

Launch Sites Locations Analysis with Folium

Estimated time needed: **40** minutes

The launch success rate may depend on many factors such as payload mass, orbit type, and so on. It may also depend on the location and proximities of a launch site, i.e., the initial position of rocket trajectories. Finding an optimal location for building a launch site certainly involves many factors and hopefully we could discover some of the factors by analyzing the existing launch site locations.

In the previous exploratory data analysis labs, you have visualized the SpaceX launch dataset using `matplotlib` and `seaborn` and discovered some preliminary correlations between the launch site and success rates. In this lab, you will be performing more interactive visual analytics using `Folium`.

Objectives

This lab contains the following tasks:

- **TASK 1:** Mark all launch sites on a map
- **TASK 2:** Mark the success/failed launches for each site on the map
- **TASK 3:** Calculate the distances between a launch site to its proximities

After completed the above tasks, you should be able to find some geographical patterns about launch sites.

Let's first import required Python packages for this lab:

```
In [2]: import piplite
        await piplite.install(['folium'])
        await piplite.install(['pandas'])
```

```
In [3]: import folium
        import pandas as pd
```

```
In [4]: # Import folium MarkerCluster plugin
        from folium.plugins import MarkerCluster
```

```
# Import folium MousePosition plugin
from folium.plugins import MousePosition
# Import folium DivIcon plugin
from folium.features import DivIcon
```

If you need to refresh your memory about folium, you may download and refer to this previous folium lab:

Generating Maps with Python

Task 1: Mark all launch sites on a map

First, let's try to add each site's location on a map using site's latitude and longitude coordinates

The following dataset with the name `spacex_launch_geo.csv` is an augmented dataset with latitude and longitude added for each site.

```
In [6]: # Download and read the `spacex_launch_geo.csv`
from js import fetch
import io

URL = 'https://cf-courses-data.s3.us.cloud-object-storage.appdomain.cloud/IBM-DS0321
resp = await fetch(URL)
spacex_csv_file = io.BytesIO((await resp.arrayBuffer()).to_py())
spacex_df=pd.read_csv(spacex_csv_file)
```

Now, you can take a look at what are the coordinates for each site.

```
In [7]: # Select relevant sub-columns: `Launch Site`, `Lat(Latitude)`, `Long(Longitude)`, `c
spacex_df = spacex_df[['Launch Site', 'Lat', 'Long', 'class']]
launch_sites_df = spacex_df.groupby(['Launch Site'], as_index=False).first()
launch_sites_df = launch_sites_df[['Launch Site', 'Lat', 'Long']]
launch_sites_df
```

```
Out[7]:
```

	Launch Site	Lat	Long
0	CCAFS LC-40	28.562302	-80.577356
1	CCAFS SLC-40	28.563197	-80.576820
2	KSC LC-39A	28.573255	-80.646895
3	VAFB SLC-4E	34.632834	-120.610745

Above coordinates are just plain numbers that can not give you any intuitive insights about where are those launch sites. If you are very good at geography, you can interpret those numbers directly in your mind. If not, that's fine too. Let's visualize those locations by pinning them on a map.

We first need to create a folium Map object, with an initial center location to be NASA Johnson Space Center at Houston, Texas.

```
In [8]: # Start Location is NASA Johnson Space Center
nasa_coordinate = [29.559684888503615, -95.0830971930759]
site_map = folium.Map(location=nasa_coordinate, zoom_start=10)
```

We could use folium.Circle to add a highlighted circle area with a text label on a specific coordinate. For example,

```
In [83]: # Create a blue circle at NASA Johnson Space Center's coordinate with a popup label
circle = folium.Circle(nasa_coordinate, radius=1000, color='#d35400', fill=True).add
# Create a blue circle at NASA Johnson Space Center's coordinate with a icon showing
marker = folium.map.Marker(
    nasa_coordinate,
    # Create an icon as a text label
    icon=DivIcon(
        icon_size=(20,20),
        icon_anchor=(0,0),
        html='<div style="font-size: 12; color:#d35400;"><b>%s</b></div>' % 'NASA JS
    )
)
site_map.add_child(circle)
site_map.add_child(marker)
```

Out[83]: Make this Notebook Trusted to load map: File -> Trust Notebook

and you should find a small yellow circle near the city of Houston and you can zoom-in to see a larger circle.

Now, let's add a circle for each launch site in data frame launch_sites

TODO: Create and add `folium.Circle` and `folium.Marker` for each launch site on the site map

An example of `folium.Circle`:

```
folium.Circle(coordinate, radius=1000, color='#000000',
fill=True).add_child(folium.Popup(...))
```

An example of `folium.Marker`:

```
folium.map.Marker(coordinate, icon=DivIcon(icon_size=(20,20),icon_anchor=
(0,0), html='<div style="font-size: 12; color:#d35400;"><b>%s</b></div>' %
'label', ))
```

In [84]:

```
# Initial the map
site_map = folium.Map(location=nasa_coordinate, zoom_start=5)
# For each launch site, add a Circle object based on its coordinate (Lat, Long) value

circle1 = folium.Circle([launch_sites_df.Lat[0],launch_sites_df.Long[0]], radius=1000)
circle2 = folium.Circle([launch_sites_df.Lat[1],launch_sites_df.Long[1]], radius=1000)
circle3 = folium.Circle([launch_sites_df.Lat[2],launch_sites_df.Long[2]], radius=1000)
circle4 = folium.Circle([launch_sites_df.Lat[3],launch_sites_df.Long[3]], radius=1000)

marker1 = folium.map.Marker(
    [launch_sites_df.Lat[0],launch_sites_df.Long[0]],
    # Create an icon as a text label
    icon=DivIcon(
        icon_size=(20,20),
        icon_anchor=(0,0),
        html='<div style="font-size: 12; color:#d35400;"><b>%s</b></div>' % launch_sites_df.Lat[0]
    )
)

marker2 = folium.map.Marker(
    [launch_sites_df.Lat[1],launch_sites_df.Long[1]],
    # Create an icon as a text label
    icon=DivIcon(
        icon_size=(20,20),
        icon_anchor=(0,0),
        html='<div style="font-size: 12; color:#d35400;"><b>%s</b></div>' % launch_sites_df.Lat[1]
    )
)

marker3 = folium.map.Marker(
    [launch_sites_df.Lat[2],launch_sites_df.Long[2]],
    # Create an icon as a text label
    icon=DivIcon(
        icon_size=(20,20),
        icon_anchor=(0,0),
        html='<div style="font-size: 12; color:#d35400;"><b>%s</b></div>' % launch_sites_df.Lat[2]
    )
)

marker4 = folium.map.Marker(
    [launch_sites_df.Lat[3],launch_sites_df.Long[3]],
    # Create an icon as a text label
    icon=DivIcon(
        icon_size=(20,20),
```

```

        icon_anchor=(0,0),
        html='<div style="font-size: 12; color:#d35400;"><b>%s</b></div>' % launch_s
    )
)
site_map.add_child(circle1)
site_map.add_child(marker1)

site_map.add_child(circle2)
site_map.add_child(marker2)

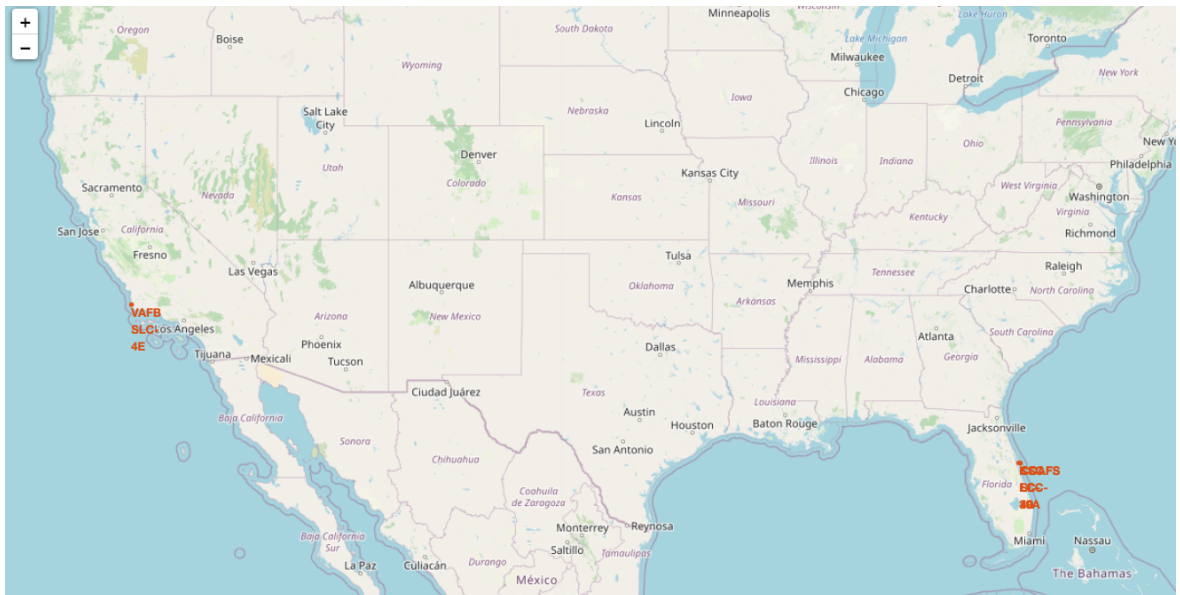
site_map.add_child(circle3)
site_map.add_child(marker3)

site_map.add_child(circle4)
site_map.add_child(marker4)

```

Out[84]: Make this Notebook Trusted to load map: File -> Trust Notebook

The generated map with marked launch sites should look similar to the following:



Now, you can explore the map by zoom-in/out the marked areas , and try to answer the following questions:

- Are all launch sites in proximity to the Equator line? no
- Are all launch sites in very close proximity to the coast? yes

Also please try to explain your findings.

Task 2: Mark the success/failed launches for each site on the map

Next, let's try to enhance the map by adding the launch outcomes for each site, and see which sites have high success rates. Recall that data frame `spacex_df` has detailed launch records, and the `class` column indicates if this launch was successful or not

```
In [85]: spacex_df.tail(10)
```

```
Out[85]:
```

	Launch Site	Lat	Long	class	marker_color
46	KSC LC-39A	28.573255	-80.646895	1	#008000
47	KSC LC-39A	28.573255	-80.646895	1	#008000
48	KSC LC-39A	28.573255	-80.646895	1	#008000
49	CCAFS SLC-40	28.563197	-80.576820	1	#008000
50	CCAFS SLC-40	28.563197	-80.576820	1	#008000
51	CCAFS SLC-40	28.563197	-80.576820	0	#FF0000
52	CCAFS SLC-40	28.563197	-80.576820	0	#FF0000
53	CCAFS SLC-40	28.563197	-80.576820	0	#FF0000
54	CCAFS SLC-40	28.563197	-80.576820	1	#008000
55	CCAFS SLC-40	28.563197	-80.576820	0	#FF0000

Next, let's create markers for all launch records. If a launch was successful (`class=1`) , then we use a green marker and if a launch was failed, we use a red marker (`class=0`)

Note that a launch only happens in one of the four launch sites, which means many launch records will have the exact same coordinate. Marker clusters can be a good way to simplify a map containing many markers having the same coordinate.

Let's first create a `MarkerCluster` object

```
In [86]: marker_cluster = MarkerCluster()
```

TODO: Create a new column in `launch_sites` dataframe called `marker_color` to store the marker colors based on the `class` value

```
In [87]: # Apply a function to check the value of `class` column
# If class=1, marker_color value will be green
# If class=0, marker_color value will be red
color=[]
for i in spacex_df['class']:
    if i ==0:
        color.append("#FF0000")
    else:
        color.append("#008000")
spacex_df['marker_color']=color
```

TODO: For each launch result in `spacex_df` data frame, add a `folium.Marker` to `marker_cluster`

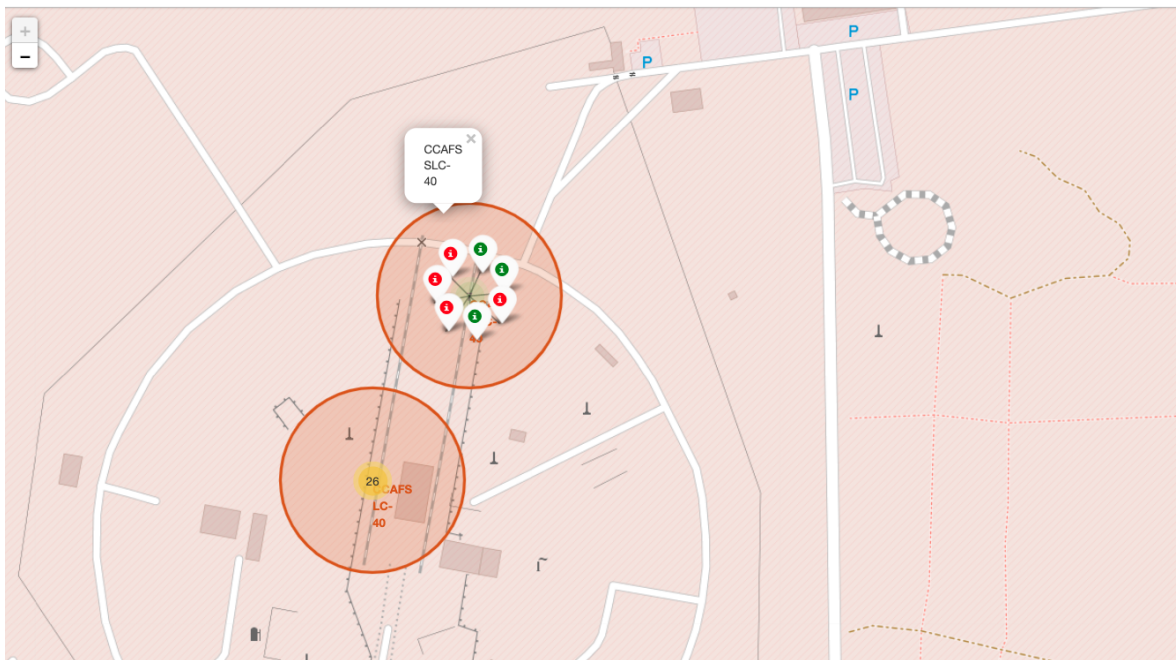
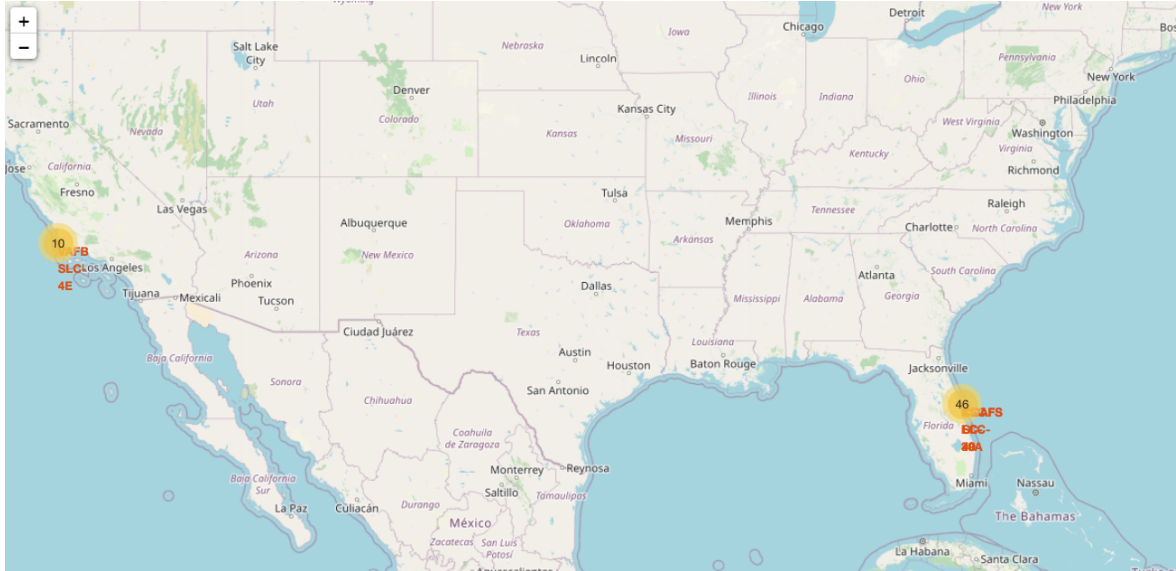
```
In [88]: # Add marker_cluster to current site_map
marker_cluster = MarkerCluster()
site_map.add_child(marker_cluster)

# for each row in spacex_df data frame
# create a Marker object with its coordinate
# and customize the Marker's icon property to indicate if this launch was succeeded
# e.g., icon=folium.Icon(color='white', icon_color=row['marker_color'])
for index, record in spacex_df.iterrows():
    # TODO: Create and add a Marker cluster to the site map
    marker = folium.Marker(
        [spacex_df.Lat[index], spacex_df.Long[index]],
        # Create an icon as a text label
        icon=folium.Icon(color='white', icon_color=record['marker_color']),
        popup=spacex_df['Launch Site'][index]
    )
    marker_cluster.add_child(marker)

site_map
```

Out[88]: Make this Notebook Trusted to load map: File -> Trust Notebook

Your updated map may look like the following screenshots:



From the color-labeled markers in marker clusters, you should be able to easily identify which launch sites have relatively high success rates.

TASK 3: Calculate the distances between a launch site to its proximities

Next, we need to explore and analyze the proximities of launch sites.

Let's first add a `MousePosition` on the map to get coordinate for a mouse over a point on the map. As such, while you are exploring the map, you can easily find the coordinates of any points of interests (such as railway)

```
In [89]: # Add Mouse Position to get the coordinate (Lat, Long) for a mouse over on the map
formatter = "function(num) {return L.Util.formatNum(num, 5)};"
mouse_position = MousePosition(
    position='topright',
    separator=' Long: ',
    empty_string='NaN',
    lng_first=False,
    num_digits=20,
    prefix='Lat:',
    lat_formatter=formatter,
    lng_formatter=formatter,
)

site_map.add_child(mouse_position)
site_map
```

Out[89]: Make this Notebook Trusted to load map: File -> Trust Notebook

Now zoom in to a launch site and explore its proximity to see if you can easily find any railway, highway, coastline, etc. Move your mouse to these points and mark down their coordinates (shown on the top-left) in order to the distance to the launch site.

Now zoom in to a launch site and explore its proximity to see if you can easily find any railway, highway, coastline, etc. Move your mouse to these points and mark down their coordinates (shown on the top-left) in order to the distance to the launch site.

```
In [90]: from math import sin, cos, sqrt, atan2, radians

def calculate_distance(lat1, lon1, lat2, lon2):
    # approximate radius of earth in km
    R = 6373.0

    lat1 = radians(lat1)
    lon1 = radians(lon1)
    lat2 = radians(lat2)
    lon2 = radians(lon2)

    dlon = lon2 - lon1
    dlat = lat2 - lat1

    a = sin(dlat / 2)**2 + cos(lat1) * cos(lat2) * sin(dlon / 2)**2
    c = 2 * atan2(sqrt(a), sqrt(1 - a))

    distance = R * c
    return distance
```

TODO: Mark down a point on the closest coastline using MousePosition and calculate the distance between the coastline point and the launch site.

```
In [91]: # find coordinate of the closet coastline
# e.g.,: Lat: 28.56367 Lon: -80.57163
distance_coastline = calculate_distance(launch_sites_df.Lat[0], launch_sites_df.Long[0],
```

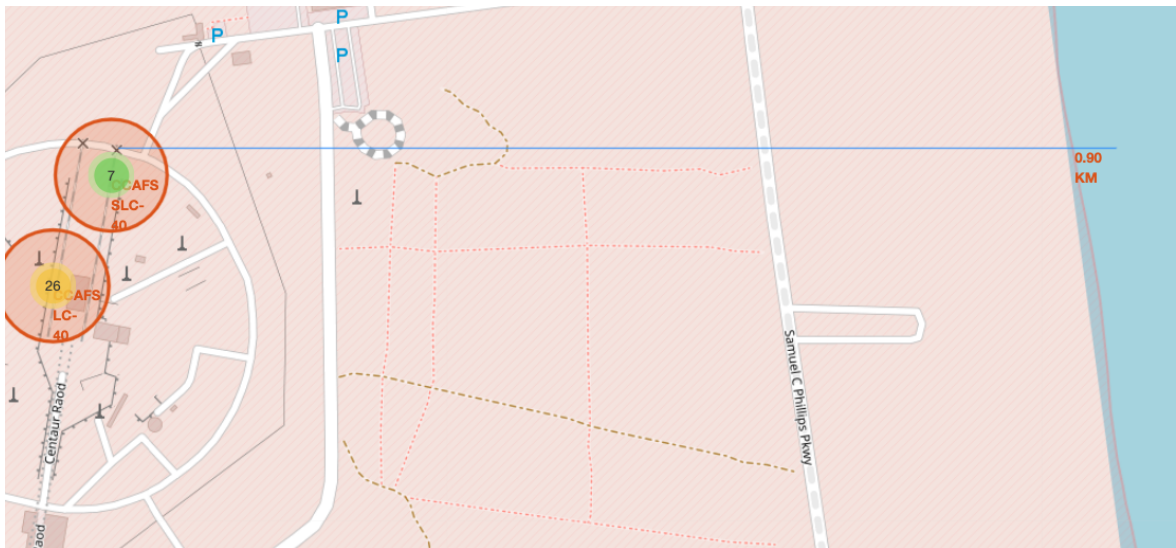
```
In [92]: # Create and add a folium.Marker on your selected closest coastline point on the map
# Display the distance between coastline point and launch site using the icon proper
# for example
distance_marker = folium.Marker(
    [28.56325, -80.56789],
    icon=DivIcon(
        icon_size=(20,20),
        icon_anchor=(0,0),
        html='<div style="font-size: 12; color:#d35400;"><b>%s</b></div>' % "{:10.2f}"
    )
)
```

TODO: Draw a PolyLine between a launch site to the selected coastline point

```
In [93]: # Create a `folium.PolyLine` object using the coastline coordinates and launch site
lines=folium.PolyLine(locations=[[launch_sites_df.Lat[0], launch_sites_df.Long[0]], [28.56325, -80.56789]])
site_map.add_child(lines)
site_map.add_child(distance_marker)
```

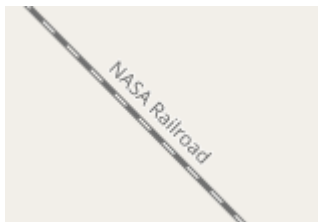
Out[93]: Make this Notebook Trusted to load map: File -> Trust Notebook

Your updated map with distance line should look like the following screenshot:

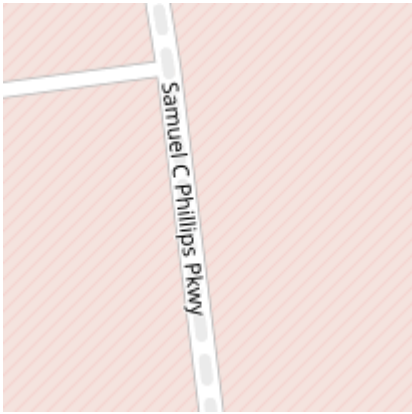


TODO: Similarly, you can draw a line between a launch site to its closest city, railway, highway, etc. You need to use `MousePosition` to find their coordinates on the map first

A railway map symbol may look like this:



A highway map symbol may look like this:



A city map symbol may look like this:



In [100...

```
# Create a marker with distance to a closest city, railway, highway, etc.
# Draw a Line between the marker to the launch site
mark={"closest city": [34.95321, -120.43625, "santa maria"], "railway": [34.63816, -120.62
```

In []:

```
distance_marker2 = folium.Marker(
    [mark['closest city'][0], mark['closest city'][1]]
    icon=DivIcon(
        icon_size=(20,20),
        icon_anchor=(0,0),
        html='<div style="font-size: 12; color:#d35400;"><b>%s</b></div>' % "{:10.2f}"
    )
)
```

In [102...

```
for i in mark.keys():
    mark[i].append(calculate_distance(launch_sites_df.Lat[3], launch_sites_df.Long[3])
    distance_marker2 = folium.Marker(
        mark[i][0:2],
        icon=DivIcon(
            icon_size=(20,20),
            icon_anchor=(0,0),
            html='<div style="font-size: 12; color:#d35400;"><b>%s</b></div>' % "{:10.2f}"
        ), popup=mark[i][2]
    )
    site_map.add_child(distance_marker2)
    lines=folium.PolyLine(locations=[[launch_sites_df.Lat[3], launch_sites_df.Long[3]]])
    site_map.add_child(lines)
```

In [103...

```
site_map
```

Out[103...] Make this Notebook Trusted to load map: File -> Trust Notebook

After you plot distance lines to the proximities, you can answer the following questions easily:

- Are launch sites in close proximity to railways?
- Are launch sites in close proximity to highways?
- Are launch sites in close proximity to coastline?
- Do launch sites keep certain distance away from cities?

Also please try to explain your findings.

In [113...

```
kk={"proximity":[],"name":[],"distance in km":[]}
for i in mark.keys():
    kk["proximity"].append(i)
    kk["name"].append(mark[i][2])
    kk["distance in km"].append(mark[i][3])
```

In [114...

```
a=pd.DataFrame(kk)
a
```

Out[114...

	proximity	name	distance in km
0	closest city	santa maria	39.037573
1	railway	santa barabara subdivision	1.238510
2	highway	ocean avenue	5.641735
3	coastline	coastline	1.346707

In []:

Next Steps:

Now you have discovered many interesting insights related to the launch sites' location using folium, in a very interactive way. Next, you will need to build a dashboard using Plotly Dash on detailed launch records.

Authors

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Change Log

Date (YYYY-MM-DD)	Version	Changed By	Change Description
2022-11-09	1.0	Pratiksha Verma	Converted initial version to Jupyterlite

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