Public transportation efficiency analysis

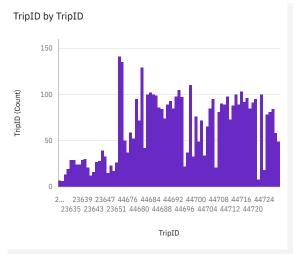
IBM COGNOS VISUALIZATION:

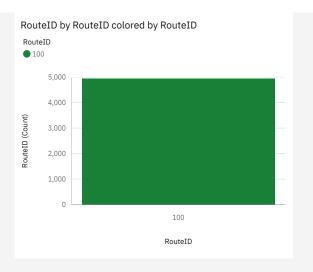
Design dashboards and reports in IBM Cognos to visualize. Continue building the analysis by creating visualizations using IBM Cognos and integrating code for data analysis.

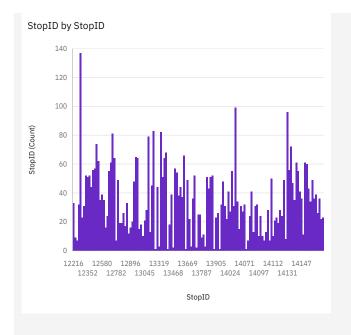
TEAMMATES:

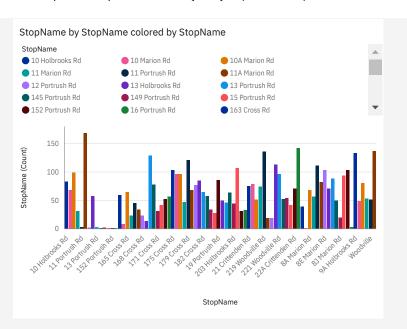
J.ABISHEK PAUL (952421104004) E.S JEBIN (952421104026) C.DEEPAKKUMAR(952421104017) S.K.STANLY (952421104053)

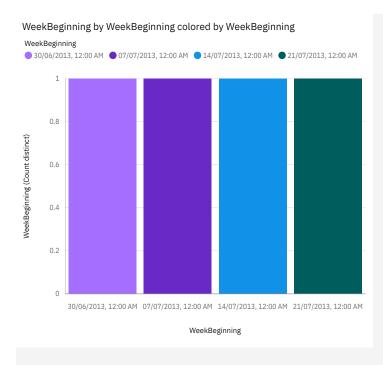
Tab 1

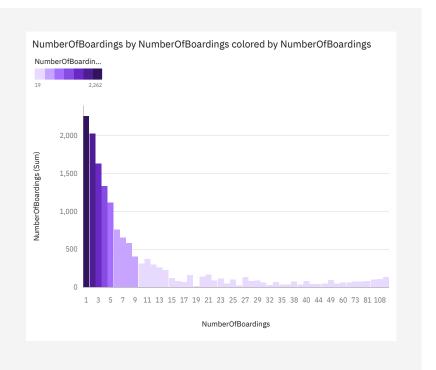






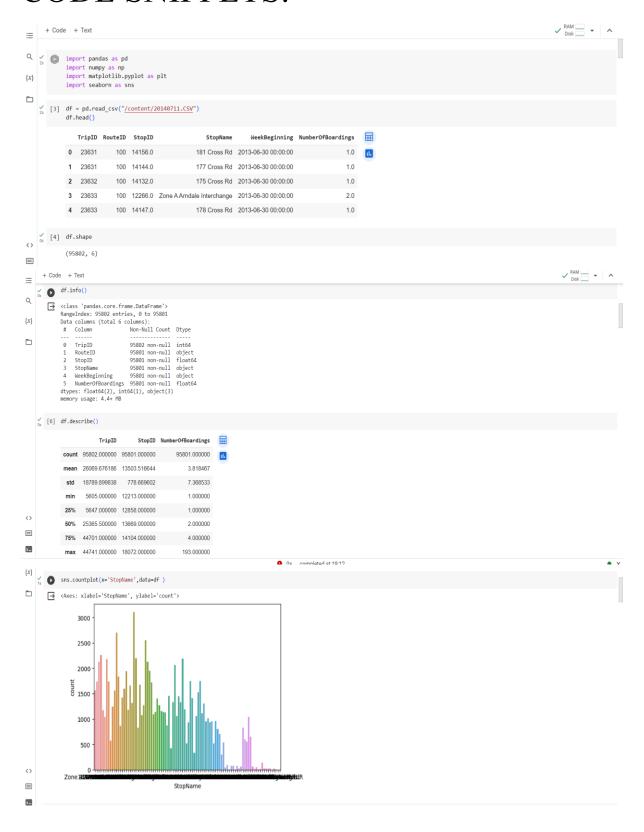






where we will delve deeper into the analysis, modeling, and recommendations to enhance public transportation efficiency. A well-structured dataset can significantly impact the project's overall success.

CODE SNIPPETS:



```
Q
      os df.isnull().sum()
\{x\}
         → TripID
RouteID
 StopID
                StopName
                WeekBeginning
               NumberOfBoardings
               dtype: int64
      \stackrel{\checkmark}{\circ} [12] for feature in df.columns:
                   if df[feature].isnull().sum()>0:
                       print(f"{feature} : {round(df[feature].isnull().mean(),4)*100}%")
               RouteID : 0.0%
StopID : 0.0%
StopName : 0.0%
               WeekBeginning: 0.0%
NumberOfBoardings: 0.0%
 <>
                                                                                                                                                                       ↓ ↑ ⑤ 首 ☆ № 目 :
      _{0\mathrm{S}}^{\checkmark} [14] ## find dublicate rows in dataset
 Q
                duplicate = df[df.duplicated()]
                duplicate
{x}
                   TripID RouteID StopID StopName WeekBeginning NumberOfBoardings
 os for i in df.columns:
                    print(f"\ \{i\}\ :\ \{len(df[i].unique())\}")
         → TripID : 182
                RouteID : 7
StopID : 166
                 StopName : 97
                 WeekBeginning : 55
                 NumberOfBoardings : 145
                                                                                                                                                                                          ✓ RAM 🚃 🕶 🐧
     + Code + Text
≔
      for feature in df.columns:
    if feature == "StopName":
Q
               If returne == Stophane:

pass
else:
bar = sns.histplot(df[feature] , kde_kors = {'bw' : 1} ,)
plt.xlabel(feature)
plt.ylabel("Stoph")
plt.title(feature)
plt.show()
{x}
\supseteq
                                                 TripID
               30000
               20000
               10000
<>
\equiv
                                                        30000 35000 40000 45000
>_
                                                RouteID
80000
               60000
               40000
               20000
                                            100C 100K
RouteID
                                                              100N
                           100
                                   100B
                                                                      100P
```

```
20000
             17500
             15000
             12500
             10000
              7500
              5000
              2500
                 12000
                                         15000
StopID
                                                  16000
                                                         17000
                                                                  18000
                         13000
                                 14000
                                     WeekBeginning
2000
             1750
             1500
             1250
           g 1000
              750
              500
              250
                                      WeekBeginning
                                    NumberOfBoardings
40000
             35000
             30000
             25000
           <u>الْمُ</u> 20000
             15000
             10000
              5000
                                     75 100 125
NumberOfBoardings
                    ò
                         25
                                                        150
                                                              175 200
                                                                                                                                                      ↑ ↓ ⊖ 📮 🕏 🖟 🗎 🗎
Q 0s
       0
            # removing outliers
            Q1 = df.quantile(0.25)
Q3 = df.quantile(0.75)
IQR = Q3 - Q1
\{x\}
            print(IQR)

→ TripID

            StopID
NumberOfBoardings
dtype: float64
                                 1246.0
            \frac{1}{15} [25] df = df[\sim((df < (Q1 - 1.5 * IQR)) |(df > (Q3 + 1.5 * IQR))).any(axis=1)]
            df.shape
            <ipython-input-25-f4e1682787c4>:1: FutureWarning: Automatic reindexing on DataFrame vs Series comparisons is deprecated and will raise ValueError in a future version. Do
df = df[~((df < (Q1 - 1.5 * IQR)) | (df > (Q3 + 1.5 * IQR))).any(axis=1)]
(87201, 6)
<>
           4
```

StopID

```
Q (27] df["StopName"].value_counts()
               11A Marion Rd
23 Findon Rd
219 Woodville Rd
                                                     2527
2319
{x}
                17 Grange Rd
220 Woodville Rd
                                                       2092
               Zone D Arndale Interchange
15 Portrush Rd
                11 East Av
               148 Portrush Rd
151 Portrush Rd
                Name: StopName, Length: 92, dtype: int64
\{x\}
      /
ls [28] ## Correlation
               plt.figure(figsize=(25,25))
ax = sns.heatmap(df.corr(), cmap = "coolwarm", annot=True, linewidth=2)
<ipython-input-28-6eb7a2dfb33e>:3: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future version, it will default to False. Sel^
ax = sns.heatmap(df.corr(), cmap = "coolwarm", annot=True, linewidth=2)
```

CONCLUSION:

we summarize the key findings, insights, and achievements of our entire project. This section serves as a culminating perspective, bringing together the various phases and efforts undertaken to address the challenges and opportunities in public transportation.

Throughout the project, we explored the public bus transport dataset, from its initial design and problem definition to the practical implementation of data preprocessing, analysis, and modeling. We tackled real-world issues related to public transportation efficiency and strove to find actionable solutions.

In this section, we encapsulate the journey by highlighting the project's key outcomes, the lessons learned, and any actionable recommendations derived from our analysis. This conclusion signifies the successful completion of our public transportation efficiency analysis and paves the way for practical applications and informed decisions in the realm of public transportation.