Extinction risk meta-analysis — Urban 2015

Urban (2015) conducted a meta-analysis of extinction risks and its relationship to climate change. He included 131 studies. In Figure 1, he plotted the number of studies reporting a certain overall proportion of extinction risk. The data (data/Urban2015_data.csv) is at a finer resolution than what needed for this figure. In fact, each study has been split into different lines according to the method and taxa used to compute the extinction risk. To reproduce Figure 1, you will need to coarse grain the data by grouping lines with the same author/year, and for each study compute the proportion of species at risk for extinction (sum the N.Ext for each study, and divide for the corresponding sum of Total.N). A close inspection of the original Figure shows that the data has been plotted in bins of unequal size (e.g., '0.5 < proportion < 1' is in one bin) so you will need to classify the various proportions into appropriate bins (0, 0-0.05, 0.05-0.1, ..., 0.5-1) before plotting. A ggplot2 version of Figure 1 of the original paper is reported in data/Urban2015_figure1.pdf. Reproduce the figure.

```
library(ggplot2)
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

u2015 <- read.csv("../data/Urban2015_data.csv", sep = '\t', stringsAsFactors = FALSE)
```

Now we need to coarse-grain the data according to the study (Author, Year). In particular, we want to compute the risk by summing all the N.Ext, and the total by summing Total.N. This can be accomplished by either cycling through the data and build a new data frame, or using the summarise function of dplyr:

```
by_study <- summarise(group_by(u2015, Author, Year), risk = sum(N.Ext), total = sum(Total.N))
# look at the results
by_study</pre>
```

```
## Source: local data frame [130 x 4]
## Groups: Author [?]
##
##
        Author Year risk total
##
         (chr) (int) (int) (int)
## 1
        Albouy
                2013
                         24
                               59
## 2
       Anciaes
                2006
                         15
                               98
## 3
      Anderson
                2009
                          1
                                6
## 4
        Araujo 2004
                        135
                             2400
                2006
                         49
                             5240
## 5
      Bakkenes
## 6
        Balint
                2011
                          4
                               18
```

```
## 7 Bambach 2013 34 168
## 8 Beaumont 2002 5 288
## 9 Beaumont 2005 0 150
## 10 Beaumont 2007 4 27
## ... ... ... ...
```

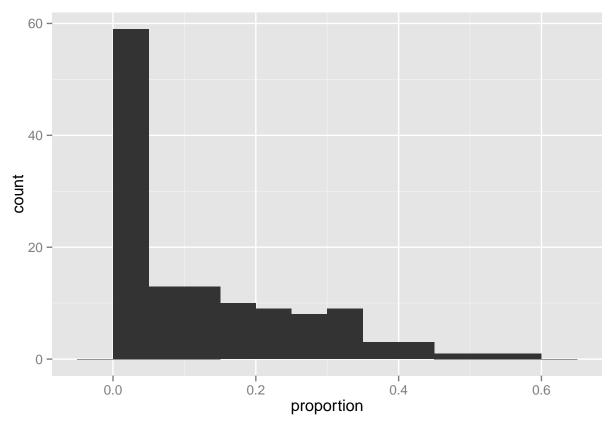
Now let's add a column expressing the proportion of species at risk of extinction:

```
by_study <- mutate(by_study, proportion = risk / total)
by_study</pre>
```

```
## Source: local data frame [130 x 5]
## Groups: Author [109]
##
##
       Author Year risk total proportion
##
        (chr) (int) (int) (int)
                                      (dbl)
## 1
       Albouy 2013
                       24
                             59 0.406779661
## 2
      Anciaes 2006
                       15
                             98 0.153061224
## 3
     Anderson 2009
                        1
                              6 0.16666667
                      135 2400 0.056250000
## 4
       Araujo 2004
## 5 Bakkenes 2006
                       49 5240 0.009351145
## 6
       Balint 2011
                        4
                             18 0.22222222
      Bambach 2013
## 7
                       34
                           168 0.202380952
## 8
     Beaumont 2002
                        5
                            288 0.017361111
## 9
               2005
                        0
                            150 0.000000000
     Beaumont
## 10 Beaumont 2007
                             27 0.148148148
## ..
```

We can plot the data using geom_bar and adjust the binwidth.

```
ggplot(data = by_study, aes(proportion)) + geom_bar(binwidth = .05)
```



However, it looks slightly different from Figure 1 in the publication given the unequal width of the bins. In order to exactly reproduce the figure, we can construct a column risk_bin that classifies the proportion of species at risk into bins of variable length. This can be accomplished using the function findInterval. For example:

head(cbind(by_study\$proportion, findInterval(by_study\$proportion, c(0, 0.000001, seq(0.05, 0.5, by = 0.000001))

```
## [,1] [,2]

## [1,] 0.406779661 10

## [2,] 0.153061224 5

## [3,] 0.166666667 5

## [4,] 0.056250000 3

## [5,] 0.009351145 2

## [6,] 0.222222222 6
```

Let's add this column to the table by_study:

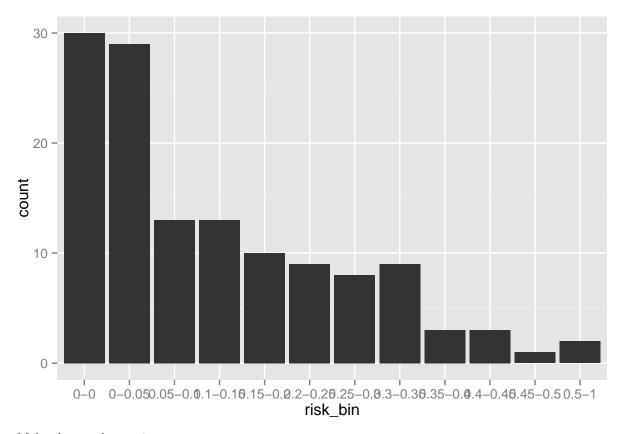
```
by_studyrisk_bin <- findInterval(by_study<math>proportion, c(0, 0.000001, seq(0.05, 0.5, by = 0.05)))
```

And transform this into a factor, with the right labels. First, build the labels for each bin:

```
leftbound <- c(0, seq(0.0, 0.45, by = 0.05), 0.5) leftbound
```

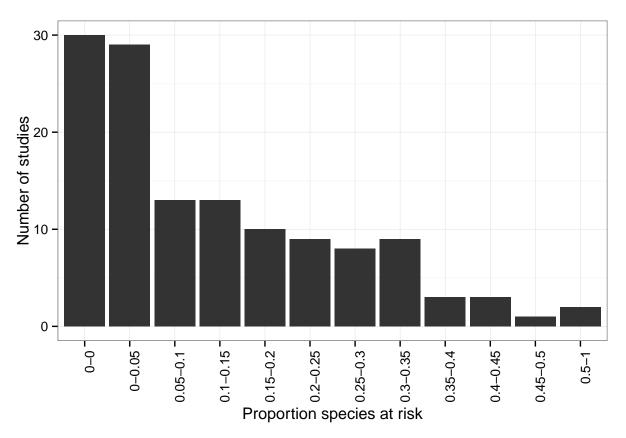
```
## [1] 0.00 0.00 0.05 0.10 0.15 0.20 0.25 0.30 0.35 0.40 0.45 0.50
```

```
rightbound <- c(0, seq(0.05, 0.5, by = 0.05), 1)
rightbound
## [1] 0.00 0.05 0.10 0.15 0.20 0.25 0.30 0.35 0.40 0.45 0.50 1.00
label_risk_bin <- paste(leftbound, rightbound, sep = "-")</pre>
label_risk_bin
## [1] "0-0"
                  "0-0.05"
                             "0.05-0.1" "0.1-0.15" "0.15-0.2" "0.2-0.25"
## [7] "0.25-0.3" "0.3-0.35" "0.35-0.4" "0.4-0.45" "0.45-0.5" "0.5-1"
# now transform bin_risk into factors
by_study$risk_bin <- factor(by_study$risk_bin, levels = 1:12, labels = label_risk_bin)</pre>
# see the result
by_study
## Source: local data frame [130 x 6]
## Groups: Author [109]
##
##
       Author Year risk total proportion risk_bin
        (chr) (int) (int) (int)
##
                                      (dbl)
## 1
       Albouy 2013
                    24 59 0.406779661 0.4-0.45
## 2
     Anciaes 2006 15
                            98 0.153061224 0.15-0.2
## 3 Anderson 2009 1
                            6 0.166666667 0.15-0.2
## 4
       Araujo 2004 135 2400 0.056250000 0.05-0.1
## 5 Bakkenes 2006 49 5240 0.009351145
                                            0-0.05
## 6
       Balint 2011
                      4 18 0.22222222 0.2-0.25
## 7
     Bambach 2013 34 168 0.202380952 0.2-0.25
## 8 Beaumont 2002
                      5 288 0.017361111
                                            0-0.05
                      0 150 0.000000000
## 9 Beaumont 2005
## 10 Beaumont 2007
                      4 27 0.148148148 0.1-0.15
## ..
          . . .
                . . .
                      . . .
After all this data reshaping, we are ready to plot!
pl <- ggplot(data = by_study, aes(x = risk_bin)) + geom_bar()</pre>
pl
```



Make the graph prettier:

```
pl <- pl + theme_bw() +
  theme(axis.text.x = element_text(angle = 90, hjust = 1)) +
  xlab("Proportion species at risk") +
  ylab("Number of studies")
pl</pre>
```



Ready to save the plot using ggsave:

```
ggsave(pl, file = "../data/Urban2015_figure1.pdf")
```

Saving 6.5 x 4.5 in image