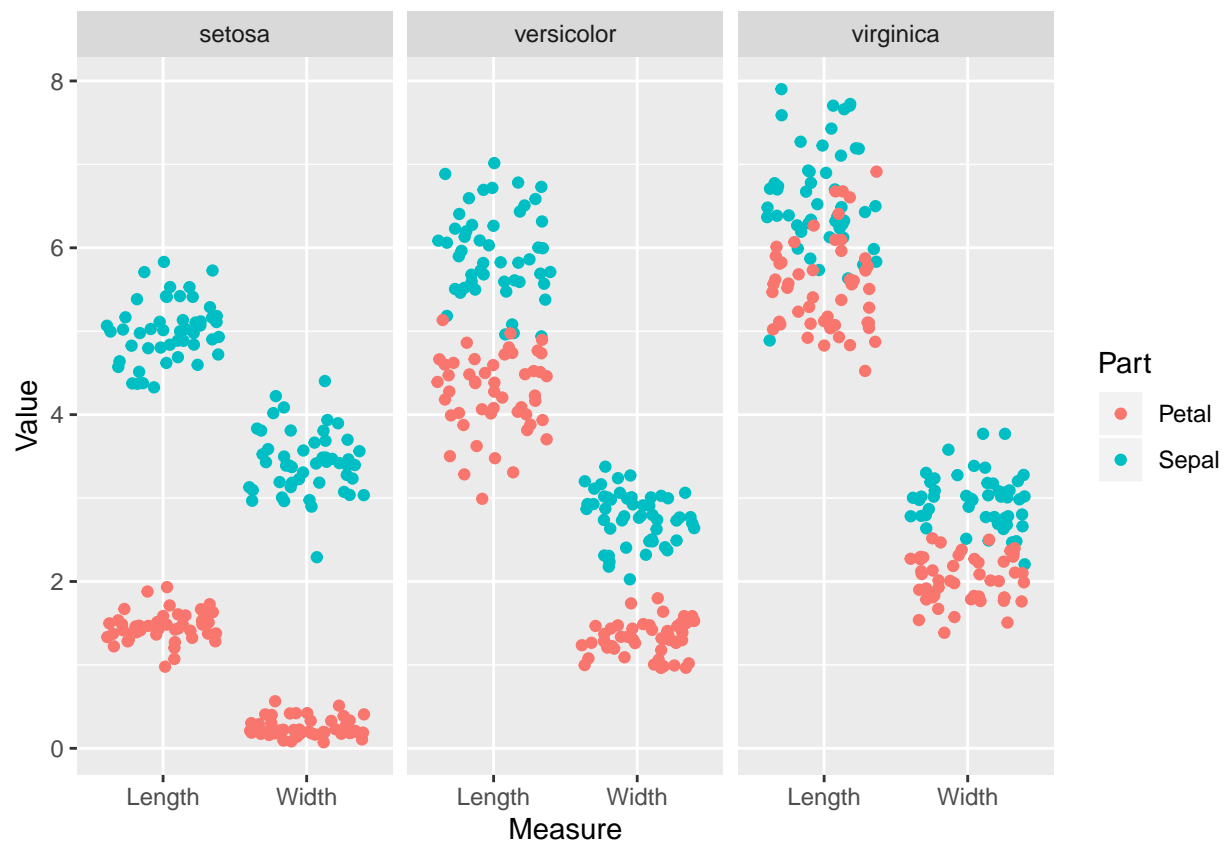
```
ggplot(iris.wide2, aes(x = Part, y = Versicolor, col = Measure)) +  
  geom_jitter()
```



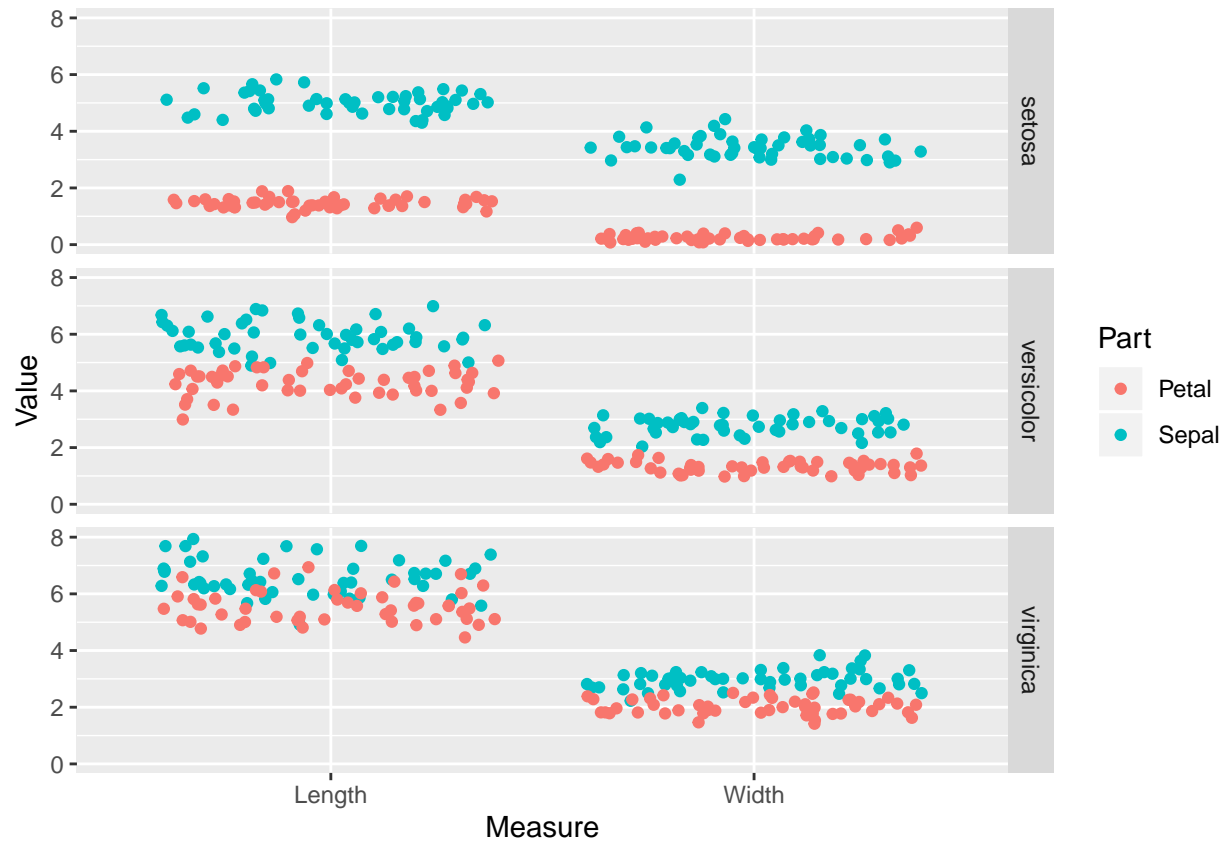
```
ggplot(iris.wide2, aes(x = Part, y = Virginica, col = Measure)) +  
  geom_jitter()
```



```
# iris.tidy
ggplot(iris.tidy, aes(x = Measure, y = Value, col = Part)) +
  geom_jitter() +
  facet_grid(. ~ Species)
```



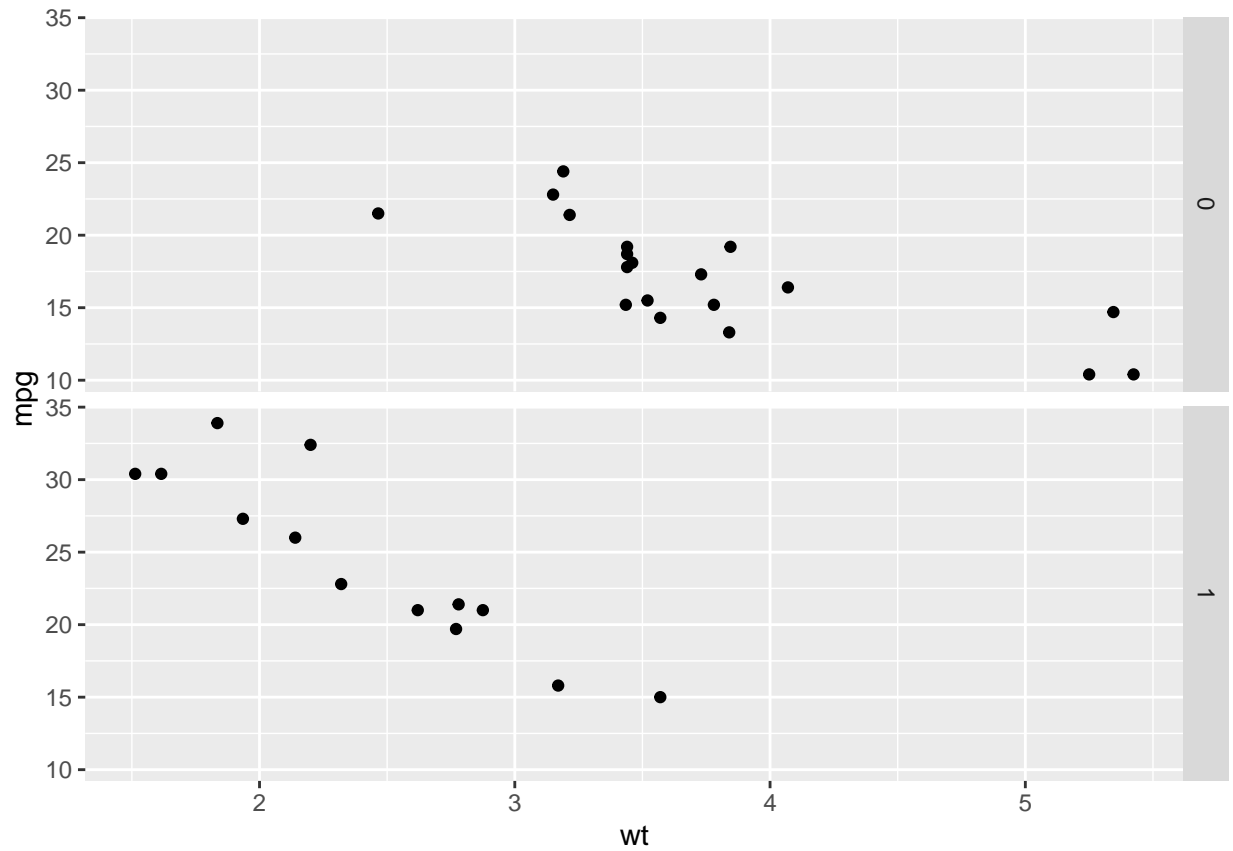
```
# iris.tidy - wrong
ggplot(iris.tidy, aes(x = Measure, y = Value, col = Part)) +
  geom_jitter() +
  facet_grid(Species ~ .)
```



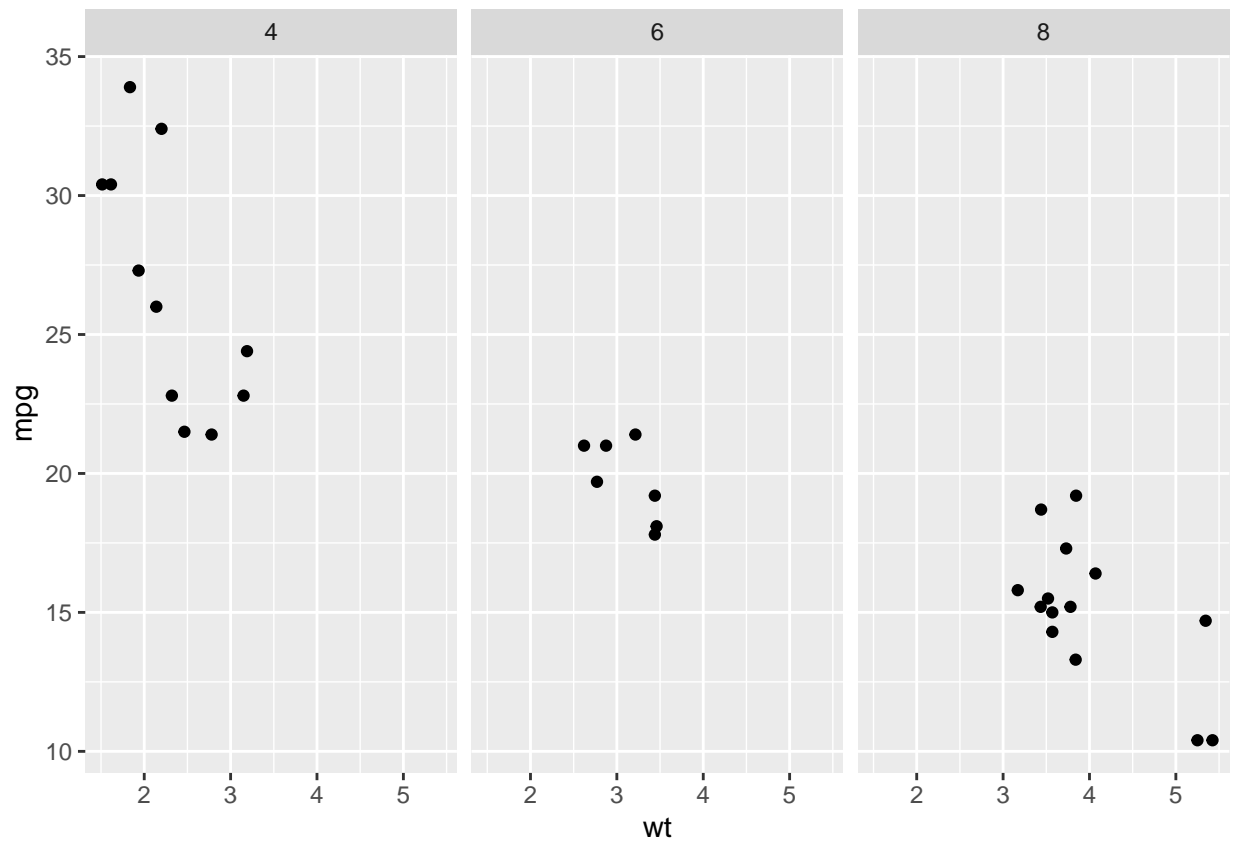
Practice

```
# Basic scatter plot
p <- ggplot(mtcars, aes(x = wt, y = mpg)) +
  geom_point()

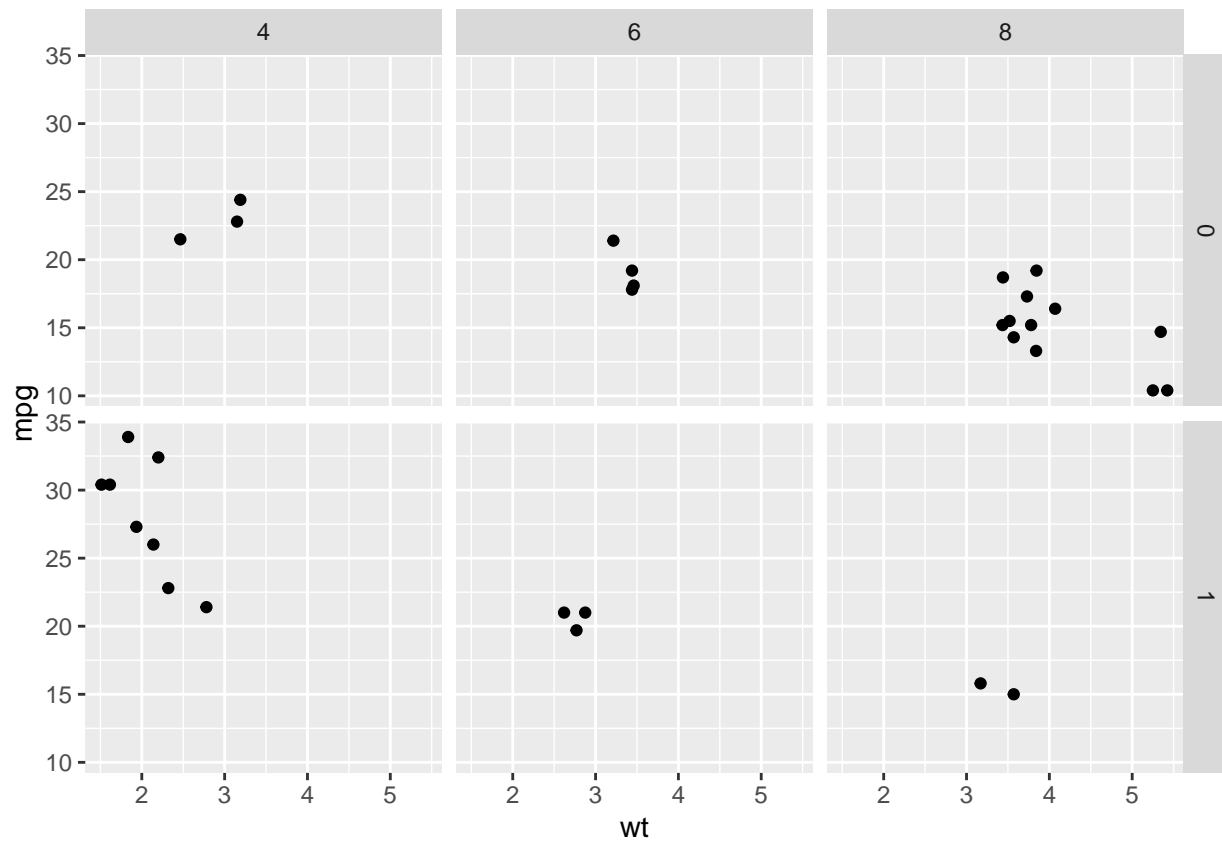
# 1 - Separate rows according to transmission type, am
p +
  facet_grid(am ~ .)
```



```
# 2 - Separate columns according to cylinders, cyl
p +
  facet_grid(. ~ cyl)
```



```
# 3 - Separate by both columns and rows
p +
  facet_grid(am ~ cyl)
```

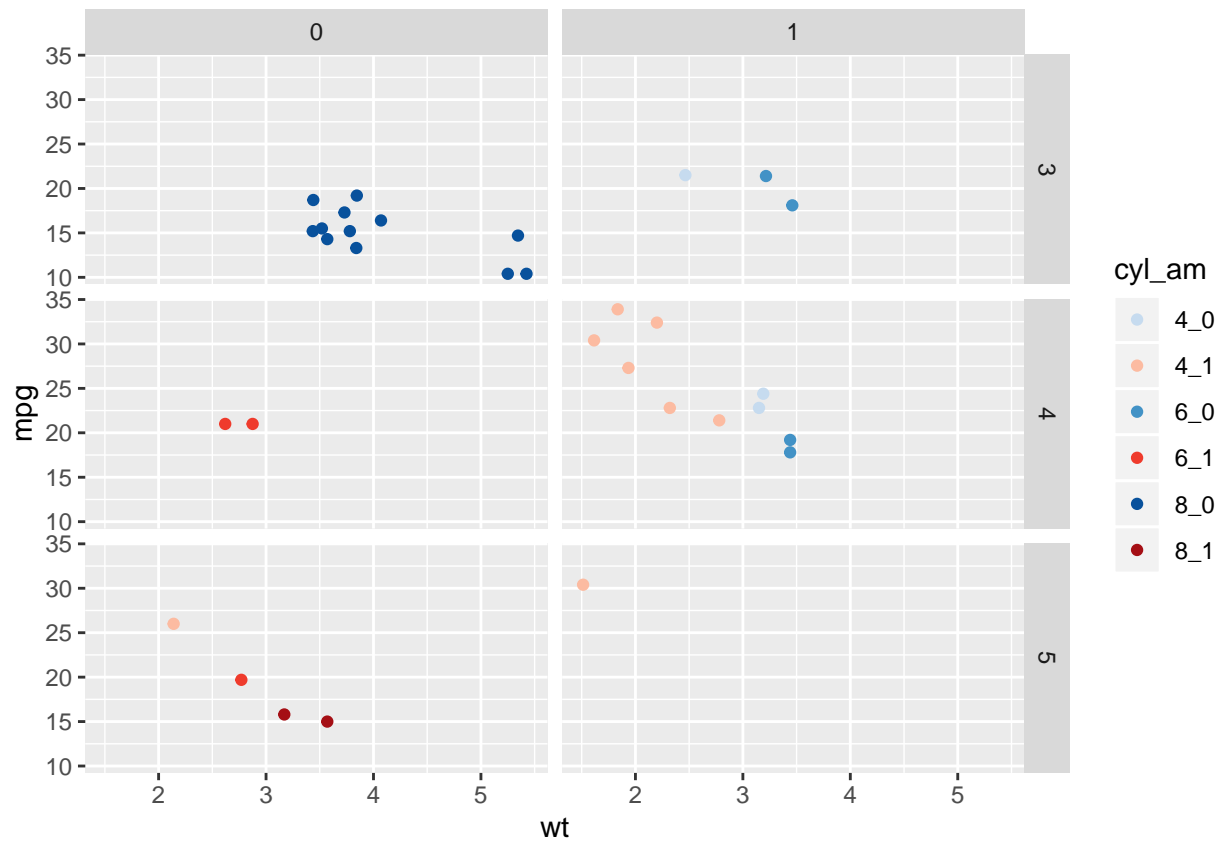


Many variables

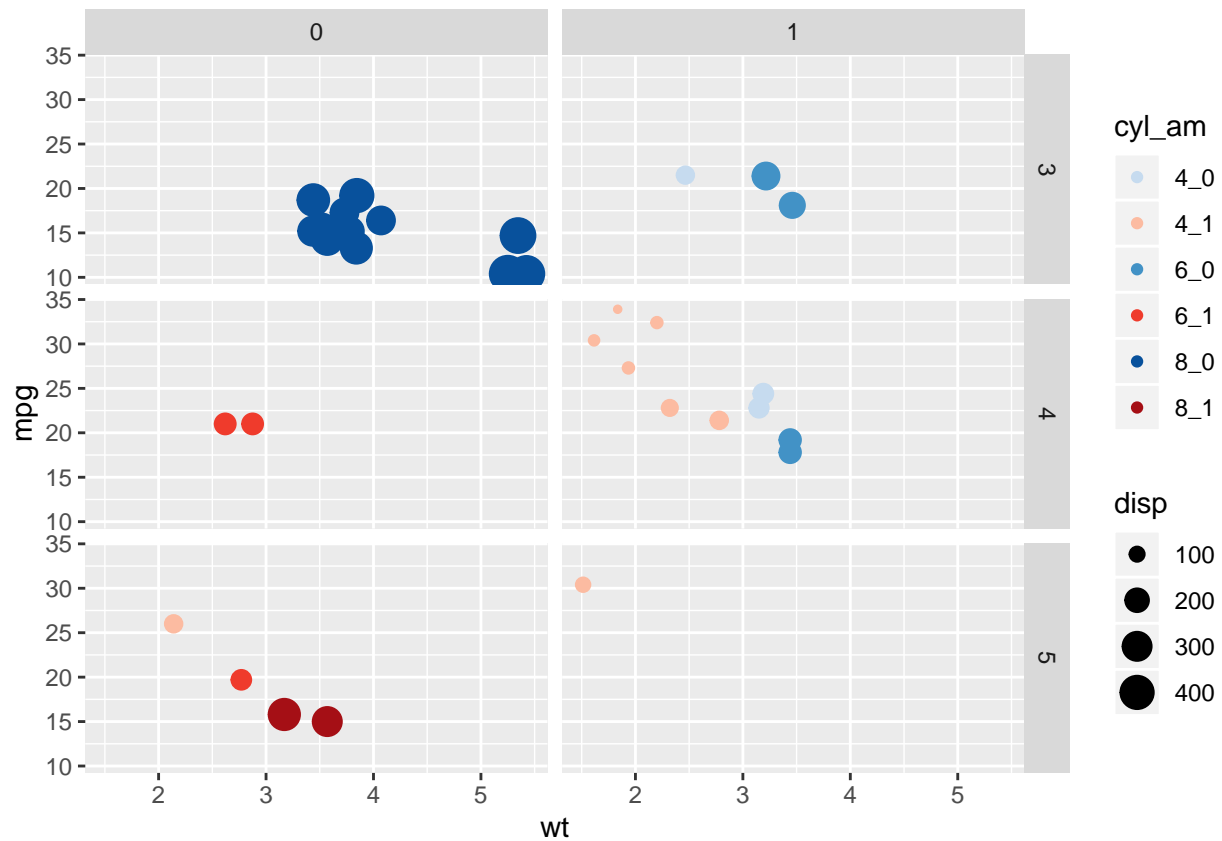
```
library("RColorBrewer")

# Code to create the cyl_am col and myCol vector
mtcars$cyl_am <- paste(mtcars$cyl, mtcars$am, sep = "_")
myCol <- rbind(brewer.pal(9, "Blues")[c(3,6,8)],
               brewer.pal(9, "Reds")[c(3,6,8)])

# Map cyl_am onto col
ggplot(mtcars, aes(x = wt, y = mpg, col = cyl_am)) +
  geom_point() +
  # Add a manual colour scale
  scale_color_manual(values = myCol)
```

```
# Also map disp to size
ggplot(mtcars, aes(x = wt, y = mpg, col = cyl_am, size = disp)) +
  geom_point() +
  # Add a manual colour scale
  scale_color_manual(values = myCol) + facet_grid(gear ~ vs)
```



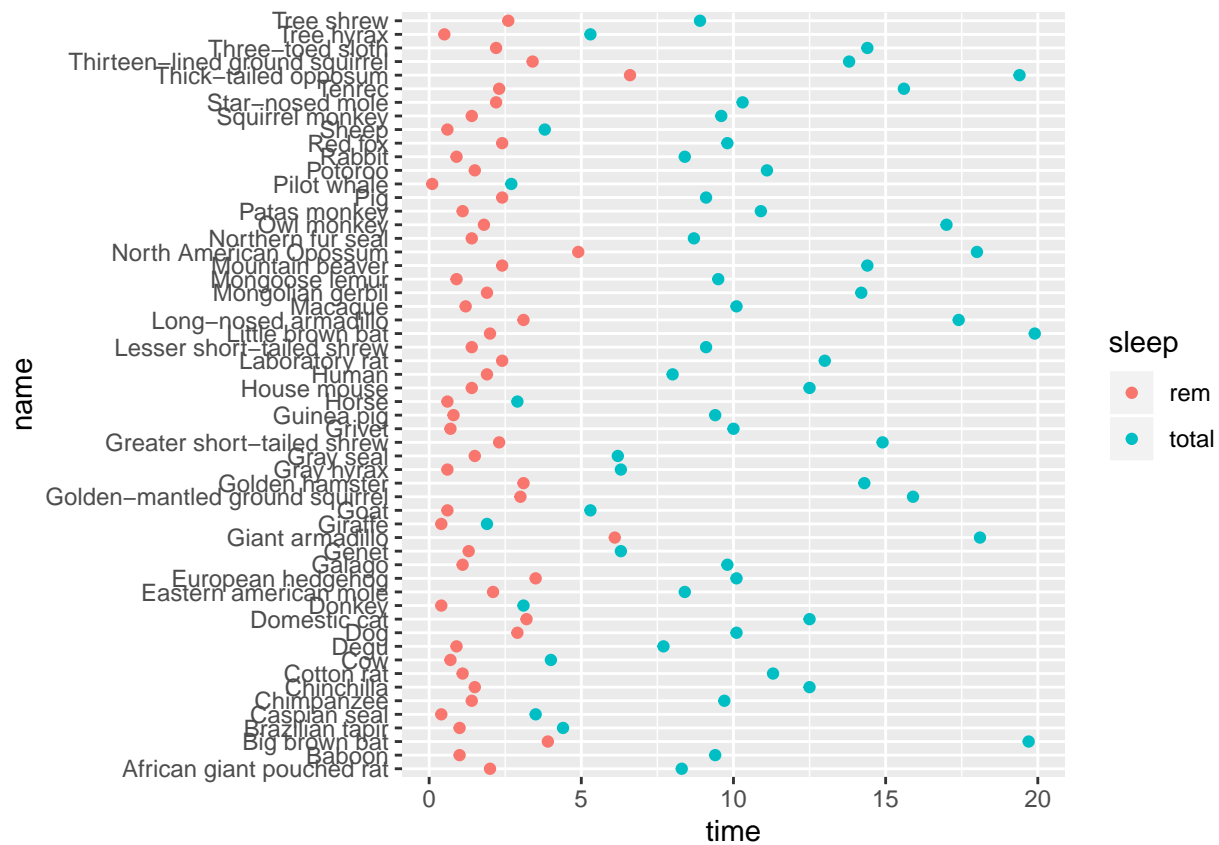
Dropping levels

Extend `facet_grid` with `scale = "free_y"` and `space = "free_y"` to leave out rows for which there is no data.

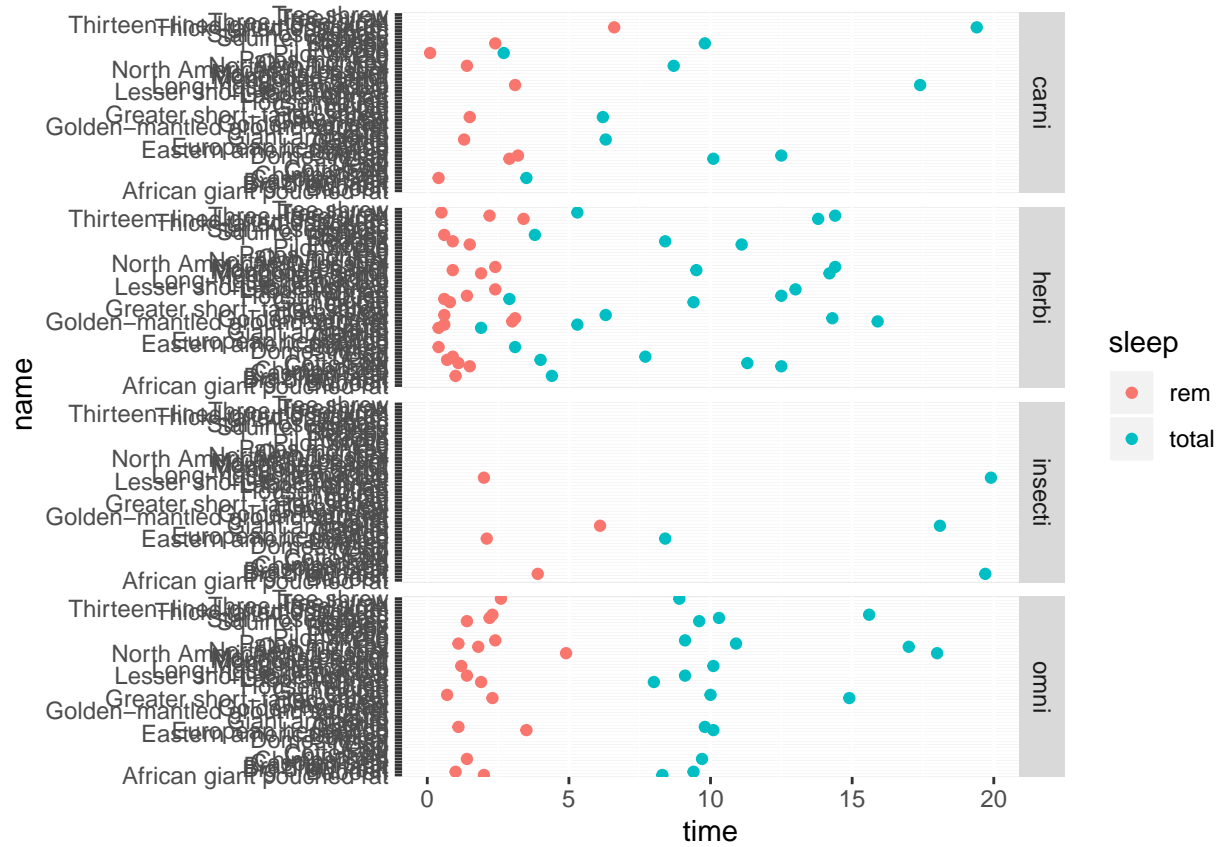
```
data(msleep)
mamsleep_clean <- na.omit(msleep[,c("vore", "name", "sleep_total", "sleep_rem")])
mamsleep <- mamsleep_clean %>% gather(sleep, time, -vore, -name) %>%
  separate(sleep, c("nouse", "sleep"), "\\_")
mamsleep$nouse <- NULL

# Basic scatter plot
p <- ggplot(mamsleep, aes(x = time, y = name, col = sleep)) +
  geom_point()

# Execute to display plot
p
```



```
# Facet rows according tovore
p +
  facet_grid(vore ~ .)
```



```
# Specify scale and space arguments to free up rows
p +
  facet_grid(vore ~ ., scale = "free_y", space = "free_y")
```

