

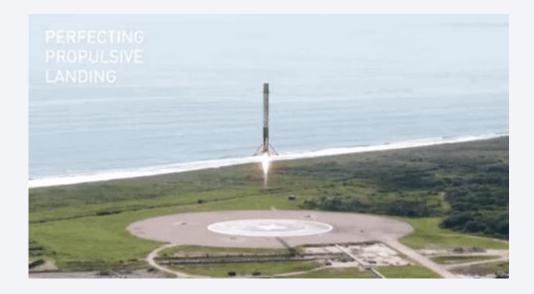
Winning Space Race with Data Science

Lena February 23, 2024



Outline

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



Executive Summary

The project is intended to collect and analyze data from rocket launches by Space X. The aim is to predict the success of a launch in order to calculate the costs of further launches.

Methodologies

- Data Collection from Space X data
- · Data Analysation and Visualization
- Machine Learning

Results

- Predictive Results using machine learning
- Analysis of relationship between different parameters

Introduction

Project background and context

The project is intended to use data from rocket launches by Space X to predict the success of further launches. These information should help the competating company Space Y to gain insights into the relevant parameters for a successful launch and therefore to reduce the costs.

- Problems you want to find answers
 - What influences the success of a rocket launch?
 - Which payload has the highest risk of failure?
 - Which locations are optimal for launch sites?
 - How can the success of rocket launch?



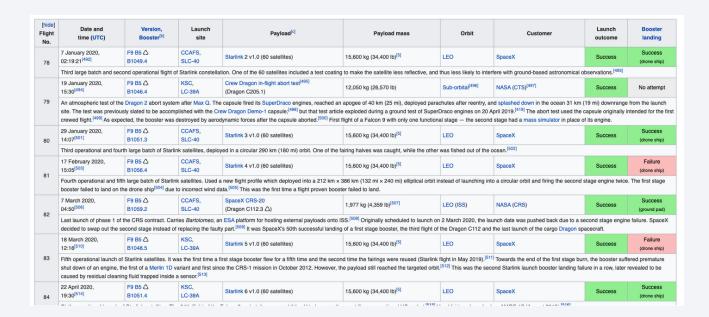
Methodology

Executive Summary

- Data collection methodology:
 - Describe how data was collected
- Perform data wrangling
 - Describe how data was processed
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
 - How to build, tune, evaluate classification models

Data Collection

- Data was collected using two different ways:
 - 1) Data was collected via the Space X API by making a request and collecting the dat
 - 2) Data was collected using wrapscaping



Data Collection – SpaceX API

- Data was collected from the Space X API via an request from a provided URL
- The data has been requested and parsed using json
- GitHub-URL:
 https://github.com/LenaLambers/R
 ocketScience/blob/main/jupyter-labs-spacex-data-collection-api.ipynb

```
Task 1: Request and parse the SpaceX launch data using the GET request

To make the requested JSON results more consistent, we will use the following static response object for this project:

[10]: static_json_url='https://cf-courses-data.s3.us.cloud-object-storage.appdomain.cloud/IBN-DS0321EN-SkillsNetwork/datasets/API_call_spacex_api.json'

We should see that the request was successfull with the 200 status response code

[11]: response.status_code

[11]: 200

Now we decode the response content as a Json using .json() and turn it into a Pandas dataframe using .json_normalize()

[14]: # Use json_normalize meethod to convert the json result into a dataframe respipson = response.json() datae.pd.json_normalize(respipson)

Using the dataframe data print the first 5 rows

[15]: # Get the head of the dataframe data print the first 5 rows
```

Data Collection - Scraping

- Data collection using webscraping from a provided URL
- Usage of beautiful soup

Github URL:
 https://github.com/LenaLam
 bers/RocketScience/blob/mai
 n/jupyter-labs webscraping.ipynb

```
[10]: # use requests.get() method with the provided static_url
    # assign the response to a object

response = requests.get(static_url).text

Create a BeautifulSoup object from the HTML response

[11]: # Use BeautifulSoup() to create a BeautifulSoup object from a response text content
    soup = BeautifulSoup(response, 'html.parser')

Print the page title to verify if the BeautifulSoup object was created properly

[12]: # Use soup.title attribute
    print(soup.title)

<title>List of Falcon 9 and Falcon Heavy launches - Wikipedia</title>
```

Data Wrangling

- Data were processed using Exploratory Data Analysis (EDA)
- Analysis using Data Frames
- Github: https://github.com/LenaLambers/RocketScience/blob/main/labs-jupyter-spacex-Data%20wrangling.ipynb

```
TASK 3: Calculate the number and occurrence of mission outcome of the orbits
   Use the method .value counts() on the column Outcome to determine the number of landing outcomes .Then as
1]: # landing outcomes = values on Outcome column
    landing outcomes = df["Outcome"].value counts()
    landing outcomes
1]: True ASDS
                 41
                 19
    True RTLS
                 14
    False ASDS
                  6
    True Ocean
                  5
    False Ocean
                  2
    None ASDS
                  2
    False RTLS
    Name: Outcome, dtype: int64
```

EDA with Data Visualization

- The following evaluations were done:
 - FlightNumber vs. Payload Mass
 - FlightNumber vs. LaunchSite
 - Launchsite vs. Payload
 - Orbit vs. Success Rate
 - Orbit vs. Flight Number
 - Orbit vs. Payload
 - Date vs. Success Rate
- Github: https://github.com/LenaLambers/RocketScience/blob/main/jupyter-labs-eda-dataviz.ipynb.jupyterlite.ipynb

EDA with SQL

- Exemplary SQL Queries:
 - Display launch sites
 - Average mass load for a specific booster version
 - date when the first successful landing outcome in ground pad was achieved
 - names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000
 - total number of successful and failure mission outcomes
- Github: https://github.com/LenaLambers/RocketScience/blob/main/jupyter-labs-eda-sql-coursera_sqllite.ipynb

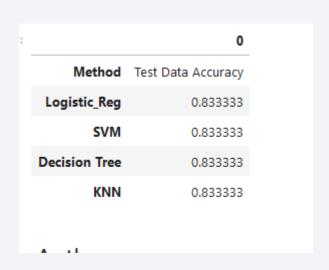
Build an Interactive Map with Folium

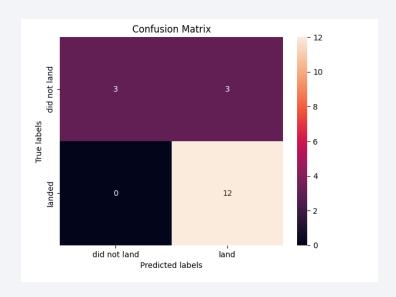
- Objects added in Folium:
 - Location of NASA Johnson Space Center
 - Location of the four launch sites
 - Marker Cluster with red and green markers for non-successful and successful launches
 - Mouse position
 - · Distance line between launch site and coast, railway and city
- Objects were added to analysis, which parameters stand for an optimal launch site
- Github: https://github.com/LenaLambers/RocketScience/blob/main/lab_jupyter_launch_site_l ocation.jupyterlite.ipynb

Predictive Analysis (Classification)

- Model building process:
 - Splitting the data into test and training
 - Fitting the model using different methods
 - Evaluate accuracy
 - Comparison of hyperparameters
 - Plotting of confusion matrix
- Comparison of Logistic Regression, SVM, Decision Tree and KNN
- Github: https://github.com/LenaLambers/RocketScience/blob/main/SpaceX_Machine_ Learning_Prediction_Part_5.jupyterlite.ipynb

Results

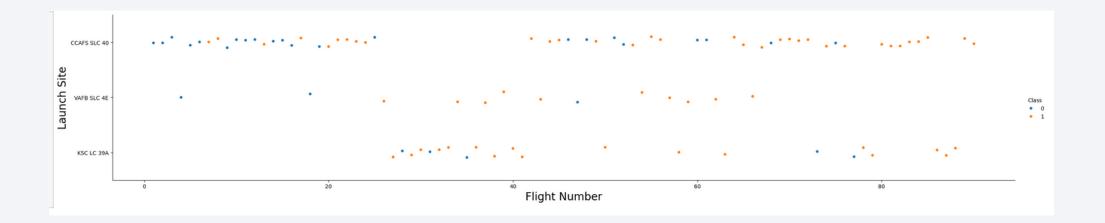




All methods have the same accuracy for predicting the data for a rocket launch.

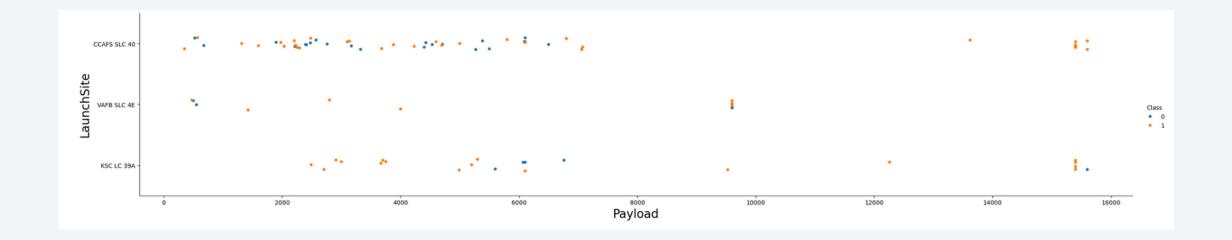


Flight Number vs. Launch Site



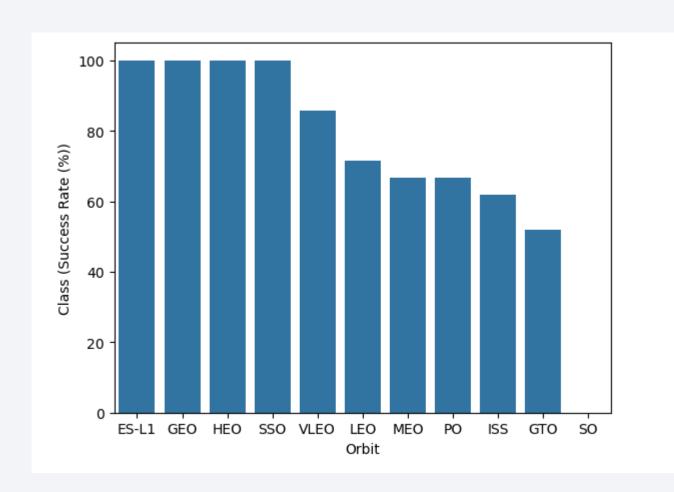
It becomes obvious, that the success rate increases with a higher number of flights. This can be explained, by the better training and planning due to longer experience of the team. There is no significant difference between the success at different launch sites

Payload vs. Launch Site



No Rockets with payload more than 10000 were launched at launch site VAFB-SLC The success rate is higher for launches with higher payload

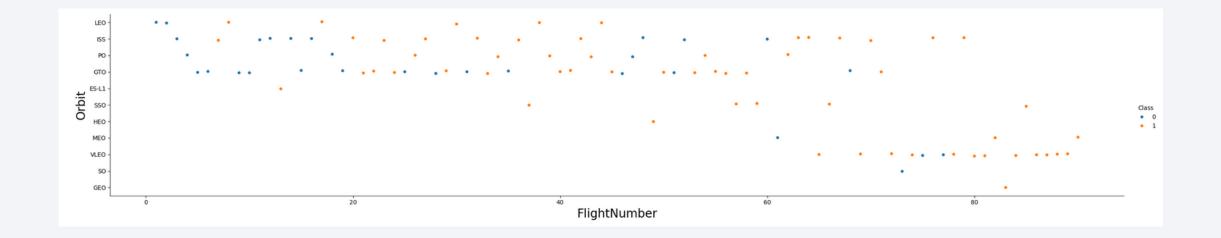
Success Rate vs. Orbit Type



The success rate was higher at orbits near the earth. There was no successfull launch for the orbit SO.

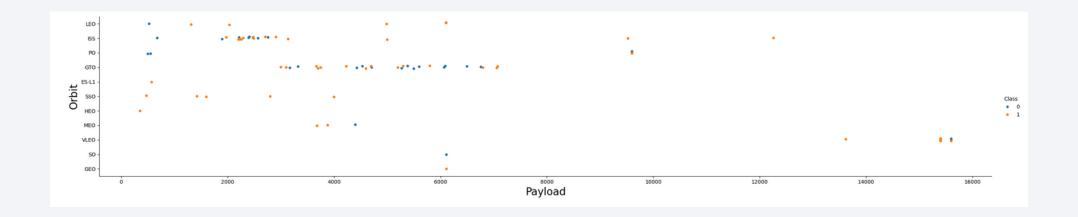
Except for this orbit, the success rate is over 50% for all orbits.

Flight Number vs. Orbit Type



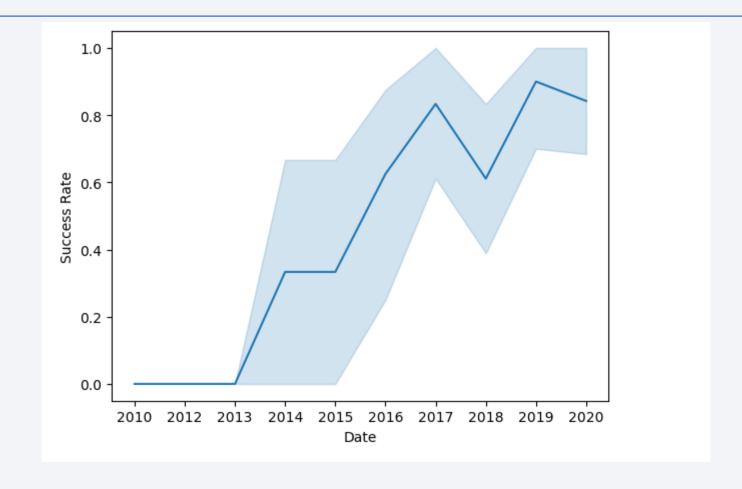
In the LEO orbit the Success appears related to the number of flights; on the other hand, there seems to be no relationship between flight number when in GTO orbit.

Payload vs. Orbit Type



With heavy payloads the successful landing or positive landing rate are more for Polar, LEO and ISS. However for GTO we cannot distinguish this well as both positive landing rate and negative landing (unsuccessful mission) are both there here.

Launch Success Yearly Trend



The success rtae significantly increases over the years, with only one exception in 2018.

All Launch Site Names

Space X uses different launch sites to launch their rockets.

The different launches site can be determined using a SQL query:

Launch_Sites

CCAFS LC-40

VAFB SLC-4E

KSC LC-39A

CCAFS SLC-40

Launch Site Names Begin with 'CCA'

| 0]: | %sql SELECT | * FROM 'SP | ACEXTBL' WHERE I | _aunch_Site LIKE | 'CCA%' LIMIT 5 | | | | [| ↑ ↓ 古 🗜 |
|-----|-------------------|---------------|------------------|------------------|--|------------------|--------------|--------------------|-----------------|---------------------|
| | * sqlite:// Done. | /my_data1. | db | | | | | | | |
| 0]: | Date | Time (UTC) | Booster_Version | Launch_Site | Payload | PAYLOAD_MASS_KG_ | Orbit | Customer | Mission_Outcome | Landing_Outcome |
| | 2010-06-04 | 18:45:00 | F9 v1.0 B0003 | CCAFS LC-40 | Dragon Spacecraft Qualification Unit | 0 | LEO | SpaceX | Success | Failure (parachute) |
| | 2010-12-08 | 15:43:00 | F9 v1.0 B0004 | CCAFS LC-40 | Dragon demo flight C1, two CubeSats, barrel of Brouere cheese | 0 | LEO (ISS) | NASA (COTS) NRO | Success | Failure (parachute) |
| | 2012-05-22 | 7:44:00 | F9 v1.0 B0005 | CCAFS LC-40 | Dragon demo flight C2 | 525 | LEO (ISS) | NASA (COTS) | Success | No attempt |
| | 2012-10-08 | 0:35:00 | F9 v1.0 B0006 | CCAFS LC-40 | SpaceX CRS-1 | 500 | LEO (ISS) | NASA (CRS) | Success | No attempt |
| | 2013-03-01 | 15:10:00 | F9 v1.0 B0007 | CCAFS LC-40 | SpaceX CRS-2 | 677 | LEO (ISS) | NASA (CRS) | Success | No attempt |
| | Task 3 | | | | | | | | | |

The results show a list of launch site, that start with the letters CCA.

Total Payload Mass

In total, the customer NASA carried a payload of 45596 kg into space using Space X rockets.

Average Payload Mass by F9 v1.1

```
Task 4

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

[20]: %sql SELECT AVG(PAYLOAD_MASS_KG_) as "AVERAGE Payload Mass (kg)", Booster_Version FROM 'SPACEXTBL' WHERE Booster_Version LIKE 'F9 v1.1%'

* sqlite:///my_datal.db
Done.

[20]: AVERAGE Payload Mass (kg) Booster_Version

2534.666666666665 F9 v1.1 B1003
```

The booster version F9 v1.1 carried an average payload of 2534.67 kg per launch.

First Successful Ground Landing Date

The first sucessful landing outcome in ground pad was performed on December 22, 2015

1 2015-12-22

```
%sql SELECT min(date) from SPACEXTBL where landing_outcome = 'Success (ground pad)'
```

Successful Drone Ship Landing with Payload between 4000 and 6000

 The following results show a list of the names of boosters which have successfully landed on drone ship and had payload mass greater than 4000 but less than 6000

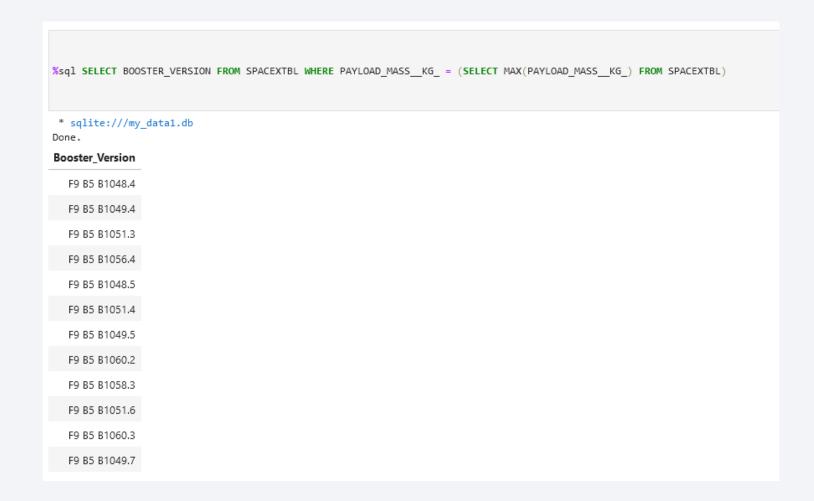


Total Number of Successful and Failure Mission Outcomes



The result list the successful and failed outcomes of all missions.

Boosters Carried Maximum Payload



The listed booster versions have carried the maximum payload mass

2015 Launch Records

| DATE | booster_version | launch_site | landing_outcome |
|------------|-----------------|-------------|----------------------|
| 2015-01-10 | F9 v1.1 B1012 | CCAFS LC-40 | Failure (drone ship) |
| 2015-04-14 | F9 v1.1 B1015 | CCAFS LC-40 | Failure (drone ship) |

There were only two failed landings in 2015, namely on October 01. and on April 14.

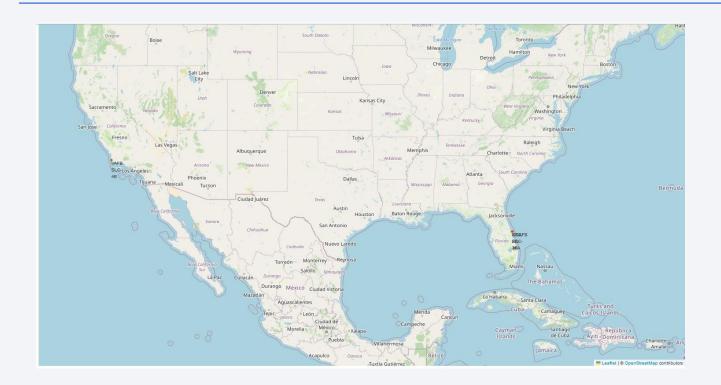
Rank Landing Outcomes Between 2010-06-04 and 2017-03-20

| landing_outcome | count_launches |
|------------------------|----------------|
| No attempt | 10 |
| Failure (drone ship) | 5 |
| Success (drone ship) | 5 |
| Controlled (ocean) | 3 |
| Success (ground pad) | 3 |
| Failure (parachute) | 2 |
| Uncontrolled (ocean) | 2 |
| Precluded (drone ship) | 1 |
| | |

• The results show a ranking of 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

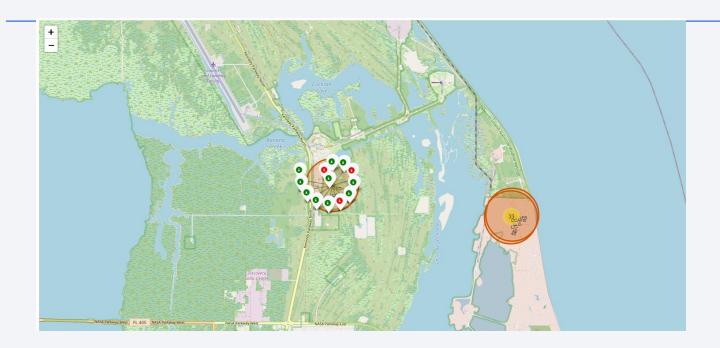


Spatial distribution of launch sites



The map shows the locations of Space X launch sites in California and Florida. We added the location as well as information on the successfusll launches from these locations. Both launch sites are located next to the coast, as well as in close distance to a railroad and are therefore optimal for rocket launches.

Launch outcomes at locations



• The map shows the two launch sites located near the Kennedy Space Center in Florida. Green markers represent successful launches from this launch sites, while red markers stand for failed launches.

Evaluation of optimal location



• To evaluate the optimal parameters for a launch site, the distances to important landmarks like coast lines and railroad were calculated. All launch sites are located at the coast, as shown in the result plot.



Classification Accuracy

| | 0 |
|---------------|--------------------|
| Method | Test Data Accuracy |
| Logistic_Reg | 0.833333 |
| SVM | 0.833333 |
| Decision Tree | 0.833333 |
| KNN | 0.833333 |

The accuracy for all four models is the same.

```
TASK 12

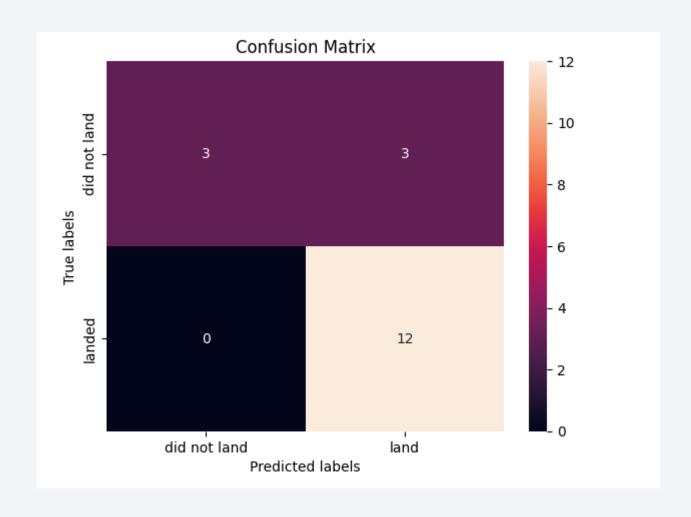
Find the method performs best:

Report = pd.DataFrame({'Method' : ['Test Data Accuracy']})

knn_accuracy=knn_cv.score(X_test, Y_test)
Decision_tree_accuracy=tree_cv.score(X_test, Y_test)
SVM_accuracy=svm_cv.score(X_test, Y_test)
Logistic_Regression=logreg_cv.score(X_test, Y_test)

Report['Logistic_Reg'] = [Logistic_Regression]
Report['SVM'] = [SVM_accuracy]
Report['Decision Tree'] = [Decision_tree_accuracy]
Report['KNN'] = [knn_accuracy]
Report.transpose()
```

Confusion Matrix



The confusion matrix shows a good prediciton for the failed launches, but has slight problems with too much false positive predictions.

Conclusions

- Data Analysis can help to predict rocket launches from Space X
- Data visualization and analysis showed an increasing successrate over time and with increasing flights
- Machine learning approaches show a good accuracy over 80% for the prediction
- Spatial analysis of location sites help to find the optimal location parameters for possible future launch sites.

