

Hands-on Lab: Interactive Visual Analytics 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 [19]:

!pip install folium pandas

```
!pip3 install folium
!pip3 install wget

#await piplite.install(['folium'])
#await piplite.install(['pandas'])
```

```
Requirement already satisfied: folium in /opt/conda/lib/pvthon3.11/site-packages (0.18.0)
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        ->folium) (2024.8.30)
        Collecting wget
          Downloading wget-3.2.zip (10 kB)
          Preparing metadata (setup.py) ... done
        Building wheels for collected packages: wget
          Building wheel for wget (setup.py) ... done
          Created wheel for wget: filename=wget-3.2-py3-none-any.whl size=9656 sha256=da0889b319bf2568af52b0d111f5c
        3f95a1f295ac8f1195944393949960f9858
          Stored in directory: /home/jupyterlab/.cache/pip/wheels/40/b3/0f/a40dbd1c6861731779f62cc4babcb234387e11d6
        97df70ee97
        Successfully built wget
        Installing collected packages: wget
        Successfully installed wget-3.2
 In []:
In [20]:
         import folium
         import wget
```

import pandas as pd

!pip install folium==0.8.3

```
Collecting folium==0.8.3
          Downloading folium-0.8.3-py2.py3-none-any.whl.metadata (3.0 kB)
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        ->folium==0.8.3) (2.2.1)
        Requirement already satisfied: certifi>=2017.4.17 in /opt/conda/lib/python3.11/site-packages (from requests
        ->folium==0.8.3) (2024.8.30)
        Downloading folium-0.8.3-py2.py3-none-any.whl (87 kB)
                                                  — 87.7/87.7 kB 10.5 MB/s eta 0:00:00
        Installing collected packages: folium
          Attempting uninstall: folium
            Found existing installation: folium 0.18.0
            Uninstalling folium-0.18.0:
              Successfully uninstalled folium-0.18.0
        Successfully installed folium-0.8.3
 In []:
In [21]: # 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
 In [ ]:
```

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

Generating Maps with Python

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 [ ]: spacex_csv_file = wget.download('https://cf-courses-data.s3.us.cloud-object-storage.appdomain.cloud/IBM-DS(
    spacex_df=pd.read_csv(spacex_csv_file)
```

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

```
In [23]: # Select relevant sub-columns: `Launch Site`, `Lat(Latitude)`, `Long(Longitude)`, `class`
    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[23]:		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

```
In []:
```

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

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

```
In [25]: # Create a blue circle at NASA Johnson Space Center's coordinate with a popup label showing its name
    circle = folium.Circle(nasa_coordinate, radius=1000, color='#d35400', fill=True).add_child(folium.Popup('N, # Create a blue circle at NASA Johnson Space Center's coordinate with a icon showing its name
    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 JSC',
        )
        site_map.add_child(circle)
        site_map.add_child(marker)
```

Out [25]: Make this Notebook Trusted to load map: File -> Trust Notebook

In []:

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(...))

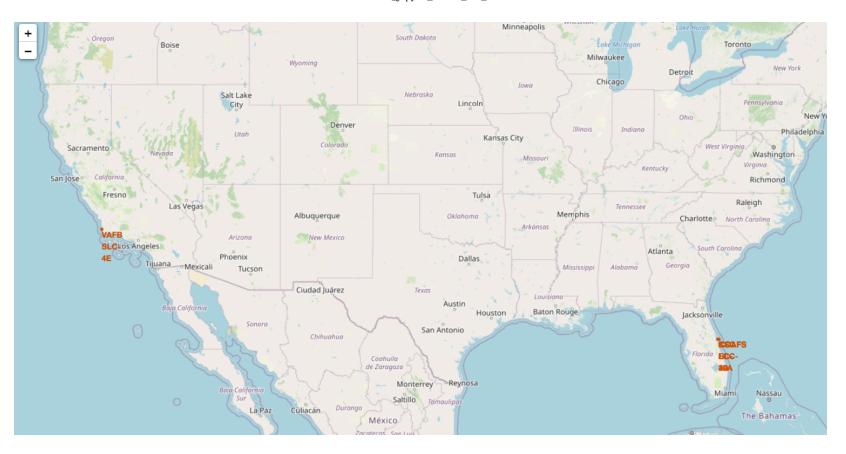
An example of folium. Marker:

```
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', ))
```

Out [26]: Make this Notebook Trusted to load map: File -> Trust Notebook

In []:

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?
- Are all launch sites in very close proximity to the coast?

Also please try to explain your findings.

Yes launch sites are in proximity to the equator and the coast. This makes sense as it takes less fuel to get into space from the equator due to the physics of Earth's rotation. The launch sites in close proximity to the coast are also logical for safety reasons.

In [29]: # 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 [31]: spacex_df

Out[31]:		Launch Site	Lat	Long	class
	0	CCAFS LC-40	28.562302	-80.577356	0
	1	CCAFS LC-40	28.562302	-80.577356	0
	2	CCAFS LC-40	28.562302	-80.577356	0
	3	CCAFS LC-40	28.562302	-80.577356	0
	4	CCAFS LC-40	28.562302	-80.577356	0
	5	CCAFS LC-40	28.562302	-80.577356	0
	6	CCAFS LC-40	28.562302	-80.577356	0
	7	CCAFS LC-40	28.562302	-80.577356	0
	8	CCAFS LC-40	28.562302	-80.577356	0
	9	CCAFS LC-40	28.562302	-80.577356	0
	10	CCAFS LC-40	28.562302	-80.577356	0
	11	CCAFS LC-40	28.562302	-80.577356	0
	12	CCAFS LC-40	28.562302	-80.577356	0
	13	CCAFS LC-40	28.562302	-80.577356	0
	14	CCAFS LC-40	28.562302	-80.577356	0
	15	CCAFS LC-40	28.562302	-80.577356	0
	16	CCAFS LC-40	28.562302	-80.577356	0
	17	CCAFS LC-40	28.562302	-80.577356	1
	18	CCAFS LC-40	28.562302	-80.577356	1
	19	CCAFS LC-40	28.562302	-80.577356	0
	20	CCAFS LC-40	28.562302	-80.577356	1
	21	CCAFS LC-40	28.562302	-80.577356	1
	22	CCAFS LC-40	28.562302	-80.577356	1

	Launch Site	Lat	Long	class
23	CCAFS LC-40	28.562302	-80.577356	0
24	CCAFS LC-40	28.562302	-80.577356	1
25	CCAFS LC-40	28.562302	-80.577356	1
26	VAFB SLC-4E	34.632834	-120.610745	0
27	VAFB SLC-4E	34.632834	-120.610745	0
28	VAFB SLC-4E	34.632834	-120.610745	1
29	VAFB SLC-4E	34.632834	-120.610745	1
30	VAFB SLC-4E	34.632834	-120.610745	1
31	VAFB SLC-4E	34.632834	-120.610745	1
32	VAFB SLC-4E	34.632834	-120.610745	0
33	VAFB SLC-4E	34.632834	-120.610745	0
34	VAFB SLC-4E	34.632834	-120.610745	0
35	VAFB SLC-4E	34.632834	-120.610745	0
36	KSC LC-39A	28.573255	-80.646895	1
37	KSC LC-39A	28.573255	-80.646895	0
38	KSC LC-39A	28.573255	-80.646895	1
39	KSC LC-39A	28.573255	-80.646895	1
40	KSC LC-39A	28.573255	-80.646895	0
41	KSC LC-39A	28.573255	-80.646895	1
42	KSC LC-39A	28.573255	-80.646895	1
43	KSC LC-39A	28.573255	-80.646895	0
44	KSC LC-39A	28.573255	-80.646895	1
45	KSC LC-39A	28.573255	-80.646895	1

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

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 [32]: marker_cluster = MarkerCluster()
```

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

```
In [33]: def assign_marker_color(launch_outcome):
    if launch_outcome == 1:
        return 'green'
    else:
        return 'red'
```

```
spacex_df['marker_color'] = spacex_df['class'].apply(assign_marker_color)
spacex_df.tail(10)
```

Out[33]:

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

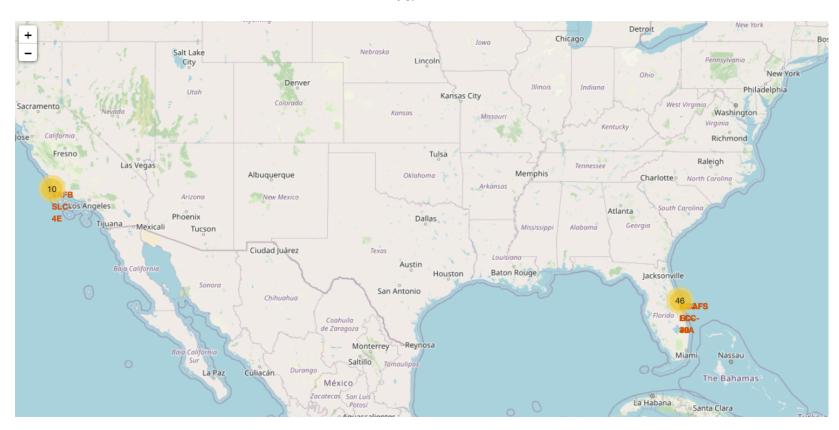
In []:

TODO: For each launch result in spacex df data frame, add a folium.Marker to marker cluster

Out [34]: Make this Notebook Trusted to load map: File -> Trust Notebook

In []:

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.

```
In [35]: # 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 [35]: Make this Notebook Trusted to load map: File -> Trust Notebook

```
In []:
```

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 [36]: # 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 [36]: Make this Notebook Trusted to load map: File -> Trust Notebook

In []:

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 [38]: 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
In []:
```

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

```
In [41]: # find coordinate of the closet coastline
    # e.g.,: Lat: 28.56367 Lon: -80.57163
    # distance_coastline = calculate_distance(launch_site_lat, launch_site_lon, coastline_lat, coastline_lon)
    launch_site_lat = 28.5631975
    launch_site_lon = -80.576820
    coastline_lat = 21.56328
    coastline_lon = -79.13276
    distance_coastline = calculate_distance(launch_site_lat, launch_site_lon, coastline_lat, coastline_lon)
    print(distance_coastline,' mi')

792.0501176057842 mi

In [42]: distance_marker = folium.Marker(
        [coastline_lat, coastline_lon],
        icon=DivIcon(
```

```
icon_size=(20,20),
    icon_anchor=(0,0),
    html='<div style="font-size: 12; color:#d35400;"><b>%s</b></div>' % "{:10.2f} KM".format(distance_color)
)
site_map.add_child(distance_marker)
```

Out [42]: Make this Notebook Trusted to load map: File -> Trust Notebook

```
In []:
```

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

```
In [43]: coordinates = [[launch_site_lat,launch_site_lon],[coastline_lat,coastline_lon]]
lines=folium.PolyLine(locations=coordinates, weight=1)
site_map.add_child(lines)
```

Out [43]: 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 betwee a launch site to its closest city, railway, highway, etc. You need to use MousePosition to find the their coordinates on the map first

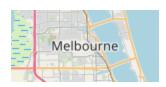
A railway map symbol may look like this:



A highway map symbol may look like this:



A city map symbol may look like this:



```
In [47]: # Create a marker with distance to a closest city, railway, highway, etc.
         # Draw a line between the marker to the launch site
         closest highway = 28.4230, -80.42905
         closest railroad = 28.4399, -80.30266
         closest city = 28.64280, -80.69651
In [ ]:
In [48]:
         distance_highway = calculate_distance(launch_site_lat, launch_site_lon, closest_highway[0], closest_highway
         print('distance_highway =',distance_highway, ' mi')
         distance_railroad = calculate_distance(launch_site_lat, launch_site_lon, closest_railroad[0], closest_rail
         print('distance_railroad =',distance_railroad, ' mi')
         distance_city = calculate_distance(launch_site_lat, launch_site_lon, closest_city[0], closest_city[1])
         print('distance_city =',distance_city, ' mi')
        distance highway = 21.25679661713841 mi
        distance_railroad = 30.104246776452985 mi
        distance city = 14.663342942873387 mi
```

```
In [ ]:
In [49]: # closest highway marker
         distance marker = folium.Marker(
            closest highway,
            icon=DivIcon(
                icon size=(20,20),
                icon anchor=(0,0),
                html='<div style="font-size: 12; color:#d35400;"><b>%s</b></div>' % "{:10.2f} KM".format(distance h
         site map.add child(distance marker)
         # closest highway line
         coordinates = [[launch site lat, launch site lon], closest highway]
         lines=folium.PolyLine(locations=coordinates, weight=1)
         site map.add child(lines)
         # closest railroad marker
         distance marker = folium.Marker(
            closest railroad,
            icon=DivIcon(
                icon size=(20,20),
                icon anchor=(0,0),
                html='<div style="font-size: 12; color:#d35400;"><b>%s</b></div>' % "{:10.2f} KM".format(distance rates)
         site map.add child(distance marker)
         # closest railroad line
         coordinates = [[launch site lat,launch site lon],closest railroad]
         lines=folium.PolyLine(locations=coordinates, weight=1)
         site map.add child(lines)
         # closest city marker
         distance marker = folium.Marker(
            closest city,
            icon=DivIcon(
                icon size=(20,20),
                icon anchor=(0,0),
                html='<div style="font-size: 12; color:#d35400;"><b>%s</b></div>' % "{:10.2f} KM".format(distance c
```

```
site_map.add_child(distance_marker)
# closest city line
coordinates = [[launch_site_lat,launch_site_lon],closest_city]
lines=folium.PolyLine(locations=coordinates, weight=1)
site_map.add_child(lines)
```

Out [49]: Make this Notebook Trusted to load map: File -> Trust Notebook

```
In [ ]:
```

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

• Are launch sites in close proximity to railways?

- Yes
- Are launch sites in close proximity to highways?
- Yes
- Are launch sites in close proximity to coastline?
- Yes
- Do launch sites keep certain distance away from cities?
- Yes, in my example about 15 miles

Also please try to explain your findings. As mentioned before, launch sites are in close proximity to equator to minimize fuel consumption by using Earth's ~ 30km/sec eastward spin to help spaceships get into orbit. Launch sites are in close proximity to coastline so they can fly over the ocean during launch, for at least two safety reasons— (1) crew has option to abort launch and attempt water landing (2) minimize people and property at risk from falling debris. Launch sites are in close proximity to highways, which allows for easily transport required people and property. Launch sites are in close proximity to railways, which allows transport for heavy cargo. Launch sites are not in close proximity to cities, which minimizes danger to population dense areas.

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 Ploty 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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