

# 巨量資料hw1

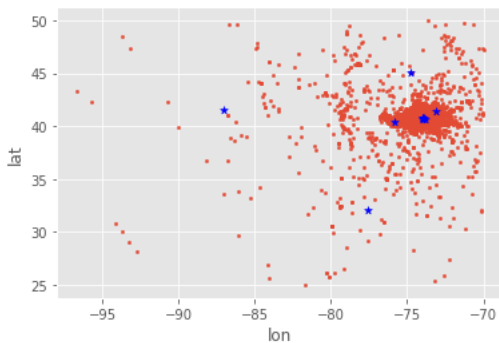
309552063吳冠潔

the scale of data : 資料總數有41859906行以及18列

analytical tools : python pandas

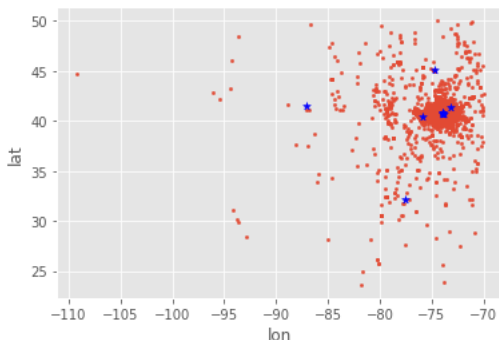
- Q1: What regions have the most pickups? What are the top-5 regions with the most pickups and drop-offs (pickups and drop-offs should be counted separately)?

## 1. pickup



由圖上可知，最多人上車的地方為經緯度約(-73.9 40.7)，最多人上車的前五個地方在經緯度(-73.96 40.78), (-73.13 41.36), (-73.996 40.73), (-73.78 40.72), (-73.98 40.76)的附近

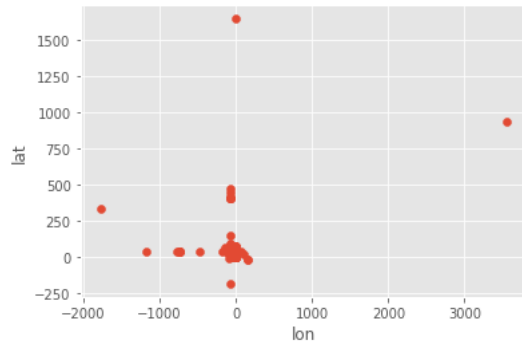
## 2. dropoff



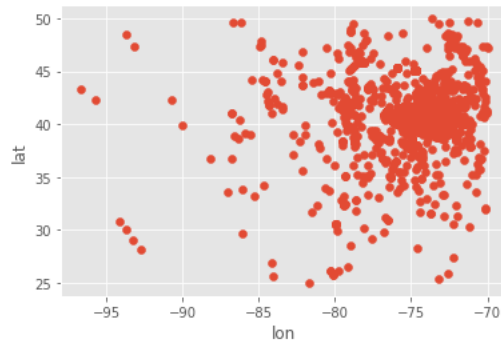
由圖上可知，最多人下車的地方為經緯度約(-73.9 40.7)，最多人下車的前五個地方在經緯度(-73.99 40.73), (-73.96 40.78), (-73.972 40.679), (-73.98 40.757), (-73.88 40.76)的附近

- how I solve this question :  
一開始先將1-3月的資料匯入到pandas中，並合併再一起，再來將pickup、dropoff的經緯度做成array，並將經緯度錯誤的資料刪除，有許多的資料經緯度不在範圍內

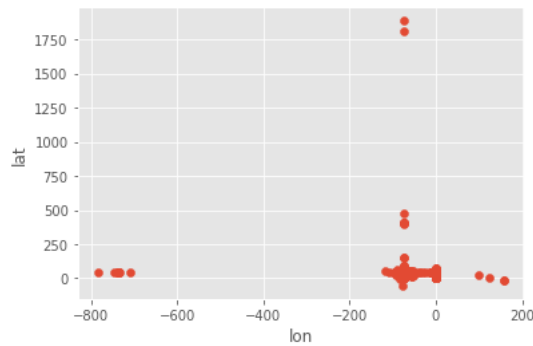
■ pickup原始未刪除錯誤經緯度的圖



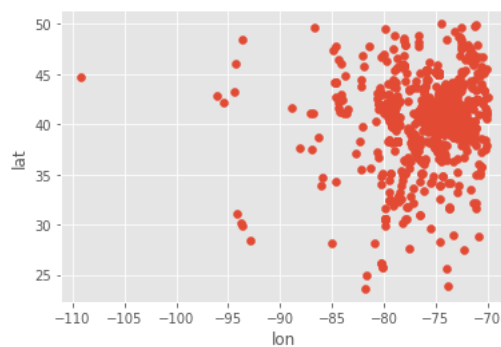
■ pickup 刪除後的圖



■ dropoff原始未刪除錯誤經緯度的圖



■ dropoff 刪除後的圖

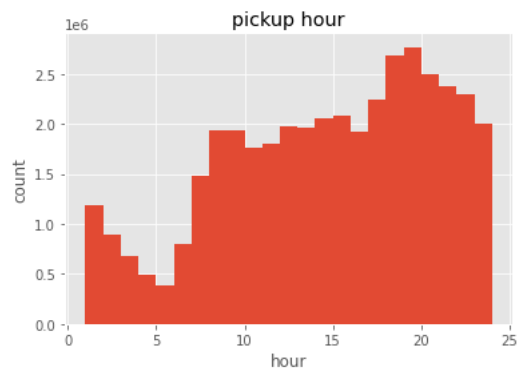


接著用kmeans取出10個cluster，並觀察其中心點，取得5個最多人上下車的區域

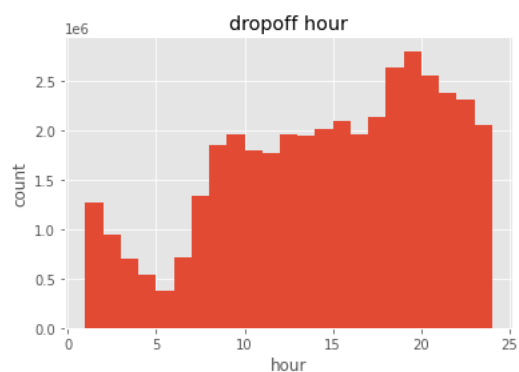
- Q2: When are the peak hours and off-peak hours for taking a taxi?
  - peak hours : 17:00~22:00
  - off-peak hours : 1:00~7:00

- how I solve this question : 取出pickup, dropoff時間中的小時資料，並繪製成直方圖，看哪些時段多人哪些時段少人

- pickup

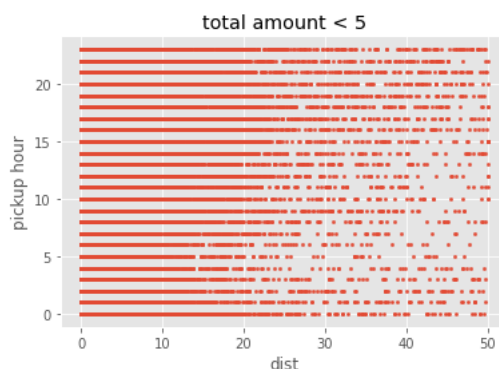


- dropoff

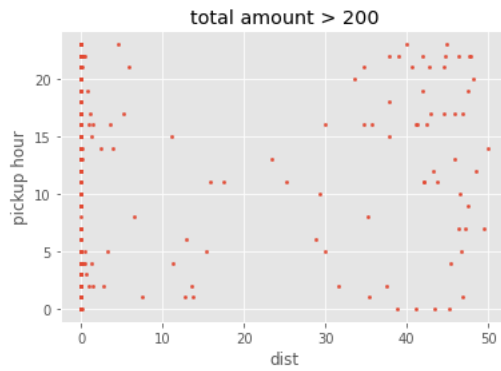


- Q3: What are the differences between big and small total amounts when taking a taxi?
  - 一開始先將small total amounts定義為5元以下，big total amounts定義為200元以上，並分別紀錄距離及pickup hour

- total amount < 5

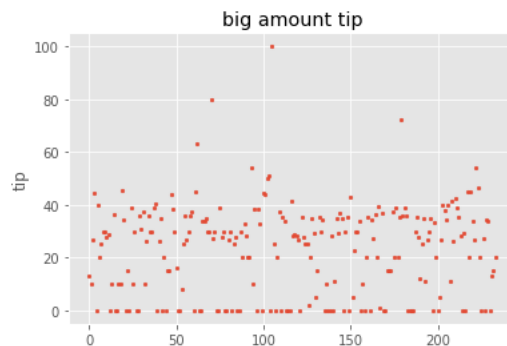


- total amount > 200



- 初始我把big/small amount分別定為100/50，但畫出來的圖很接近，可以看出基本大家的乘車費都在那之間，所以我進一步將其範圍縮小，我拿距離及上車時間作為判斷標準，因為搭計程車中最會影響費用的為距離，另一個我用上車時間做為標準，想看出是否在不同時間收費的標準會有不同，在定義small amount時我從50，縮減到10，再進一步縮減到5，但是數量都還是很大，在兩張圖中也看不出顯著差異；於是我進一步將total amount > 200所拿到的tip畫出來

- tip (total amount > 200)



- 車費超過200元的，在小費上都會給很多，光是小費就高於定義的small amount，可以看出，雖然距離相差不多，但車費相差多也有出自於客人給的小費比較多
- Difficulties Encountered
  - 一開始我使用pandas，並用geopy來獲取正確的地理位置，但會有runtime error的問題，後來改用pyspark，但因為不熟悉pyspark，導致出現很多error，需要查一些資料才能解決問題，但在使用kmeans時會有記憶體問題出現，所以最後還是用pandas，並用kmeans來取得最多上下車地點，雖然無法得出確切地址，但還是能獲得其中心經緯度
- source code

```

1 import pandas as pd
2 import numpy as np
3 from glob import glob
4 from sklearn.cluster import KMeans
5 import time
6 import matplotlib.pyplot as plt
7 from matplotlib import style
8 import time
9 style.use("ggplot")
10
11 files = glob('yellow_tripdata_2009*.csv')
12 print(files)
13 df = pd.concat((pd.read_csv(file) for file in files))
14
15 print(df.columns)
16 df.head(1)
17
18 # Q1
19 # pickup
20 sLon = df['Start_Lon']
21 sLat = df['Start_Lat']
22
23 plt.scatter(sLon, sLat)
24 plt.xlabel('lon')
25 plt.ylabel('lat')
26 plt.show()
27
28 # lon(-70~-130) lat(23.5~50)
29 # delete data that out of range
30 delIndex1 = np.append(np.where(sLon < -130), np.where(sLon > -70))
31 delIndex2 = np.append(np.where(sLat < 23.5), np.where(sLat > 50))
32 delIndex = np.append(delIndex1, delIndex2)
33
34 tpickupLoc = np.column_stack((sLon, sLat))
35 pickupLoc = np.delete(tpickupLoc, delIndex, axis=0)
36
37 plt.scatter(pickupLoc[:,0], pickupLoc[:,1])
38 plt.xlabel('lon')
39 plt.ylabel('lat')
40 plt.show()
41
42 # use kmeans
43 km_fit = KMeans(n_clusters=10).fit(pickupLoc)
44 cluster_centers = km_fit.cluster_centers_
45 print(cluster_centers)
46
47 plt.scatter(pickupLoc[:,0], pickupLoc[:,1], s=5)
48 plt.scatter(cluster_centers[:,0], cluster_centers[:,1], marker='*', c='b')
49 plt.xlabel('lon')
50 plt.ylabel('lat')
51 plt.show()
52
53 # dropoff
54 eLon = df['End_Lon']
55 eLat = df['End_Lat']
56
57 plt.scatter(eLon, eLat)
58 plt.xlabel('lon')
59 plt.ylabel('lat')
60 plt.show()
61
62 # lon(-70~-130) lat(23.5~50)
63 # delete data that out of range
64 delIndex1 = np.append(np.where(eLon < -130), np.where(eLon > -70))
65 delIndex2 = np.append(np.where(eLat < 23.5), np.where(eLat > 50))
66 delIndex = np.append(delIndex1, delIndex2)
67
68 tdropoffLoc = np.column_stack((eLon, eLat))
69 dropoffLoc = np.delete(tdropoffLoc, delIndex, axis=0)
70
71 plt.scatter(dropoffLoc[:,0], dropoffLoc[:,1])
72 plt.xlabel('lon')
73 plt.ylabel('lat')
74 plt.show()
75
76 # use kmeans
77 km_fit = KMeans(n_clusters=10).fit(dropoffLoc)
78 cluster_centers = km_fit.cluster_centers_
79 print(cluster_centers)
80
81 plt.scatter(dropoffLoc[:,0], dropoffLoc[:,1], s=5)
82 plt.scatter(cluster_centers[:,0], cluster_centers[:,1], marker='*', c='b')

```

```

83 plt.xlabel('lon')
84 plt.ylabel('lat')
85 plt.show()
86
87 # Q2
88 time_format = '%Y-%m-%d %H:%M:%S'
89
90 time_hour = []
91 for i in range(1, 25):
92     time_hour.append(i)
93 print(time_hour)
94
95 def get_hour(t):
96     h = time.strptime(t, time_format).tm_hour
97     return h
98
99 pickupTime = list(map(get_hour, df['Trip_Pickup_DateTime']))
100 dropoffTime = list(map(get_hour, df['Trip_Dropoff_DateTime']))
101
102 plt.hist(pickupTime, bins=time_hour)
103 plt.title('pickup hour')
104 plt.xlabel('hour')
105 plt.ylabel('count')
106 plt.show()
107
108 plt.hist(dropoffTime, bins=time_hour)
109 plt.title('dropoff hour')
110 plt.xlabel('hour')
111 plt.ylabel('count')
112 plt.show()
113
114 # Q3
115 t_totalAmt = df['Total_Amt']
116 t_dist = df['Trip_Distance']
117 totalAmt = t_totalAmt.to_numpy()
118 dist = t_dist.to_numpy()
119
120 small_x = []
121 small_y = []
122 for i in range(len(totalAmt)):
123     if totalAmt[i] < 5:
124         small_x.append(dist[i])
125         small_y.append(pickupTime[i])
126
127 plt.scatter(small_x, small_y, s=5)
128 plt.title('total amount < 5')
129 plt.xlabel('dist')
130 plt.ylabel('pickup hour')
131 plt.show()
132
133 big_x = []
134 big_y = []
135 for i in range(len(totalAmt)):
136     if totalAmt[i] > 200:
137         big_x.append(dist[i])
138         big_y.append(pickupTime[i])
139
140 plt.scatter(big_x, big_y, s=5)
141 plt.title('total amount > 200')
142 plt.xlabel('dist')
143 plt.ylabel('pickup hour')
144 plt.show()
145
146 t_tip = df['Tip_Amt']
147 tip = t_tip.to_numpy()
148 big_tip = []
149 tx = []
150
151 j = 0
152 for i in range(len(totalAmt)):
153     if totalAmt[i] > 200:
154         tx.append(j)
155         j += 1
156         big_tip.append(tip[i])
157
158 plt.scatter(tx, big_tip, s=7)
159 plt.title('big amount tip')
160 plt.ylabel('tip')
161 plt.show()

```