# **Project Objective:**

Using user data, 'takehome\_users.csv', and user engagement data, 'takehome\_user\_engagement.csv', find the predictors of 'Adopted User'.

'Adopted User' is defined as one who has logged into the product on three separate days in at least one sevenday period.

```
In [1]: # import relevant libraries and packages
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

%matplotlib inline
sns.set(style = 'whitegrid', font_scale = 1.8)

# ignore warnings
import warnings
warnings.filterwarnings('ignore')
```

```
In [2]: # check the files encoding

import chardet

rawdata = open('takehome_users.csv', 'rb').read()

result = chardet.detect(rawdata)
    charenc = result['encoding']
    print(charenc)
```

ISO-8859-1

Load user data

```
In [3]: | df 1 = pd.read csv('takehome users.csv', encoding = 'ISO-8859-1')
        df 1.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 12000 entries, 0 to 11999
        Data columns (total 10 columns):
             Column
                                         Non-Null Count Dtype
             _____
                                                        ----
        - - -
                                         -----
         0
             object id
                                         12000 non-null int64
         1
             creation time
                                         12000 non-null object
         2
             name
                                         12000 non-null object
         3
             email
                                         12000 non-null object
         4
             creation source
                                         12000 non-null object
         5
             last session creation time 8823 non-null
                                                        float64
             opted in to mailing list
                                         12000 non-null int64
         6
         7
             enabled_for_marketing_drip 12000 non-null int64
         8
             org id
                                         12000 non-null int64
         9
             invited_by_user_id
                                         6417 non-null
                                                         float64
        dtypes: float64(2), int64(4), object(4)
        memory usage: 937.6+ KB
```

#### Load user engagement data

```
In [4]: | df 2 = pd.read csv('takehome user engagement.csv', encoding = 'ISO-8859-1')
        df_2.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 207917 entries, 0 to 207916
        Data columns (total 3 columns):
             Column
                        Non-Null Count
                                         Dtype
             -----
                         -----
                                         ----
            time stamp 207917 non-null object
         0
             user id
                        207917 non-null int64
         2
             visited
                        207917 non-null int64
        dtypes: int64(2), object(1)
        memory usage: 4.8+ MB
```

```
In [5]:
          # check df_1
          df_1.head()
Out[5]:
              object_id creation_time
                                          name
                                                                             creation_source last_session_c
                            2014-04-22
                                        Clausen
           0
                      1
                                                 AugustCClausen@yahoo.com
                                                                              GUEST INVITE
                                                                                                           1
                              03:53:30
                                         August
                            2013-11-15
                                          Poole
                      2
                                                     MatthewPoole@gustr.com
                                                                                 ORG INVITE
                                                                                                           1
                              03:45:04
                                       Matthew
                           2013-03-19
                                          Bottrill
           2
                      3
                                                     MitchellBottrill@gustr.com
                                                                                 ORG_INVITE
                                                                                                           1
                              23:14:52
                                        Mitchell
                            2013-05-21
                                        Clausen
           3
                                                 NicklasSClausen@yahoo.com
                                                                              GUEST_INVITE
                                                                                                           1
                              08:09:28
                                        Nicklas
                            2013-01-17
                                           Raw
                      5
                                                       GraceRaw@yahoo.com
                                                                              GUEST INVITE
                              10:14:20
                                          Grace
In [6]:
          # check df_2
          df_2.head()
Out[6]:
                                  user_id visited
                     time_stamp
              2014-04-22 03:53:30
                                        1
                                                 1
              2013-11-15 03:45:04
                                        2
                                                 1
              2013-11-29 03:45:04
                                        2
              2013-12-09 03:45:04
                                        2
                                                 1
              2013-12-25 03:45:04
                                        2
                                                 1
```

Process user engagement data to find the 'adopted users'.

## Out[7]:

	time_stamp	user_id	visited	date
0	2014-04-22 03:53:30	1	1	2014-04-22 03:53:30
1	2013-11-15 03:45:04	2	1	2013-11-15 03:45:04
2	2013-11-29 03:45:04	2	1	2013-11-29 03:45:04
3	2013-12-09 03:45:04	2	1	2013-12-09 03:45:04
4	2013-12-25 03:45:04	2	1	2013-12-25 03:45:04
5	2013-12-31 03:45:04	2	1	2013-12-31 03:45:04
6	2014-01-08 03:45:04	2	1	2014-01-08 03:45:04
7	2014-02-03 03:45:04	2	1	2014-02-03 03:45:04
8	2014-02-08 03:45:04	2	1	2014-02-08 03:45:04
9	2014-02-09 03:45:04	2	1	2014-02-09 03:45:04
10	2014-02-13 03:45:04	2	1	2014-02-13 03:45:04
11	2014-02-16 03:45:04	2	1	2014-02-16 03:45:04
12	2014-03-09 03:45:04	2	1	2014-03-09 03:45:04
13	2014-03-13 03:45:04	2	1	2014-03-13 03:45:04
14	2014-03-31 03:45:04	2	1	2014-03-31 03:45:04
15	2013-03-19 23:14:52	3	1	2013-03-19 23:14:52
16	2013-05-22 08:09:28	4	1	2013-05-22 08:09:28
17	2013-01-22 10:14:20	5	1	2013-01-22 10:14:20
18	2013-12-19 03:37:06	6	1	2013-12-19 03:37:06
19	2012-12-20 13:24:32	7	1	2012-12-20 13:24:32

```
In [8]: # create new coulum 'delta_t_weeks'
df_2['delta_t_weeks'] = 0
df_2.head(20)
```

## Out[8]:

	time_stamp	user_id	visited	date	delta_t_weeks
0	2014-04-22 03:53:30	1	1	2014-04-22 03:53:30	0
1	2013-11-15 03:45:04	2	1	2013-11-15 03:45:04	0
2	2013-11-29 03:45:04	2	1	2013-11-29 03:45:04	0
3	2013-12-09 03:45:04	2	1	2013-12-09 03:45:04	0
4	2013-12-25 03:45:04	2	1	2013-12-25 03:45:04	0
5	2013-12-31 03:45:04	2	1	2013-12-31 03:45:04	0
6	2014-01-08 03:45:04	2	1	2014-01-08 03:45:04	0
7	2014-02-03 03:45:04	2	1	2014-02-03 03:45:04	0
8	2014-02-08 03:45:04	2	1	2014-02-08 03:45:04	0
9	2014-02-09 03:45:04	2	1	2014-02-09 03:45:04	0
10	2014-02-13 03:45:04	2	1	2014-02-13 03:45:04	0
11	2014-02-16 03:45:04	2	1	2014-02-16 03:45:04	0
12	2014-03-09 03:45:04	2	1	2014-03-09 03:45:04	0
13	2014-03-13 03:45:04	2	1	2014-03-13 03:45:04	0
14	2014-03-31 03:45:04	2	1	2014-03-31 03:45:04	0
15	2013-03-19 23:14:52	3	1	2013-03-19 23:14:52	0
16	2013-05-22 08:09:28	4	1	2013-05-22 08:09:28	0
17	2013-01-22 10:14:20	5	1	2013-01-22 10:14:20	0
18	2013-12-19 03:37:06	6	1	2013-12-19 03:37:06	0
19	2012-12-20 13:24:32	7	1	2012-12-20 13:24:32	0

```
In [9]: # get list of users who have logged at least 3 times in 7 days - 'adopted' use
         lst user = df 2['user id'].value counts().index
         t 0 = 0
         lst_user_a = []
         log max = 0
         for user in 1st user:
             df_0 = df_2[df_2['user_id'] == user].copy()
             df 0.reset index(drop = True, inplace = True)
             t_0 = df_0['date'].iloc[0]
             for i in range(len(df 0)):
                 df_0['delta_t_weeks'].iloc[i] = (df_0['date'].iloc[i] - t_0).days // 7
             log_max = df_0['delta_t_weeks'].value_counts().values.max()
             if log_max >= 3:
                 lst user a.append(user)
In [10]: | # print number of all users and of adopted users
         print('All users: ', len(lst user))
         print('Adopted users: ', len(lst_user_a))
         All users: 8823
         Adopted users: 1472
In [11]: | # check for user id = 1
         1 in 1st user a
Out[11]: False
In [12]: # check for user_id = 2
         2 in lst_user_a
Out[12]: True
In [13]: # checks out
```

Use list of 'adopted users' to separate user data in 'Adopted' and 'Non-Adopted' users.

```
In [14]:
         mask user a = df 1['object id'].isin(lst user a)
         # adopted users
         df_1a = df_1[mask_user_a].copy()
         df 1a.reset index(inplace = True, drop = True)
         df 1a.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 1472 entries, 0 to 1471
         Data columns (total 10 columns):
              Column
                                          Non-Null Count Dtype
         - - -
              -----
                                           -----
                                                          ----
          0
                                                           int64
              object id
                                          1472 non-null
          1
              creation time
                                          1472 non-null
                                                          object
          2
                                          1472 non-null
                                                          object
              name
          3
              email
                                          1472 non-null
                                                          object
          4
              creation_source
                                          1472 non-null
                                                           object
          5
              last session creation time 1472 non-null
                                                          float64
          6
              opted in to mailing list
                                          1472 non-null
                                                           int64
          7
              enabled_for_marketing_drip 1472 non-null
                                                           int64
          8
              org id
                                          1472 non-null
                                                           int64
          9
              invited_by_user_id
                                          836 non-null
                                                           float64
         dtypes: float64(2), int64(4), object(4)
         memory usage: 115.1+ KB
         # non-adopted users
In [15]:
         df 1b = df 1[\sim mask user a].copy()
         df 1b.reset index(inplace = True, drop = True)
         df 1b.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 10528 entries, 0 to 10527
         Data columns (total 10 columns):
          #
              Column
                                          Non-Null Count Dtype
              ____
                                           _____
          0
              object id
                                          10528 non-null int64
                                          10528 non-null object
          1
              creation time
          2
                                          10528 non-null object
              name
          3
              email
                                          10528 non-null object
          4
              creation source
                                          10528 non-null object
          5
              last session creation time 7351 non-null
                                                          float64
          6
              opted in to mailing list
                                          10528 non-null int64
          7
              enabled for marketing drip
                                          10528 non-null int64
          8
              org id
                                          10528 non-null int64
              invited_by_user_id
                                          5581 non-null
                                                          float64
         dtypes: float64(2), int64(4), object(4)
         memory usage: 822.6+ KB
```

```
In [16]: # adopted users
df_1a.head()
```

#### Out[16]:

	object_id	creation_time	name	email	creation_source	last_sessioı
0	2	2013-11-15 03:45:04	Poole Matthew	MatthewPoole@gustr.com	ORG_INVITE	
1	10	2013-01-16 22:08:03	Santos Carla	CarlaFerreiraSantos@gustr.com	ORG_INVITE	
2	20	2014-03-06 11:46:38	Helms Mikayla	lqyvjilf@uhzdq.com	SIGNUP	
3	42	2012-11-11 19:05:07	Pinto Giovanna	GiovannaCunhaPinto@cuvox.de	SIGNUP	
4	43	2012-12-31 07:13:17	Gomes Beatrice	vgqmpljq@yyyxt.com	GUEST_INVITE	
4						<b>&gt;</b>

In [17]: # non-adopted users
df\_1b.head()

### Out[17]:

	object_id	creation_time	name	email	creation_source	last_sessi
0	1	2014-04-22 03:53:30	Clausen August	AugustCClausen@yahoo.com	GUEST_INVITE	
1	3	2013-03-19 23:14:52	Bottrill Mitchell	MitchellBottrill@gustr.com	ORG_INVITE	
2	4	2013-05-21 08:09:28	Clausen Nicklas	NicklasSClausen@yahoo.com	GUEST_INVITE	
3	5	2013-01-17 10:14:20	Raw Grace	GraceRaw@yahoo.com	GUEST_INVITE	
4	6	2013-12-17 03:37:06	Cunha Eduardo	EduardoPereiraCunha@yahoo.com	GUEST_INVITE	
4						+

Compare the composition of key features - 'creation\_source', 'opted\_in\_to\_mailing\_list', and 'enabled\_for\_marketing\_drip' - in 'Adopted' and 'Non-Adopted' users.

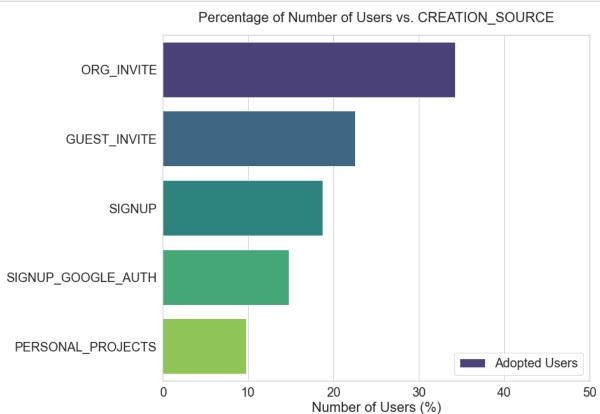
```
In [18]: # define a function to plot the value count of a feature from a dataset
         data feat = 'creation source'
         data_name = 'Adopted Users' # data_name will be used as a legend in the plot
         count feat = 0
         n pts = 25 # number of values to plot from the value count
         fig size x = 12
         fig size y = 10
         legend_box_x = 1.0
         legend box y = 0.1
         x max = 100
         d x = 10
         # define function
         def plot feat count(count feat, data feat, data name, n pts, fig size x, fig s
         ize_y, x_max, d_x, legend_box_x, legend_box_y):
             plt.figure(figsize = (fig size x, fig size y))
             sns.barplot(x = count_feat.iloc[0:plt_pts].values, y = count_feat.iloc[0:p
         lt_pts].index, palette = 'viridis', label = data_name)
             plt.xlim(0, x max)
             plt.xticks(np.arange(0, x_max + d_x , d_x))
             plt.xlabel('Number of Users (%)')
             plt.title('Percentage of Number of Users vs. ' + data feat.upper(), pad =
         20)
             plt.legend(bbox_to_anchor = (legend_box_x, legend_box_y))
             plt.show()
```

#### Examine 'creation\_source'

```
In [19]: # get and print feature value count for data a
         data = df 1a
         data feat = 'creation source'
         count feat = data[data feat].value counts(normalize = True) * 100
         print(count feat)
         print('\n')
         print(data_feat + ' - total number of values: ', len(count_feat))
         ORG INVITE
                               34.239130
         GUEST INVITE
                               22.554348
         SIGNUP
                               18.682065
         SIGNUP_GOOGLE_AUTH
                               14.741848
                               9.782609
         PERSONAL PROJECTS
         Name: creation_source, dtype: float64
         creation_source - total number of values: 5
```

```
In [20]: # plot

data_name = 'Adopted Users'
plt_pts = len(count_feat)
x_max = 50
d_x = 10
plot_feat_count(count_feat, data_feat, data_name, plt_pts, fig_size_x, fig_size_y, x_max, d_x, legend_box_x, legend_box_y)
```



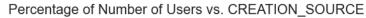
```
In [21]: # get and print feature value count for data_b
         data = df 1b
         count feat = data[data feat].value counts(normalize = True) * 100
         print(count_feat)
         print('\n')
         print(data_feat + ' - total number of values: ', len(count_feat))
         ORG INVITE
                               35.619301
         PERSONAL_PROJECTS
                               18.683511
         GUEST INVITE
                               17.391717
         SIGNUP
                               17.211246
         SIGNUP_GOOGLE_AUTH
                               11.094225
         Name: creation_source, dtype: float64
```

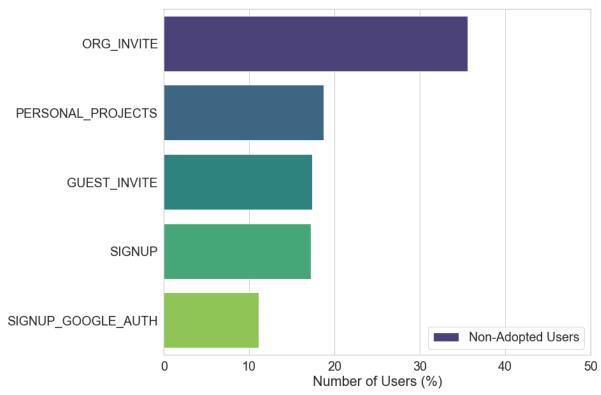
creation\_source - total number of values: 5

```
In [22]: # plot

data_name = 'Non-Adopted Users'
plt_pts = len(count_feat)

plot_feat_count(count_feat, data_feat, data_name, plt_pts, fig_size_x, fig_size_y, x_max, d_x, legend_box_x, legend_box_y)
```





#### Examine 'opted\_in\_to\_mailing\_list'

```
In [23]: # get and print feature value count for data_a

data = df_1a
    data_feat = 'opted_in_to_mailing_list'
    count_feat = data[data_feat].value_counts(normalize = True) * 100

print(count_feat)
    print('\n')
    print(data_feat + ' - total number of values: ', len(count_feat))

0    73.845109
1    26.154891
Name: opted_in_to_mailing_list, dtype: float64

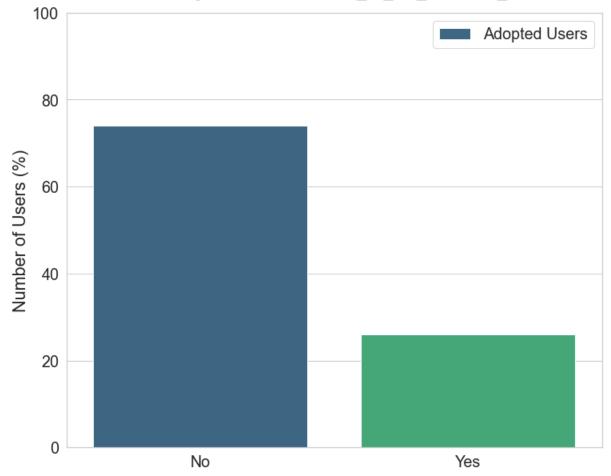
opted_in_to_mailing_list - total number of values: 2
```

```
In [24]: x_plt = ['No', 'Yes']
y_plt = [74, 26]

data_name = 'Adopted Users'

plt.figure(figsize = (fig_size_x, fig_size_y))
sns.barplot(x_plt, y_plt, palette = 'viridis', label = data_name)
plt.ylim(0, 100)
#plt.xticks(np.arange(0, x_max + d_x , d_x))
plt.ylabel('Number of Users (%)')
plt.title('Percentage of Users - ' + data_feat.upper(), pad = 20)
plt.legend()
plt.show()
```

## Percentage of Users - OPTED\_IN\_TO\_MAILING\_LIST



```
In [25]: # get and print feature value count for data_ab

data = df_1b

count_feat = data[data_feat].value_counts(normalize = True) * 100

print(count_feat)
print('\n')
print(data_feat + ' - total number of values: ', len(count_feat))

0    75.218465
1    24.781535
Name: opted_in_to_mailing_list, dtype: float64

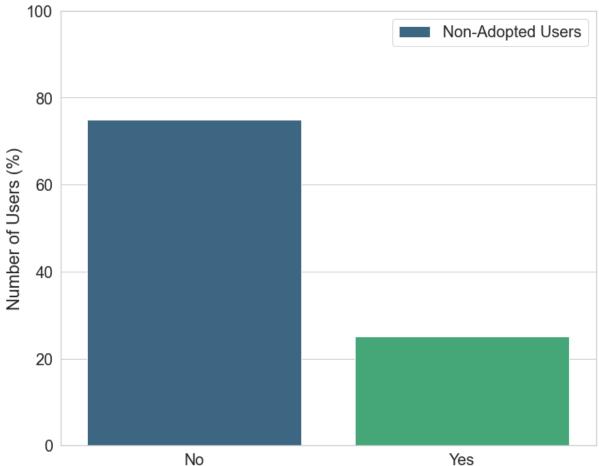
opted_in_to_mailing_list - total number of values: 2
```

```
In [26]: x_plt = ['No', 'Yes']
y_plt = [75, 25]

data_name = 'Non-Adopted Users'

plt.figure(figsize = (fig_size_x, fig_size_y))
sns.barplot(x_plt, y_plt, palette = 'viridis', label = data_name)
plt.ylim(0, 100)
#plt.xticks(np.arange(0, x_max + d_x , d_x))
plt.ylabel('Number of Users (%)')
plt.title('Percentage of Users - ' + data_feat.upper(), pad = 20)
plt.legend()
plt.show()
```





Examine 'enabled\_for\_marketing\_drip'

```
In [27]: # get and print feature value count for data_a

data = df_1a
    data_feat = 'enabled_for_marketing_drip'
    count_feat = data[data_feat].value_counts(normalize = True) * 100

print(count_feat)
    print('\n')
    print(data_feat + ' - total number of values: ', len(count_feat))

0    84.578804
1    15.421196
Name: enabled_for_marketing_drip, dtype: float64

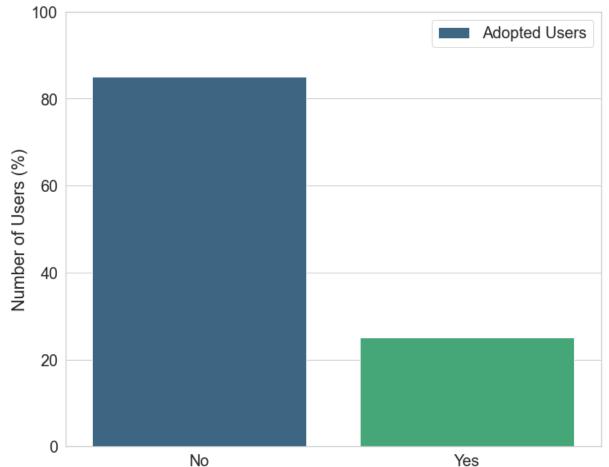
enabled_for_marketing_drip - total number of values: 2
```

```
In [28]: x_plt = ['No', 'Yes']
y_plt = [85, 25]

data_name = 'Adopted Users'

plt.figure(figsize = (fig_size_x, fig_size_y))
sns.barplot(x_plt, y_plt, palette = 'viridis', label = data_name)
plt.ylim(0, 100)
#plt.xticks(np.arange(0, x_max + d_x , d_x))
plt.ylabel('Number of Users (%)')
plt.title('Percentage of Users - ' + data_feat.upper(), pad = 20)
plt.legend()
plt.show()
```

# Percentage of Users - ENABLED\_FOR\_MARKETING\_DRIP



```
In [29]: # get and print feature value count for data_b

data = df_1b

count_feat = data[data_feat].value_counts(normalize = True) * 100

print(count_feat)
print('\n')
print(data_feat + ' - total number of values: ', len(count_feat))

0    85.134878
1    14.865122
Name: enabled_for_marketing_drip, dtype: float64

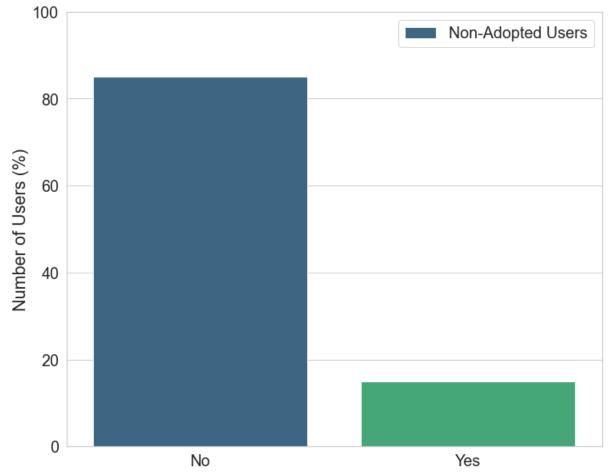
enabled_for_marketing_drip - total number of values: 2
```

```
In [30]: x_plt = ['No', 'Yes']
y_plt = [85, 15]

data_name = 'Non-Adopted Users'

plt.figure(figsize = (fig_size_x, fig_size_y))
sns.barplot(x_plt, y_plt, palette = 'viridis', label = data_name)
plt.ylim(0, 100)
#plt.xticks(np.arange(0, x_max + d_x , d_x))
plt.ylabel('Number of Users (%)')
plt.title('Percentage of Users - ' + data_feat.upper(), pad = 20)
plt.legend()
plt.show()
```

# Percentage of Users - ENABLED\_FOR\_MARKETING\_DRIP



Examine 'org\_id'

```
In [31]: # get and print feature value count for data a
         data = df 1a
         data feat = 'org id'
         count_feat = data[data_feat].value_counts(normalize = True) * 100
         n pts = 10
         print(count_feat[0:n_pts])
         print('\n')
         print(data feat + ' - total number of values: ', len(count feat))
         7
               1.086957
         2
               0.951087
         3
               0.951087
         4
               0.951087
         9
               0.951087
         1
               0.883152
         13
               0.815217
               0.747283
         a
               0.747283
         62
               0.679348
         Name: org_id, dtype: float64
         org_id - total number of values: 398
In [32]: # get and print feature value count for data_a
         data = df 1b
         count_feat = data[data_feat].value_counts(normalize = True) * 100
         print(count_feat[0:n_pts])
         print('\n')
         print(data_feat + ' - total number of values: ', len(count_feat))
         0
               2.925532
         1
               2.089666
         2
               1.776216
         3
               1.462766
               1.377280
         4
               1.234802
               1.120821
         9
               1.044833
         7
               0.978343
         10
               0.911854
         Name: org_id, dtype: float64
         org_id - total number of values: 417
```

# Findings:

From the features which were examined there are no true predictors for 'Adopted' vs. 'Non-Adopted' users.

There is a need for collecting more user data with new features in order to acomplish our objective.