Assignment3

May 29, 2019

1 Assignment 3 - Building a Custom Visualization

In this assignment you must choose one of the options presented below and submit a visual as well as your source code for peer grading. The details of how you solve the assignment are up to you, although your assignment must use matplotlib so that your peers can evaluate your work. The options differ in challenge level, but there are no grades associated with the challenge level you chose. However, your peers will be asked to ensure you at least met a minimum quality for a given technique in order to pass. Implement the technique fully (or exceed it!) and you should be able to earn full grades for the assignment.

Ferreira, N., Fisher, D., & Konig, A. C. (2014, April). Sample-oriented task-driven visualizations: allowing users to make better, more confident decisions. In Proceedings of the SIGCHI Conference on Human Factors in Computing Systems (pp. 571-580). ACM. (video)

In this paper the authors describe the challenges users face when trying to make judgements about probabilistic data generated through samples. As an example, they look at a bar chart of four years of data (replicated below in Figure 1). Each year has a y-axis value, which is derived from a sample of a larger dataset. For instance, the first value might be the number votes in a given district or riding for 1992, with the average being around 33,000. On top of this is plotted the 95% confidence interval for the mean (see the boxplot lectures for more information, and the yerr parameter of barcharts).

Figure 1 from (Ferreira et al, 2014).

A challenge that users face is that, for a given y-axis value (e.g. 42,000), it is difficult to know which x-axis values are most likely to be representative, because the confidence levels overlap and their distributions are different (the lengths of the confidence interval bars are unequal). One of the solutions the authors propose for this problem (Figure 2c) is to allow users to indicate the y-axis value of interest (e.g. 42,000) and then draw a horizontal line and color bars based on this value. So bars might be colored red if they are definitely above this value (given the confidence interval), blue if they are definitely below this value, or white if they contain this value.

Figure 2c from (Ferreira et al. 2014). Note that the colorbar legend at the bottom as well as the arrows are not required in the assignment descriptions below.

Easiest option: Implement the bar coloring as described above - a color scale with only three colors, (e.g. blue, white, and red). Assume the user provides the y axis value of interest as a parameter or variable.

Harder option: Implement the bar coloring as described in the paper, where the color of the bar is actually based on the amount of data covered (e.g. a gradient ranging from dark blue for the

distribution being certainly below this y-axis, to white if the value is certainly contained, to dark red if the value is certainly not contained as the distribution is above the axis).

Even Harder option: Add interactivity to the above, which allows the user to click on the y axis to set the value of interest. The bar colors should change with respect to what value the user has selected.

Hardest option: Allow the user to interactively set a range of y values they are interested in, and recolor based on this (e.g. a y-axis band, see the paper for more details).

Note: The data given for this assignment is not the same as the data used in the article and as a result the visualizations may look a little different.

In [43]: # Use the following data for this assignment: import pandas as pd import numpy as np np.random.seed(12345) df = pd.DataFrame([np.random.normal(32000,200000,3650), np.random.normal(43000,100000,3650), np.random.normal(43500,140000,3650), np.random.normal(48000,70000,3650)], index=[1992,1993,1994,1995]) df 2 0 1 Out [43]: 1992 -8941.531897 127788.667612 -71887.743011 -79146.060869 -51896.094813 198350.518755 -123518.252821 -129916.759685 1993 1994 152336.932066 192947.128056 389950.263156 -93006.152024 1995 -69708.439062 -13289.977022 -30178.39099155052.181256 4 5 6 7 425156.114501 310681.166595 50581.575349 88349.230566 1992 216119.147314 49845.883728 149135.648505 1993 62807.672113 5529.230706 -32989.370488 1994 100818.575896 223942.967178 1995 152883.621657 12930.835194 63700.461932 64148.489835 8 9 3640 1992 185804.513522 281286.947277 171938.760289 23365.577348 -109686.264981 1993 -44566.520071 1994 -66721.580898 165085.806360 47826.269111 . . . 1995 -29316.268556 59645.677367 -13901.388118 3643 3641 3642 3644 1992 150650.759924 203663.976475 -377877.158072 -197214.093861 1993 101032.122475 117648.199945 160475.622607 -13759.888342

107329.726875 199250.734156 -36792.202754

1994

74735.174090

```
1995
               50173.686673 53965.990717 4128.990173
                                                             72202.595138
                       3645
                                      3646
                                                     3647
                                                                    3648
         1992 24185.008589 -56826.729535 -67319.766489 113377.299342
                                                                         -4494.878
         1993 -37333.493572 103019.841174 179746.127403 13455.493990 34442.898
         1994 -71861.846997
                             26375.113219 -29328.078384
                                                           65858.761714 -91542.003
         1995 39937.199964 139472.114293 59386.186379 73362.229590 28705.082
         [4 rows x 3650 columns]
In [44]: %matplotlib notebook
         import matplotlib.pyplot as plt
         from scipy import stats
         import matplotlib.colors as col
         import matplotlib.cm as cm
In [45]: #mean, standard dev
         mean = df.mean(axis=1)
         std = df.std(axis=1)
In [46]: years = list(df.index)
         y_pos = np.arange(len(years))
In [47]: n = df.shape[1]
In [48]: #confidence interval of 95%
         verr = std/np.sqrt(n) * stats.norm.ppf(1-0.05/2)
         conf_ints = [stats.norm.interval(0.95, loc=mu, scale=se) for mu, se in zig
In [49]: def compute_probs(y, conf_int):
             if y < np.min(conf_int):</pre>
                 result = 1.0
             elif y > np.max(conf_int):
                result = 0.0
             else:
                 result = (np.max(conf_int) - y)/(np.max(conf_int) - np.min(conf_ir
             return result
In [50]: #value of horizontal line
         y = 43000
In [51]: probs = [compute_probs(y, ci) for ci in conf_ints]
In [52]: cmap = cm.get_cmap('YlOrRd')
         cpick = cm.ScalarMappable(cmap=cmap, norm=col.Normalize(vmin=0, vmax=1.0))
         cpick.set_array([])
In [53]: rects = plt.bar(y_pos, mean, align='center', yerr=yerr, capsize=5, color =
<IPython.core.display.Javascript object>
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<IPython.core.display.HTML object>
In [54]: plt.axhline(y=y, zorder=1, color='k')
         yt = plt.gca().get_yticks()
         yt = np.append(yt, y)
         plt.gca().set_yticks(yt)
Out [54]: [<matplotlib.axis.YTick at 0x7fdd2d3aa860>,
          <matplotlib.axis.YTick at 0x7fdd2d0d9e10>,
          <matplotlib.axis.YTick at 0x7fdd2d21ad68>,
          <matplotlib.axis.YTick at 0x7fdd2d097ba8>,
          <matplotlib.axis.YTick at 0x7fdd2d09c5c0>,
          <matplotlib.axis.YTick at 0x7fdd2d09cf28>,
          <matplotlib.axis.YTick at 0x7fdd2d0a3940>,
          <matplotlib.axis.YTick at 0x7fdd2d12e358>]
In [55]: plt.xticks(y_pos, years)
Out [55]: ([<matplotlib.axis.XTick at 0x7fdd2d0e4128>,
           <matplotlib.axis.XTick at 0x7fdd2d0b1470>,
           <matplotlib.axis.XTick at 0x7fdd2d213278>,
           <matplotlib.axis.XTick at 0x7fdd2d1edda0>],
          <a list of 4 Text xticklabel objects>)
In [56]: cbar = plt.colorbar(cpick, orientation="vertical")
In [57]: [plt.gca().spines[loc].set_visible(False) for loc in ['top', 'right']]
         plt.show()
In [ ]:
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