hw1_rl3089

February 3, 2020

1 HW1 Applied Machine Learning

```
[5]: import pandas as pd
  import matplotlib.pyplot as plt
  import numpy as np
  fire=pd.read_csv("fire_archive_V1_96617.csv")
```

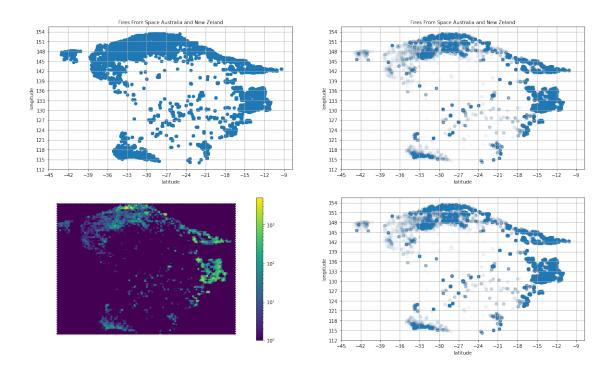
Task 1: Density Plots #1.1 Plot the longitude vs latitude several ways within a single figure (each in its own axes):

```
[48]: fig, axes = plt.subplots(2, 2, figsize=(20, 12))
     x=fire['latitude']
     y=fire['longitude']
     #first graph
     axes[0,0].scatter(x,y)
     axes[0,0].set_xlabel("latitude")
     axes[0,0].set_ylabel("longitude")
     axes[0,0].grid()
     axes[0,0].set_title('Fires From Space Australia and New Zeland',size=10)
     my_x_{ticks} = np.arange(-45, -7, 3)
     my_y_ticks = np.arange(112, 155, 3)
     axes[0,0].set_xticks(my_x_ticks)
     axes[0,0].set_yticks(my_y_ticks)
     #second graph
     axes[0,1].scatter(x,y,alpha=.01)
     axes[0,1].set_xlabel("latitude")
     axes[0,1].set_ylabel("longitude")
     axes[0,1].grid()
     axes[0,1].set_title('Fires From Space Australia and New Zeland', size=10)
     axes[0,1].set_xticks(my_x_ticks)
     axes[0,1].set_yticks(my_y_ticks)
     #third graph
```

```
X=[x.tolist(),y.tolist()]
im3=axes[1,0].hexbin(x, y, bins='log')
axes[1,0].axis("off")
plt.colorbar(im3,ax=axes[1,0])

#fourth graph
fire1=fire.sample(frac=0.2)
x1=fire1['latitude']
y1=fire1['longitude']

axes[1,1].scatter(x,y,alpha=.01)
axes[1,1].set_xlabel("latitude")
axes[1,1].set_ylabel("longitude")
axes[1,1].set_ylabel("longitude")
axes[1,1].set_title('Fires From Space Australia and New Zeland',size=10)
axes[1,1].set_xticks(my_x_ticks)
axes[1,1].set_yticks(my_y_ticks)
```

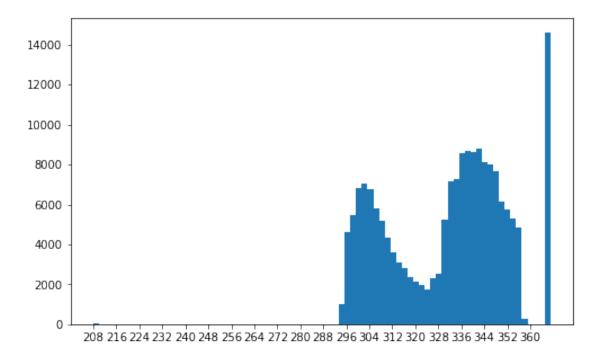


2.2
The anomalies (measurements) most located in:
#1 Latitude (-33,-27) with Longitude(151,154)
#2 Latitude (-18,-12) with Longitude(129,137)
#3 Latitude (-18,-12) with Longitude(141,144)
Task 2: Visualizing class membership
2.0

Visualize the distribution of Brightness temperature I-4 as a histogram (with appropriate settings).

Through the graph below, we are certain on the value of 267, it is saturated.

```
<matplotlib.axis.XTick at 0x11ed5a908>,
<matplotlib.axis.XTick at 0x11ed5ae80>,
<matplotlib.axis.XTick at 0x11ed4d438>,
<matplotlib.axis.XTick at 0x11ed5a780>,
<matplotlib.axis.XTick at 0x11ed627b8>,
<matplotlib.axis.XTick at 0x11ed4d860>,
<matplotlib.axis.XTick at 0x11ed4de48>,
<matplotlib.axis.XTick at 0x11ed38400>,
<matplotlib.axis.XTick at 0x11ed38978>],
<a list of 20 Text xticklabel objects>)
```

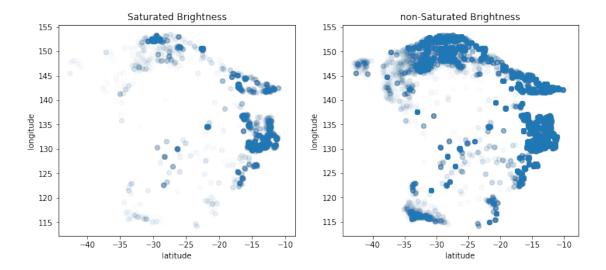


2.1 Do a small multiples plot of whether the brightness is saturated

```
[55]: fig, axe = plt.subplots(1, 2,figsize=(12,5))
    fire1=fire.loc[fire['bright_ti4']==367]
    axe[0].scatter(fire1['latitude'],fire1['longitude'],alpha=.01)
    axe[0].set_title("Saturated Brightness")
    axe[0].set_xlabel("latitude")
    axe[0].set_ylabel("longitude")

fire2=fire.loc[fire['bright_ti4']<367]
    axe[1].scatter(fire2['latitude'],fire2['longitude'],alpha=.01)
    axe[1].set_title("non-Saturated Brightness")
    axe[1].set_xlabel("latitude")
    axe[1].set_ylabel("longitude")</pre>
```

[55]: Text(0, 0.5, 'longitude')



Answer:

we can find that in the area whose longitude is below 125, brightness is rarely saturated. However, several groups of non-saturated brightness exists in this area.

Saturated brightness located upon logitude 145 is less aggregated than non-saturated brightness

2.2 Plot both groups in the same axes with different colors.

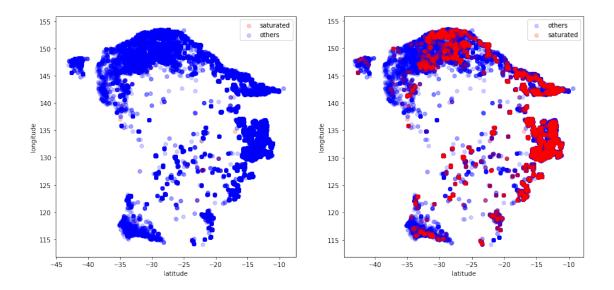
```
[53]: fig, axe = plt.subplots(1, 2,figsize=(15,7))
    fire1=fire.loc[fire['bright_ti4']==367]
    fire2=fire.loc[fire['bright_ti4']<367]</pre>
    axe[0].

scatter(fire1['latitude'],fire1['longitude'],c="r",label="saturated",alpha=0.
     →2)
    axe[0].

scatter(fire2['latitude'],fire2['longitude'],c="b",label="others",alpha=0.2)
    axe[0].set xlabel("latitude")
    axe[0].set_ylabel("longitude")
    axe[0].legend(loc='best')
    axe[1].
     axe[1].

→scatter(fire1['latitude'],fire1['longitude'],c="r",label="saturated",alpha=0.
    axe[1].set_xlabel("latitude")
    axe[1].set_ylabel("longitude")
    axe[1].legend(loc='best')
```

[53]: <matplotlib.legend.Legend at 0x121f81d68>



Answer:

The first graph is mainly covered by the blue points—"non-saturated". Those "saturated" points can be telled in the second graph, as its quantity is far smaller that of "non-saturated" ones.

By plotting both groups in the same axes, we are able to tell their location differences much easier.

2.3 Can you find a better way to compare the two distributions?

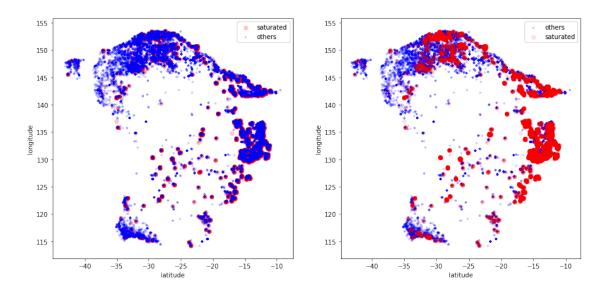
```
[57]: fig, axe = plt.subplots(1, 2,figsize=(15,7))
     fire1=fire.loc[fire['bright_ti4']==367]
     fire2=fire.loc[fire['bright_ti4']<367]</pre>
     axe[0].

→scatter(fire1['latitude'],fire1['longitude'],c="r",label="saturated",alpha=0.
      →2)
     axe[0].

→scatter(fire2['latitude'],fire2['longitude'],c="b",label="others",alpha=0.
      \rightarrow2, marker='+', s=20)
     axe[0].set xlabel("latitude")
     axe[0].set ylabel("longitude")
     axe[0].legend(loc='best')
     axe[1].
      →scatter(fire2['latitude'],fire2['longitude'],c="b",label="others",alpha=0.
      \rightarrow2, marker='+', s=20)
     axe[1].
      →scatter(fire1['latitude'],fire1['longitude'],c="r",label="saturated",alpha=0.
      →1)
     axe[1].set_xlabel("latitude")
     axe[1].set_ylabel("longitude")
```

axe[1].legend(loc='best')

[57]: <matplotlib.legend.Legend at 0x11daddf28>



Answer:

By adjusting the marker size and changed the "others" symbol to "+", we can better tell the differences.