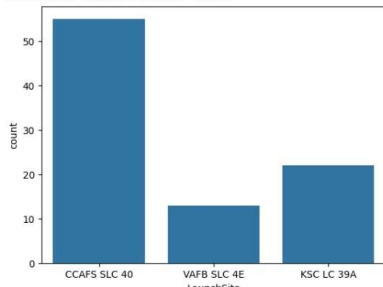


Data Collection and Preprocessing Phase

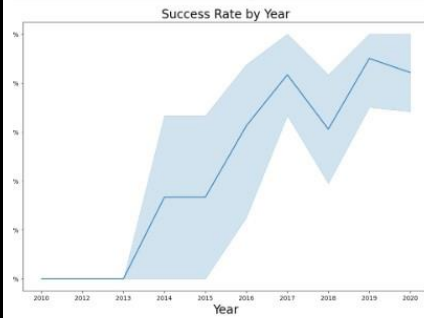
Date	15 July 2024
Team ID	740087
Project Title	SpaceX Falcon 9 First Stage Landing Success Predictor
Maximum Marks	6 Marks

Data Exploration and Preprocessing Template

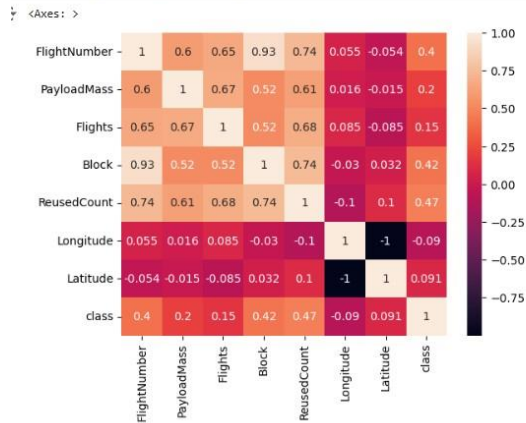
Identifies data sources, assesses quality issues like missing values and duplicates, and implements resolution plans to ensure accurate and reliable analysis.

Section	Description																																																																																
Data Overview	statistics																																																																																
	<div><pre>df.describe()</pre><table><thead><tr><th></th><th>FlightNumber</th><th>PayloadMass</th><th>Flights</th><th>Block</th><th>ReusedCount</th><th>Longitude</th><th>Latitude</th><th>class</th></tr></thead><tbody><tr><td>count</td><td>90.000000</td><td>90.000000</td><td>90.000000</td><td>90.000000</td><td>90.000000</td><td>90.000000</td><td>90.000000</td><td>90.000000</td></tr><tr><td>mean</td><td>45.500000</td><td>6104.959412</td><td>1.788889</td><td>3.500000</td><td>1.655556</td><td>-86.366477</td><td>29.449963</td><td>0.666667</td></tr><tr><td>std</td><td>26.124701</td><td>4694.671720</td><td>1.213172</td><td>1.595288</td><td>1.710254</td><td>14.149518</td><td>2.141306</td><td>0.474045</td></tr><tr><td>min</td><td>1.000000</td><td>350.000000</td><td>1.000000</td><td>1.000000</td><td>0.000000</td><td>-120.610829</td><td>28.561857</td><td>0.000000</td></tr><tr><td>25%</td><td>23.250000</td><td>2510.750000</td><td>1.000000</td><td>2.000000</td><td>0.000000</td><td>-80.603956</td><td>28.561857</td><td>0.000000</td></tr><tr><td>50%</td><td>45.500000</td><td>4701.500000</td><td>1.000000</td><td>4.000000</td><td>1.000000</td><td>-80.577366</td><td>28.561857</td><td>1.000000</td></tr><tr><td>75%</td><td>67.750000</td><td>8912.750000</td><td>2.000000</td><td>5.000000</td><td>3.000000</td><td>-80.577366</td><td>28.608058</td><td>1.000000</td></tr><tr><td>max</td><td>90.000000</td><td>15600.000000</td><td>6.000000</td><td>5.000000</td><td>5.000000</td><td>-80.577366</td><td>34.632093</td><td>1.000000</td></tr></tbody></table></div>		FlightNumber	PayloadMass	Flights	Block	ReusedCount	Longitude	Latitude	class	count	90.000000	90.000000	90.000000	90.000000	90.000000	90.000000	90.000000	90.000000	mean	45.500000	6104.959412	1.788889	3.500000	1.655556	-86.366477	29.449963	0.666667	std	26.124701	4694.671720	1.213172	1.595288	1.710254	14.149518	2.141306	0.474045	min	1.000000	350.000000	1.000000	1.000000	0.000000	-120.610829	28.561857	0.000000	25%	23.250000	2510.750000	1.000000	2.000000	0.000000	-80.603956	28.561857	0.000000	50%	45.500000	4701.500000	1.000000	4.000000	1.000000	-80.577366	28.561857	1.000000	75%	67.750000	8912.750000	2.000000	5.000000	3.000000	-80.577366	28.608058	1.000000	max	90.000000	15600.000000	6.000000	5.000000	5.000000	-80.577366	34.632093
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Bivariate Analysis



Multivariate Analysis



Outliers and Anomalies

Identification and treatment of outliers.

Data Preprocessing Code Screenshots

Loading Data

```
[1] df=pd.read_csv("/content/spacex_launch_data.csv")
```

Code to load the dataset

Handling Missing Data

```
[8] df.isnull().sum()/df.count()*100
```

```
FlightNumber    0.000
Date            0.000
BoosterVersion  0.000
PayloadMass     0.000
Orbit           0.000
LaunchSite      0.000
Outcome         0.000
Flights         0.000
GridFins        0.000
Reused          0.000
Legs            0.000
LandingPad     40.625
Block           0.000
ReusedCount     0.000
Serial          0.000
Longitude       0.000
Latitude        0.000
dtype: float64
```

Removing columns

```
[9] df.drop(["LandingPad"],axis=1,inplace=True)
```

<p>Data Transformation</p>	<pre> orbit_le=LabelEncoder().fit(df['Orbit']) df['Orbit']=orbit_le.transform(df['Orbit']) launchsite_le=LabelEncoder().fit(df['LaunchSite']) df['LaunchSite']=launchsite_le.transform(df['LaunchSite']) longitude_le=LabelEncoder().fit(df['Longitude']) df['Longitude']=longitude_le.transform(df['Longitude']) latitude_le=LabelEncoder().fit(df['Latitude']) df['Latitude']=latitude_le.transform(df['Latitude']) gridfins_le=LabelEncoder().fit(df['GridFins']) df['GridFins']=gridfins_le.transform(df['GridFins']) legs_le=LabelEncoder().fit(df['Legs']) df['Legs']=legs_le.transform(df['Legs']) 44] Scaler=preprocessing.StandardScaler() x_train=Scaler.fit_transform(x_train) x_test=Scaler.transform(x_test) </pre>
<p>Feature Engineering</p>	<pre> landing_outcomes=df["Outcome"].value_counts() landing_outcomes Outcome True ASDS 41 None None 19 True RTLS 14 False ASDS 6 True Ocean 5 False Ocean 2 None ASDS 2 False RTLS 1 Name: count, dtype: int64 19] for i,outcome in enumerate(landing_outcomes.keys()): print(i,outcome) 0 True ASDS 1 None None 2 True RTLS 3 False ASDS 4 True Ocean 5 False Ocean 6 None ASDS 7 False RTLS 20] bad_outcomes=set(landing_outcomes.keys()[[1,3,5,6,7]]) bad_outcomes {'False ASDS', 'False Ocean', 'False RTLS', 'None ASDS', 'None None'} 21] landing_class=[0 if i in set(bad_outcomes) else 1 for i in df["Outcome"]] 22] df["class"]=landing_class df[["class"]].head(8) class </pre>
<p>Save Processed Data</p>	<pre> filename="project.pkl" pickle.dump(lr,open(filename,'wb')) pickle.dump(orbit_le, open('orbit_le.pkl', 'wb')) pickle.dump(launchsite_le, open('launchsite_le.pkl', 'wb')) pickle.dump(Scaler, open('scaler.pkl', 'wb')) </pre>