

Outline

- Executive Summary
- Introduction
- Methodology
- Results
- Conclusion
- Appendix



Executive Summary

- Utilize data science methodologies to define and formulate a real-world business problem
- Utilize your data analysis tools to load a dataset, clean it, and find out interesting insights from it

Introduction

In this capstone, we will predict if the Falcon 9 first stage will land successfully. SpaceX advertises Falcon 9 rocket launches on its website, with a cost of 62 million dollars; other providers cost upward of 165 million dollars each, much of the savings is because SpaceX can reuse the first stage. Therefore if we can determine if the first stage will land, we can determine the cost of a launch. This information can be used if an alternate company wants to bid against SpaceX for a rocket launch.



Methodology

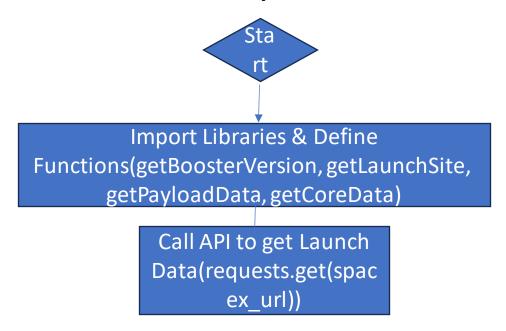
Executive Summary

- Data Collection Methodology
 - Describe how data was collected
- Perform Data Warngling
 - Describe how data was processed
- Perform exploratory data analysis using Visulization and SQL
- Perform interactive visual analytics using Flolium and Plotly Dash
- Perform Predictive Analytics using classification models
 - How to build, tune evaluate classification models

Data Collection

- Describe how Data sets were collected
- Present data collection process use key phrases and flowchart

Data Collection – SpaceX API



https://github.com/susmitapattnaik/CapStone_Project/blob/main/jupyter-labs-spacex-data-collection-api%20(2).ipynb

Data Collection - Scraping

Data Wrangling

- https://github.com/susmitapattnaik/CapStone_Project/blob/main/la bs-jupyter-spacex-data_wrangling_jupyterlite.jupyterlite%20(1).ipynb
- Calculate the number of launches on each site
- Calculate the number and occurrence of each orbit
- Calculate the number and occurrence of mission outcome per orbit type
- Create a landing outcome label from Outcome column

EDA with Data Visualization

EDA with SQL

- %sql select distinct Launch_Site from SPACEXTABLE
- %sql select * from SPACEXTABLE where Launch_Site like 'CCA%' LIMIT 5
- %sql select sum(PAYLOAD_MASS_KG_) as payloadmass from SPACEXTABLE where "Customer" = 'NASA (CRS)'
- %sql select AVG("PAYLOAD_MASS_KG_") AS avgpayload from SPACEXTABLE where "Booster_Version" = 'F9 v1.1'
- %sql select min(Date) from SPACEXTABLE where "Landing_Outcome" = 'Success'
- %sql select distinct "Booster_Version" from SPACEXTABLE where "Landing_Outcome" = 'Success (drone ship)'
- %sql select "Mission Outcome", count(1) from SPACEXTABLE
- %sql select BOOSTER_VERSION as boosterversion from SPACEXTBL where PAYLOAD_MASS__KG_=(select max(PAYLOAD_MASS__KG_) from SPACEXTBL);
- %sql select substr(Date, 4, 2), Landing Outcome, Booster Version, Launch Site from SPACEXTBL where substr(date, 1, 4)='2015'
- %sql SELECT LANDING_OUTCOME FROM SPACEXTBL WHERE DATE BETWEEN '2010-06-04' AND '2017-03-20' ORDER BY DATE DESC;

Github Link:

https://github.com/susmitapattnaik/CapStone_Project/blob/main/jupyter-labs-eda-sql-coursera_sqllite.ipynb

Build and Interactive map with Folium

- Github Link <u>https://github.com/susmitapattnaik/CapStone_Project/blob/main/la</u>
 <u>b jupyter launch site location.jupyterlite.ipynb</u>
- This Exercise includes using markers, Circles and lines on Folium maps to identify launch sites and study

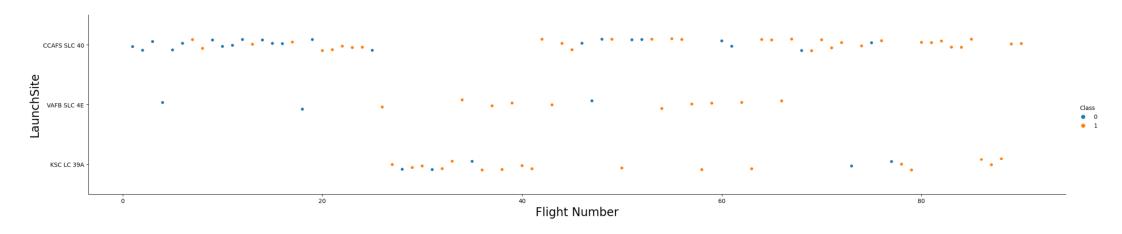
Build a dashboard with Plotly Dash

Predictive Analysis(Classification)

Results

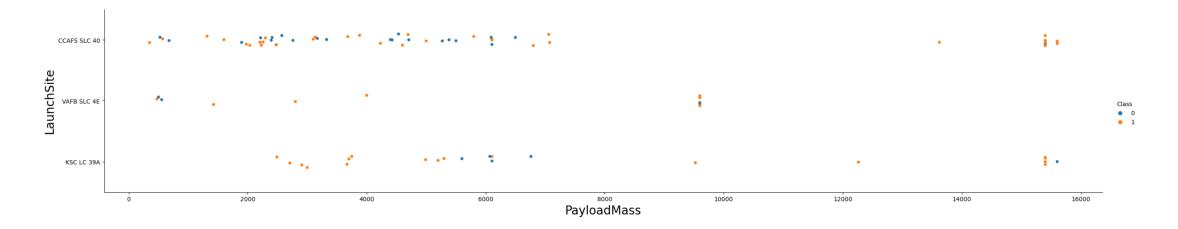


Flight Number Vs. Launch Site



```
sns.catplot(y="LaunchSite", x="FlightNumber", hue="Class", data=df, aspect = 5)
plt.xlabel("Flight Number", fontsize=20)
plt.ylabel("LaunchSite", fontsize=20)
plt.show()
```

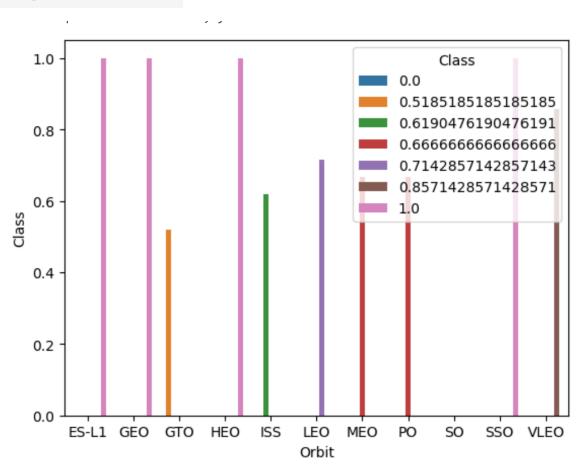
PayLoad Vs. Launch Site



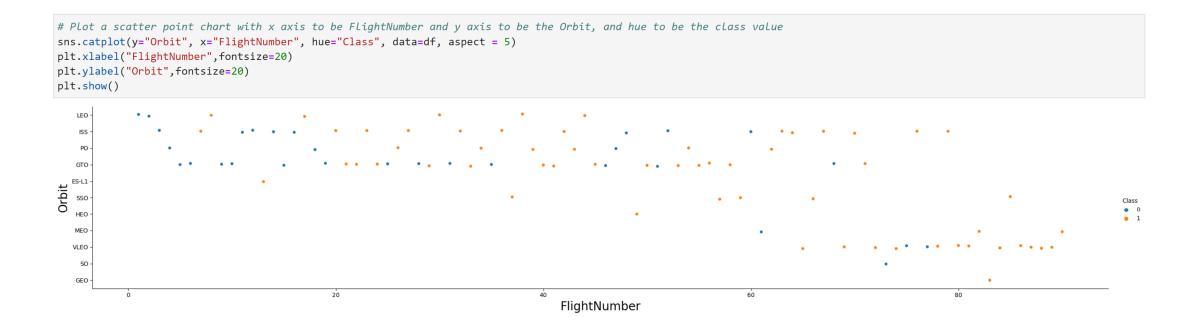
```
### TASK 2: Visualize the relationship between Payload and Launch Site
sns.catplot(y="LaunchSite", x="PayloadMass", hue="Class", data=df, aspect = 5)
plt.xlabel("PayloadMass",fontsize=20)
plt.ylabel("LaunchSite",fontsize=20)
plt.show()
```

Success Rate Vs. Orbit Type

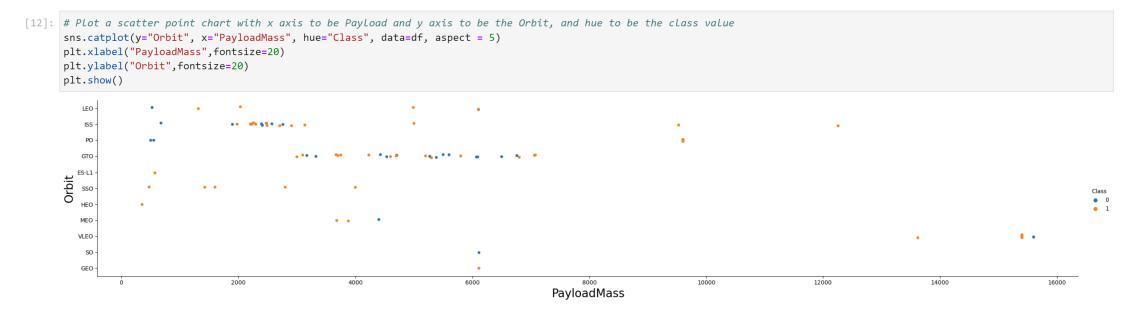
TASK 3: Visualize the relationship between success rate of each orbit type orbit_success = df.groupby('Orbit').mean() orbit_success.reset_index(inplace=True) sns.barplot(x="Orbit",y="Class",data=orbit_success,hue='Class')



Flight Number Vs. Orbit Type



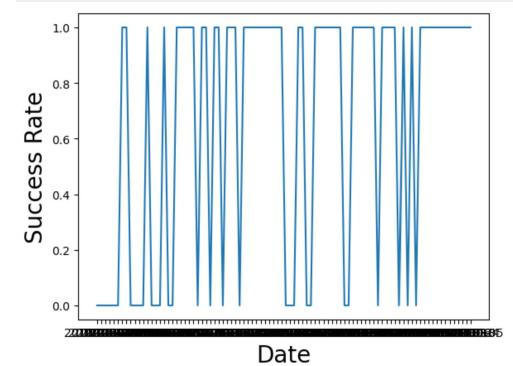
PayLoad Vs. Orbit Type



With heavy payloads the successful landing or positive landing rate are more for Polar, LEO and ISS.

Launch Success Yearly Trend

```
[6]: # Plot a line chart with x axis to be the extracted year and y axis to be the success rate
sns.lineplot(data=df, x="Date", y="Class")
plt.xlabel("Date",fontsize=20)
plt.ylabel("Success Rate",fontsize=20)
plt.show()
```



All Launch Site Names

```
[10]: %sql select distinct Launch_Site from SPACEXTABLE
       * sqlite:///my_data1.db
      Done.
[10]:
       Launch_Site
       CCAFS LC-40
       VAFB SLC-4E
        KSC LC-39A
      CCAFS SLC-40
```

Launch Site names begin with CCA

Task 2
Display 5 records where launch sites begin with the string 'CCA'

F9 v1.0 B0007 CCAFS LC-40

2013-01-03

15:10:00

12]: %sql select * from SPACEXTABLE where Launch_Site like 'CCA%' LIMIT 5 * sqlite:///my_data1.db Done. Date Time (UTC) Booster Version Launch Site Payload PAYLOAD MASS KG Orbit Customer Mission_Outcome Landing_Outcome 2010-04-06 18:45:00 Dragon Spacecraft Qualification Unit F9 v1.0 B0003 CCAFS LC-40 0 **LEO** SpaceX Failure (parachute) 2010-08-12 F9 v1.0 B0004 CCAFS LC-40 Dragon demo flight C1, two CubeSats, barrel of Brouere cheese Success Failure (parachute) 15:43:00 0 LEO (ISS) NASA (COTS) NRO Dragon demo flight C2 2012-05-22 07:44:00 F9 v1.0 B0005 CCAFS LC-40 525 LEO (ISS) NASA (COTS) Success No attempt 2012-08-10 00:35:00 F9 v1.0 B0006 CCAFS LC-40 SpaceX CRS-1 500 LEO (ISS) NASA (CRS) Success No attempt

SpaceX CRS-2

677 LEO (ISS)

NASA (CRS)

Success

No attempt

Total PayLoad Mass

Task 3 Display the total payload mass carried by boosters launched by NASA (CRS)

Average PayLoad Mass by F9 v1.1

lask 4 Display average payload mass carried by booster version F9 v1.1

First Successful Ground Landing Date

List the date when the first succesful landing outcome in ground pad was acheived.

Hint:Use min function

```
[26]: %sql select min(Date) from SPACEXTABLE where "Landing_Outcome" = 'Success'
    * sqlite:///my_data1.db
    Done.
[26]: min(Date)
    2018-03-12
```

Successful Drone Ship Landing with Payload between 4000 and 6000

List the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000

```
[28]: %sql select distinct "Booster Version" from SPACEXTABLE where "Landing Outcome" = 'Success (drone ship)'
        * sqlite:///my_data1.db
       Done.
[28]: Booster_Version
          F9 FT B1021.1
           F9 FT B1022
          F9 FT B1023.1
           F9 FT B1026
          F9 FT B1029.1
          F9 FT B1021.2
          F9 FT B1029.2
          F9 FT B1036.1
          F9 FT B1038.1
          F9 B4 B1041.1
          F9 FT B1031.2
          F9 B4 B1042.1
          F9 B4 B1045.1
          F9 B5 B1046.1
```

Total Number of Successful and Failure Mission Outcomes

Task 7

List the total number of successful and failure mission outcomes

```
[29]: %sql select "Mission_Outcome", count(1) from SPACEXTABLE

* sqlite://my_data1.db
Done.
[29]: Mission_Outcome count(1)

Success 101
```

Boosters Carried Maximum Payload

F9 B5 B1060.3 F9 B5 B1049.7

▼ Task 8 List the names of the booster_versions which have carried the maximum payload mass. Use a subquery [30]: %sql select BOOSTER_VERSION as boosterversion from SPACEXTBL where PAYLOAD_MASS__KG_=(select max(PAYLOAD_MASS__KG_) from SPACEXTBL); * sqlite:///my_data1.db Done. boosterversion F9 B5 B1048.4 F9 B5 B1049.4 F9 B5 B1051.3 F9 B5 B1056.4 F9 B5 B1048.5 F9 B5 B1051.4 F9 B5 B1049.5 F9 B5 B1060.2 F9 B5 B1058.3 F9 B5 B1051.6

2015 Launch Records

List the records which will display the month names, failure landing_outcomes in drone ship ,booster versions, launch_site for the months in year 2015. 1

Note: SQLLite does not support monthnames. So you need to use substr(Date, 4, 2) as month to get the months and substr(Date, 7, 4) = '2015' for year.

]:	substr(Date, 4, 2)	Landing_Outcome	Booster_Version	Launch_Site
	5-	Failure (drone ship)	F9 v1.1 B1012	CCAFS LC-40
	5-	Controlled (ocean)	F9 v1.1 B1013	CCAFS LC-40
	5-	No attempt	F9 v1.1 B1014	CCAFS LC-40
	5-	Failure (drone ship)	F9 v1.1 B1015	CCAFS LC-40
	5-	No attempt	F9 v1.1 B1016	CCAFS LC-40
	5-	Precluded (drone ship)	F9 v1.1 B1018	CCAFS LC-40
	5-	Success (ground pad)	F9 FT B1019	CCAFS LC-40

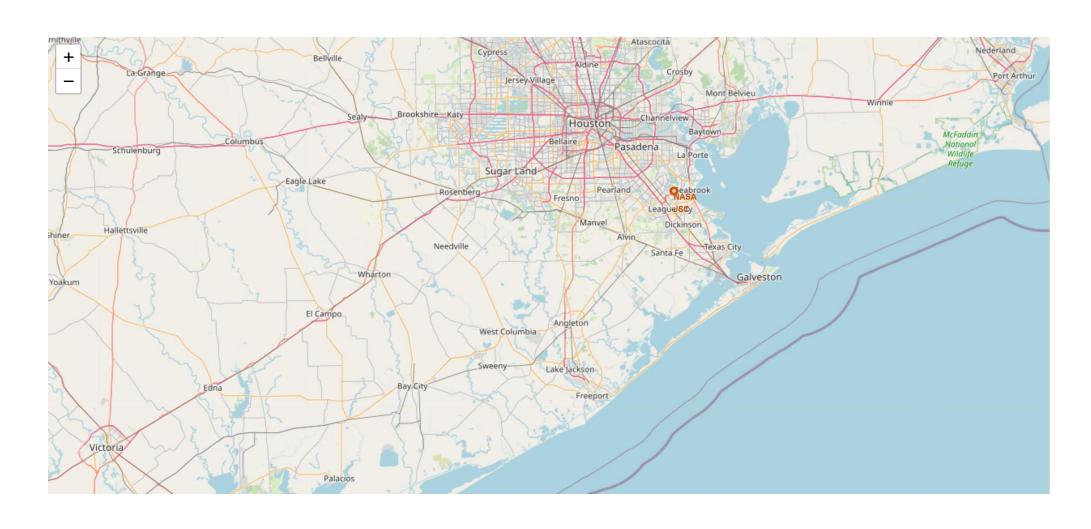
Rank Landing Outcomes Between 2010-06-04 and 2017-03-20

Rank the count of landing outcomes (such as Failure (drone ship) or Success (ground pad)) between the date 2010-06-04 and 2017-03-20, in descending order.

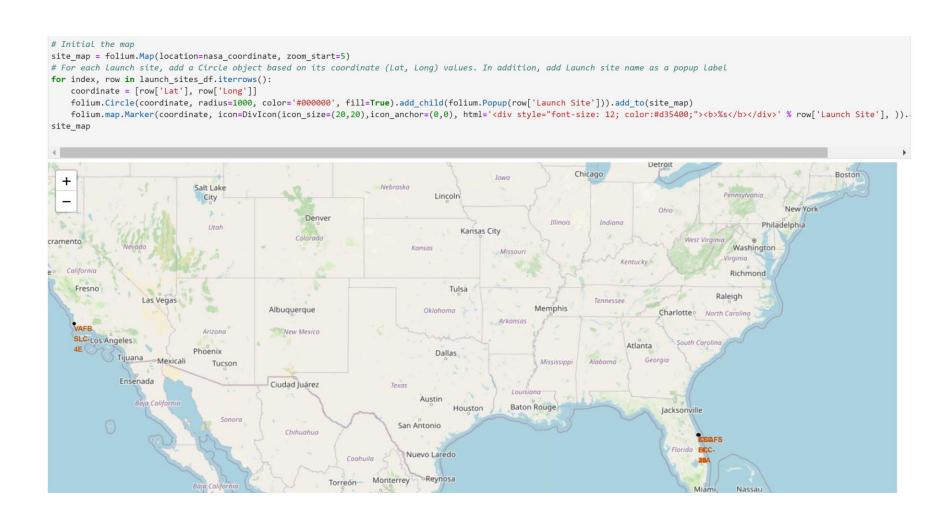


Section 3 Launch Sites **Proximities Analysis**

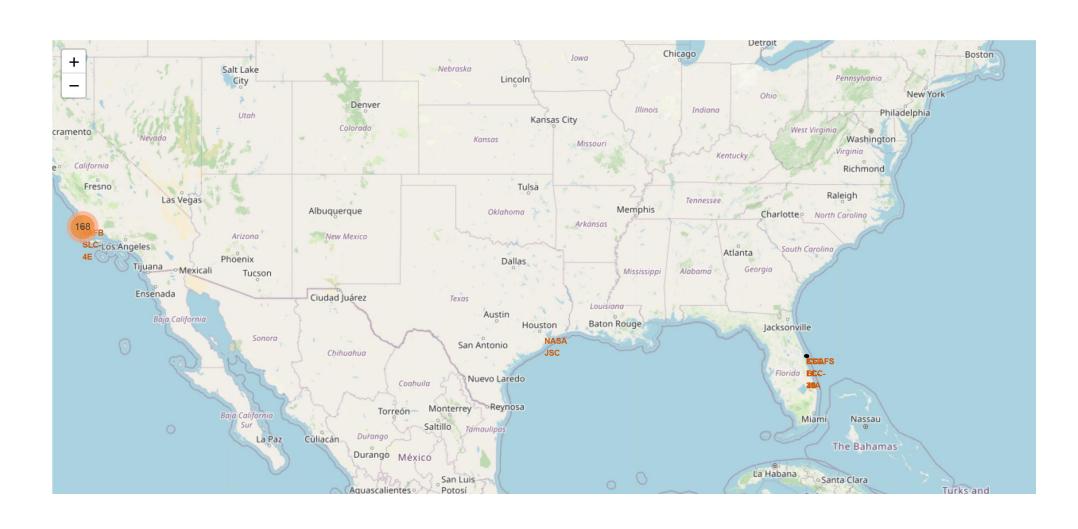
Folium Map Screenshot 1

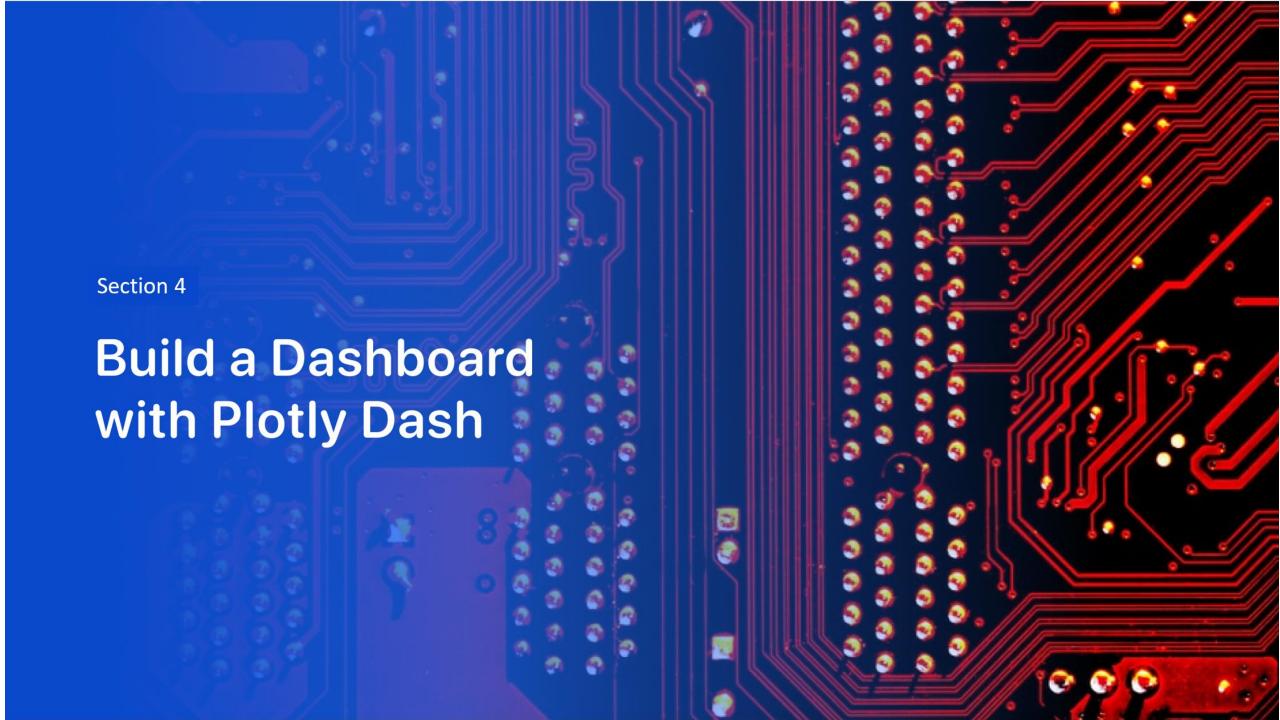


Folium Map Screen Shot 2



Folium Map Screen Shot 3





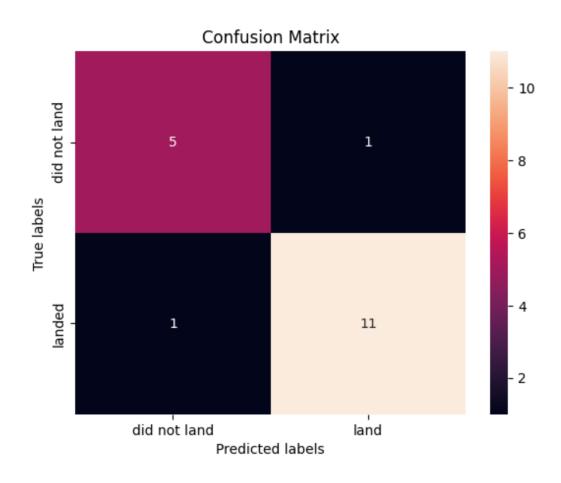
Dashboard Screenshot -1

Dashboard Screenshot -2

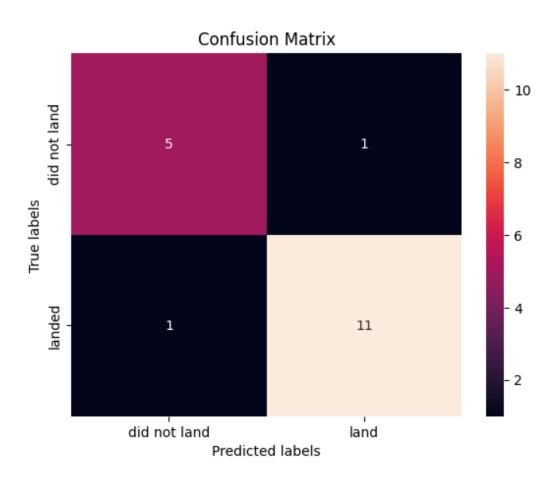
Dashboard Screenshot -3



Classification Accuracy



Confusion Matrix



Conclusions

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TASK 12

Find the method performs best:

```
In [45]:
    predictors = [knn_cv, svm_cv, logreg_cv, tree_cv]
    best_predictor = ""
    best_result = 0
    for predictor in predictors:
        predictor.score(X_test, Y_test)
```

Out[45]: 0.83333333333333334

Appendix

