Visualization of large multivariate datasets with the tabplot package

Martijn Tennekes and Edwin de Jonge

June 15, 2011

Abstract

The tableplot is a powerful visualization method to explore and analyse large multivariate datasets. In this vignette, the implementation of tableplots in R is described.

1 Introduction

The tableplot is a visualization method that is used to explore and analyse large datasets. By using tableplots, data analysts are able to observe the relationships between the variables, discover strange data patterns, and check the occurrence and selectivity of missing values.

An example of a tableplot applied to the diamonds dataset of the ggplot2 package is illustrated in Figure ??. Each column represents a variable. The whole data set is ordered according to one or more columns (in this case, carat), and then grouped into row bins. Algorithm ?? describes the creation of a tableplot into detail.

2 Getting started with the tableplot function

To illustrate the tabplot package, we will use the diamonds dataset that is provided in the ggplot2 package. This dataset contains information about 53,940 diamonds. There are 7 continuous variables and 3 categorical. In order to illustrate the visualization of missing values, we add several NA's.

- > require(ggplot2)
 > data(diamonds)
 > is.na(diamonds\$price) <- diamonds\$cut == "Ideal"
 > is.na(diamonds\$cut) <- (runif(nrow(diamonds)) > 0.8)
 - A tableplot is simply created by the function tableplot:
- > tableplot(diamonds)

Algorithm 1 Create tableplot

```
Input: Tabular dataset t, column i_s of which the distribution is of interest.
   1: t' \leftarrow \text{sort } t \text{ according to the values of column } i_s.
   2: Divide t' into n row bins according to the order of t'.
          for each column i do
                      if i is numeric then
                                 m_{ib} \leftarrow \text{mean value per bin } b
                                 c_{ib} \leftarrow \text{fraction of missing values per bin } b
   6:
                      end if
   7:
                     if i is categorical then
   8:
   9:
                                  f_{ijb} \leftarrow \text{frequency of each category } j \text{ (including missing values)}
                      end if
10:
11: end for
12: for each column i do
13:
                      if i is numeric then
                                  Plot a bar chart of the mean values \{m_{i1}, m_{i2}, \ldots, m_{in}\}. A loga-
14:
                                  rithmic scale can be used. The fraction of missing values \{c_{i1},
                                  c_{i2}, \ldots, c_{in} determines the lightness of the bar colour. The light-
                                  er the colour, the more missing values occur in bin b. If all values
                                  are missing, a light red bar of full length is drawn.
                      end if
15:
16:
                      if i is categorical then
                                  Plot a stacked bar chart according to the frequencies \{f_{i1b}, f_{i2b}, 
17:
                                  \ldots} for each bin b. Each category is shown is a distinct colour.
                                  If there are missing values, they are depicted by a red colour.
                      end if
18:
19: end for
```

The result is depicted in Figure ??. By default, all variables of the dataset are depicted. With the argument colNames, we can specify which variables are plotted.

Further, the dataset is by default sorted according to the values of the first variable. With the argument **sortCol**, we can specify on which variables the data is sorted.

The result is illustrated in Figure ??.

Setting an appropriate number of row bins (argument nbins) is important, like in a histogram. A good number of row bins is a trade of between good polished but meaningless data, and detaild, but noisy data. In practice, we found out that the default number of 100 usually is a good starting

point.

The percentages near the vertical axis indicate which subset of the data in terms of units (row) is depicted. The range from 0% to 100% in Figure ?? means that all units of the data are plotted.

Suppose we want to focus our attention to the 5% most expensive diamonds. This is possible by setting the from argument to 0, and the to argument to 5:

Observe that in the obtained tableplot in Figure ??, the number of row bins is still 100, so that the number of units per row bin is now 27 instead of 540. Therefore, much more detail can be observed in this tableplot.

The vertical axis contains two sets of tick marks. The set of small tick marks correspond with the row bins. The set of large tick marks correspond with the percentages. The number of tick marks between from and to is determined by R's function pretty.

3 Customizing the tableplot

3.1 Continuous variables

For each bin of a continuous variable, the mean value is calculated (see Algorithm??). When the distribution of these mean values is exponential, it is useful to apply a logarithmic transformation (see ?). The argument scales is default set to the auto-detection mode "auto", and can also set to linear mode "lin" or logarithmic mode "log".

Observe that the x-axes of the variables depth and table in Figure $\ref{eq:condition}$ are broken. The x-axis of a variable i is broken if either

```
0 < max(m_{i1}, m_{i2}, \dots, m_{in}) AND bias_brokenX \cdot max(m_{i1}, m_{i2}, \dots, m_{in}) < min(m_{i1}, m_{i2}, \dots, m_{in}) OR 0 > min(m_{i1}, m_{i2}, \dots, m_{in}) AND bias_brokenX \cdot min(m_{i1}, m_{i2}, \dots, m_{in}) > max(m_{i1}, m_{i2}, \dots, m_{in}),
```

where bias_brokenX is a bias parameter that should be a number between 0 and 1. If bias_brokenX=1 then the above conditions are always false, which implies that the x-axes are never broken. On the other hand, if bias_brokenX=0 then the x-axes are always broken. By default, bias_brokenX=0.8, which mean that an x-axis is broken if (in case of a variable with positive values) the minimum value is at least 0.8 times the maximum value. In the diamonds dataset, this applies to the variables depth and table.

3.2 Categorical variables

The color palettes of categorical variables can be customized with the argument pals. Several qualitative palettes from literature (?, ?) are implemented. They are stored in the list tabplotPalettes and can be shown by

> tableplot_showPalettes()

The default palette is a combination of Set1 and Set2. It has the advantage each category has a unique color for varibles with up to 16 categories.

Suppose we want a to use the default palette for the variable cut, but starting with the seventh color, pink. Further we want the color blind friendly palette for the variable color, but without the first color (black), and a custom palette, say a rainbow palette, for the variable clarity:

```
> tableplot(diamonds, pals = list(7, "col_blind_friendly(2)", rainbow(8)))
```

3.3 The tabplot object

The function tableplot returns a tabplot object, that can be used to make minor changes to the tableplot, for instance the order of columns or the color palettes. Of course, these changes can also be made by generating a new tableplot, such as in the examples above. However, if it takes considerable time to generate a tableplot, then it is practical to make minor changes immediately.

The output of the tableplot function can be assigned to a variable. The graphical output can be omitted by setting the argument plot to FALSE.

```
> tab <- tableplot(diamonds, plot = FALSE)
```

The tabplot object is a list that contains all information to depict a tableplot. The generic functions summary and plot can be applied to the tabplot object.

> summary(tab)

gene	general variable1					variable2			
variables:10		name		:carat		name		:cut	
objects	:53940	type		:numeric		type	:categorical		
bins	:100	sort		:decreasing		sort	:NA		
from	:0%	scale_	_init :auto ca		catego	gories:6			
to	:100%	scale_	final	:lin					
variable3			variable4				variable5		
name	:color		name)	:clarit	У	name	:depth	
type	:catego	rical	type)	:catego	rical	type	:numeric	
sort	:NA		sort	;	:NA		sort	:NA	

```
categories:7
                          categories:8
                                                    scale_init :auto
                                                    scale_final:lin
      variable6
                             variable7
                                                    variable8
name
           :table
                      name
                                  :price
                                             name
                                                         :x
           :numeric
                                                         :numeric
type
                      type
                                  :numeric
                                              type
sort
           :NA
                                  :NA
                                                         :NA
                       sort
                                              sort
                                              scale_init :auto
scale_init :auto
                      scale_init :auto
scale_final:lin
                       scale_final:lin
                                              scale_final:lin
      variable9
                             variable10
                                  :z
name
           :y
                      name
           :numeric
                                  :numeric
type
                      type
sort
           :NA
                       sort
                                  :NA
scale_init :auto
                      scale_init :auto
scale_final:lin
                      scale_final:lin
```

> plot(tab)

The function changeTabplot is used to make minor changes to a tabplot object. Suppose we want the columns in the order of ??, and we want to change all color palettes to default starting with the first color.

4 Graphical User Interface tableGUI

The function tableGUI starts a Graphical User Interface (GUI).

> tableGUI()

5 Final remarks

Conclusions

- A First appendix on the tabplot package
- B Second appendix