Report_Draft_Analysis Geography_v5

December 5, 2024

0.1 Geographic Impacts

The geographic distribution of sales across the US was examined. The results were very surprising and could either be due to very poor data collection or very unusual sales patterns.

The sales data was analysed in a variety of ways (box plot, scatter plot with regression line, top and bottom 5, a choropleth map) but all supported the following: - The spread of sales number and \$ value is very small when the big differences in population across states are considered - For example the 4th highest sales are in the US Minor Outlying Islands which has a population of 300, and spending as much as Alabama which has a population of over 5 million - And, the largest state, California, with a population of nearly 39 million has the lowest sales total - The spread of customers across states is also very even, the interquartile range is approximately 80 to 95

This is in contrast to the number of physical store per state which do correlate well with population size. So it does seem likely that there is a problem with the quality of the sales data collected. Before making any business recommendations, the data collection pipeline and sources should be investigated; or worst case the operational information systems that provided the source data.

```
[1]: # Retrieve the completed working dataframes for analysis
%store -r sales_df stock_df customer_reviews_df customers_df states_df
```

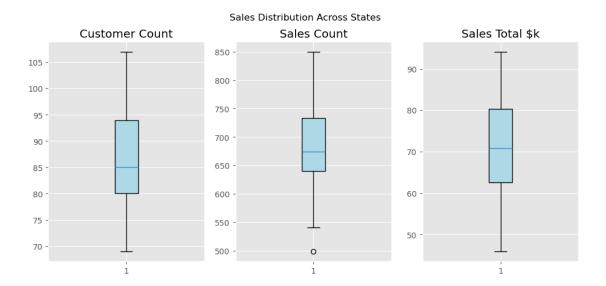
```
import pandas as pd
import matplotlib.pyplot as plt
plt.style.use('ggplot')

# Aggregate customer sales counts and total value
cust_sales = pd.merge(customer_reviews_df, customers_df, on='customer_id')
cust_sales = pd.merge(cust_sales, sales_df, on='uniq_id')

cust_summary = cust_sales[['customer_id', 'state_ISO']].drop_duplicates()
groups = cust_sales.groupby('customer_id')
cust_summary['sales_count'] = groups['uniq_id'].transform('count')
cust_summary['sales_total$k'] = groups['sale_price'].transform('sum')
cust_summary['sales_total$k'] = (cust_summary['sales_total$k'] / 1000).round(2)

# Aggregate the customer counts and sales totals by state
states_working = states_df.copy()
```

```
states_working['customer_count'] = customers_df.
 ⇒groupby('state_ISO')['state_ISO'].transform('count')
states_working.reset_index()
states working['sales count'] = cust summary.
 agroupby('state_ISO')['sales_count'].sum().reset_index()['sales_count']
states_working['sales_total$k'] = cust_summary.
 Groupby('state_ISO')['sales_total$k'].sum().reset_index()['sales_total$k']
# Some analysis of these figures geographical
# Too many states for a meaningful bar plot for each states ...
# Distribution of sales across states, using box plots
fig, (ax0, ax1, ax2) = plt.subplots(1, 3, figsize = (12, 5))
fig.suptitle('Sales Distribution Across States')
ax0.set title('Customer Count')
ax1.set_title('Sales Count')
ax2.set_title('Sales Total $k')
box0 = ax0.boxplot(states_working['customer_count'], patch_artist=True)
for patch in box0['boxes']:
   patch.set(facecolor='lightblue')
box1 = ax1.boxplot(states working['sales count'], patch artist=True)
for patch in box1['boxes']:
   patch.set(facecolor='lightblue')
box2 = ax2.boxplot(states_working['sales_total$k'], patch_artist=True)
for patch in box2['boxes']:
   patch.set(facecolor='lightblue')
plt.show()
# Examine top and bottom 5 states by sales value
top5 = states_working.nlargest(5, 'sales_total$k')
bottom5 = states_working.nsmallest(5, 'sales_total$k')
print('Top Five States by Sales Value')
display(top5[['state_name', 'population', 'customer_count', 'sales_count', ")
print('Bottom Five States by Sales Value')
display(bottom5[['state_name', 'population', 'customer_count', 'sales_count',
# some correlations with size, populations ...
```



Top Five States by Sales Value

	state_name	population	customer_count	sales_count	\
9	Florida	22610726	90	838	
14	Indiana	6862199	83	715	
34	North Dakota	783926	79	802	
53	US Minor Outlying Islands	300	79	716	
0	Alabama	5108468	96	774	

sales_total\$k
9 94.10
14 93.25
34 89.19
53 86.74
0 86.14

Bottom Five States by Sales Value

	state_name	population	customer_count	sales_count	sales_total\$k
4	California	38965193	79	546	45.95
16	Kansas	2940546	79	541	49.78
43	Texas	30503301	76	705	54.34
30	New Jersey	9290841	102	499	55.79
22	Michigan	10037261	69	630	55.90

```
[3]: # Compare the sales value per state against the population
# Scatter plot and regression line
import scipy.stats
x = states_working['population'] / 1000000
```

```
y = states_working['sales_total$k']

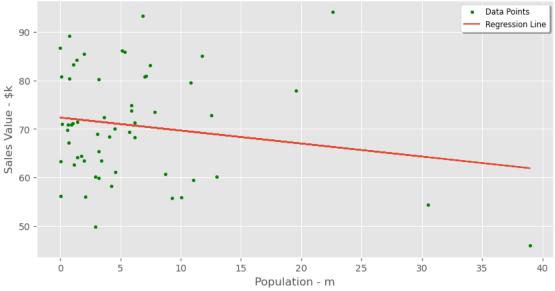
# Calculate the regression line
slope, intercept, r, p, stderr = scipy.stats.linregress(x, y)

# Scatter plot with fir line
fig, ax = plt.subplots(figsize = (10, 5))

ax.set_title('Sales per State vs Population')
ax.set_xlabel('Population - m')
ax.set_ylabel('Sales Value - $k')

ax.scatter(x, y, label= 'Data Points', color = 'g', s = 10)
ax.plot(x, intercept + slope * x, label='Regression Line')
ax.legend(fontsize='small', loc='upper right', shadow=True, facecolor='white')
plt.show()
```

Sales per State vs Population



```
[4]: # Sanity check of physical stores per state against the population
# Scatter plot and regression line

import scipy.stats

x = states_working['population'] / 1000000
y = states_working['stores_total']
```

```
# Calculate the regression line
slope, intercept, r, p, stderr = scipy.stats.linregress(x, y)

# Scatter plot with fir line
fig, ax = plt.subplots(figsize = (10, 5))

ax.set_title('Stores per State vs Population')
ax.set_xlabel('Population - m')
ax.set_ylabel('Number of Stores')

ax.scatter(x, y, label= 'Data Points', color = 'g', s = 10)
ax.plot(x, intercept + slope * x, label='Regression Line')
ax.legend(fontsize='small', loc='upper right', shadow=True, facecolor='white')
plt.show()
```

