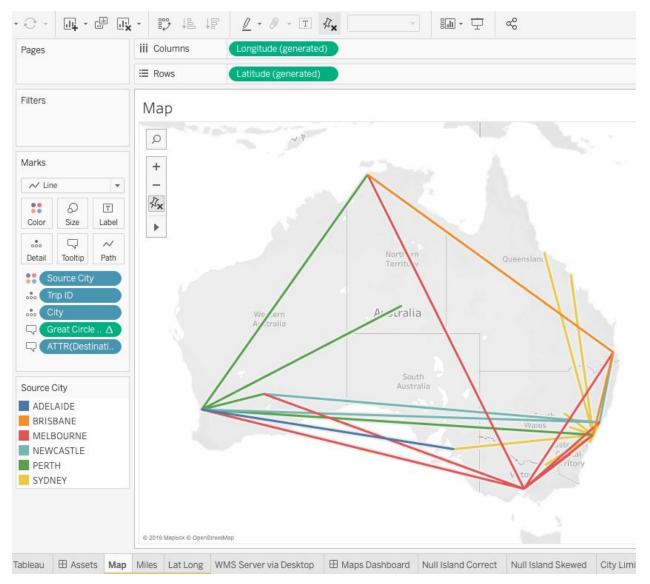
Assignment 3: Tableau Maps, Dashboards, and Animation

Team member: Chuyue Wu, Menghong Han, Xiaofan Sun, Yifan He

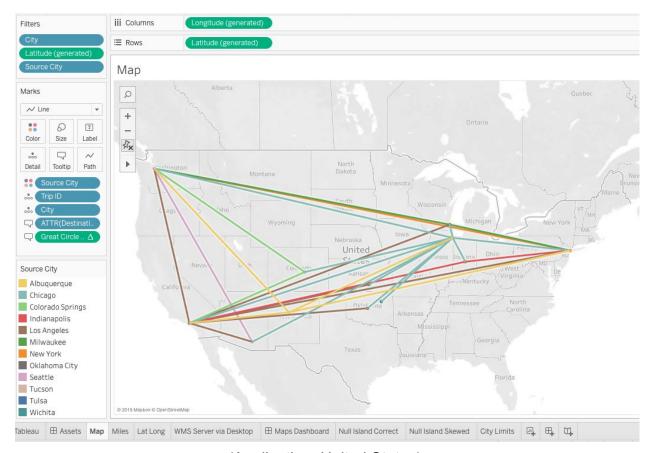
A. Chapter 9, page 272, displaying routes and calculating distances



(Replication: Australia)

Brief Introduction of Replication:

This is our screenshots from replication. It shows six main source cities in Australia and flight routes between these source cities and destination cities.



(Application: United States)

Brief Introduction of Application:

This is our screenshots from application. The new dataset is from the openflights website (https://openflights.org/data.html). We download airports.dat and routes.dat, merge them by airport ID, select US data, delete unnecessary columns, and melt them in the same structure as the replication data does. Then, we write them out of R in a csv file, import it into Tableau dataset, and repeat the plot process in Tableau again. To make the plot more clear, I filter 50 largest US cities by population but only 12 cities have location information in Tableau(latitude generated). Compared to dataset of Australia which has only 6 source cities, we believe 12 Source cities are enough.

Analogous Features:

Firstly, after clearing data in R, two datasets have the same structure. That's to say, two datasets both have Trip ID, city, Dest/Orig, so we can repeat the plot process easily in application.

Secondly, when plotting our own data, we didn't use all city information in the dataset, but select some important cities in the US. In replication, we also only plot 6 main cities in Australia. It's a good way to focus on big cities in two countries and make the plot more clearly.

Insight:

From the replication plot in Australia, we could find that source cities in the east coast of Australia has more flight routes such as Sydney and Melbourne. Some cities in the west coast of Australia and north coast of Australia are popular destination cities such as Perth and Darwin, but they are not as busy as Sydney and Melbourne. That makes sense because harbour cities have beautiful scenery, warm climate and convenient import and export system, which could make both people increase and economy develop.

From the application plot in the US, we could find that in the US, there are also some popular cities along the coasts such as New York City, Los Angeles and Seattle. However, due to the extreme long distance from the west coast to the east coast in the US, there are something different from Australia. We could find from the plot that the busiest source city that has most flight routes is Chicago. Chicago is located in the middle of the US, which has been a famous north-south transportation hub in the United States all the time. From the plot, we could tell that Chicago still plays an important role in today's air transportation system.

B. Chapter 10, page 308, creating a template

Screenshot for template

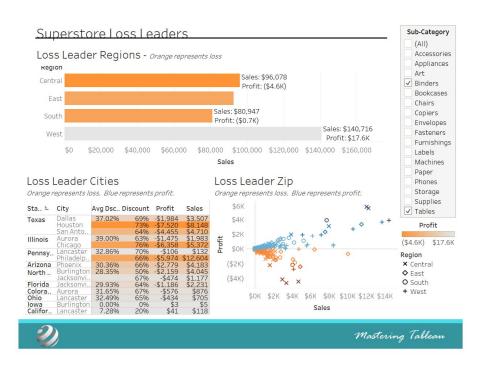
Template

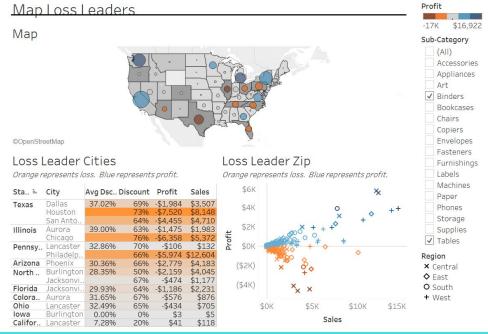


C.

For template: We use template completed in B and adjust the color of lower bar

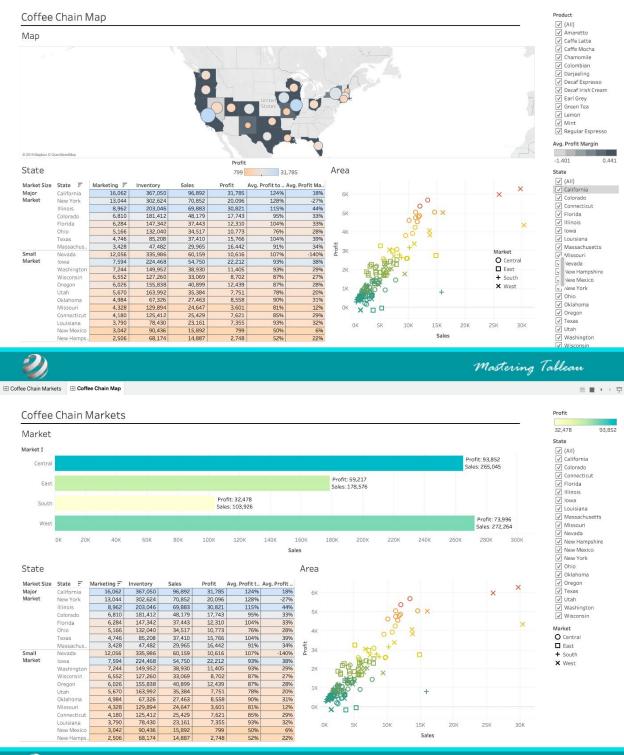
Screenshots for replication





2)

Screenshots for application



Coffee Chain Markets ## Coffee Chain Map

Description of the new application data

Brief introduction:

The Coffee Chain dataset is from Chapter 8 of Mastering Tableau 2019.2, downloaded from

https://www.packtpub.com/big-data-and-business-intelligence/mastering-tableau-20191-second-edition. The data provides operation information(including budget, expenses, sales, profits and so on) of different shops across varying areas of a coffee chain.

Analogous features:

Superstore dataset in the book calculates sales, profit generated by stores in each city of a region, with detailed information like discount and sub-category. In the 2 dashboards created using Superstore data, we can compare sales and profit in city, state and region level.

Just like Superstore data, Coffee Chain data we found also contains operation, product and geographic information of each chain store. Thus, we can make comparison between coffee chain in different areas based on dashboards made from this data.

Insights

For replication

First, among all the stores in the dataset, there's only half of them made a profit. Generally, west region did a better job than other areas, with the highest sales and profit in total.

Second, in this case, profit is not necessarily correlated with sales. There are not a few stores with high amount of sales failed to make a breakeven ultimately.

Finally, stores in categories related to paper(envelopes, labels, paper, copiers) are more likely to generate a profit. In contrast, furniture stores selling tables, chairs and bookcases can be hard to make a breakeven.

For application

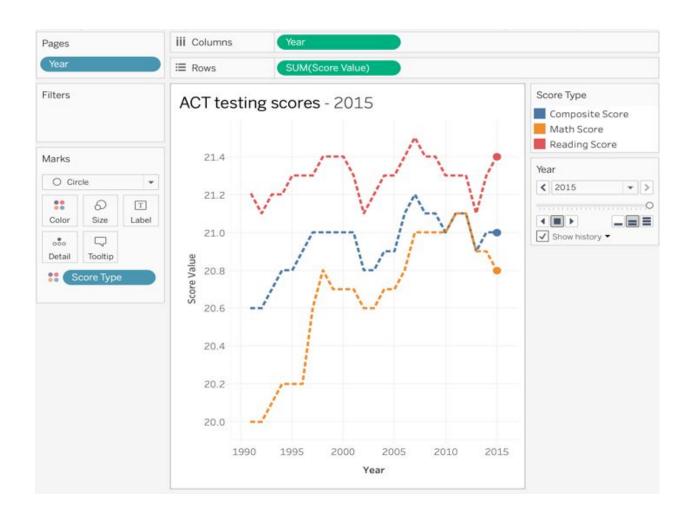
In overall, most coffee stores gain a profit. Central and western area have higher total sales and profit. It might be because the central and western regions have larger marketing size.

In this coffee shops case, profit is highly correlated with sales. Shops selling more tend to make more profit, no matter what types of coffee they sell.

Another finding is that among all the products sold by the coffee chain, mint and regular espresso are least popular. Regular espresso is only available in east area..

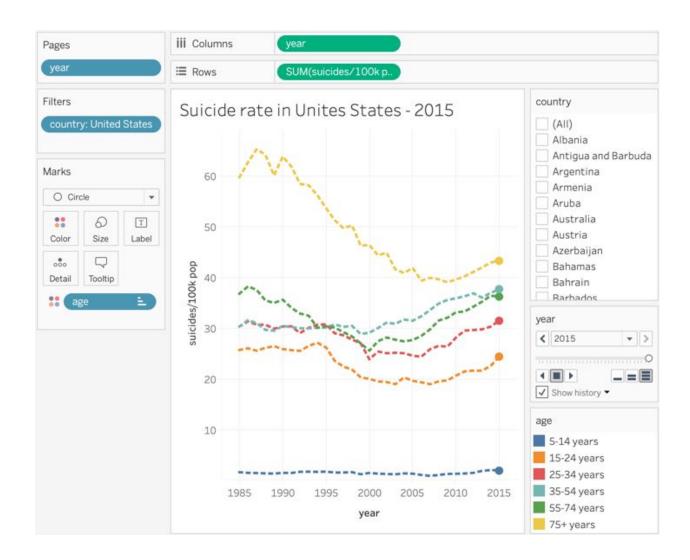
D. Chapter 10, page 322, creating an animation with Tableau

Screenshots for replication



Screenshots for application

(Suicide rate/100k pop in United States 1985-2015)



Description of the new application data

Brief introduction:

The dataset is downloaded from

https://www.kaggle.com/russellyates88/suicide-rates-overview-1985-to-2016, which

includes 4 columns: country (select "United States", year, age group, suicide rate/100k pop.

	country T	# year	Ψ	A age	Ψ	# suicides/100k pop T
	4	111111	85 2016	15-24 years 35-54 years Other (4)	17% 17% 67%	
		1985				0 225
1	United States		1985	35-54 years		22.77
2	United States		1985	35-54 years		7.58
3	United States		1986	35-54 years		23.57
4	United States		1986	35-54 years		8.06
5	United States		1987	35-54 years		23.28
6	United States		1987	35-54 years		7.73
7	United States		1988	35-54 years		22.4
8	United States		1988	35-54 years		7.3
9	United States		1989	35-54 years		22.61
10	United States		1989	35-54 years		6.91

Analogous feature:

The ACT test score data in the book includes 3 columns, which are score type (group), score value (value) and year (time). With the data we can create an animation and see the different types of score value change along with time.

Similarly, the new dataset I chose also includes 3 columns, which are age group (group), suicide rate/100k pop (value), year (time). With the data we can create an animation and see the different suicide rate in different age group change along with time.

Insight

For replication

We saw all kinds of ACT testing scores increase from 1990 to 2015 although fluctuations exit, among which math score has the maximum increase but still ranks third by the end of 2015 after reading score and composite score. That is to say, reading is easier for students compared to math and compositing.

For replication

From 1985 to 2015, the suicide rate of United States decreases and all of them are below 45 (/100k pop), which is a good sign. Generally speaking, the suicide rate correlated with age, the older the higher the suicide rate is, except for the age group of 35-54, which is higher than he age group of 55-74 after 1995. Maybe we should pay special attention to the 35-54 age group when promoting suicide prevention measures.

The animation via tableau enables us to gain a clearer and more direct view of the dynamic trend throughout the years and see the components of interested variables in details. However, an animation can certainly make for engaging communication but does not work well for deeper dives into the data. For example, we cannot figure out the reason behind why people among 35-54 have a relatively high suicide rate. Therefore, other visualization and analysis tools should be considered to dig further.