

Class 5: Data Viz with ggplot

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Today we are exploring the **ggplot** package and how to make nice figures in R.

There are lots of ways to make figures and plot in R. These include:

- so called “base” R
- and add on packages like **ggplot2**

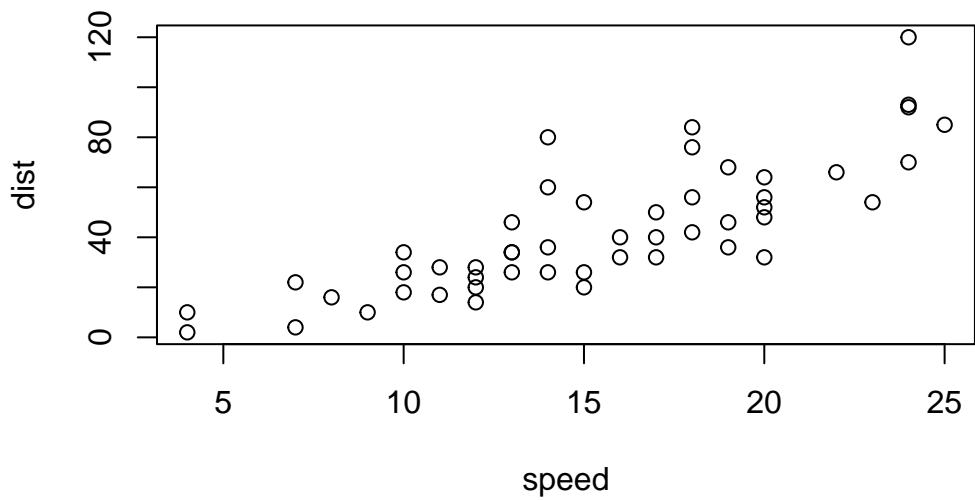
Here is a simple “base” R plot.

```
head(cars)
```

	speed	dist
1	4	2
2	4	10
3	7	4
4	7	22
5	8	16
6	9	10

We can simply pass to the `plot()` function.

```
plot(cars)
```



Key-point: Base R is quick but not so nice looking in some folks eyes

Let's see how we can plot this with **ggplot2**...

1st I need to install this add-on package. For this we use the `install.packages()` function - **WE DO THIS IN THE CONSOLE, NOT our report**. This is a one time only deal.

2nd We need to load the package with the `library()` function every time we want to use it.

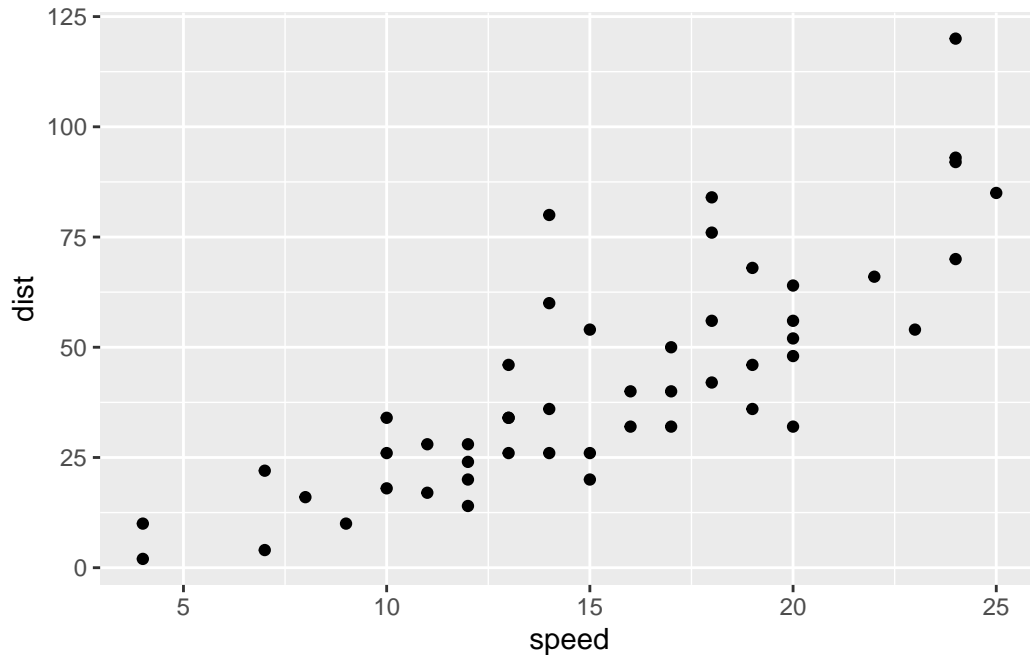
```
library(ggplot2)
ggplot(cars)
```



Every ggplot is composed of at least 3 layers:

- **data** (i.e. a data.frame with the things you want to plot)
- aesthetics **aes()** that map the columns of data to your plot features (i.e. aesthetics)
- geoms like **geom_point()** that srt how the plot appears

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



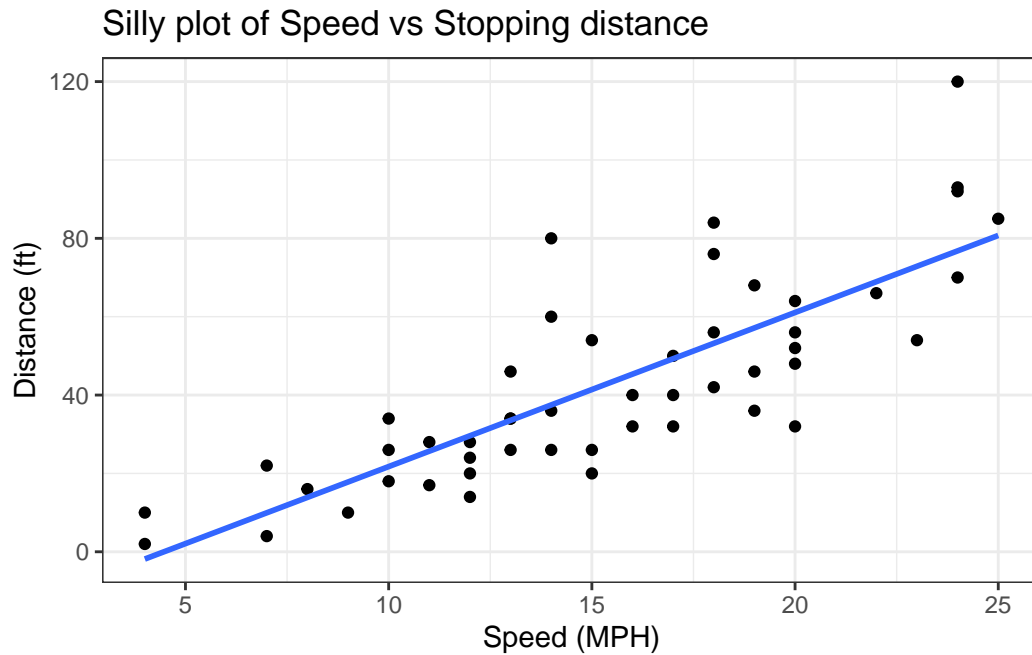
Key point: For simple “canned” graphs, base R is quicker, but as things get more custom and elaborate, then ggplot wins out...

Let’s add more layers to our ggplot

Add a line showing the relationship between x and y Add a title Add custom axis labels “Speed (MPH)” and “Distance(ft)” Change the theme...

```
ggplot(cars) +
  aes(x=speed, y=dist) +
  geom_point() +
  geom_smooth(method="lm", se=FALSE) +
  labs(title = "Silly plot of Speed vs Stopping distance",
       x = "Speed (MPH)",
       y = "Distance (ft)") +
  theme_bw()
```

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



Going further

Read some gene expression data

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

head(genes)
```

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

Q1. How many genes are in this wee dataset?

```
nrow(genes)
```

```
[1] 5196
```

```
ncol(genes)
```

```
[1] 4
```

Q2. How many “up” regulated genes are there?

```
sum( genes$State == "up" )
```

```
[1] 127
```

A useful function for counting up occurrences of things in a vector is the `table()` function.

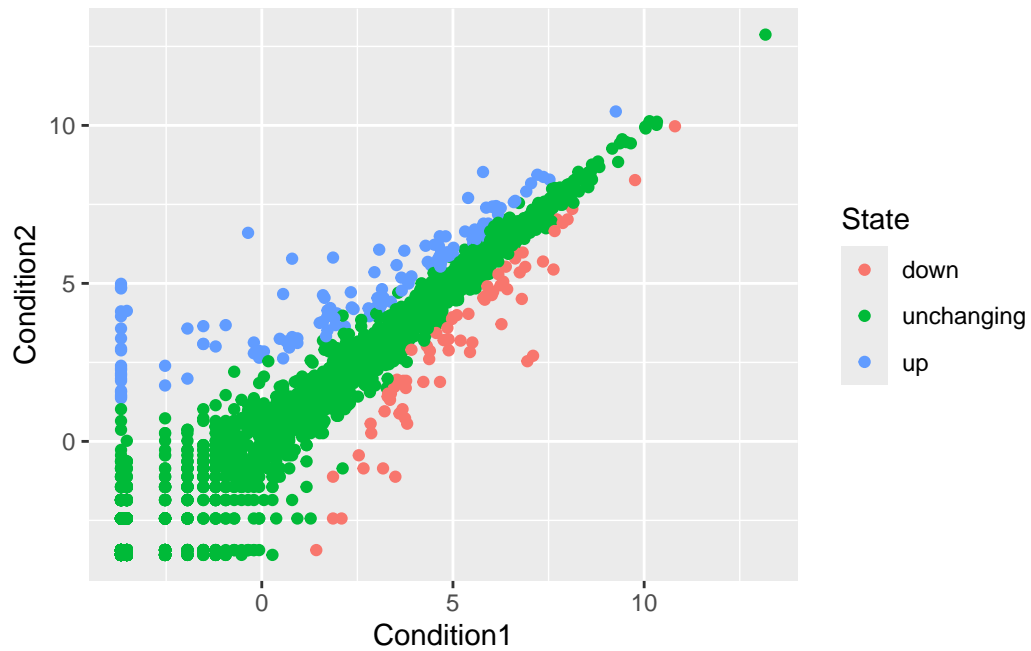
```
table( genes$State )
```

down	unchanging	up
72	4997	127

Make a v1 figure

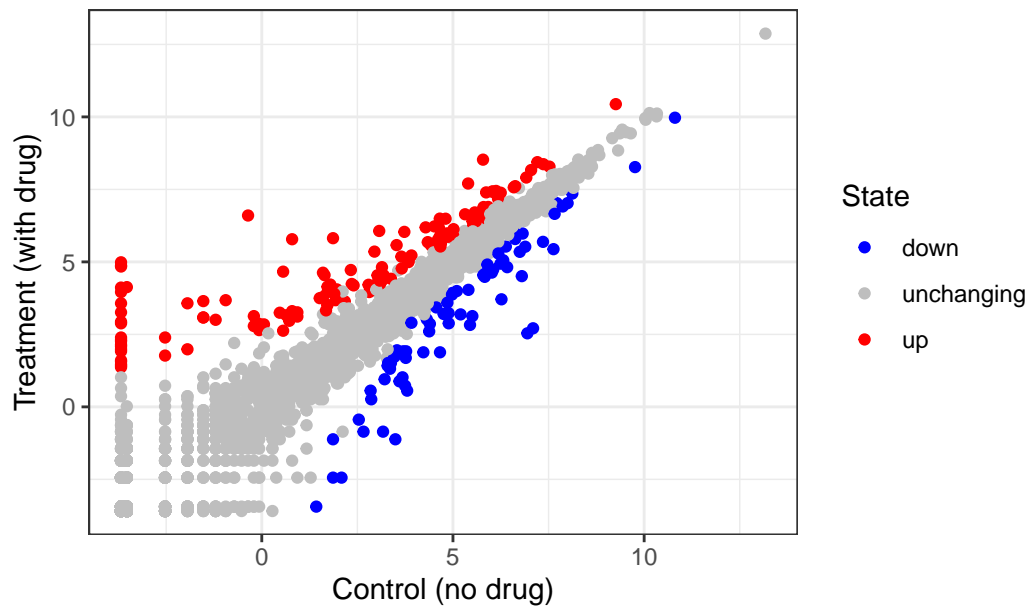
```
p <- ggplot(genes) +  
  aes(x=Condition1,  
      y=Condition2,  
      col=State) +  
  geom_point()
```

```
p
```



```
p + scale_colour_manual(values = c("blue", "gray", "red")) +
  labs(title="Expression changes upon drug treatment",
        x= "Control (no drug)",
        y= "Treatment (with drug)") +
  theme_bw()
```

Expression changes upon drug treatment



More Plotting

Read in the gapminder dataset

```
# File location online
url <- "https://raw.githubusercontent.com/jennybc/gapminder/master/inst/extdata/gapminder.tsv"
gapminder <- read.delim(url)
```

Let's have a wee peak

```
head(gapminder, 3)
```

	country	continent	year	lifeExp	pop	gdpPercap
1	Afghanistan	Asia	1952	28.801	8425333	779.4453
2	Afghanistan	Asia	1957	30.332	9240934	820.8530
3	Afghanistan	Asia	1962	31.997	10267083	853.1007

```
tail(gapminder, 3)
```


	country	continent	year	lifeExp	pop	gdpPercap
1702	Zimbabwe	Africa	1997	46.809	11404948	792.4500
1703	Zimbabwe	Africa	2002	39.989	11926563	672.0386
1704	Zimbabwe	Africa	2007	43.487	12311143	469.7093

Q4. How many different country values are in this dataset?

```
length( table(gapminder$country) )
```

```
[1] 142
```

Q5. How many different continent values are in this dataset?

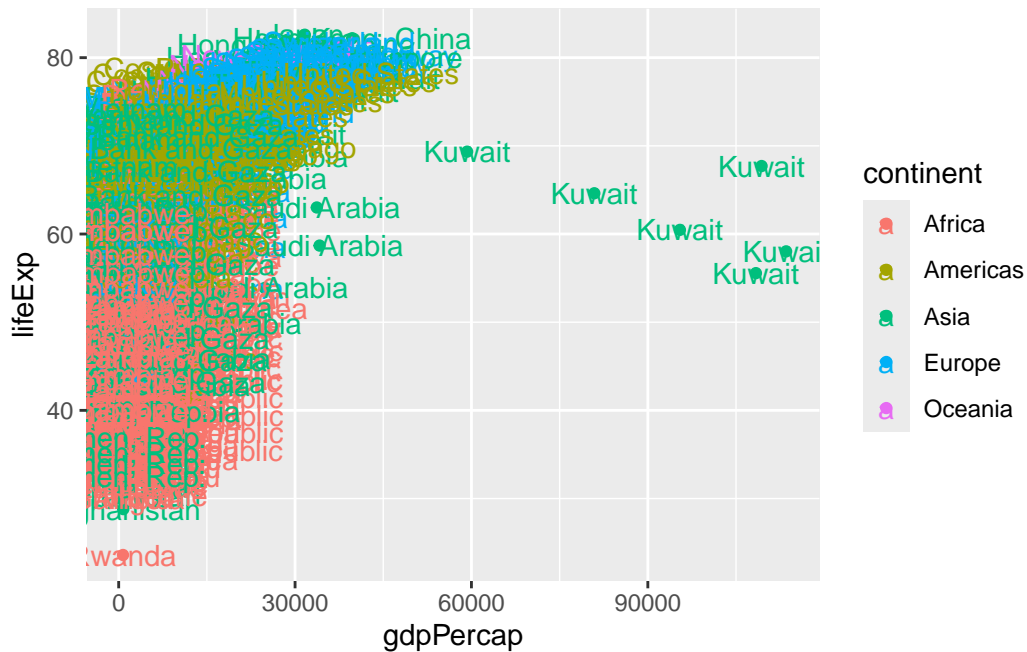
```
length( unique(gapminder$continent) )
```

```
[1] 5
```

```
ggplot(gapminder) +
  aes(gdpPercap, lifeExp, col= continent) +
  geom_point()
```



```
ggplot(gapminder) +
  aes(gdpPercap, lifeExp, col= continent, label= country) +
  geom_point() +
  geom_text()
```

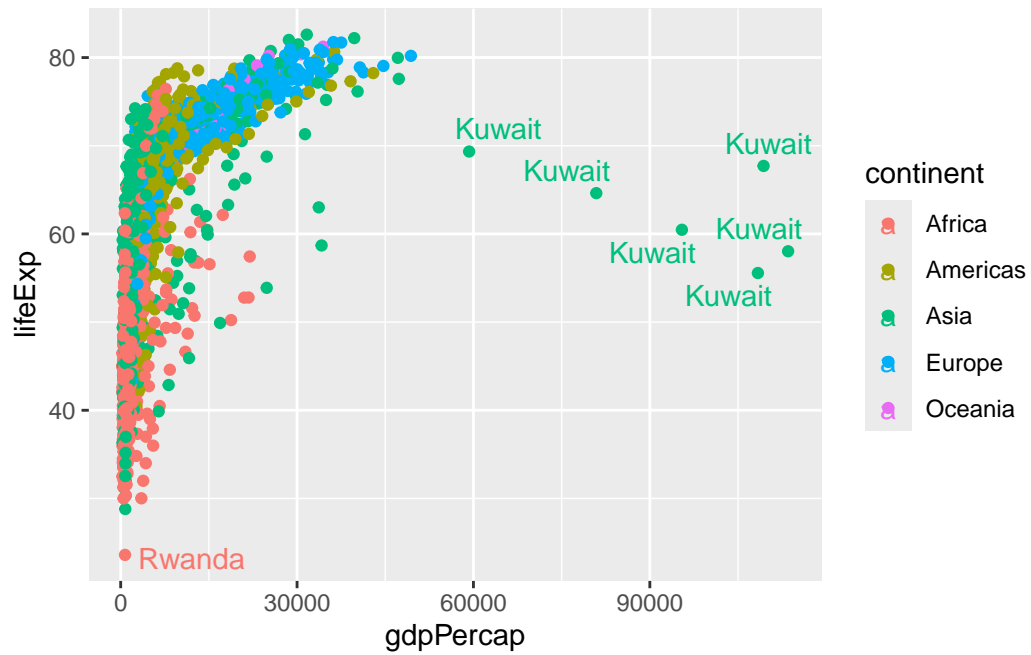


I can use the **ggrepel** package to make more sensible labels here.

```
library(ggrepel)

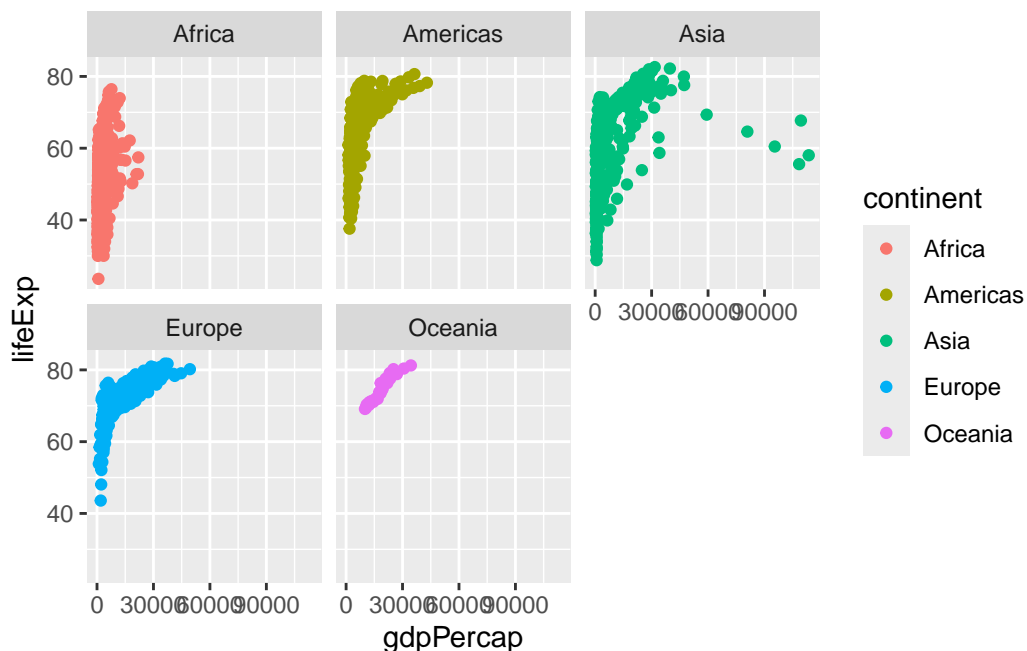
ggplot(gapminder) +
  aes(gdpPercap, lifeExp, col= continent, label= country) +
  geom_point() +
  geom_text_repel()
```

Warning: ggrepel: 1697 unlabeled data points (too many overlaps). Consider increasing max.overlaps



I want a separate panel per continent

```
ggplot(gapminder) +
  aes(gdpPercap, lifeExp, col= continent, label= country) +
  geom_point() +
  facet_wrap(~continent)
```



ggplot2 offers several advantages over base R plots:

The main advantages of ggplot over base R plot are:

1. **Layered grammar:** ggplot2 builds plots by adding layers (data, aesthetics, geoms, themes), making complex plots easier to construct and modify step-by-step [1], [2], [3].
2. **Consistency:** The same syntax and logic apply across different plot types, reducing the need to learn new functions for each visualization [1], [2], [3].
3. **Publication-quality visuals:** ggplot2 produces attractive, polished figures suitable for scientific publications with sensible defaults [1], [2], [3].
4. **Customization:** It is easier to customize legends, colors, themes, and annotations, and to combine multiple plots [1].
5. **Data mapping:** Direct mapping of data columns to visual aesthetics (color, size, shape) is built-in and intuitive [2], [3].
6. **Reproducibility:** ggplot2 code is scriptable and reproducible, making it easy to regenerate or update figures [1], [2].

Base R plots are quick for simple, exploratory graphics, but become fiddly and time-consuming for refined, complex, or publication-ready figures [1], [2], [3], [5], [4].

Which of these advantages do you think is most important for your work?