

Advanced graphics: practical 1

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First install the ggplot2 package

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

then install the course package

```
##Windows & Linux users
install.packages("nclRggplot2",
                 repos="http://R-Forge.R-project.org")
##Apple users
install.packages("nclRggplot2",
                 repos="http://R-Forge.R-project.org",
                 type="source")
```

To load the package, use

```
library(nclRggplot2)
```

1 Practical 1

To get you familiar with the underlying ggplot2 concepts, we'll recreate some standard graphics. Some of these plots aren't particularly useful, we are just using them for illustration purposes.

To BEGIN WITH, install the package

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

then load the package:

```
library(ggplot2)
## Loading required package: methods
```

Next we load the beauty data set:¹

```
library(nclRggplot2)
data(Beauty)
```

When loading in data, it's always a good idea to carry out a sanity check. I tend to use the commands

```
head(Beauty)
colnames(Beauty)
dim(Beauty)
```

¹ Details of the beauty data set can be found at the end of this practical.

2 Scatter plots

Scatter plots are created using the point geom. Let's start with a basic scatter plot

```
ggplot(data = Beauty) + geom_point(aes(x = age, y = beauty))
```

To save typing, we can also store the plot as a variable:

```
g = ggplot(data = Beauty)
g1 = g + geom_point(aes(x = age, y = beauty))
```

In this practical, we are creating the plots in a slightly verbose way.

To view this plot, type g1.

The arguments x and y are called aesthetics. For geom_point, these parameters are required. This particular geom has other aesthetics: shape, colour, size and alpha.² Here are some things to try out.

² These aesthetics are usually available for most geoms.

- Experiment with other aesthetics. For example,

```
g + geom_point(aes(x = age, y = beauty, colour = gender))
```

or

```
g + geom_point(aes(x=age, y=beauty,
                   alpha=evaluation, colour=gender))
```

Some aesthetics, like shape must be discrete. So use shape = factor(tenured).

- Are there any differences between numeric values like tenured and characters like gender for some aesthetics? What happens if you convert tenured to a factor in the colour aesthetic. For example, colour = factor(tenured).
- What happens if you set colour (or some other aesthetic) outside of the aes function? For example, compare

```
g + geom_point(aes(x = age, y = beauty, colour = "blue"))
```

to

```
g + geom_point(aes(x = age, y = beauty), colour = "blue")
```

- What happens when you set an aesthetic to a constant value. For example, colour=2. What happens if you put this argument outside of the aes function?

3 Box plots

The box plot geom has the following aesthetics: x, y, colour, fill, linetype, weight, size and alpha. We can create a basic boxplot using the following commands:

Adjustment	Description
dodge	Adjust position by overlapping to the side
fill	Stack overlapping elements; standardise stack height
identity	Do nothing
jitter	Jitter points
stack	Stack overlapping elements

Table 1: Position adjustments - table 4.5 in the ggplot2 book.

```
g + geom_boxplot(aes(x = gender, y = beauty))
```

Similar to the point geom, we can add in aesthetics:

```
g + geom_boxplot(aes(x=gender, y=beauty,
                     colour=factor(tenured)))
```

Why do you think we have to convert tenured to a discrete factor?

As before, experiment with the different aesthetics. For some of the aesthetics, you will need to convert the continuous variables to discrete variables. For example, this will give an error:

```
g + geom_boxplot(aes(x = gender, y = beauty, colour = tenured))
```

while this is OK

```
g + geom_boxplot(aes(x=gender, y=beauty, colour=factor(tenured)))
```

Make sure you play about with the different aesthetics.

4 Combining plots

The key idea with ggplot2 is to think in terms of layers not in terms of plot “types”. For example,

In the lectures we will discuss what this means.

```
g + geom_boxplot(aes(x=gender, y=beauty,
                     colour=factor(tenured))) +
  geom_point(aes(x=gender, y=beauty))
```

- What happens to the plot if you swap the order of the `geom_boxplot` and `geom_point` function calls?
- In this case, `geom_point` isn't that great. Try using `geom_jitter`:

We have a bit too much data for `geom_jitter`, but you get the point.

```
g + geom_boxplot(aes(x=gender, y=beauty,
                     colour=factor(tenured))) +
  geom_jitter(aes(x=gender, y=beauty))
```

tenured	minority	age	evaluation	gender	students	beauty
0	1	36	4.3	Female	43	0.202
1	0	59	4.5	Male	20	-0.826
1	0	51	3.7	Male	55	-0.660
1	0	40	4.3	Female	46	-0.766
0	0	31	4.4	Female	48	1.421

Table 2: The first five rows of the beauty data set. There are a total of 463 course evaluations.

5 Bar plots

The bar geom has the following aesthetics: `x`, `colour`, `fill`, `size`, `linetype`, `weight` and `alpha`. Here is a command to get started:

```
g + geom_bar(aes(x = factor(tenured)))
```

- As before, try different aesthetic combinations. Convert parameters to discrete versions as needed using `factor(...)`.
- Let's get a bit more fancy. First, we round ages to the nearest decade:

```
Beauty$dec = signif(Beauty$age, 1)
```

then plot:

```
g = ggplot(data = Beauty)
g + geom_bar(aes(x = gender, fill = factor(dec)))
```

We can adjust the layout of this bar plot using ggplot's position adjustments. The five possible adjustments are listed in table 1. The default adjustment is `stack`

```
g + geom_bar(aes(x=gender, fill=factor(dec)),
             position="stack")
```

- Try the other adjustments.

6 The beauty data set

This data set is from a study where researchers were interested in whether a lecturers' attractiveness affected their course evaluation.³ This is a cleaned version of the data set and contains the following variables:

- `evaluation` - the questionnaire result.
- `tenured` - does the lecturer have tenure; 1 == Yes. In R, this value is continuous.

- minority - does the lecturer come from an ethnic minority (in the USA).
- age.
- gender.
- students - number of students in the class.
- beauty - each of the lecturers' pictures was rated by six undergraduate students: three women and three men. The raters were told to use a 10 (highest) to 1 rating scale, to concentrate on the physiognomy of the professor in the picture, to make their ratings independent of age, and to keep 5 in mind as an average. The scores were then normalised.

Table 2 gives the first few rows of the data set.