Figures using ggplot2

January 23, 2019

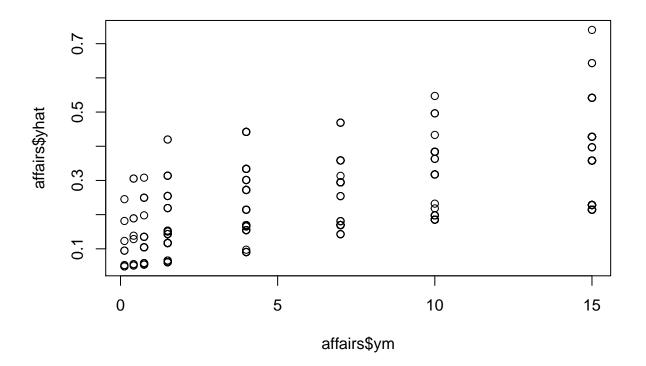
Figures using ggplot2

The graphical toolbox in R is particularly impressive. I very much like the default plotting library that ships with R: it is clear, simple, and you can build pretty much any figure you can imagine, because it allows to add and manipulate every single element of the figure.

Let's take the model we built before with some minor changes:

and create an easy plot of the predictions:

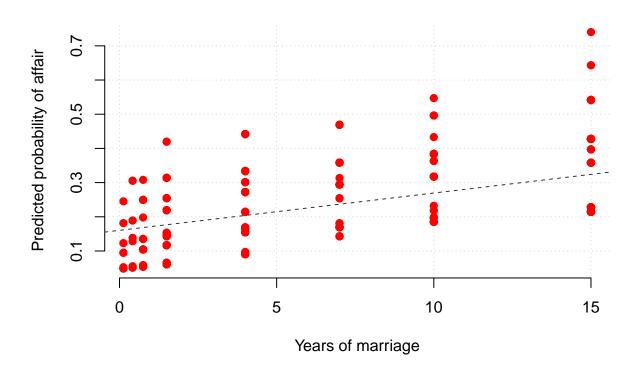
```
affairs$yhat <- predict(sample_model, newdata=affairs, type="response")
plot(affairs$ym, affairs$yhat)</pre>
```



and with some effort we could customize the picture a bit:

```
pch=19,
   bty="n",
   main="", xlab="", ylab="",
   panel.first=grid())
title(xlab="Years of marriage",
     ylab="Predicted probability of affair",
     main="Partial effect of years of marriage")
abline(lm(affairs$yhat ~ affairs$ym),
     lty=2,
     col="gray20")
legend(9.5, 0, "Best fit", col="gray20", lty=2, bty="n")
```

Partial effect of years of marriage

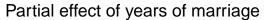


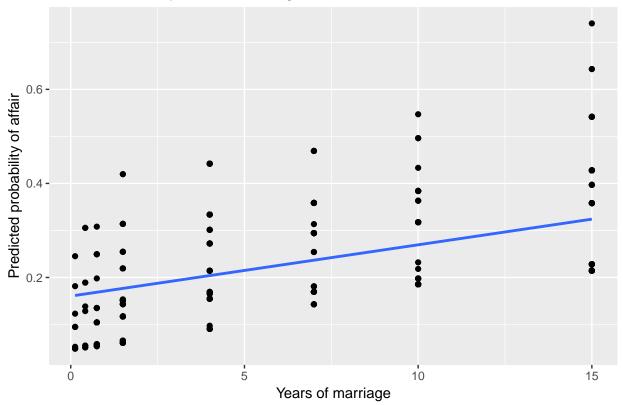
However, its flexibility comes at a cost, and figures usually require a lot of work.

The ggplot2, which implements the "grammar of graphics" approach to building data visualization, has rightfully gained a great deal of popularity, and it is the library that we will use here.

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

Before going into details, let's start



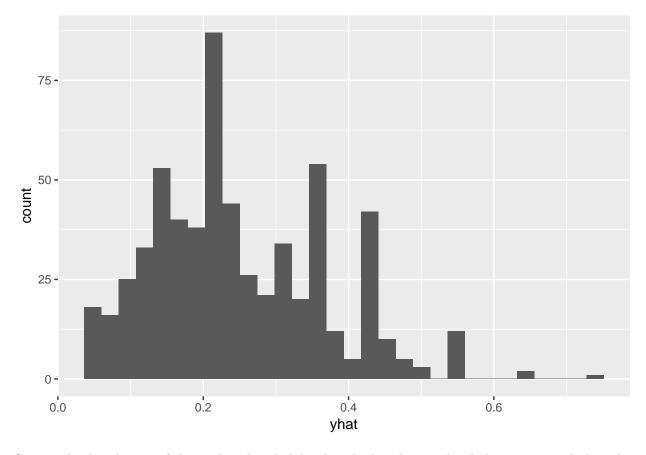


The advantages of ggplot2 as, I hope, can be seen above are that 1) it is much easier to create a publication-ready figure and 2) the syntax is also cleaner. If nothing else, ggplot2 offers "sensible" defaults for plots but in exchange the approach to building plots is slightly different from other languages —although it is currently adopted in many other tools.

The structure of a figure in ggplot2 is simple. We first define the data frame that we will be using, the basic aesthetics for the figure (which variables go to which axis, whether the dataset is grouped). We use the ggplot function for that. Then, we add the different layers of the figure that correspond to the different elements of ghe graph.

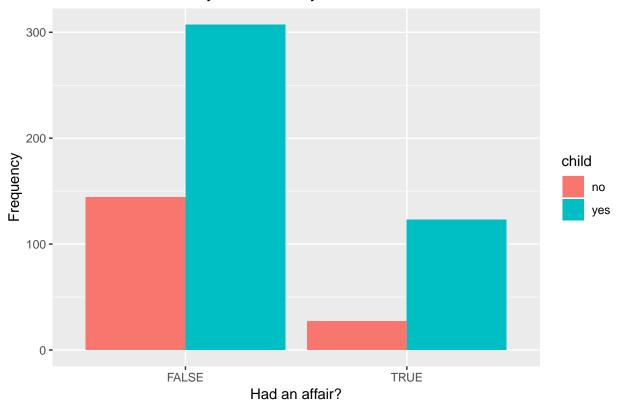
For instance, to define a histogram to plot the distribution of our predicted variable, all we need is one variable living in a dataset:

```
p <- ggplot(affairs, aes(x=yhat))
p + geom_histogram()</pre>
```



Getting the distribution of the predicted probability by whether the man has kids is easy: we declare that data is grouped by child and, in addition, that the bar corresponding to each of the values of child will have a different color. In addition, we add labels and a title (because we love good practices):

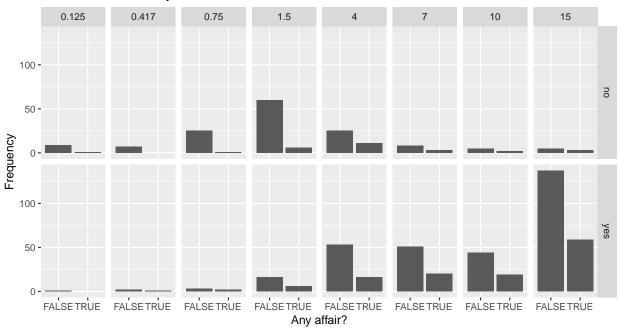
Likelihood of affair by whether they had children



ggplot2 that makes it very easy to get conditional plots. For instance, our theoretical model above posed an interaction between years of marriage and whether the man had children. This is, the expected number of affairs for a given time ellapsed in marriage is expected to be different depending on whether the man has children or not. Therefore, in the exploration of our data, we would like to see the number of affairs by ym and by children. For instance, we could look at:

```
p <- ggplot(affairs, aes(x=nbaffairs > 0, group=child))
p + geom_bar(position="dodge") +
   facet_grid(child ~ ym) +
   labs(title="Likelihood of affair by children", x="Any affair?", y="Frequency")
```





What we see in the representation above is a number of facets. The upper section represents the distribution for the cases in which the man didn't have children. The bottom section, the cases for which he did. In each of the vertical cells we see different durations of marriage. We see that men without children seem more likely have affairs right after being married but then the difference between the yes and no bars decrease over time after reaching a peak in around 1.5 years of marriage. However, for men with children, the difference grows over time.

We can now compare this result to what we estimated in the model above. To do that, we will repeat the same process we did before. First we create a fake data frame that captures the effect we are interested in, and then we apply our model to the data:

```
##
          ym child religious
                                     phat
## 1
       0.125
                             1 0.1814869
                 no
## 2
       0.417
                             1 0.1890527
                 no
## 3
       0.750
                             1 0.1979732
                 no
## 4
       1.500
                             1 0.2192134
                 no
## 5
       4.000
                             1 0.3013084
                 no
       7.000
## 6
                               0.4192001
                 no
  7
      10.000
                               0.5470998
##
                 no
## 8
      15.000
                             1 0.7402603
                 no
## 9
       0.125
                yes
                             1 0.4075765
## 10
       0.417
                             1 0.4101427
                yes
       0.750
                             1 0.4130751
## 11
                yes
## 12
       1.500
                             1 0.4197021
## 13
       4.000
                             1 0.4419892
                yes
```

```
## 14 7.000 yes 1 0.4690376
## 15 10.000 yes 1 0.4962690
## 16 15.000 yes 1 0.5416242
```

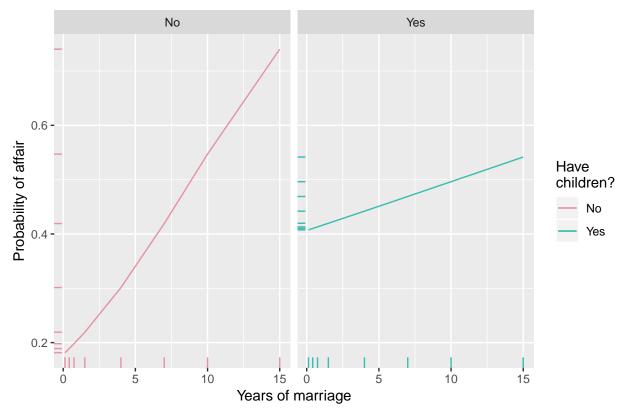
First some housekeeping:

```
library(scales)
levels(fake_data$child) <- c("No", "Yes")
cols <- colorspace:::rainbow_hcl(2)
print(cols)</pre>
```

[1] "#E495A5" "#39BEB1"

And now we plot things with some additional details to make it look ready for a presentation:

Predicted values of the main model



How to save a figure? The easiest way is probably to use the ggsave wrapper:

ggsave("my-pretty-figure.png", pq)