0-Outbrain-Data Visualizations

January 21, 2022

0.1 Data visualization

0.1.1 importing the data visualization libraries

```
[1]: import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
import os
import gc # We're gonna be clearing memory a lot
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline

p = sns.color_palette()

# our bucket directory
DATA_BUCKET_FOLDER = "gs://akhilbucket/data/"
```

```
[2]: df_train = pd.read_csv(DATA_BUCKET_FOLDER+'clicks_train.csv')
df_test = pd.read_csv(DATA_BUCKET_FOLDER+'clicks_test.csv')
```

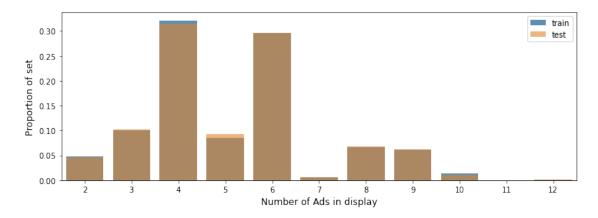
/opt/conda/miniconda3/lib/python3.8/site-packages/seaborn/_decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

```
warnings.warn(
```

/opt/conda/miniconda3/lib/python3.8/site-packages/seaborn/_decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

warnings.warn(

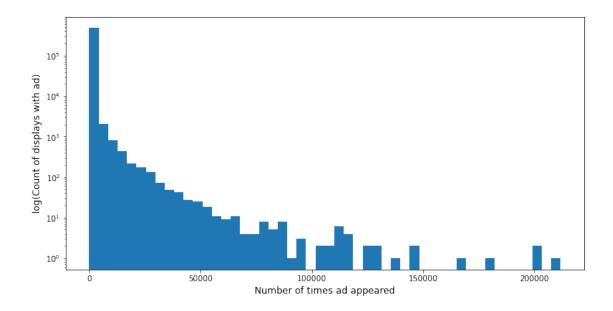
[3]: Text(0, 0.5, 'Proportion of set')



This looks like a perfect split to me! So we can assume the distribution to be the same within the sets - no weird trickery going on here.

What about adverts? How many adverts are there that are very often used, and how many are rare?

Ads that appear less than 2 times: 22.69% Ads that appear less than 10 times: 61.74% Ads that appear less than 50 times: 86.93% Ads that appear less than 100 times: 91.04% Ads that appear less than 1000 times: 97.45%



Here we can see that a huge number of ads appear just a few times in the training set (so much that we have to use a log graph to show it), with two-thirds having less than 10 appearances. This shows us that we have to be able to predict whether someone will click on an ad not just based on the past data of that specific ad, but also by linking it to other adverts.

I think this is the underlying challenge that Outbrain has for us.

Before we move on, let's check how many ads in the test set are not in the training set.

```
[5]: ad_prop = len(set(df_test.ad_id.unique()).intersection(df_train.ad_id.

→unique())) / len(df_test.ad_id.unique())

print('Proportion of test ads in test that are in training: {}%'.

→format(round(ad_prop * 100, 2)))
```

Proportion of test ads in test that are in training: 82.87%

1 Events

Let's move on to the events file. I'm not going to cover the timestamp or the location, since these have already been beautifully explored - you can click the links to view those EDAs.

```
[8]: events = pd.read_csv(DATA_BUCKET_FOLDER+'events.csv')
    print('Shape:', events.shape)
    print('Columns', events.columns.tolist())
    events.head()
```

```
/opt/conda/miniconda3/lib/python3.8/site-
packages/IPython/core/interactiveshell.py:3185: DtypeWarning: Columns (4) have
mixed types.Specify dtype option on import or set low_memory=False.
has_raised = await self.run_ast_nodes(code_ast.body, cell_name,
```

```
Shape: (23120126, 6)
Columns ['display_id', 'uuid', 'document_id', 'timestamp', 'platform',
'geo_location']
```

```
[8]:
        display_id
                                    document_id timestamp platform geo_location
                              uuid
     0
                                                                    3
                 1
                   cb8c55702adb93
                                          379743
                                                         61
                                                                         US>SC>519
                                                         81
     1
                 2 79a85fa78311b9
                                         1794259
                                                                    2
                                                                         US>CA>807
     2
                 3 822932ce3d8757
                                                        182
                                                                    2
                                                                         US>MI>505
                                         1179111
     3
                 4 85281d0a49f7ac
                                         1777797
                                                        234
                                                                    2
                                                                         US>WV>564
     4
                   8d0daef4bf5b56
                                                                    2
                                                                             SG>00
                                          252458
                                                        338
```

```
[]: plat = events.platform.value_counts()

print(plat)
print('\nUnique values of platform:', events.platform.unique())
```

This is very interesting, notice how 1, 2 and 3 are repeated twice in the platform, once as floats, and once as strings.

This might become useful in the future (leak? :P) - but for now, I'm just going to treat them as the same thing.

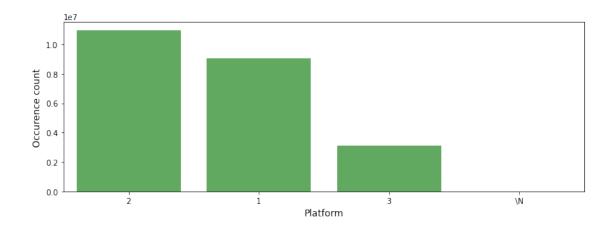
```
[9]: events.platform = events.platform.astype(str)
plat = events.platform.value_counts()

plt.figure(figsize=(12,4))
sns.barplot(plat.index, plat.values, alpha=0.8, color=p[2])
plt.xlabel('Platform', fontsize=12)
plt.ylabel('Occurence count', fontsize=12)
```

/opt/conda/miniconda3/lib/python3.8/site-packages/seaborn/_decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

[9]: Text(0, 0.5, 'Occurence count')

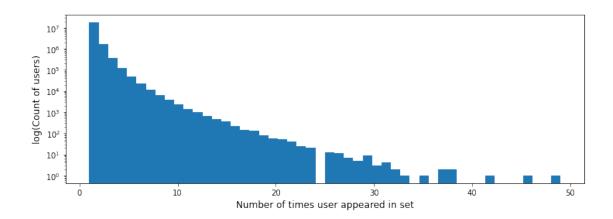
warnings.warn(



It's still unclear what the platform means, but it's possible that it's things like computers, phones, tablets etc.

The \Ns and string numbers may have come from the way that a file was parsed while creating the dataset, but it's still a mystery. If anyone has any other ideas, I'd love to hear them!

Let's do some quick analysis on the UUID next.



Here we see a distribution much like the ad ids, with 88% of users being unique - there will be little scope of building user-based recommendation profiles here.

I'd love to look at things like whether the same user ever clicks on the same ad twice, or whether a user transverses the training & testing set, but we're limited by the Kernels memory limit here:

2 Categories

Outbrain has some content classification algorithms, and they have provided us with the output from these classifications. Let's take a look at some of the most popular classifications.

```
[12]: topics = pd.read_csv(DATA_BUCKET_FOLDER+'documents_topics.csv')
    print('Columns:',topics.columns.tolist())
    print('Number of unique topics:', len(topics.topic_id.unique()))
    topics.head()
```

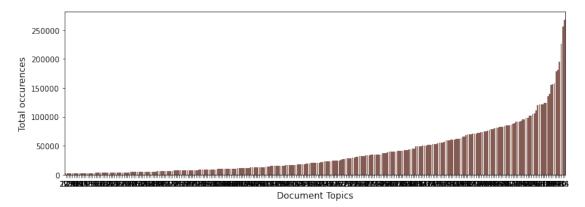
Columns: ['document_id', 'topic_id', 'confidence_level'] Number of unique topics: 300

```
[12]:
         document_id topic_id confidence_level
              1595802
                             140
                                           0.073113
      0
      1
              1595802
                              16
                                           0.059416
      2
              1595802
                             143
                                           0.045421
      3
             1595802
                             170
                                           0.038867
      4
              1524246
                             113
                                           0.196450
```

```
Number of topics that appear more than 10000 times: 201 Number of topics that appear more than 50000 times: 86 Number of topics that appear more than 100000 times: 22 Number of topics that appear more than 200000 times: 3
```

/opt/conda/miniconda3/lib/python3.8/site-packages/seaborn/_decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

warnings.warn(



plt.show()

```
Columns: ['document_id', 'category_id', 'confidence_level']
Number of unique categories: 97
Number of categories that appear more than 1000 times: 88
Number of categories that appear more than 10000 times: 65
Number of categories that appear more than 50000 times: 35
Number of categories that appear more than 100000 times: 15
```

/opt/conda/miniconda3/lib/python3.8/site-packages/seaborn/_decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

warnings.warn(

