

Space X Landing Prediction



IBM Data Science Capstone

Fethi BENMAKHLOUF

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PRESNTATION OUTLINE

- ❑ Executive Summary
- ❑ Introduction
- ❑ Methodology
- ❑ Results
 - Visualization – Charts
 - Dashboard
- ❑ Discussion
 - Findings & Implications
- ❑ Conclusion
- ❑ Appendix

EXECUTIVE SUMMARY



Data collection,



Data Wrangling



Explatory Data Analysis with SQL Pandas and PyPlot



Dashboard with Folium Plotly & Dash



Apply Machine learning algorithms on data cleaned

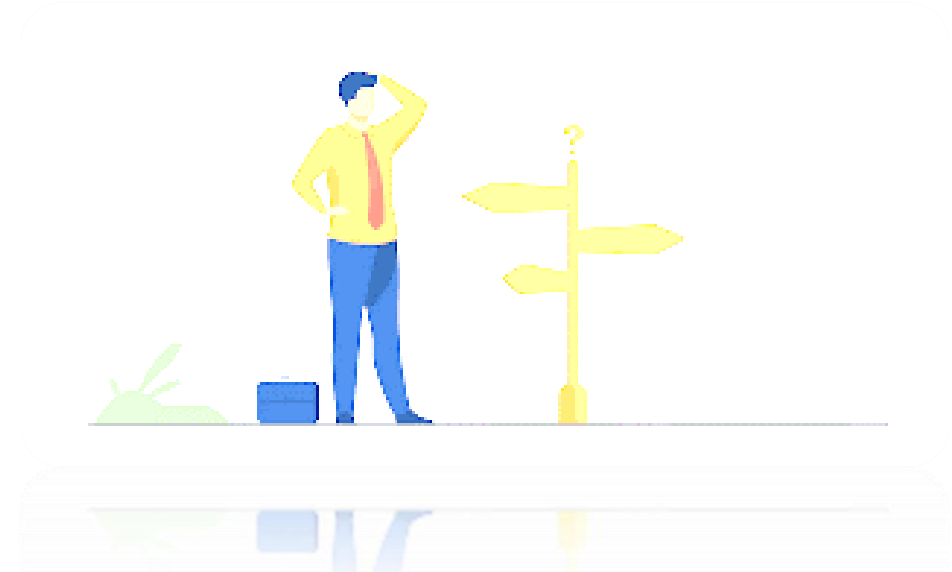
INTRODUCTION

This analysis aims to get help aero spatial researchers in their information mining process to collect informations and factors to have successful rocket landing

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.

What factors influences a successful rocket landing ?

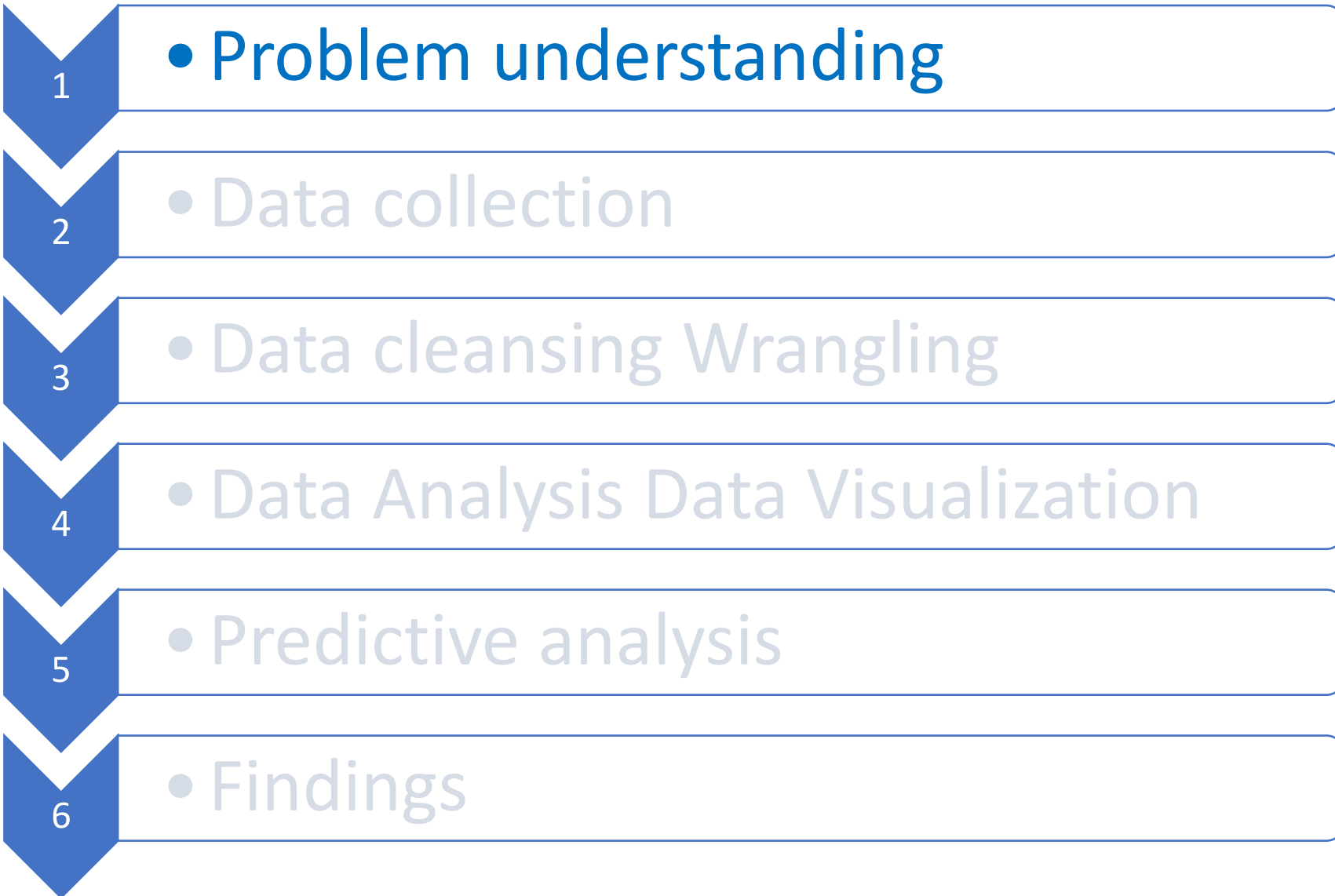
What conditions should be available to achieve to get the best results and ensure the best rocket success landing rate?



METHODOLOGY



METHODOLOGY



Problem



What cause a success?



What cause a failure?



What kind of data we need?



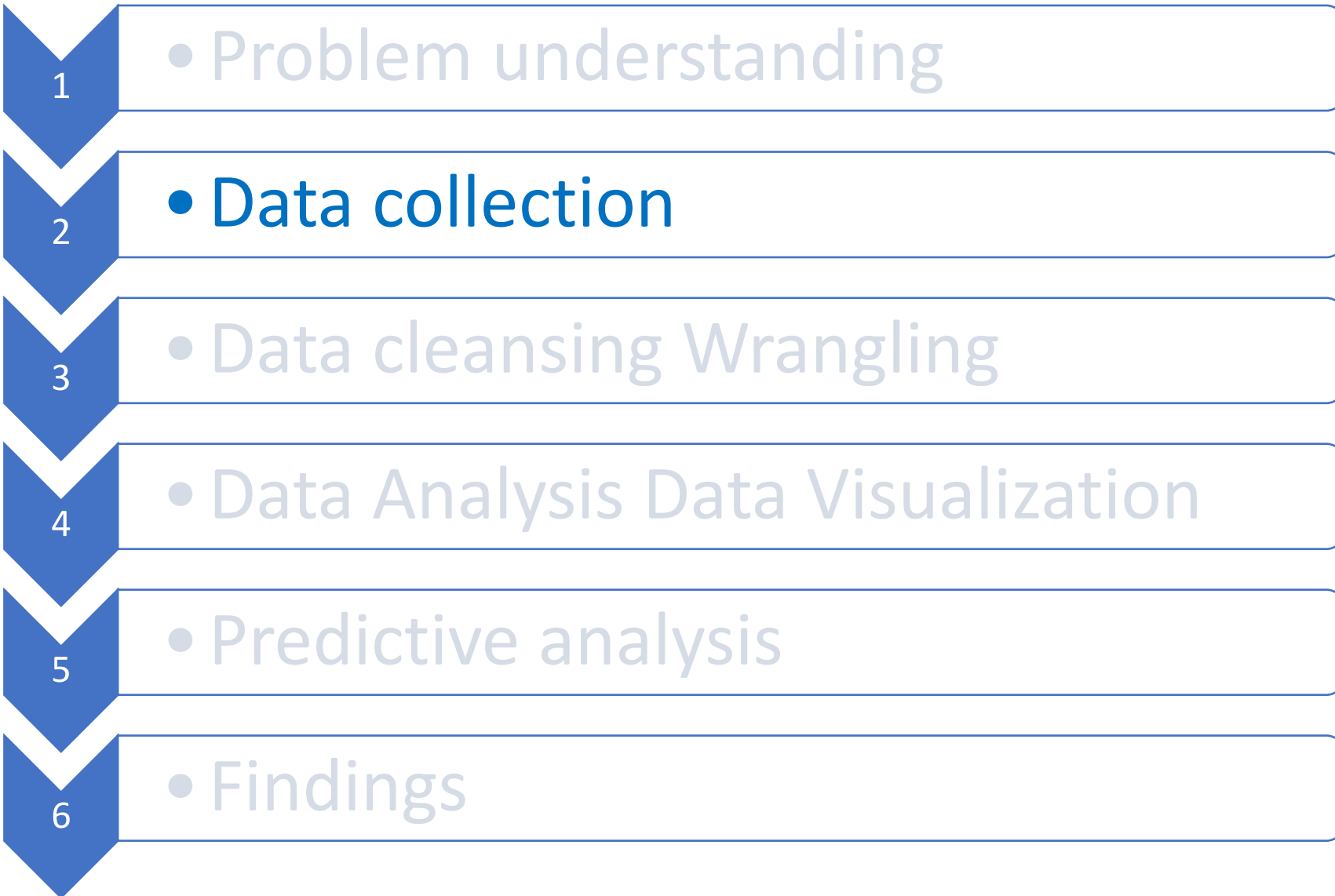
How to obtain it?



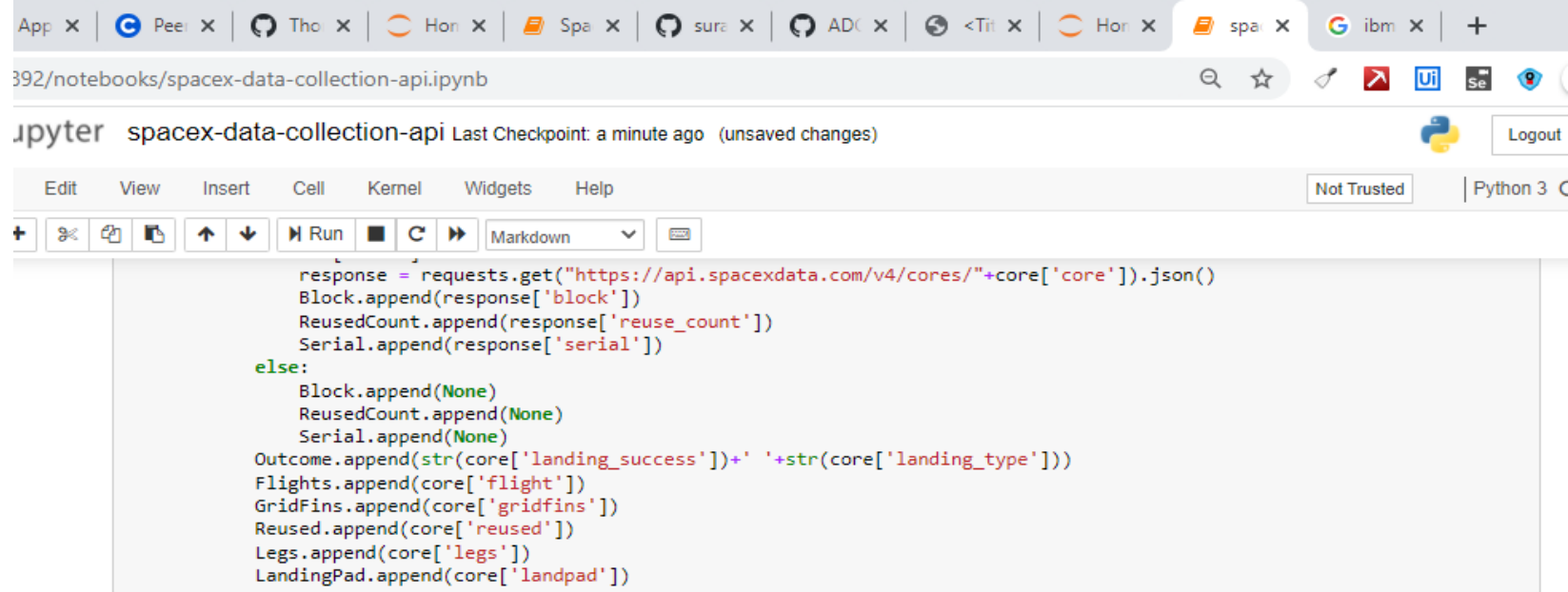
Data scraping it?



METHODOLOGY



Data collection



The screenshot shows a Jupyter Notebook titled 'spacex-data-collection-api'. The code in the cell is as follows:

```
response = requests.get("https://api.spacexdata.com/v4/cores/"+core['core']).json()
Block.append(response['block'])
ReusedCount.append(response['reuse_count'])
Serial.append(response['serial'])
else:
    Block.append(None)
    ReusedCount.append(None)
    Serial.append(None)
Outcome.append(str(core['landing_success'])+' '+str(core['landing_type']))
Flights.append(core['flight'])
GridFins.append(core['gridfins'])
Reused.append(core['reused'])
Legs.append(core['legs'])
LandingPad.append(core['landpad'])
```

Now let's start requesting rocket launch data from SpaceX API with the following URL:

```
In [6]: spacex_url="https://api.spacexdata.com/v4/launches/past"
response = requests.get(spacex_url).json()
```

Check the content of the response

```
In [7]: response = requests.get(spacex_url)
response
```

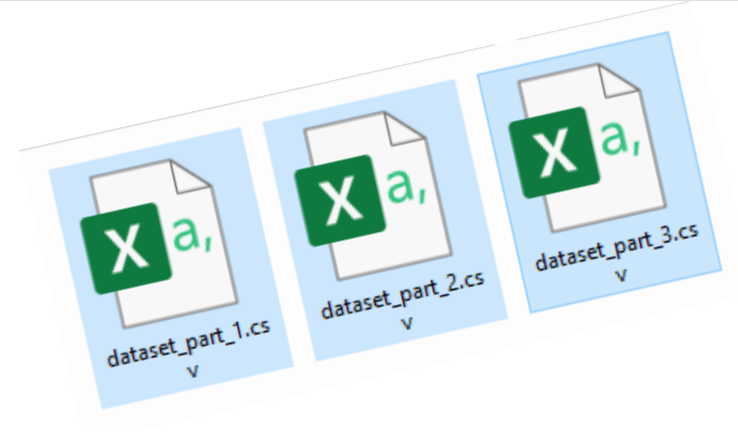
```
Out[7]: <Response [200]>
```



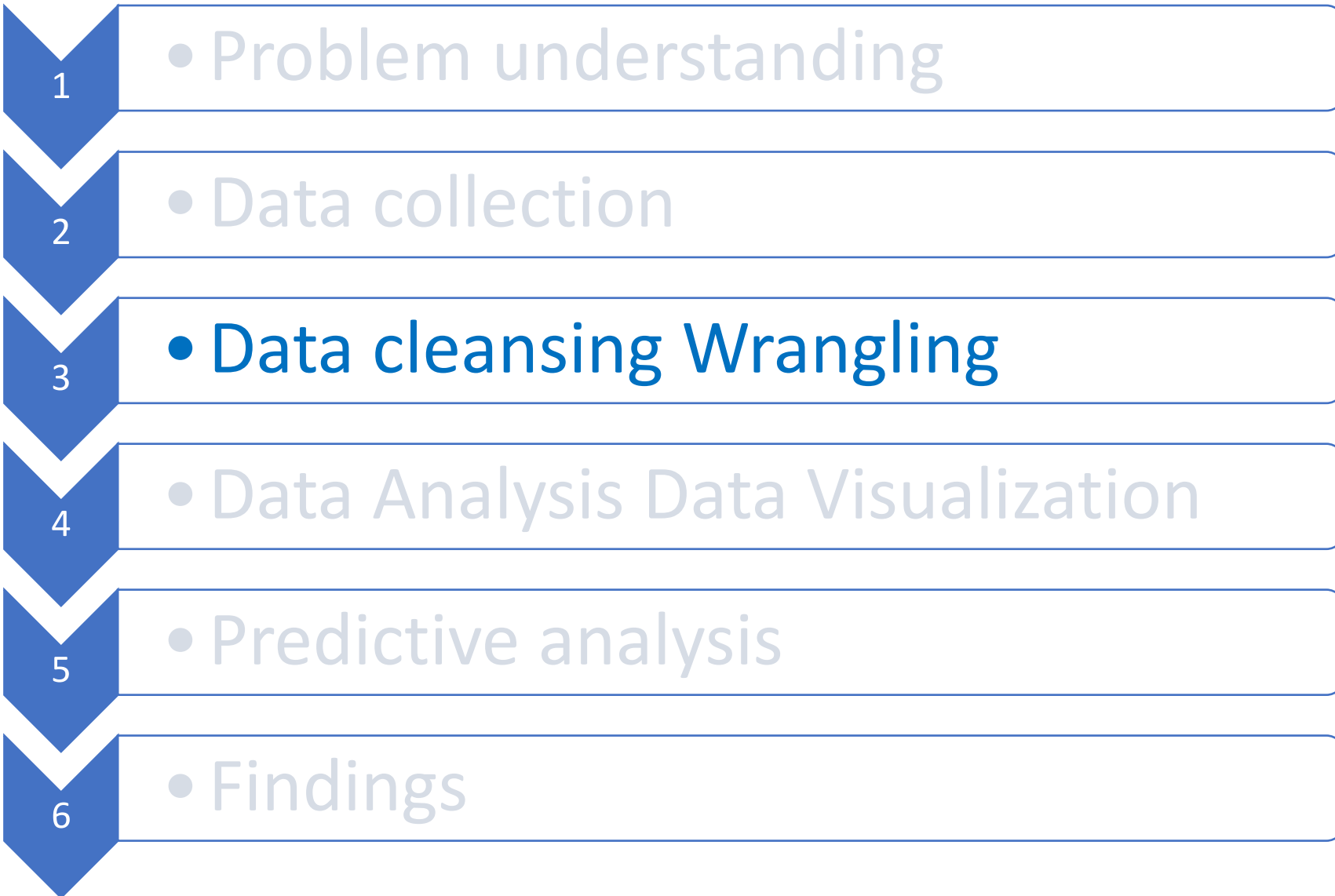
HTML API



Data scraping with BeautifulSoup Library



METHODOLOGY



```
Out[11]:
```

	Class
0	0
1	0
2	0
3	0
4	0
5	0
6	1
7	1

Data Wrangling

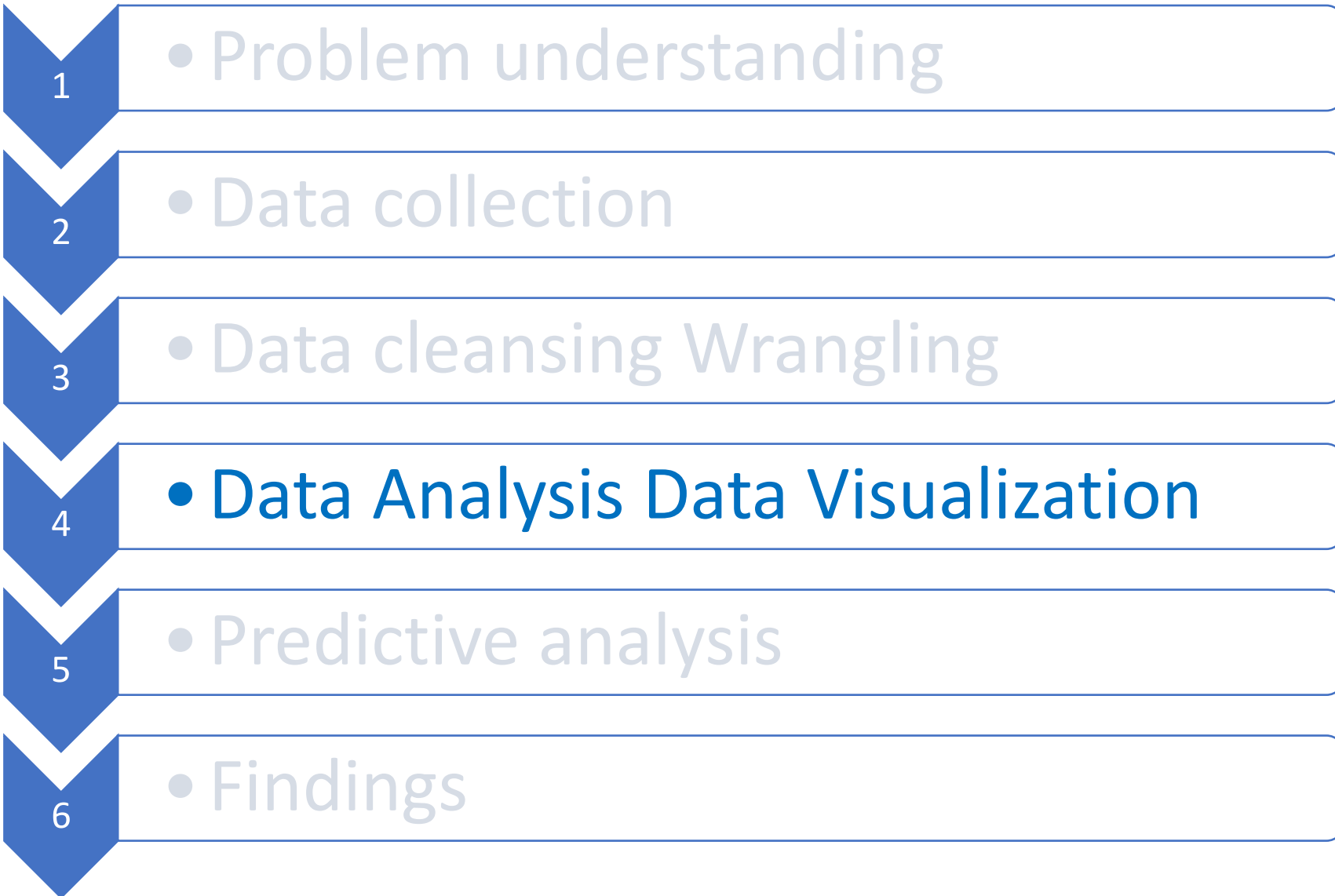
```
In [12]: df.head(5)
```

```
Out[12]:
```

	FlightNumber	Date	BoosterVersion	PayloadMass	Orbit	LaunchSite	Outcome	Flights	GridFins	Reused	Legs	LandingPad	Block	ReusedCount	Serial
0	1	2010-06-04	Falcon 9	6104.959412	LEO	CCAFS SLC 40	None None	1	False	False	False	NaN	1.0	0	B0003
1	2	2012-05-22	Falcon 9	525.000000	LEO	CCAFS SLC 40	None None	1	False	False	False	NaN	1.0	0	B0005
2	3	2013-03-01	Falcon 9	677.000000	ISS	CCAFS SLC 40	None None	1	False	False	False	NaN	1.0	0	B0007
3	4	2013-09-29	Falcon 9	500.000000	PO	VAFB SLC 4E	False Ocean	1	False	False	False	NaN	1.0	0	B1003
4	5	2013-12-03	Falcon 9	3170.000000	GTO	CCAFS SLC 40	None None	1	False	False	False	NaN	1.0	0	B1004

convert those outcomes into Training Labels
with 1 means the booster successfully
landed 0 means it was unsuccessful.

METHODOLOGY



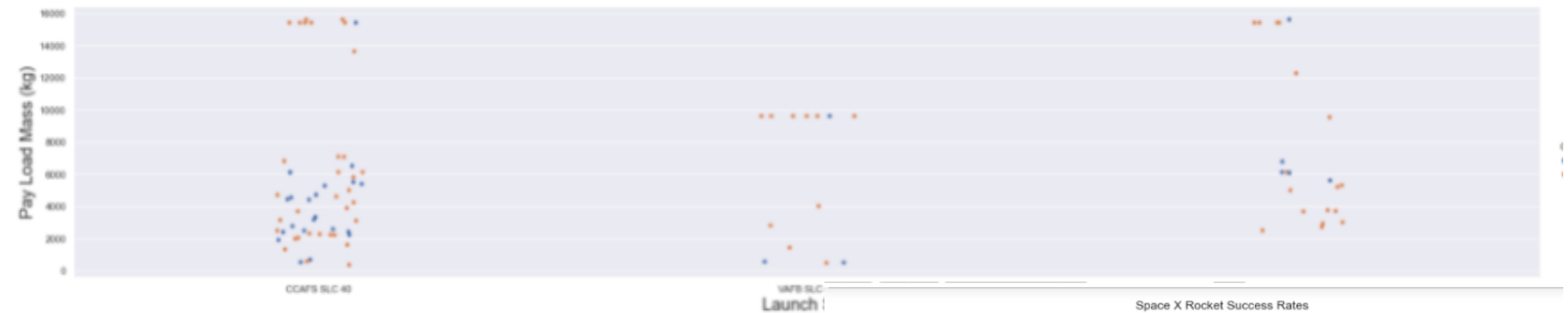
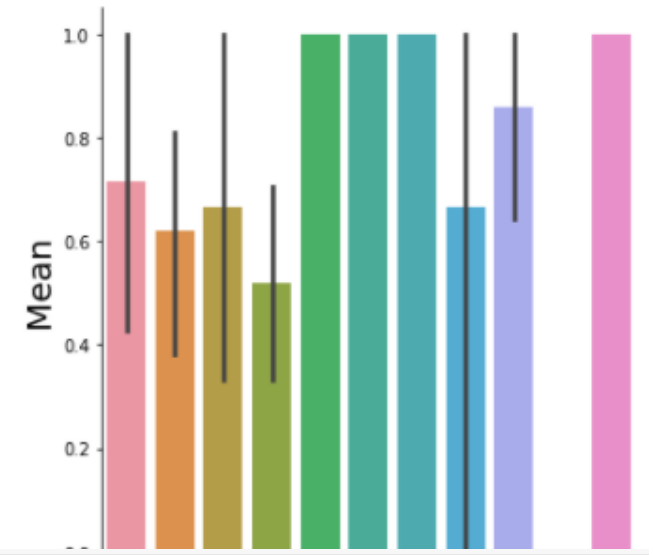
Analysis and Visualization



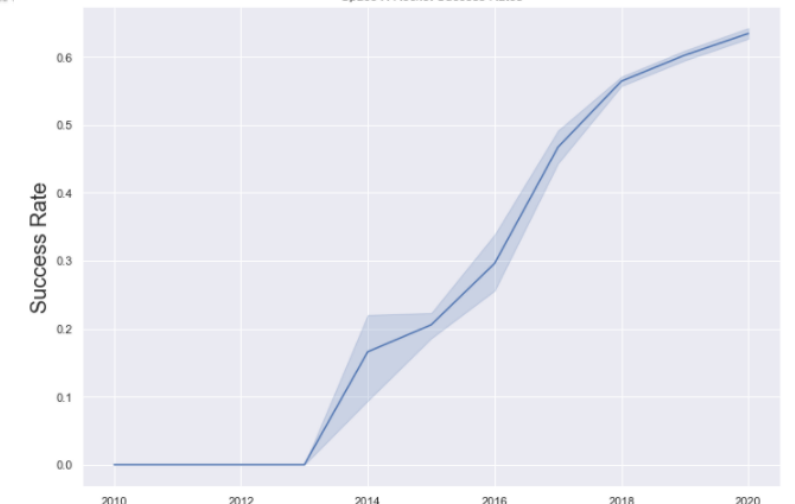
Collecting inferential statistics with SQL & Py Pandas



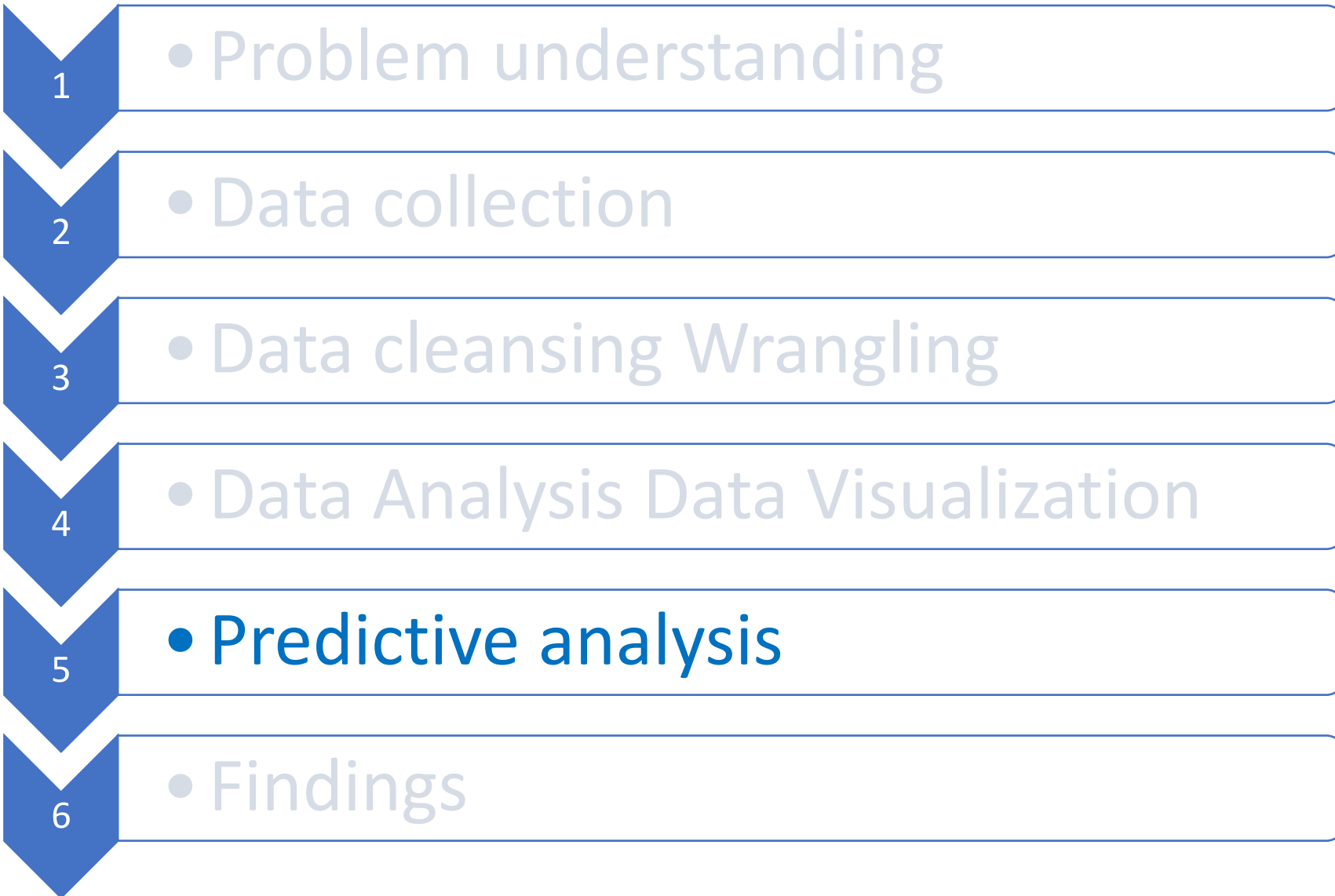
Visualizing Data using Pyplot



Identify correlations
Eliminate redundancy



METHODOLOGY



Predictive Analysis

BUILDING MODEL

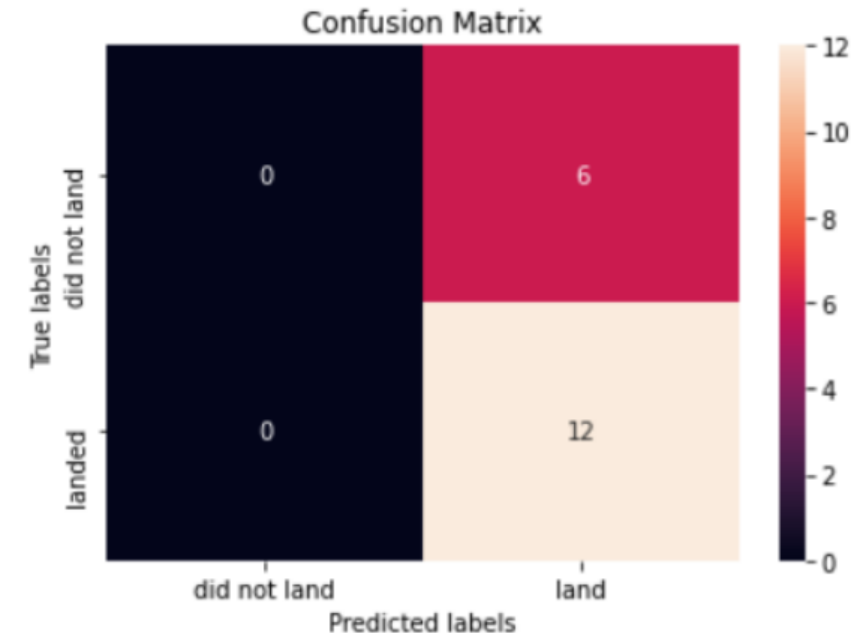
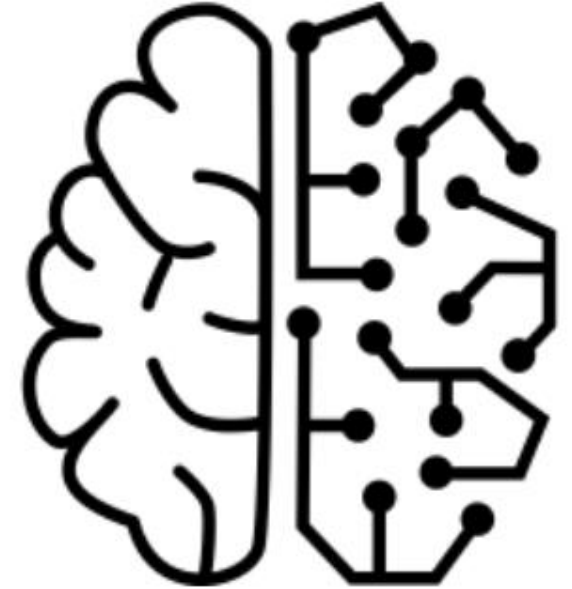
- Load our dataset into NumPy and Pandas
- Transform Data (scalar)
- Split data into training and test data sets
- Check the number of test samples
- Decide which type of machine learning algorithms to run
- Set parameters and algorithms to GridSearchCV
- Fit datasets into the GridSearchCV objects and train our dataset.

EVALUATING MODEL

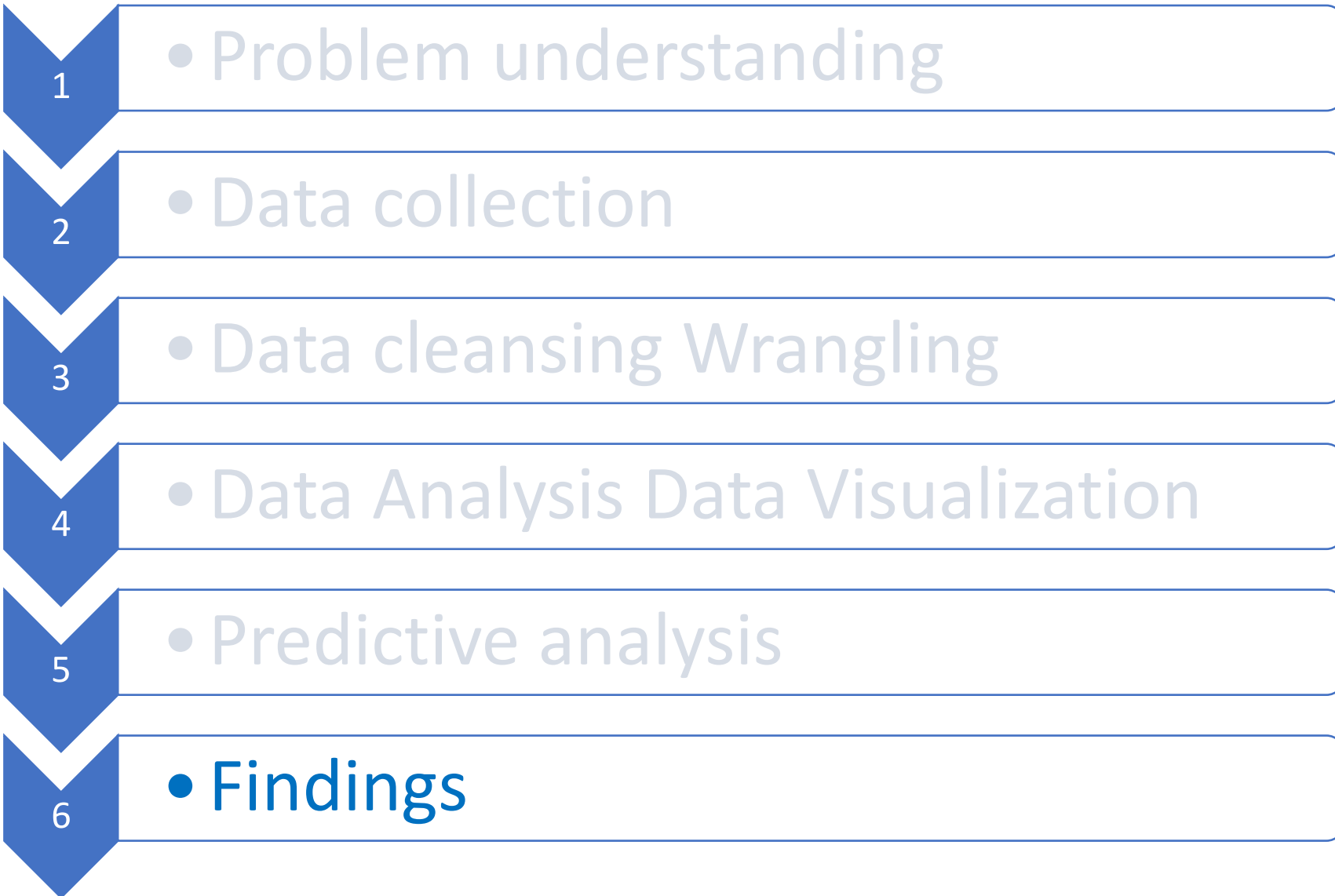
- compute the accuracy for each model
- Get tuned hyperparameters for each type of algorithms
- Plot Confusion Matrix IMPROVING MODEL
- Feature Engineering
- Algorithm Tuning

FINDING THE BEST MODEL






- The model with the highest accuracy score is retained



METHODOLOGY



DISCUSSION

-  Best Algorithms: Tree Classifier
-  Best launch site : KSC LC-39A
-  Best Orbite : GEO,HEO,SSO,ES-L1
-  Best payloads : with low weight
-  Success rate increase by year after year.