

Lecture 20 - Plotting continuous data with ggplot2

Most of today will be spent working through examples of a variety of plots for continuous data that can be generated using R and the ggplot2 package. We'll work through the provided code and talk about what is happening and why it works. Make sure to ask any questions you may have.

```
# only ever install packages one time
```

```
install.packages("ggplot2")
```

```
install.packages("cowplot")
```

```
# load a package every time you wish to use it
```

```
library(ggplot2)
```

```
library(cowplot)
```

```
# read in data
```

```
mpg = read.table("mpg.txt", header=TRUE, sep="\t", stringsAsFactors=FALSE)
```

```
dim(mpg)
```

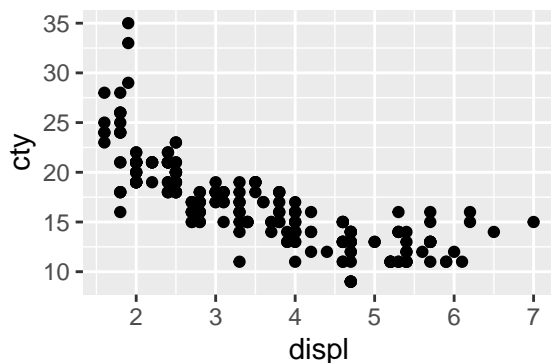
```
## [1] 234 9
```

```
head(mpg)
```

```
##   manufacturer model displ year cyl      trans drv  cty   hwy  
## 1         audi   a4    1.8 1999   4  auto(l5)  f   18   29  
## 2         audi   a4    1.8 1999   4 manual(m5)  f   21   29  
## 3         audi   a4    2.0 2008   4 manual(m6)  f   20   31  
## 4         audi   a4    2.0 2008   4  auto(av)   f   21   30  
## 5         audi   a4    2.8 1999   6  auto(l5)  f   16   26  
## 6         audi   a4    2.8 1999   6 manual(m5)  f   18   26
```

```
# plot of displacement (engine size) vs. city miles per gallon (cty)
```

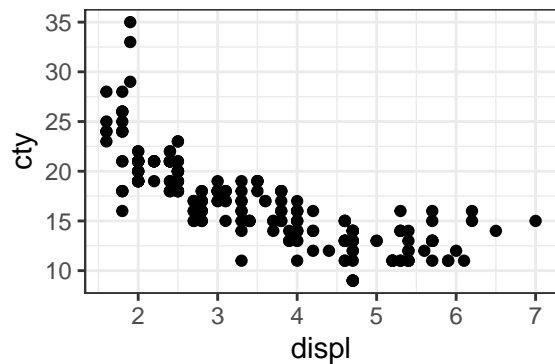
```
ggplot(data = mpg,  
       aes(x = displ, y = cty)) +  
  geom_point()
```



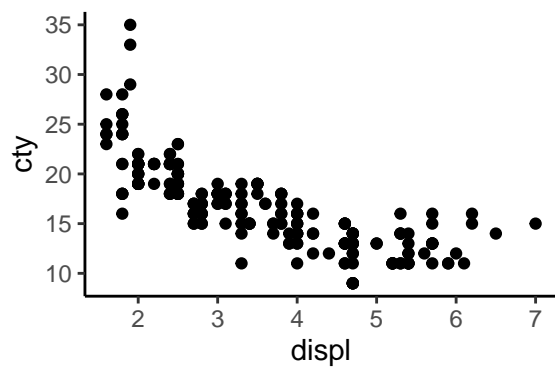
```
# remove grey background
```

```
ggplot(data = mpg,  
       aes(x = displ, y = cty)) +
```

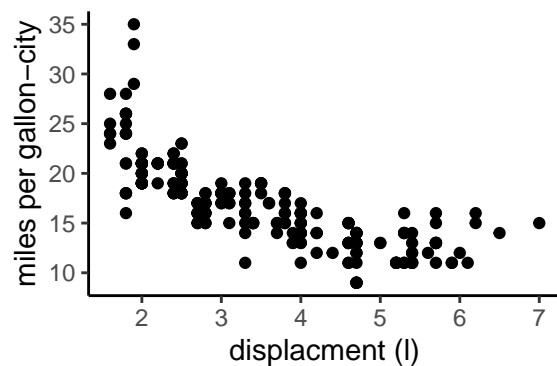
```
geom_point() +  
theme_bw()
```



```
# remove grey background and gridlines  
ggplot(data = mpg,  
       aes(x = displ, y = cty)) +  
geom_point() +  
theme_classic()
```



```
# change the x and y labels; cartesian coordinates are default  
ggplot(data = mpg,  
       aes(x = displ, y = cty)) +  
geom_point() +  
xlab("displacement (l)") +  
ylab("miles per gallon-city") +  
theme_classic()
```

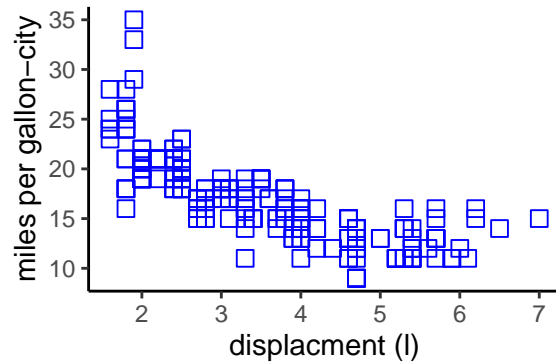


```
# arguments to geom_point() can alter the appearance of the points  
ggplot(data = mpg,
```

```

    aes(x = displ, y = cty)) +
  geom_point(color = "blue", shape = 22, size = 3) +
  xlab("displacement (l)") +
  ylab("miles per gallon-city") +
  theme_classic()

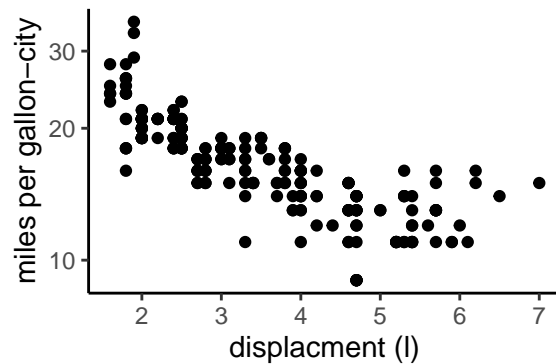
```



```

# log transform the y-axis
ggplot(data = mpg,
  aes(x = displ, y = cty)) +
  geom_point() +
  xlab("displacement (l)") +
  ylab("miles per gallon-city") +
  scale_y_log10() +
  theme_classic()

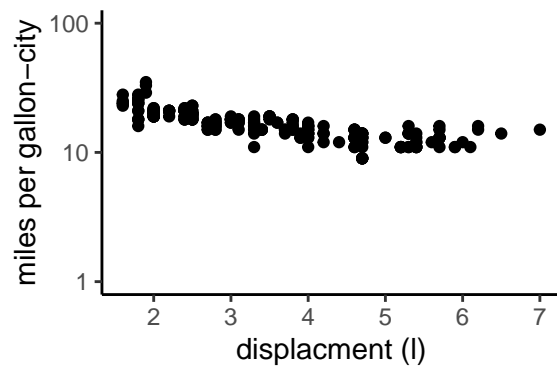
```



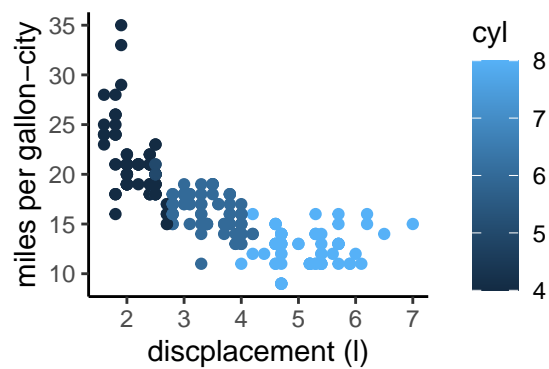
```

# arguments to scale can also customize the range and tick locations
ggplot(data = mpg,
  aes(x = displ, y = cty)) +
  geom_point() +
  theme_classic() +
  xlab("displacement (l)") +
  ylab("miles per gallon-city") +
  scale_y_log10(limits=c(1,100),
    breaks=c(1,10,100))

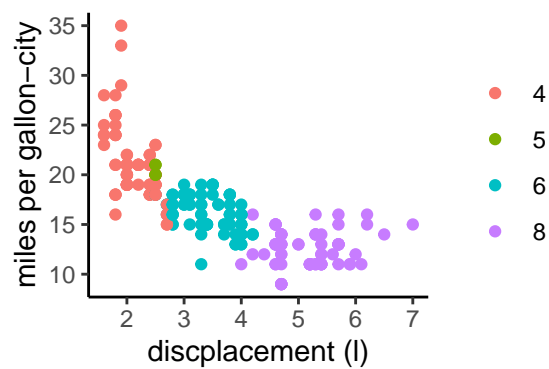
```



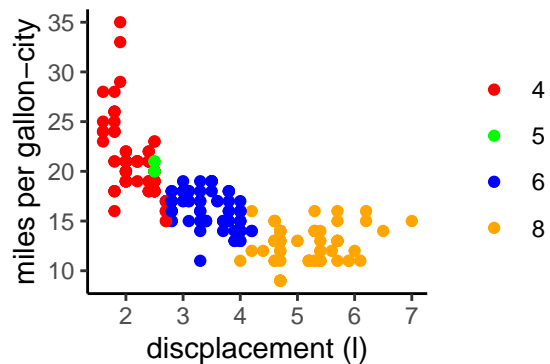
```
# we can also color code points based on continuous or categorical variables
# continuous
ggplot(data = mpg,
       aes(x = displ, y = cty, color = cyl)) +
  geom_point() +
  xlab("displacement (l)") +
  ylab("miles per gallon-city") +
  theme_classic()
```



```
# categorical
ggplot(data = mpg,
       aes(x = displ, y = cty, color = as.factor(cyl))) +
  geom_point() +
  xlab("displacement (l)") +
  ylab("miles per gallon-city") +
  theme_classic() +
  theme(legend.title=element_blank())
```

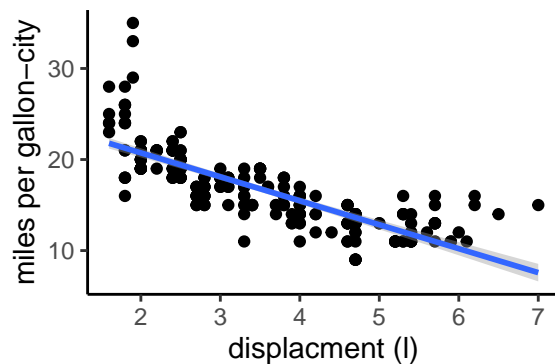


```
# categorical - change display colors
ggplot(data = mpg,
       aes(x = displ, y = cty, color = as.factor(cyl))) +
  geom_point() +
  xlab("displacement (l)") +
  ylab("miles per gallon-city") +
  scale_color_manual(values = c('red', 'green', 'blue', 'orange')) +
  theme_classic() +
  theme(legend.title=element_blank())
```



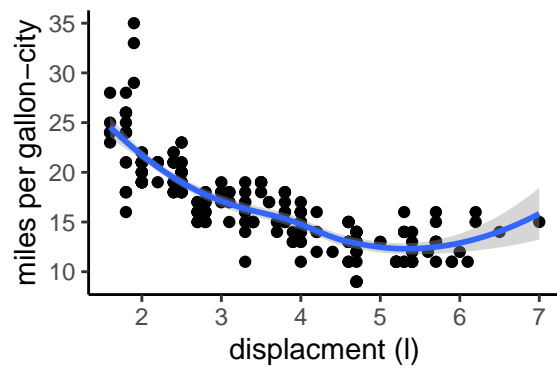
```
# add a linear trendline with a new layer
ggplot(data = mpg,
       aes(x = displ, y = cty)) +
  geom_point() +
  xlab("displacement (l)") +
  ylab("miles per gallon-city") +
  stat_smooth(method="lm") +
  theme_classic()
```

`geom_smooth()` using formula 'y ~ x'



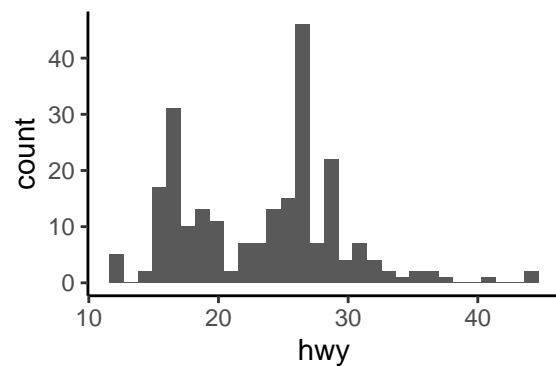
```
# add a spline with a new layer
ggplot(data = mpg,
       aes(x = displ, y = cty)) +
  geom_point() +
  xlab("displacement (l)") +
  ylab("miles per gallon-city") +
  stat_smooth(method="loess") +
  theme_classic()
```

`geom_smooth()` using formula 'y ~ x'

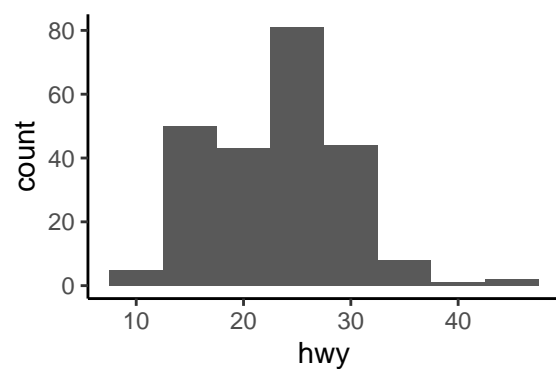


```
# histogram of mpg hwy
ggplot(data = mpg, aes(x = hwy)) +
  geom_histogram() +
  theme_classic()
```

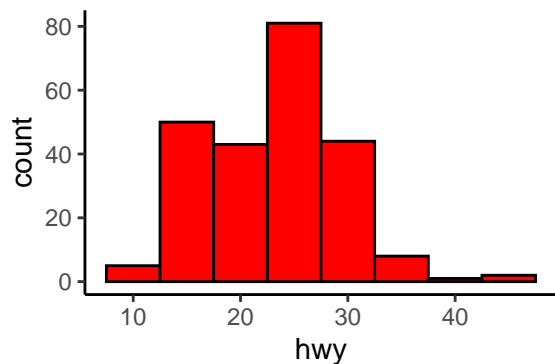
```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



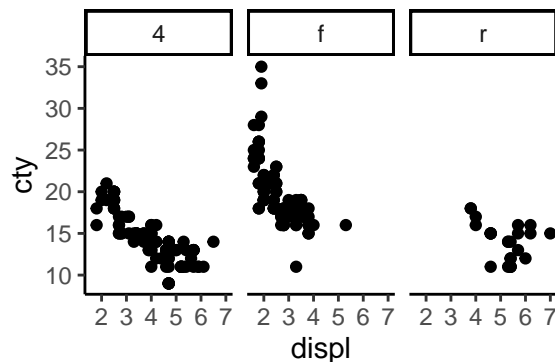
```
# change number of bins
ggplot(data = mpg, aes(x = hwy)) +
  geom_histogram(binwidth = 5) +
  theme_classic()
```



```
# change number of bins
# specify color in geom_histogram
ggplot(data = mpg, aes(x = hwy)) +
  geom_histogram(binwidth = 5, fill = "red", color = "black") +
  theme_classic()
```



```
# faceting allows the same plot across different categories to be generated
ggplot(data = mpg, aes(x = displ, y = cty)) +
  geom_point() +
  facet_wrap(~drv) +
  theme_classic()
```



```
# cowplot allows for multiple panels with different plot types
# makes sure you install the cowplot package with install.packages("cowplot")
# and load the package with library(cowplot)
```

```
#store the plots as variables
plot1 <- ggplot(data = mpg,
  aes(x = displ, y = cty, color = as.factor(cyl))) +
  geom_point() +
  xlab("displacement (l)") +
  ylab("miles per gallon-city") +
  scale_color_manual(values = c('red','green','blue','orange')) +
  theme_classic() +
  theme(legend.title=element_blank())
```

```
plot2 <- ggplot(data = mpg, aes(x = hwy)) +
  geom_histogram(binwidth = 5, fill = "red", color = "black") +
  theme_classic()
```

```
#put the subplots together in a variable called "fig1"
fig1 <- plot_grid(plot1, plot2,
  labels = c("a", "b"),
  rel_widths = c(1, 0.85),
  ncol = 2,
  nrow = 1)
```

```
#save your figure to an external file  
ggsave(filename = "Fig1.pdf",  
        plot = fig1,  
        width = 8,  
        height = 5,  
        dpi = 300)
```

Challenge

Practice using the syntax demonstrated above by writing a script to generate the following plots using the mpg data.

1. A scatter plot of miles per gallon city versus miles per gallon highway. Color code the points by 'drv' (four-wheel drive vs. front-wheel drive vs. rear-wheel drive). Add a linear trendline to the plot.
2. A “density plot” of engine displacement.
3. A faceted series of scatter plots across different manufacturers. Each plot should show engine size (displacement, displ) vs. highway miles. Make each manufacturer a different color.