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Introduction & Problem Statement





- 1. How many of you have used vending machines this year?
 - a. Did you ever wonder how companies know when to restock a vending machine?
- 2. ABC Vending is a small vending machine company with 5 vending machines across Mississauga
- 3. They Sell Four Categories of Products
 - a. Carbonated Beverages
 - b. Food
 - c. Non Carbonated Beverages
 - d. Water
- 4. In order to help ABC Vending with their business
 - a. Executive Dashboard
 - b. Logistic Mapping Supply Chain
 - c. Predictive Modelling Through Python



Excel Data Analysis

 The team used excel to slice and dice different data analytics techniques to achieve both exploratory data analysis and predictive modelling



ABC Analysis

• For predictive modelling we utilized the ABC Analysis to use historical data to predict what SKUs can be categorized based on the % of Cumulative Inventory:

| Product | Sum of MQty | % of Total Inventory 💌 | % of Cum. Inventory 🔻 ABC 🔻 |
|---------------------------|-------------|------------------------|-----------------------------|
| Coca Cola - Zero Sugar | 524 | 8% | 8% A |
| Monster Energy Origina | 385 | 6% | 14% A |
| Poland Springs Water | 311 | 5% | 19% A |
| KitKat - Crisp Wafers | 267 | 4% | 23% A |
| Sunkist Soda - Orange | 217 | 3% | 26% A |
| Red Bull - Original | 158 | 2% | 29% A |
| Coca Cola - Regular | 156 | 2% | 31% A |
| Wonderful Pistachios - | 147 | 2% | 33% A |
| CheezIt - Original | 138 | 2% | 35% A |
| SunChips Multigrain - F | 131 | 2% | 37% A |
| Robert Irvine's - Fit Cru | 121 | 2% | 39% A |
| Oreo Mini | 111 | 2% | 41% A |
| Snapple Diet Tea - Peac | 109 | 2% | 43% A |
| SunChips Multigrain - S | 106 | 2% | 44% A |
| Takis - Hot Chilli Peppe | 97 | 1% | 46% A |
| Goldfish Baked - Chedo | 92 | 1% | 47% A |
| Snapple Diet Tea - Lem | 88 | 1% | 48% A |
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Excel to PowerBI Transition

• We used both the exploratory data and predictive modelling done in Excel to create various KPIs that would help a Executive staff member make decision regarding the business in a strategic and astute manner.

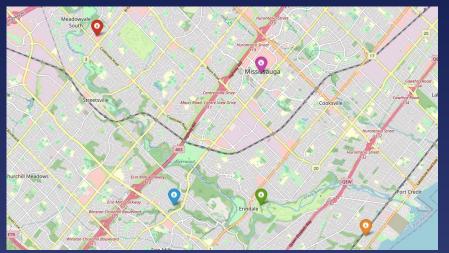
| KPI | What Impact does it Have? | |
|---------------------------------|---|--|
| Count ABC SKU's | It will inform the team the quantity of SKUs in each bucket - once you click on a certain Class - it will give you the list of SKUs along with the quantity sold. | |
| Sum of dollars sold by Category | This will indicate which category is performing the best | |
| Sum of Price by Vendor | This will ensure all executive are aware of the top performing vendors, this will help in making decision in the future regarding vendor partnerships | |
| Sum of Transaction By Type | This will indicated what method of transactions we need to focus on | |
| ArcGIS Map | This will indicate where the machines are, and gives the users the ability to slice and dice based on the location | |
| Profitability Analysis | A quick view of different profitability metrics related to the company | |

To Access the Dashboard - Click Here.



Python Programming: Mapping

- Goal is to determine optimal geo-spatial coordinates for vending machines and the optimal positioning for the warehouse facility
- The vending machines were given arbitrary longitudes and latitudes
 - Each location was given a weight to determine the warehouse location
 - Red Marker (Brunswick Sq Mall), Purple Marker (Earle Asphalt), Blue Marker (EB Public Library), Yellow Marker (GuttenPlans), Green Marker (Warehouse)





Python Programming: Mapping pt. 2

- Example presented is the optimal route between the Warehouse facility and Earle Asphalt:
 - O Distance between two locations found to be 4.35 km or 2.70 miles
 - o For drive mode assuming 50 km/h average speed it would take 5.22 minutes

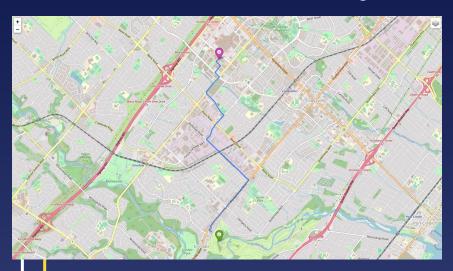
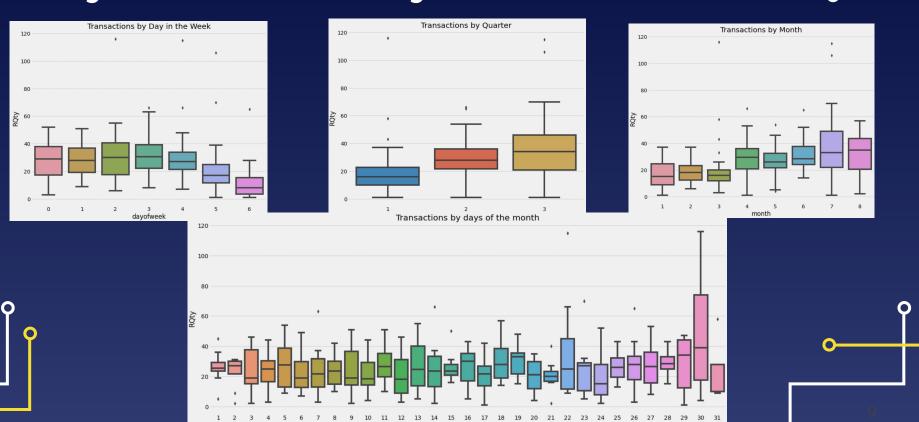






Figure: Optimal Route Using 'bike' mode

Python Initial Analysis



Python Automatic Inventory System

```
#dfTPS=DataFrame
dfabc["TPS"]=TPS
dfabc["ABC%"]=(dfabc['Total Amount']/dfabc['TPS'])*100
number rows abc=len(dfabc)
dfabc.sort values(by=['ABC%'])
import numpy as np
conditions = [
    (dfabc['ABC%'] >= 2),#60%
    (dfabc['ABC%'] >= 1) & (dfabc['ABC%'] < 2),#61-75
    (dfabc['ABC%'] < 1)#rest
values = ['A','B','C']
dfabc['abc']=np.select(conditions, values)
import numpy as np
conditions1 = [
    (dfabc['abc'] == 'A'),
    (dfabc['abc'] == 'B'),
    (dfabc['abc'] == 'C')
values1 = [50, 25, 10]
dfabc['Initial stock']=np.select(conditions1, values1)
dfabc['Stock to BE']=np.select(conditions1, values1)
```

```
        Product
        Total Amount
        TPS
        ABC% abc
        Initial_stock

        0
        Autumns Granola Bar - Cinnamon Almond
        17
        6340
        0.268139
        C
        10

        1
        Bai Antioxidant - Brasilia BB
        4
        6340
        0.063091
        C
        10

        2
        Bai Antioxidant - Kula Watermelon
        5
        6340
        0.078864
        C
        10
```

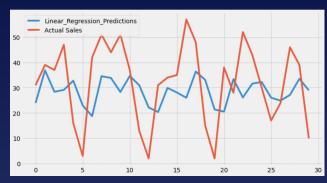
```
# checking if anything is out of stock?
for i in range(162):
   if dfabc['abc'][i]=='A' and dfabc['Stock_to_BE'][i]<20:
      print("order 50 more of.....",dfabc['Product'][i] )

if dfabc['abc'][i]=='B' and dfabc['Stock_to_BE'][i]<15:
   print("order 30 more of.....",dfabc['Product'][i] )

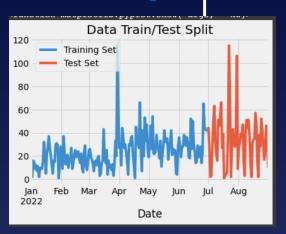
if dfabc['abc'][i]=='C' and dfabc['Stock_to_BE'][i]<5:
   print("order 20 more of......",dfabc['Product'][i] )</pre>
```

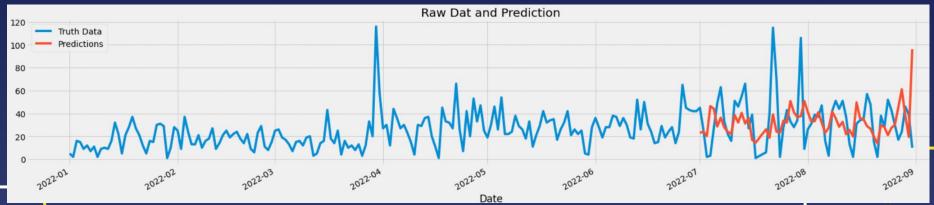
```
order 20 more of...... Cheetos Baked - Flaming Hot order 20 more of..... Chesters Fries Flaming hot order 20 more of..... Good Health Veggie Stix - Sea Salt
```

Python Predictive Modelling









Conclusion

ABC vending machine with 8 months dataset, improvements were achieved through some descriptive and predictive analysis.

Five fronts were maintained including:

- Creating a executive Dashboard.
- Creating Different graphs for monitoring
- Rcoil Analysis (Most sales in that vending machine Row)
- Mapping (Optimal route)
- Machine learning to predict future demand