Final Assignment WorkEarly Report

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March 2023

1 Introduction

This is a brief report regarding the Final WorkEarly Data Science School Project (https://github.com/Workearly/Final-Assignment). A dataset that contains Liquor Sales in the state of Iowa in USA between 2012-2020 is given. The purpose is to find the most popular item per zipcode and the percentage of sales per store in the period between 2016-2019.

2 Brief Workflow

- Write a simple MySQL query to collect data from finance_liquor_sales.
 sql dataset that includes the sales that took place between 01-01-2016 and 31-12-2019.
- Save MySQL output to a csv file, called Liquor_Sales_2016_2019.csv.
- Write a Python code to read, inspect, process the data and produce the desired outputs.
- Visualize the outputs using matplotlib and seaborn
- Re-process the data and visualize the outputs using Tableau.

3 Data Processing

3.1 Get 2016-2019 data using MySQL

The finance_liquor_sales.sql database was loaded in MySQL Workbench and then a simple query was written in order to get the sales data between 01-01-2016 and 31-12-2019. The query as well as the output is demonstrated in Fig. 1. At first, the database was loaded and inspected in order to understand its structure. Then we use the WHERE and BETWEEN commands to keep the sales between 2016-2019.

After obtaining the results above, we export them to the Liquor_Sales_2016_2019.csv file, in order to further process them via Python.

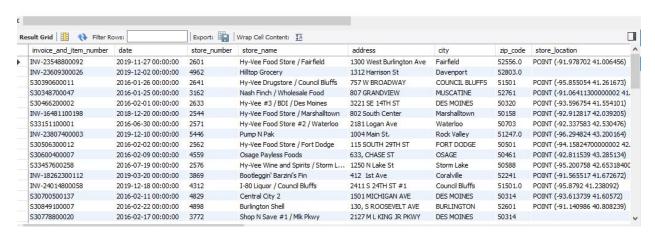


Figure 1: MySQL Query with corresponding results

3.2 Data Processing using Python/Pandas and Visualiza-

At first, we import the essential packages and read the csv file using Pandas:

```
import numpy as np
import pandas as pd
import seaborn as sns
from matplotlib import pyplot as plt

####### import Liquor_Sales_2016_2019.csv which includes
####### Liquor Sales
####### in the state of Iowa in USA between 2016-2019.

file2read='Liquor_Sales_2016_2019.csv'
df=pd.read_csv(file2read)
```

At next, we perform a first inspection of the dataframe's content, in order to better understand it. For this reason, we print the columns' names and get the 10 first rows' values (4 columns selected to appear through the Nc variable):

```
Nc=4
########## first inspection
###### get columns' names:

print(df.columns, '\n')
###### select to display Nc columns
pd.set_option('display.max_columns', Nc)
####### check headers (first 10 rows)'
print(df.head(10))
```

The output appeared is shown in Fig. 2:

Furthermore, we reveal the popularity of each item's in terms of sales (Fig. 3:

```
###### show most popular (top sales) item IDs in Iowa
print(df['item_number'].value_counts())
```

Then, in order to reveal the summary of bottles sold per item ID and zip code, we use the *groupby* operation, summing the bottles sold per each group. The produced dataframe's indices as well as the bottles' summaries are transformed to (parallel) numpy arrays and properly grouped in a single *data1* numpy array.

Figure 2: Dataset columns' names

```
In [3]: print(df['item_number'].value_counts())
86251
48099
          4
43031
          4
77487
          3
43034
          3
67557
          2
86112
          2
67586
          2
          2
56193
          2
67524
75087
           2
35917
          1
43040
          1
38089
          1
67526
          1
86843
          1
82187
          1
86739
```

Figure 3: Top sales item IDs

```
####### group by zip code and item number, sum of bottles
####### sold per item in each zip code

bottles_items_by_zip=df.groupby(['zip_code','item_number'\
,'item_description'])['bottles_sold'].sum()

#### to numpy

zipcodes=np.array(bottles_items_by_zip.index.\
get_level_values(0),dtype=np.intc)
itemids=np.array(bottles_items_by_zip.index.\
get_level_values(1),dtype=np.intc)
itemdesc=np.array(bottles_items_by_zip.index.\
get_level_values(2))
bottles1=bottles_items_by_zip.to_numpy(dtype=np.intc)
data1=np.array([zipcodes,itemids,itemdesc,bottles1])
```

To visualize this data we use a *seaborn* package scatterplot:

```
axes1=sns.scatterplot(x=zipcodes,y=bottles1,hue=itemdesc,\
                     data=itemids)
axes1.set(xlabel='Zip_Code', ylabel='#_of_bottles_sold')
plt.title('Bottles_sold_by_zip_code_and_item')
plt.legend(ncol=4,shadow=True, bbox_to_anchor=(0.5, -0.55),
             loc='lower_center', borderaxespad=0)
plt.setp(axes1.get_legend().get_texts(), fontsize='7')
plt.subplots_adjust(bottom=0.35)
plt.xlim(min(zipcodes),max(zipcodes))
plt.ylim(-10, \mathbf{max}(bottles1)+100)
major_ticks = np.arange(min(zipcodes), max(zipcodes), 250)
minor_ticks = np.arange(min(zipcodes), max(zipcodes), 50)
axes1.set_xticks(major_ticks)
axes1.set_xticks(minor_ticks, minor=True)
\begin{array}{ll} axes1.\,grid\,(\,which='minor\,'\,,\ alpha\,{=}\,0.2)\\ axes1.\,grid\,(\,which='major\,'\,,\ alpha\,{=}\,0.5) \end{array}
manager = plt.get_current_fig_manager()
manager.window.showMaximized()
```

The following figure appears (Fig. 4):

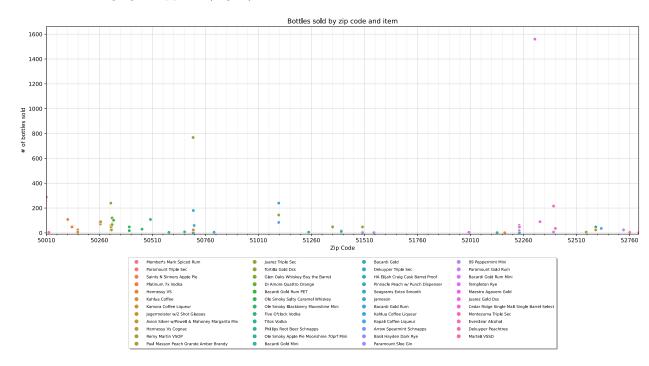


Figure 4: Bottles sold by zip code and item.

Then we create a horizontal bar chart of the top 20 stores by the number of bottles sold in descending order (the top number can be configured through the TopN variable. In this part of code, we call the customly created <code>create_horz_barplot</code> function.

```
####### group by store (number), sum of sales
####### group by store (number), sum of sales
####### (in bottles) per store
TopN=20
saleskey='(bottles)'
bottles_by_store=df.groupby(['store_number','city'])\
['bottles_sold'].sum()
TotalSalesBottles=np.sum(bottles_by_store)
percofbottles_by_store=100*bottles_by_store/TotalSalesBottles
percofbottles_by_store_sort=percofbottles_by_store.sort_values\
(ascending=False).round(decimals=2)
palette='GnBu_d'
create_horz_barplot(percofbottles_by_store_sort,TopN,palette,saleskey)
```

The code written to implement the create_horz_barplot function is demonstrated below:

```
##### function visualize with Seaborn horizontal barplot
def create_horz_barplot(df,TopN,palette,saleskey):
    storeids=np.array(df.index.get_level_values(0),dtype=np.intc)
    store_city=np.array(df.index.get_level_values(1))
    storeinfo=np.column_stack((storeids, store_city))
    storeinfoall=np.apply\_along\_axis(lambda d: str(d[0]) + ', \_'
    + d[1], 1, storeinfo)
    percvals=df.to_numpy()
    plt.figure()
    axes=sns.barplot(y=storeinfoall[0:TopN],x=percvals[0:TopN],\
    palette=palette)
    axes.set(xlabel='Sales_Percentage_(%)', ylabel='Store,_City')
    plt.title('Top_'+str(TopN)+'_Stores_by_sales_'+saleskey+'_\
___percentage')
    axes.grid()
    manager = plt.get_current_fig_manager()
    manager.window.showMaximized()
```

The produced barchart is depicted in Fig. 5 below:

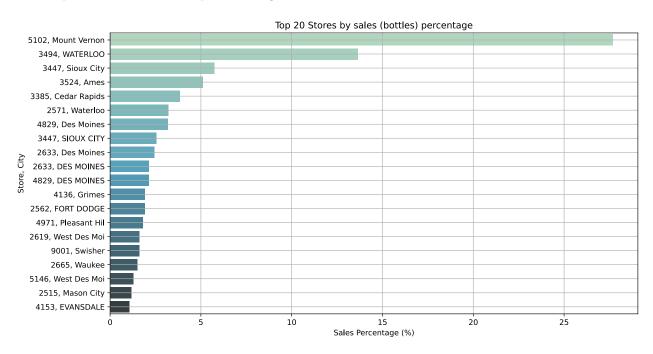


Figure 5: Top 20 stores by sales (number of bottles) percentage

The same process was followed to reveal the first 20 stores by sales in dollars (percentages):

```
####### group by store (number), sum of sales ($) per store
TopN=20
saleskey='($)'
sales_by_store=df.groupby(['store_number','city'])['sale_dollars'].sum()
TotalSales=np.sum(sales_by_store)
###### get percentages
percofsales_by_store=100*sales_by_store/TotalSales
percofsales_by_store_sort=percofsales_by_store.sort_values\
(ascending=False).round(decimals=2)
palette='YlOrBr'
create_horz_barplot(percofsales_by_store_sort,TopN,palette,saleskey)
```

The produced barchart is depicted in Fig. 6 below:

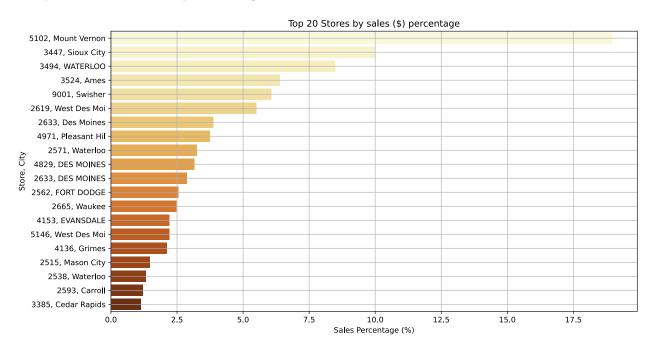


Figure 6: Top 20 stores by sales (\$) percentage

The same process was followed to reveal the first 20 stores by sales in terms of gallons (percentages):

```
TopN=20
saleskey='(gallons)'
gallons_by_store=df.groupby(['store_number','city'])\
['volume_sold_gallons'].sum()
TotalGallons=np.sum(gallons_by_store)
##### get percentages
percofgallons_by_store=100*gallons_by_store/TotalGallons
percofgallons_by_store_sort=percofgallons_by_store.sort_values\
(ascending=False).round(decimals=2)
palette='icefire'
create_horz_barplot(percofgallons_by_store_sort,TopN,palette,saleskey)
```

with the following barchart produced (Fig. 7):

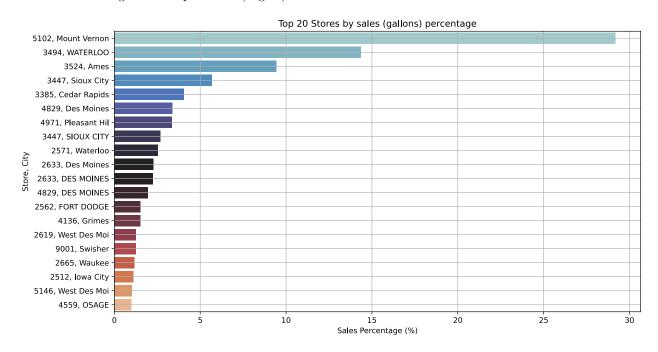


Figure 7: Top 20 stores by sales (gallons) percentage

The whole code written to implement the processing and visualization above is presented below:

```
import numpy as np
import pandas as pd
import seaborn as sns
from matplotlib import pyplot as plt
###### import Liquor_Sales_2016_2019.csv which includes Liquor Sales
\#\#\#\# in the state of Iowa in USA between 2016-2019.
file2read='Liquor_Sales_2016_2019.csv'
df=pd.read_csv(file2read)
Nc=4
\#\#\#\#\#\#\# first inspection –
\#\#\#\#\# get columns 'names:
print(df.columns, '\n')
##### select to display Nc columns
pd.set_option('display.max_columns', Nc)
###### check headers (first 10 rows);
print (df. head (10))
###### show most popular (top sales) item IDs in Iowa
print(df['item_number'].value_counts())
####### group by zip code and item number, sum of bottles
\#\#\#\#\# sold per item in each zip code
bottles_items_by_zip=df.groupby(['zip_code','item_number','item_description'])\
    ['bottles_sold'].sum()
\#\#\# to numpy
zipcodes=np.array(bottles_items_by_zip.index.get_level_values(0),dtype=np.intc)
itemids=np.array(bottles_items_by_zip.index.get_level_values(1),dtype=np.intc)
itemdesc=np.array(bottles_items_by_zip.index.get_level_values(2))
bottles1=bottles_items_by_zip.to_numpy(dtype=np.intc)
data1=np.array([zipcodes, itemids, itemdesc, bottles1])
###### visualize with Seaborn scatterplot
axes1=sns.scatterplot(x=zipcodes,y=bottles1, hue=itemdesc,\
                  data=itemids)
axes1.set(xlabel='Zip_Code', ylabel='#_of_bottles_sold')
plt.title('Bottles_sold_by_zip_code_and_item')
plt.legend(ncol=4,shadow=True,bbox_to_anchor=(0.5, -0.55),
           loc='lower_center', borderaxespad=0)
plt.setp(axes1.get_legend().get_texts(), fontsize='7')
```

```
plt.subplots_adjust(bottom=0.35)
plt.xlim(min(zipcodes),max(zipcodes))
plt.ylim(-10, \mathbf{max}(bottles1)+100)
major_ticks = np.arange(min(zipcodes), max(zipcodes), 250)
minor_ticks = np.arange(min(zipcodes), max(zipcodes), 50)
axes1.set_xticks(major_ticks)
axes1.set_xticks(minor_ticks, minor=True)
axes1.grid(which='minor', alpha=0.2) axes1.grid(which='major', alpha=0.5)
manager = plt.get_current_fig_manager()
manager.window.showMaximized()
#### function visualize with Seaborn horizontal barplot
def create_horz_barplot (df, TopN, palette, saleskey):
    storeids=np.array(df.index.get_level_values(0),dtype=np.intc)
    store_city=np.array(df.index.get_level_values(1))
    storeinfo=np.column_stack((storeids, store_city))
    storeinfoall = np. apply_along_axis(lambda d: str(d[0]) + ', ' + 
                                     d[1], 1, storeinfo)
    percvals=df.to_numpy()
    plt.figure()
    axes=sns.barplot(y=storeinfoall[0:TopN],x=percvals[0:TopN],palette=palette)
    axes.set(xlabel='Sales_Percentage_(%)', ylabel='Store,_City')
    plt.title('Top_'+str(TopN)+'_Stores_by_sales_'+saleskey+'_percentage')
    axes.grid()
    manager = plt.get_current_fig_manager()
    manager.window.showMaximized()
###### group by store (number), sum of sales (in bottles) per store
TopN=20
saleskey='(bottles)'
bottles_by_store=df.groupby(['store_number', 'city'])['bottles_sold'].sum()
TotalSalesBottles=np.sum(bottles_by_store)
percofbottles_by_store=100*bottles_by_store/TotalSalesBottles
percofbottles_by_store_sort=percofbottles_by_store.sort_values
    (ascending=False).round(decimals=2)
palette='GnBu_d'
create_horz_barplot (percofbottles_by_store_sort, TopN, palette, saleskey)
####### group by store (number), sum of sales ($) per store
TopN=20
saleskey='(\$)'
sales_by_store=df.groupby(['store_number', 'city'])['sale_dollars'].sum()
TotalSales=np.sum(sales_by_store)
```

```
\#\#\#\# get percentages
\verb|percofsales_by_store| = 100*sales_by_store/TotalSales|
percofsales_by_store_sort=percofsales_by_store.sort_values
    (ascending=False).round(decimals=2)
palette='YlOrBr'
create_horz_barplot(percofsales_by_store_sort, TopN, palette, saleskey)
####### group by store (number), sum of sold gallons per store
TopN=20
saleskey='(gallons)'
gallons_by_store=df.groupby(['store_number','city'])\
    ['volume_sold_gallons'].sum()
TotalGallons=np.sum(gallons_by_store)
##### qet percentages
percofgallons_by_store=100*gallons_by_store/TotalGallons
percofgallons_by_store_sort=percofgallons_by_store.sort_values
    (ascending=False).round(decimals=2)
palette='icefire'
create_horz_barplot(percofgallons_by_store_sort, TopN, palette, saleskey)
```

3.3 Visualization using Tableau

In this part, we exploit Tableau Public in order to visualize some aspects of the Liquor_Sales_2016_2019.csv dataset. Fig. 8 shows the data loaded in Tableau DataSource.

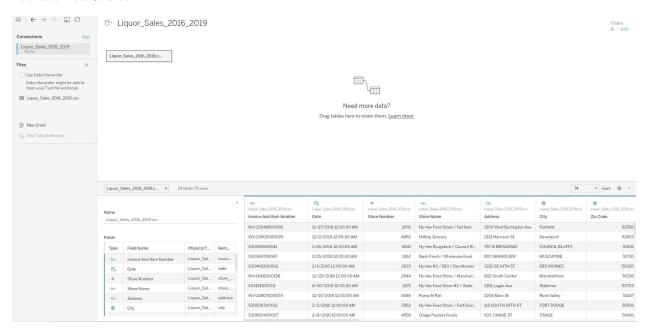


Figure 8: Tableau DataSource.

Figure 9 depicts the number of bottles sold, grouped by per Zip code and item (descending). The corresponding link to this sheet can be found in here.

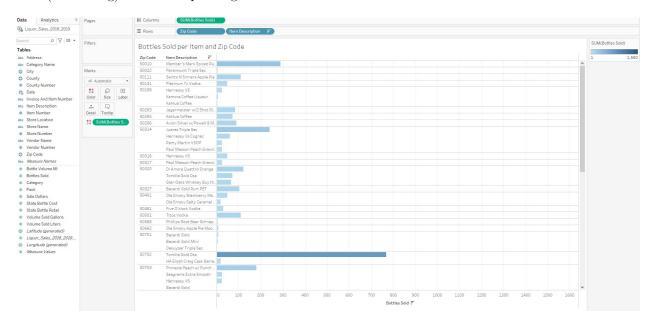


Figure 9: Bottles sold per zip code and item.

Figure 10 depicts the sales value in dollars per item store (raw \$ values, not percentage). The corresponding link to this sheet can be found in here.

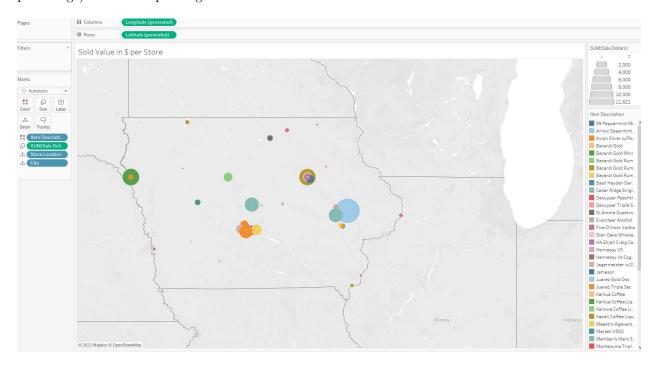


Figure 10: Sold value in (\$) per store.

Figure 11 depicts the sales value in gallons per item store (raw gallons' values, not percentage). The corresponding link to this sheet can be found in here.

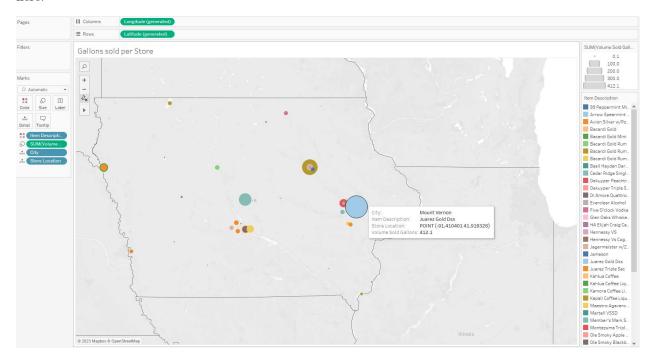


Figure 11: Sold value in gallons per store.

Figure 12 depicts the number of bottles sold per item store (number of bottles, not percentage). The corresponding link to this sheet can be found in here.

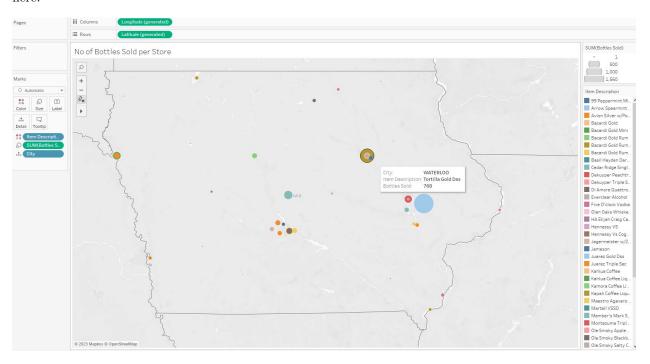


Figure 12: Sold number of bottles per store.

Figure 13 depicts the percentage of total bottles sold per item store (horizontal bar plot, descending order). The corresponding link to this sheet can be found in here.

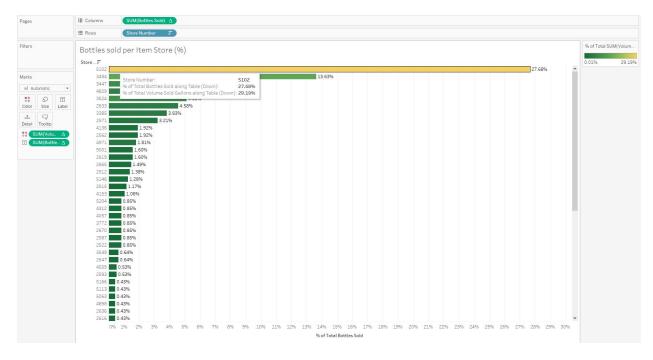


Figure 13: Sold percentage of bottles per store.

Figure 14 demonstrates the percentage of total sales in dollars per item store (circular area marks). The corresponding link to this sheet can be found in here.

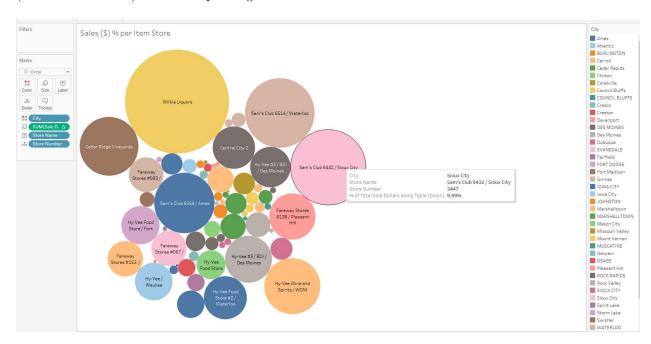


Figure 14: Percentage of dollars sales per store.

Finally, we create a dashboard which displays the whole diagrams/charts displayed in Dashboard, also shown in Fig. 15. Action highlights between different sheets have been added.

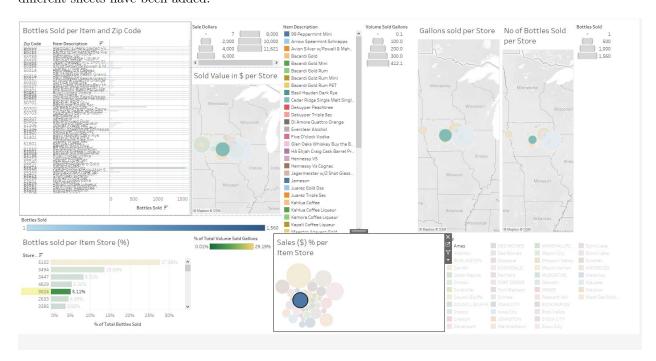


Figure 15: Dashboard