Tutorial 9: To what extent do you think we should let the data speak for themselves?

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Introduction

In the world full data we live in today, there are countless data sets for us to explore. Endless data sets are released every day, every second. However, the dataset in its raw form may vary. Some data that includes many unnecessary "noises", while some data are very simple and include just the right amount of information we need. In most cases, the raw data sets are not "clean". The raw data sets may have missing values, incorrect values, and even incorrect formats. These traits often cause issues when we try to manipulate the data or create visuals with the data. Therefore, before we start creating visuals and manipulations, we often need to "clean" the data. This ensure that the data sets only include the values we need and would not causes any erros later down the road. However, how much data should be cleaned? To what extend should we manipulate the original data? This short essay will discuss the topic of "To what extent do you think we should let the data speak for themselves"?

Why clean the data?

As mentioned in the article (AU 2020), Randy mentioned that the goal of cleaning data is: 1. Fix things that will make your analysis algorithm choke, 2. Reduce Unwanted Variation, and 3. Eliminate Bias (where you can). I personally think that these points are very valid and in order to obtain our goal, these three methods and reason should be applied to data sets. As mentioned in the introduction, when a data contains many different names, data types, empty cells, it would often cause the algorithm to choke and malfunction. Imagine if I asked you to find all the "King" cards from this deck of cards, but there are many cards that are named "King", "kIng", "KINg", "KING". Which one would you pick out? You might think that all these words spell king and that some are typos so you pick them all out. However, algorithms are not smart like human beings, when we ask it to pick out "King", it would only pick out "King". As a result having many different versions of the word "King" spelled in a data set would cause the algorithm to miss all the other variations of the spelling. This then brings us directly to the second point, reduce unwanted variations. This is one major time consuming aspect of data cleaning. Lastly, we want to ensure that there are no bias in the data collected. Some examples may be cleaning out extreme outliers when trying to model something. As mentioned before, algorithms are not smart like human, they would not see the outliers as outliers and rather see it as part of the data. This could cause the model to be in accurate when running on data sets that doesn't include outliers

Too much data

A good example that Randy brought up is the largest data generation/collection system. The Larger Hadron Collider collects so much data that it is not even possible to store in the raw form, and for that reason physicists spend countless hours trying to break it down and even just to decide which data to keep which data to drop. Sometimes it is just impossible to keep and have everything. I think this would be something interesting in the future as AI and machine learning gets better and better. From the article (Jordan 2019), he mentioned about the mistakes that in the medical field causing countless of misjudgments. I hope that in the near future, this tedious work could be handed over to machine learning. However, this would mean that the algorithms from the machine learning needs to be really well trained, and also minimizing bias. With my current experience, ChatGPT can do the basic cleaning, but in terms of deciding which one data to keep and which data to drop, ChatGPT still has a long ways to go.

Too much cleaning?

However, Randy also mentioned that if we are doing an linguistic study? The examples that he mentioned is the amount of misspellings in the data, if we cleaned it all up, a linguistic

study wouldn't be able to discover all the different variations of how people might have spelled a word. This brings up a good point that mentioned in class countless times and we are often reminded to practice. Reproducibility! It is important to make sure that even if we cleaned a data set for our own use, the original remains untouched. This allows the next person to take the same data set and manipulate it in ways that would be beneficial for their own goal.

Conclusions

Finally, my personal thoughts on this topic is that we should allow the data to speak as much as it could. Even if we need to collect and do our own work, we must respect the data set and allow reproducibility for the next person. The original data should remain untouched as much as possible.

AU, RANDY. 2020. "Data Cleaning IS Analysis, Not Grunt Work." https://counting.substack.com/p/data-cleaning-is-analysis-not-grunt.

Jordan, Michael I. 2019. "Artificial Intelligence—The Revolution Hasn't Happened Yet." *Harvard Data Science Review* 1 (1).