```
In [1]: import pandas as pd
  import numpy as np
  import matplotlib.pyplot as plt
  import seaborn as sns
  %matplotlib inline
```

Conclusion

By making use of the BigQuery and programming in Python, I summarized the finding in this cell. Due to the time limit, I did not comment the code below. The important tables and visualizations can be found in the following section.

- Top five regions with the most users in 2015 is India, Athens, Italy, Denmark and LA
- · Most of user profile does not contain the information of ages.
- The averge active duration(from creation to the last edit) is around 235 days, however, it has a
 large variations, popular posts can last much longer than normal ones.
- Through 12 months of 2015, people tend to produce less posts in January and Feburary(might due to holidays, more research needs).
- The score, view_count, favoriate_num are quite correalted with each other
- Surprisingly, the comments_num does not increase with number of score, view_count and favoriate_num(can be interpreted that a good post does not need too much explain)

Load and preprocess data retireved by BigQuery

Out[2]:

	userID	post_cnt	avg_far	avg_com	avg_ans	avg_scr	avg_view
(4304022	21	3.0	1.809524	1.571429	2.142857	1956.571429
	1 1531040	23	0.0	1.565217	1.470588	-0.043478	89.058824
	2 3395829	27	1.0	1.740741	1.250000	1.185185	392.125000
;	3 4235960	122	0.5	2.762295	1.681818	0.786885	209.136364
	4 4801826	9	1.0	0.222222	0.800000	0.000000	112.600000

In [9]: posts.describe()

Out[9]:

	userID	post_cnt	avg_far	avg_com	avg_ans	avg_scr	a١
count	1.600000e+04	16000.000000	16000.000000	16000.000000	16000.000000	16000.000000	16000
mean	3.104228e+06	22.366875	0.710942	1.919714	1.406654	0.724918	282
std	1.633933e+06	57.621019	1.160697	1.183553	0.866429	1.989294	992
min	1.363000e+03	1.000000	0.000000	0.000000	0.125000	-8.000000	12
25%	1.643535e+06	6.000000	0.000000	1.176471	0.928571	0.200000	87
50%	3.349431e+06	11.000000	0.666667	1.714286	1.272727	0.500000	145
75%	4.569840e+06	22.000000	1.000000	2.400000	1.625000	0.944444	262
max	6.306298e+06	2372.000000	37.500000	28.000000	13.000000	115.000000	62835

```
In [3]: locationinfo = location.groupby(['location']).count().reset_index()
    locationinfo = locationinfo.rename(columns={'userID':"user num"})
    locationinfo.sort_values(by=['user num'],ascending=False).head()
```

Out[3]:

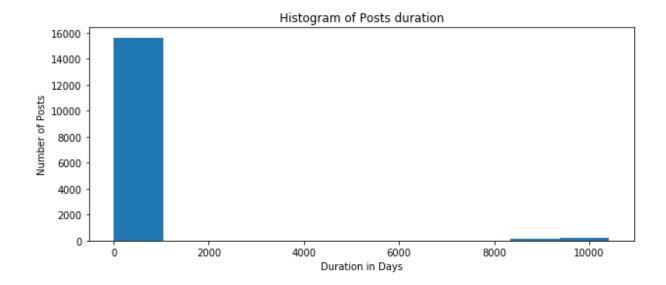
	location	user num
888	Pune, Maharashtra, India	271
51	Athens	233
520	Italy	222
317	Denmark	218
616	Los Angeles	197

Explore the active date of posts

Out[4]:

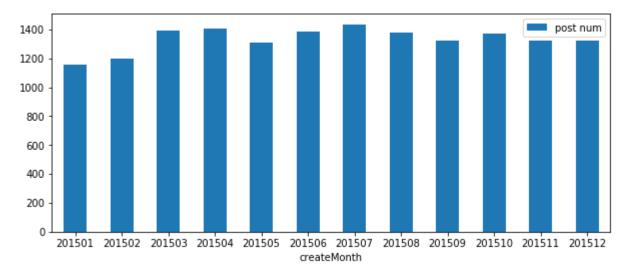
	creation_date	last_activity_date	createMonth	create	lastedit	dur
0	2015-12-01 08:55:53.743 UTC	2015-12-04 11:52:25.647 UTC	201512	20151201	20151204	3
1	2015-02-04 12:38:37.197 UTC	2015-02-04 12:38:37.197 UTC	201502	20150204	20150204	0
2	2015-12-02 22:50:38.273 UTC	2015-12-17 08:06:23.237 UTC	201512	20151202	20151217	15
3	2015-01-08 09:17:39.027 UTC	2015-01-10 20:15:07.260 UTC	201501	20150108	20150110	2
4	2015-08-18 05:42:56.653 UTC	2015-08-18 09:31:58.830 UTC	201508	20150818	20150818	0

```
dates['dur'].plot(kind='hist', figsize=(10, 4))
In [5]:
         count, bin edges = np.histogram(dates['dur'], 30)
         print(count)
        print(bin edges)
        plt.title('Histogram of Posts duration') # add a title to the histogram
        plt.ylabel('Number of Posts') # add y-label
        plt.xlabel('Duration in Days') # add x-Label
        plt.show()
        print('The average duration date:', dates['dur'].mean())
        [15458
                  107
                         54
                                6
                                       0
                                             0
                                                   0
                                                         0
                                                                0
                                                                      0
                                                                            0
                                                                                   0
              0
                    0
                          0
                                0
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                                             0
                                                   0
                                                          0
                                                                0
                                                                      0
                                                                            0
                                                                                   0
                                            491
              0
                   44
                         86
                              114
                                      82
              0.
                    347.3
                            694.6
                                   1041.9
                                            1389.2
                                                    1736.5
                                                             2083.8
                                                                     2431.1
                                                                             2778.4
           3125.7
                                   4167.6
                                            4514.9
                                                    4862.2
                                                                             5904.1
                   3473.
                           3820.3
                                                             5209.5
                                                                     5556.8
                                    7293.3
                                                    7987.9
                                                            8335.2
                                                                     8682.5
                                                                             9029.8
           6251.4
                   6598.7
                           6946.
                                            7640.6
          9377.1
                  9724.4 10071.7 10419. ]
```



The average duration date: 234.7088125

```
In [6]: monthposts = dates.groupby(['createMonth']).count().reset_index()
    monthposts = monthposts.rename(columns={'creation_date':"post num"})
    ax = monthposts.plot.bar(x='createMonth', y='post num', rot=0, figsize=(10, 4))
```

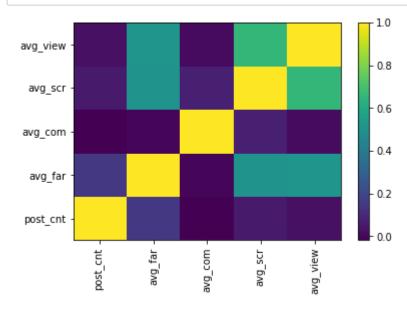


Correlation and Heatmap

Out[7]:

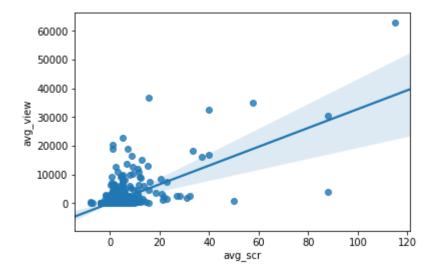
	post_cnt	avg_far	avg_com	avg_scr	avg_view
post_cnt	1.000000	0.144992	-0.019277	0.052184	0.024247
avg_far	0.144992	1.000000	-0.004022	0.502795	0.514834
avg_com	-0.019277	-0.004022	1.000000	0.070266	0.011941
avg_scr	0.052184	0.502795	0.070266	1.000000	0.656394
avg_view	0.024247	0.514834	0.011941	0.656394	1.000000

```
In [8]: grouped_pivot = correlation
        fig, ax = plt.subplots()
        im = ax.pcolor(grouped pivot)
        #label names
        row_labels = grouped_pivot.columns
        col_labels = grouped_pivot.index
        #move ticks and labels to the center
        ax.set_xticks(np.arange(grouped_pivot.shape[1]) + 0.5, minor=False)
        ax.set_yticks(np.arange(grouped_pivot.shape[0]) + 0.5, minor=False)
        #insert labels
        ax.set_xticklabels(row_labels, minor=False)
        ax.set_yticklabels(col_labels, minor=False)
        #rotate label if too long
        plt.xticks(rotation=90)
        fig.colorbar(im)
        plt.show()
```



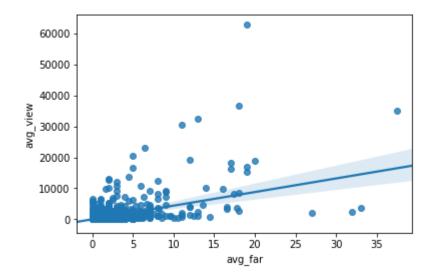
```
In [11]: sns.regplot(x="avg_scr", y="avg_view", data=posts)
```

Out[11]: <matplotlib.axes._subplots.AxesSubplot at 0x2577bb889e8>





Out[12]: <matplotlib.axes._subplots.AxesSubplot at 0x2577afdada0>



In [13]: sns.regplot(x="avg_com", y="avg_view", data=posts)

Out[13]: <matplotlib.axes._subplots.AxesSubplot at 0x25779a67080>

