# Class 5 Data Visualization ggplot2

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### **Using GGPLOT**

The ggplot2 package needs to be installed as it does not come with R "out of the box". We use the install.packages() function to do this.

To use ggplot I need to load it up before I can call any of the functions in the package. I do this with the library() function.

```
library(ggplot2)
ggplot()
```

10

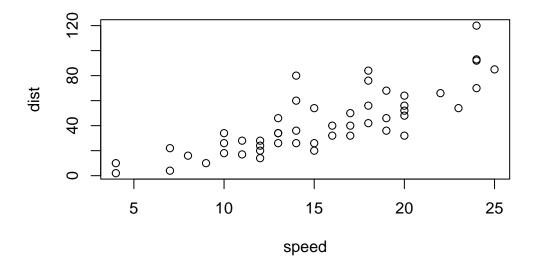
All ggplot figures have at least 3 things: - data (the stuff we want to plot) - aesthetic mapping (aes values) - geoms

```
ggplot(cars) +
  aes(x=speed, y=dist) +
  geom_point()
```



ggplot is not the only graphing system in R there are lots of others. There is even "base R" graphics.

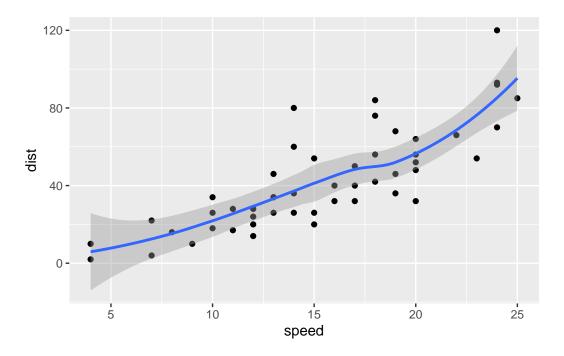
plot(cars)



Can you add a trend line layer to help show the relationship between the plot variables with the <code>geom\_smooth()</code> function?

```
ggplot(cars) +
  aes(x=speed, y=dist) +
  geom_point() +
  geom_smooth()
```

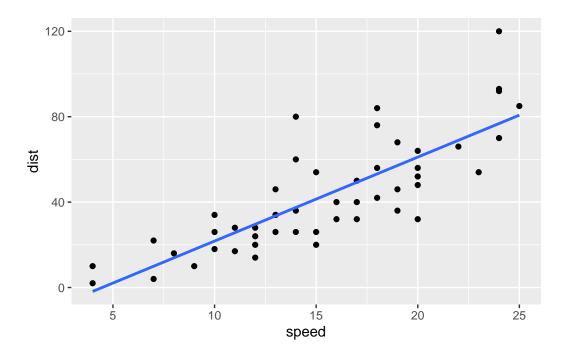
 $<sup>\</sup>ensuremath{\text{`geom\_smooth()`}}\ using method = 'loess' and formula = 'y ~ x'$ 



Argue with <code>geom\_smooth()</code> to add a straight line from a linear model without the shaded standard error region?

```
ggplot(cars) +
  aes(x=speed, y=dist) +
  geom_point() +
  geom_smooth(method="lm", se=FALSE)
```

<sup>`</sup>geom\_smooth()` using formula = 'y ~ x'

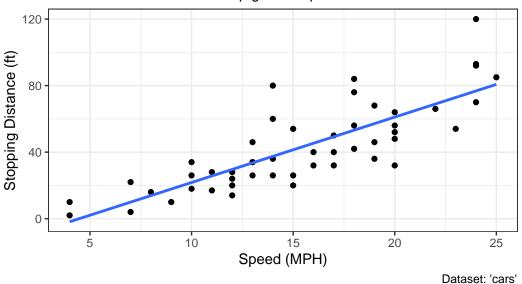


Can you finish this plot by adding various label annotations with the labs() function and changing the plot look to a more conservative "black & white" theme by adding the theme\_bw() function?

<sup>`</sup>geom\_smooth()` using formula = 'y ~ x'

# Speed and Stopping Distances of Cars

Distance in feet it takes to stop given a speed in MPH



Let's turn for a moment to more relevant example data set. The code below reads the results of a differential expression analysis where a new anti-viral drug is being tested.

```
url <- "https://bioboot.github.io/bimm143_S20/class-material/up_down_expression.txt"
genes <- read.delim(url)
head(genes)</pre>
```

```
Gene Condition1 Condition2
                                        State
       A4GNT -3.6808610 -3.4401355 unchanging
1
2
       AAAS
             4.5479580 4.3864126 unchanging
3
       AASDH
                         3.4787276 unchanging
              3.7190695
4
       AATF
              5.0784720
                         5.0151916 unchanging
5
       AATK
              0.4711421
                        0.5598642 unchanging
6 AB015752.4 -3.6808610 -3.5921390 unchanging
```

Use the nrow() function to find out how many genes are in this dataset.

```
nrow(genes)
```

[1] 5196

Use the colnames() function and the ncol() function on the genes data frame to find out what the column names are (we will need these later) and how many columns there are.

Use the table() function on the State column of this data.frame to find out how many 'up' regulated genes there are.

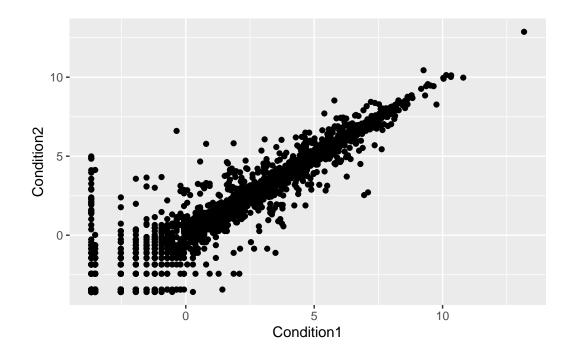
```
down unchanging up
72 4997 127
```

Using your values above and 2 significant figures. What fraction of total genes is up-regulated in this dataset?

We can make a first basic scatter plot of this dataset genes by following the same recipe we have already seen, namely:

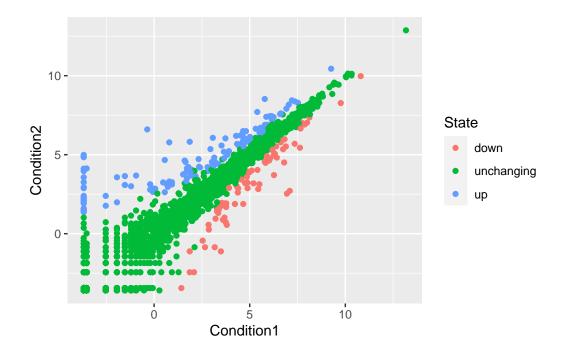
- Pass the genes data.frame as input to the ggplot() function.
- Then use the aes() function to set the x and y aesthetic mappings to the Condition1 and Condition2 columns.
- Finally add a geom\_point() layer to add points to the plot.
- Don't forget to add layers step-wise with the + operator at the end of each line.

```
ggplot(genes) +
  aes(x=Condition1, y=Condition2) +
  geom_point()
```



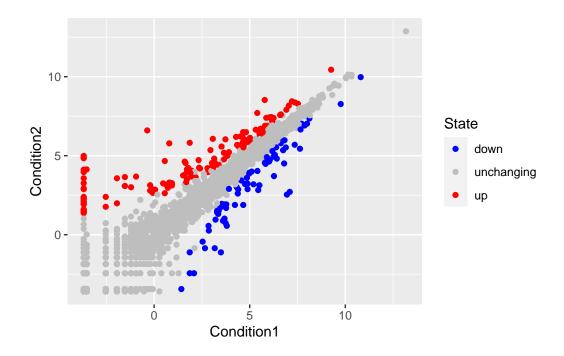
There is extra information in this dataset, namely the State column, which tells us whether the difference in expression values between conditions is statistically significant. Let's map this column to point color:

```
p <- ggplot(genes) +
  aes(x=Condition1, y=Condition2, col=State) +
  geom_point()
p</pre>
```



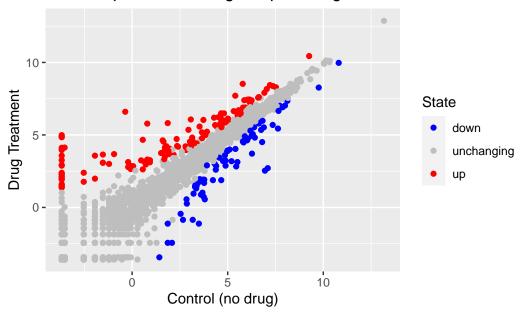
I am not a big fan of these default colors so let's change them up by adding another layer to explicitly specify our color scale. Note how we saved our previous plot as the object p and can use it now to add more layers:

```
p + scale_colour_manual( values=c("blue", "gray", "red"))
```



Add some plot annotations to the p object with the labs() function:

## Gene Expression Changes Upon Drug Treatment



The gapminder dataset contains economic and demographic data about various countries since 1952. It must be installed before it can be used.

#### install.packages("gapminder") library("gapminder)

This dataset covers many years and many countries. Before we make some plots we will use some dplyr code to focus in on a single year. It must be installed before it can be used.

The following line takes gapminder data frame and filters to contain only the rows with a year value of 2007.

```
library(gapminder)
library(dplyr)
```

Attaching package: 'dplyr'

The following objects are masked from 'package:stats':

filter, lag

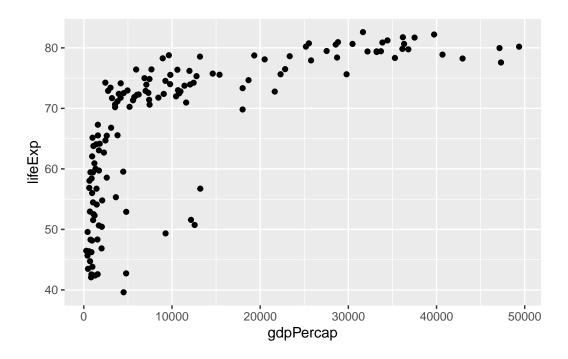
The following objects are masked from 'package:base':

```
intersect, setdiff, setequal, union
```

```
gapminder_2007 <- gapminder %>% filter(year==2007)
```

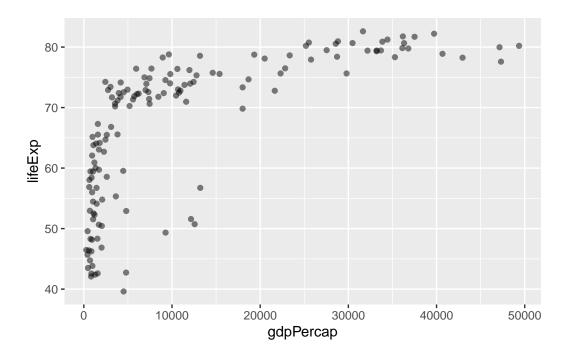
Consider the gapminder\_2007 dataset which contains the variables GDP per capita gdpPercap and life expectancy lifeExp for 142 countries in the year 2007. Below is a scatterplot of the gapminder\_2007 dataset.

```
ggplot(gapminder_2007) +
  aes(x=gdpPercap, y=lifeExp) +
  geom_point()
```



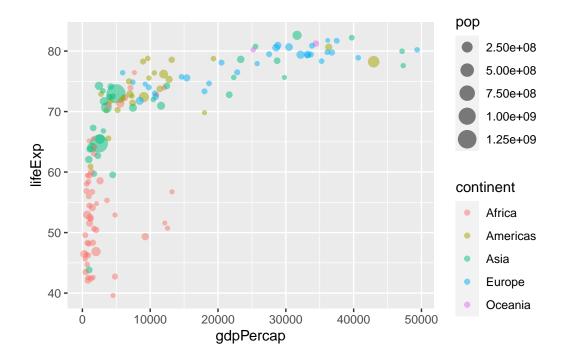
There are quite a few points that are nearly on top of each other in the above plot. Add an alpha=0.4 argument to your geom\_point() call to make the points slightly transparent.

```
ggplot(gapminder_2007) +
  aes(x=gdpPercap, y=lifeExp) +
  geom_point(alpha=0.5)
```



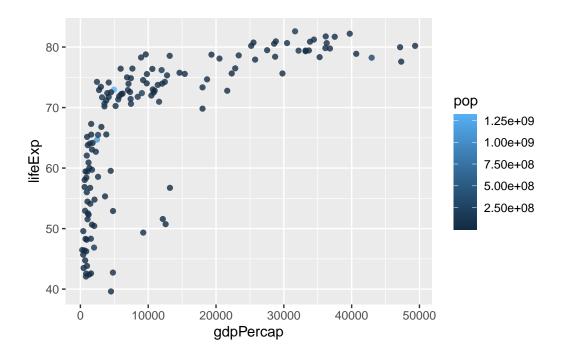
By mapping the continent variable to the point color aesthetic and the population pop (in millions) through the point size argument to aes() we can obtain a much richer plot that now includes 4 different variables from the data set:

```
ggplot(gapminder_2007) +
  aes(x=gdpPercap, y=lifeExp, color=continent, size=pop) +
  geom_point(alpha=0.5)
```



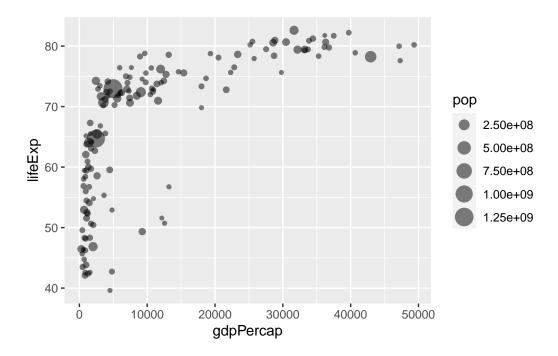
See how the plot looks like if we color the points by the numeric variable population pop:

```
ggplot(gapminder_2007) +
  aes(x = gdpPercap, y = lifeExp, color = pop) +
  geom_point(alpha=0.8)
```

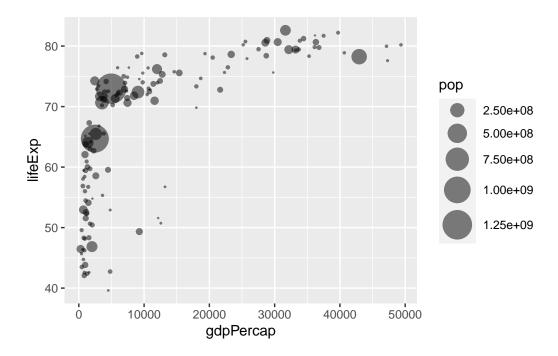


Plot the GDP per capita (x=gdpPercap) vs. the life expectancy (y=lifeExp) and set the point size based on the population (size=pop) of each country.

```
ggplot(gapminder_2007) +
  aes(x = gdpPercap, y = lifeExp, size = pop) +
  geom_point(alpha=0.5)
```



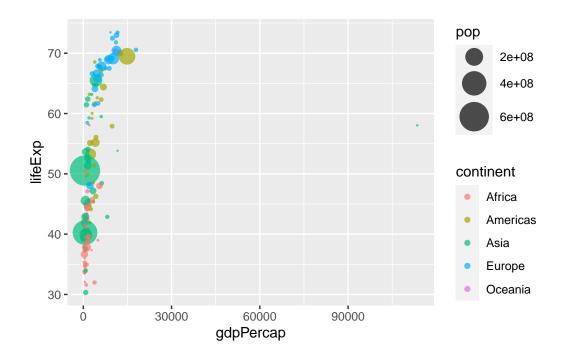
To reflect the actual population differences by the point size we can use the scale\_size\_area() function instead:



Adapt the code you have learned thus far to reproduce our gapminder scatter plot for the year 1957.

```
gapminder_1957 <- gapminder %>% filter(year==1957)

ggplot(gapminder_1957) +
  aes(x = gdpPercap, y = lifeExp, color = continent, size = pop) +
  geom_point(alpha = 0.7) +
  scale_size_area(max_size = 10)
```



Compare the plot of 1957 to that of 2007 by including the layer facet\_wrap(~year):

