(../04-dplyr/index.html)

Intro to R and RStudio for Genomics (../)



Data Visualization with ggplot2

Overview

Teaching: 60 min **Exercises:** 30 min

Questions

•

Objectives

- •
- Describe the role of data, aesthetics, and geoms in ggplot functions.
- Choose the correct aesthetics and alter the geom parameters for a scatter plot, histogram, or box plot.
- Layer multiple geometries in a single plot.
- Customize plot scales, titles, subtitles, themes, fonts, layout, and orientation.
- Apply a facet to a plot.
- Apply additional ggplot2-compatible plotting libraries.
- Save a ggplot to a file.
- List several resources for getting help with ggplot.
- List several resources for creating informative scientific plots.

We start by loading the package ggplot2.

R

library(ggplot2)

Plotting with ggplot2

ggp1ot2 is a plotting package that makes it simple to create complex plots from data in a data frame. It provides a more programmatic interface for specifying what variables to plot, how they are displayed, and general visual properties. Therefore, we only need minimal changes if the underlying data change or if we decide to change from a bar plot to a scatter plot. This helps in creating publication quality plots with minimal amounts of adjustments and tweaking.

ggplot2 functions like data in the 'long' format, i.e., a column for every dimension, and a row for every observation. Well-structured data will save you lots of time when making figures with ggplot2

ggplot graphics are built step by step by adding new elements. Adding layers in this fashion allows for extensive flexibility and customization of plots.

To build a ggplot, we will use the following basic template that can be used for different types of plots:

```
R
ggplot(data = <DATA>, mapping = aes(<MAPPINGS>)) + <GEOM_FUNCTION>()
```

• use the ggplot() function and bind the plot to a specific data frame using the data argument

```
R
ggplot(data = variants)
```

• define a mapping (using the aesthetic (aes) function), by selecting the variables to be plotted and specifying how to present them in the graph, e.g. as x/y positions or characteristics such as size, shape, color, etc.

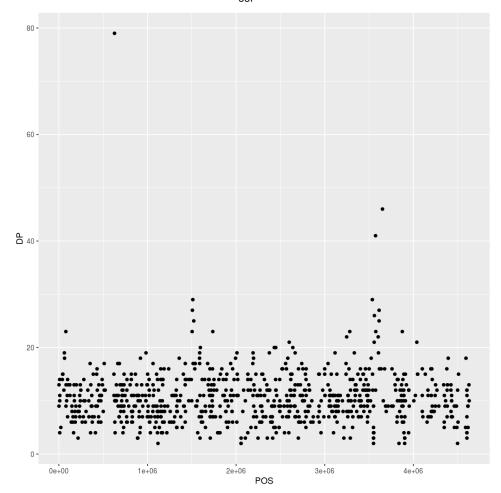
```
R
ggplot(data = variants, aes(x = POS, y = DP))
```

• add 'geoms' – graphical representations of the data in the plot (points, lines, bars). ggplot2 offers many different geoms; we will use some common ones today, including:

```
* `geom_point()` for scatter plots, dot plots, etc.
* `geom_boxplot()` for, well, boxplots!
* `geom_line()` for trend lines, time series, etc.
```

To add a geom to the plot use the + operator. Because we have two continuous variables, let's use <code>geom_point()</code> first:

```
R
ggplot(data = variants, aes(x = POS, y = DP)) +
  geom_point()
```



The + in the ggplot2 package is particularly useful because it allows you to modify existing ggplot objects. This means you can easily set up plot templates and conveniently explore different types of plots, so the above plot can also be generated with code like this:

```
# Assign plot to a variable
coverage_plot <- ggplot(data = variants, aes(x = POS, y = DP))

# Draw the plot
coverage_plot +
    geom_point()</pre>
```

Notes

- Anything you put in the ggplot() function can be seen by any geom layers that you add (i.e., these are universal plot settings). This includes the x- and y-axis mapping you set up in aes().
- You can also specify mappings for a given geom independently of the mappings defined globally in the ggplot() function.
- The + sign used to add new layers must be placed at the end of the line containing the *previous* layer. If, instead, the + sign is added at the beginning of the line containing the new layer, ggplot2 will not add the new layer and will return an error message.

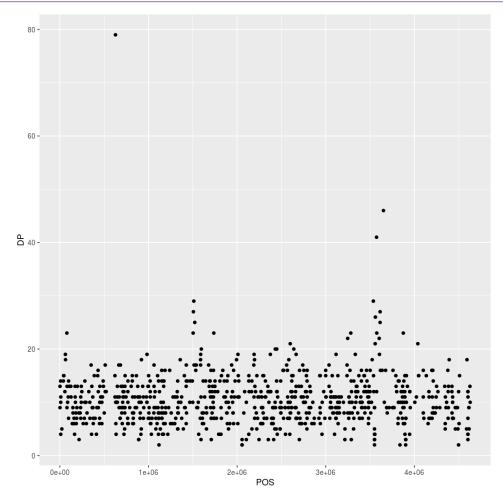
```
# This is the correct syntax for adding layers
coverage_plot +
geom_point()

# This will not add the new layer and will return an error message
coverage_plot
+ geom_point()
```

Building your plots iteratively

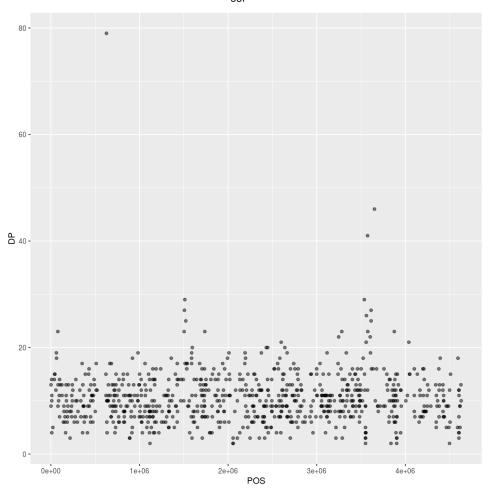
Building plots with ggplot2 is typically an iterative process. We start by defining the dataset we'll use, lay out the axes, and choose a geom:

```
R
ggplot(data = variants, aes(x = POS, y = DP)) +
  geom_point()
```



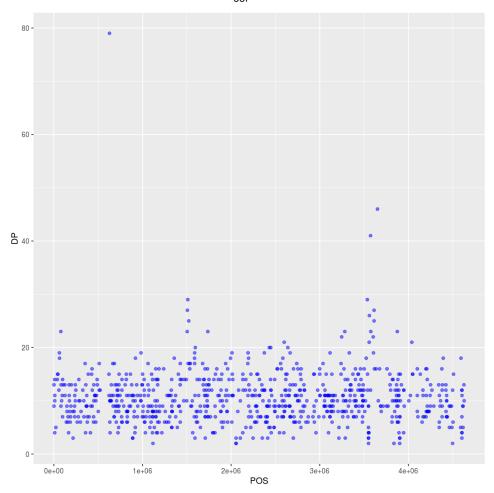
Then, we start modifying this plot to extract more information from it. For instance, we can add transparency (alpha) to avoid overplotting:

```
R
ggplot(data = variants, aes(x = POS, y = DP)) +
  geom_point(alpha = 0.5)
```



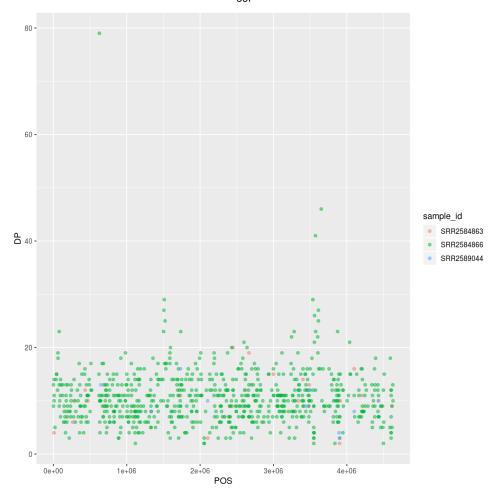
We can also add colors for all the points:

```
R
ggplot(data = variants, aes(x = POS, y = DP)) +
geom_point(alpha = 0.5, color = "blue")
```



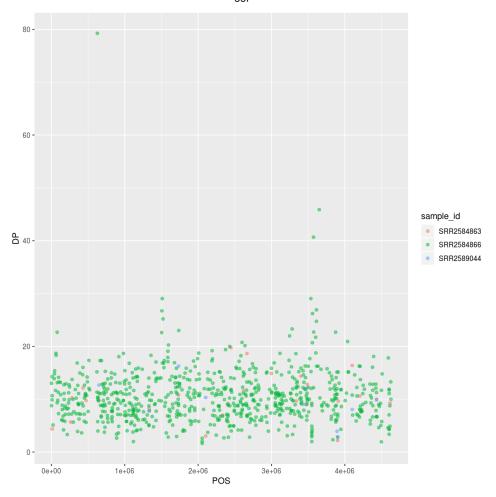
Or to color each species in the plot differently, you could use a vector as an input to the argument **color**. ggplot2 will provide a different color corresponding to different values in the vector. Here is an example where we color with sample_id:

```
R
ggplot(data = variants, aes(x = POS, y = DP, color = sample_id)) +
  geom_point(alpha = 0.5)
```



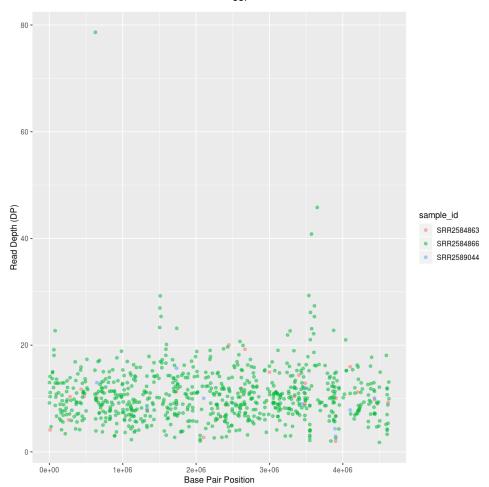
Notice that we can change the geom layer and colors will be still determined by sample_id

```
R
ggplot(data = variants, aes(x = POS, y = DP, color = sample_id)) +
    geom_jitter(alpha = 0.5)
```



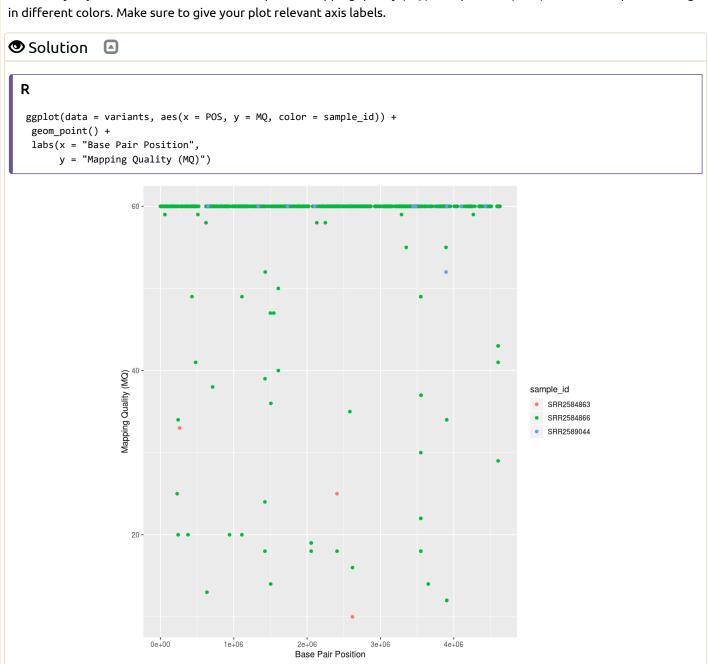
To make our plot more readable, we can add axis labels:

```
ggplot(data = variants, aes(x = POS, y = DP, color = sample_id)) +
  geom_jitter(alpha = 0.5) +
  labs(x = "Base Pair Position",
      y = "Read Depth (DP)")
```



Challenge

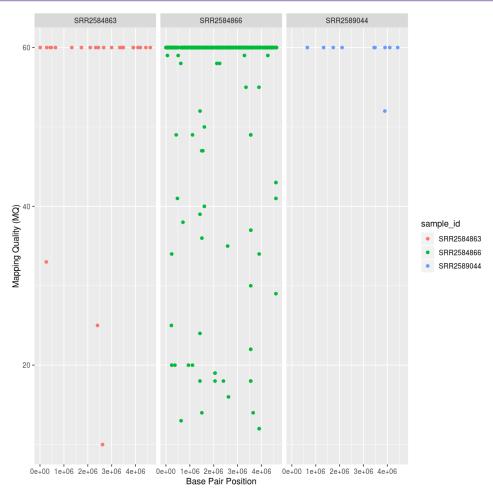
Use what you just learned to create a scatter plot of mapping quality (MQ) over position (POS) with the samples showing



Faceting

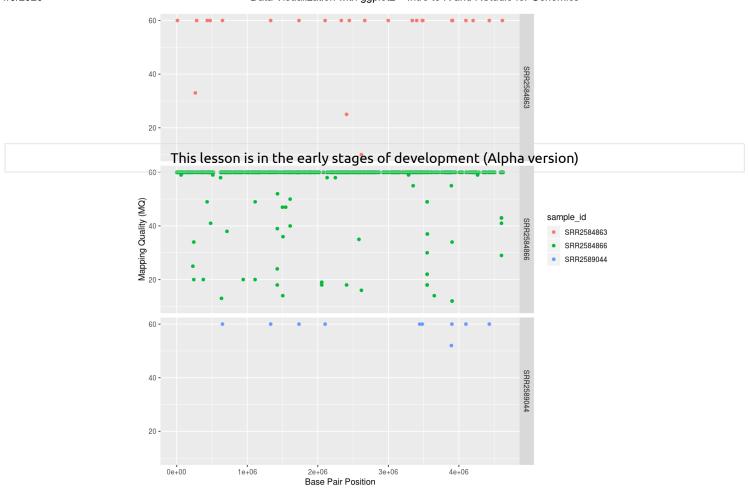
ggplot2 has a special technique called faceting that allows the user to split one plot into multiple plots based on a factor included in the dataset. We will use it to split our mapping quality plot into three panels, one for each sample.

```
ggplot(data = variants, aes(x = POS, y = MQ, color = sample_id)) +
geom_point() +
labs(x = "Base Pair Position",
    y = "Mapping Quality (MQ)") +
facet_grid(. ~ sample_id)
```



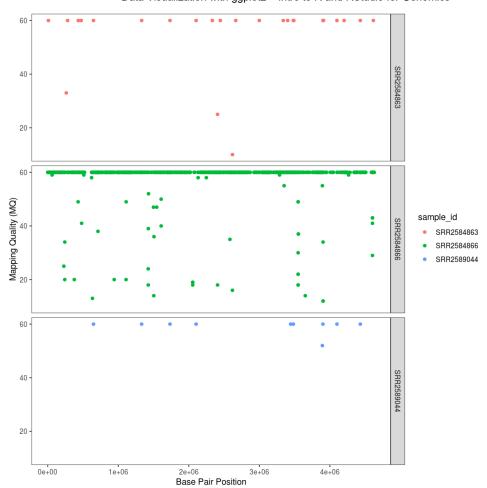
This looks ok, but it would be easier to read if the plot facets were stacked vertically rather than horizontally. The facet_grid geometry allows you to explicitly specify how you want your plots to be arranged via formula notation (rows ~ columns; a . can be used as a placeholder that indicates only one row or column).

```
ggplot(data = variants, aes(x = POS, y = MQ, color = sample_id)) +
geom_point() +
labs(x = "Base Pair Position",
    y = "Mapping Quality (MQ)") +
facet_grid(sample_id ~ .)
```



Usually plots with white background look more readable when printed. We can set the background to white using the function theme_bw(). Additionally, you can remove the grid:

```
ggplot(data = variants, aes(x = POS, y = MQ, color = sample_id)) +
  geom_point() +
  labs(x = "Base Pair Position",
        y = "Mapping Quality (MQ)") +
  facet_grid(sample_id ~ .) +
  theme_bw() +
  theme(panel.grid = element_blank())
```



Challenge

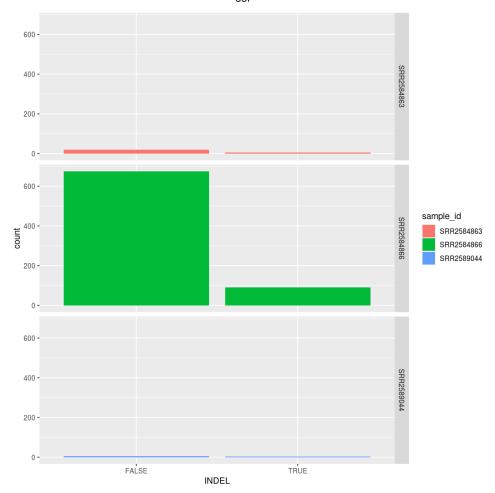
Use what you just learned to create a scatter plot of PHRED scaled quality (QUAL) over position (POS) with the samples



Barplots

We can create barplots using the geom_bar geom. Let's make a barplot showing the number of variants for each sample that are indels.

```
ggplot(data = variants, aes(x = INDEL, fill = sample_id)) +
 geom_bar() +
 facet_grid(sample_id ~ .)
```



Challenge Since we already have the sample id labels on the individual plot facets, we don't need the legend. Use the help file for geom_bar and any other online resources you want to use to remove the legend from the plot. Solution R ggplot(data = variants, aes(x = INDEL, color = sample_id)) + geom_bar(show.legend = F) + facet_grid(sample_id ~ .) 600 -400 -200 -600 -400 200 -600 400 -200 -FALSE TRUE INDEL

ggplot2 themes

In addition to theme_bw(), which changes the plot background to white, ggplot2 comes with several other themes which can be useful to quickly change the look of your visualization. The complete list of themes is available at https://ggplot2.tidyverse.org/reference/ggtheme.html (https://ggplot2.tidyverse.org/reference/ggtheme.html). theme_minimal() and theme_light() are popular, and theme_void() can be useful as a starting point to create a new hand-crafted theme.

The ggthemes (https://jrnold.github.io/ggthemes/reference/index.html) package provides a wide variety of options (including an Excel 2003 theme). The ggplot2 extensions website (https://www.ggplot2-exts.org) provides a list of packages that extend the capabilities of ggplot2, including additional themes.

Challenge

With all of this information in hand, please take another five minutes to either improve one of the plots generated in this exercise or create a beautiful graph of your own. Use the RStudio ggplot2 cheat sheet

(https://github.com/rstudio/cheatsheets/raw/master/data-visualization-2.1.pdf) for inspiration. Here are some ideas:

- See if you can change the size or shape of the plotting symbol.
- Can you find a way to change the name of the legend? What about its labels?
- Try using a different color palette (see http://www.cookbook-r.com/Graphs/Colors_(ggplot2)/).

• Key Points		
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•		

```
(../04-
dplyr/index.html)
(../XX
knitr-market
```

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