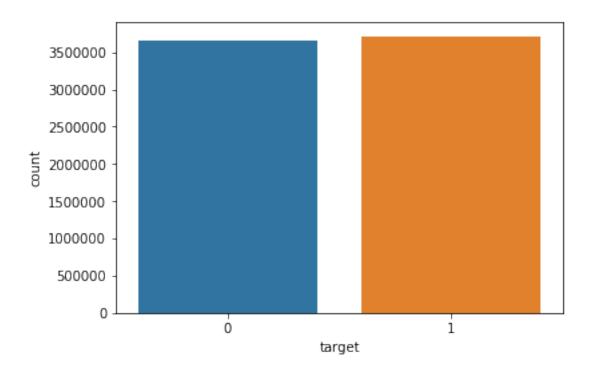
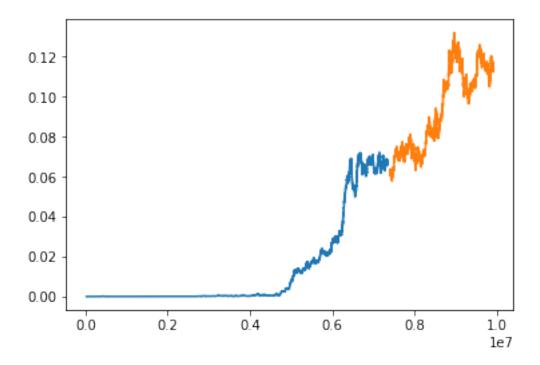
## EDA

June 3, 2020

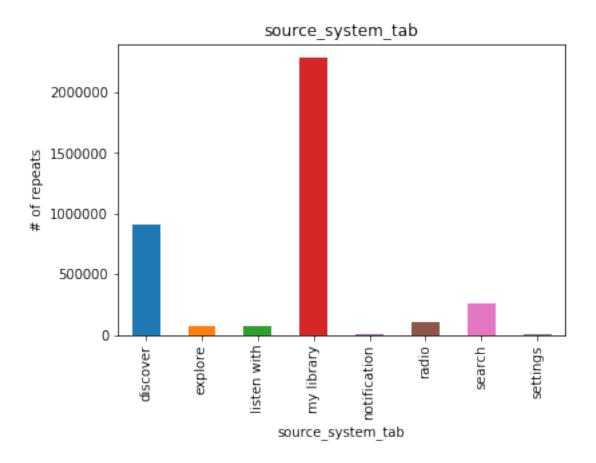
```
Explantory Data Analysis (EDA)
In [1]: import numpy as np
                        import pandas as pd
                        import seaborn as sns
                        import matplotlib.pyplot as plt
                        # load data
                        train = pd.read_csv('train.csv')
                        test = pd.read_csv('test.csv')
                        songs = pd.read_csv('songs.csv')
                        members = pd.read_csv('members.csv',parse_dates=["registration_init_time","expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expiration_expirati
                        songs_extra_info = pd.read_csv('song_extra_info.csv')
0.1 songs_extra_info
In [2]: def isrc_to_year(isrc):
                                    if type(isrc) == str:
                                                if int(isrc[5:7]) > 17:
                                                            return 1900 + int(isrc[5:7])
                                                else:
                                                            return 2000 + int(isrc[5:7])
                                    else:
                                                return np.nan
                        songs_extra_info['song_year'] = songs_extra_info['isrc'].apply(isrc_to_year)
In [3]: songs_extra_info = songs_extra_info[["song_id","song_year"]]
                        train = train.merge(songs_extra_info, on = 'song_id', how = 'left')
                        test = test.merge(songs_extra_info, on = 'song_id', how = 'left')
0.2 train V.S. test
In [4]: plt.figure()
                        sns.countplot(train['target'])
Out[4]: <matplotlib.axes._subplots.AxesSubplot at 0x1a1759a978>
```

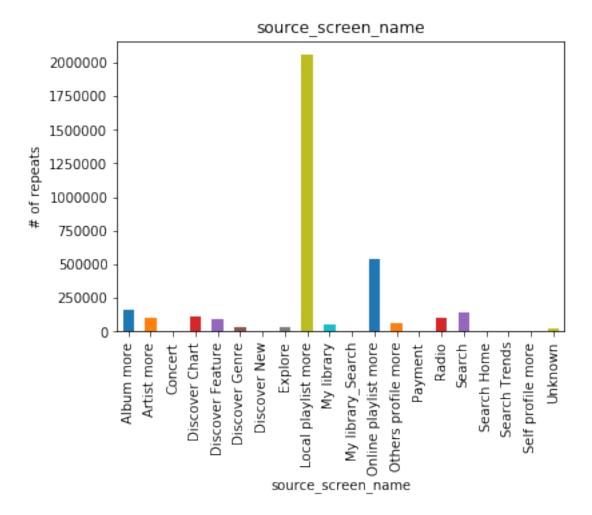


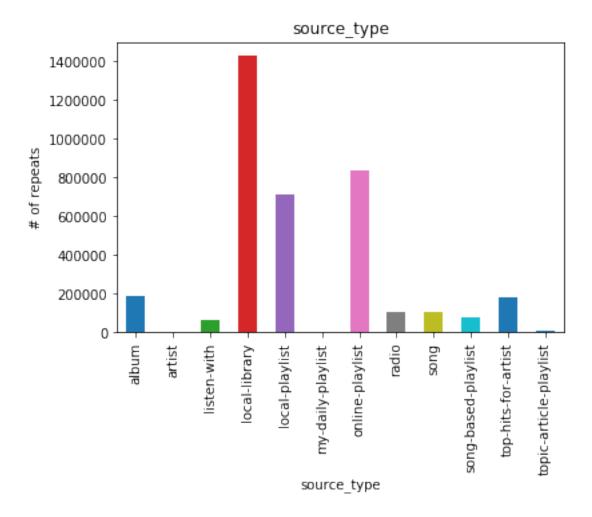
• So, *target* in train set is balanced.

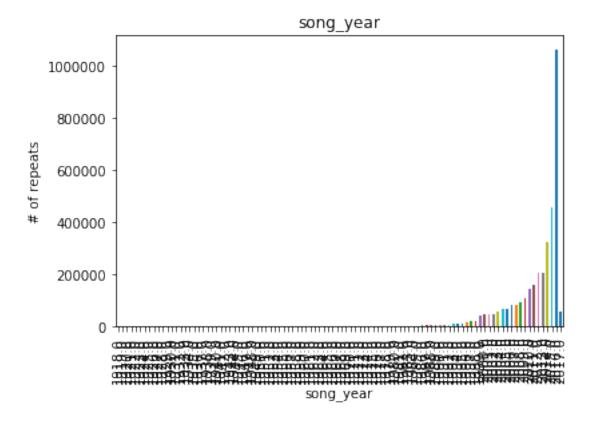


• Data is chronologically ordered - so we should choose last 25% of *train* set as the validation set instead of random shuffling for modeling since it is a good representation of the *test* set.









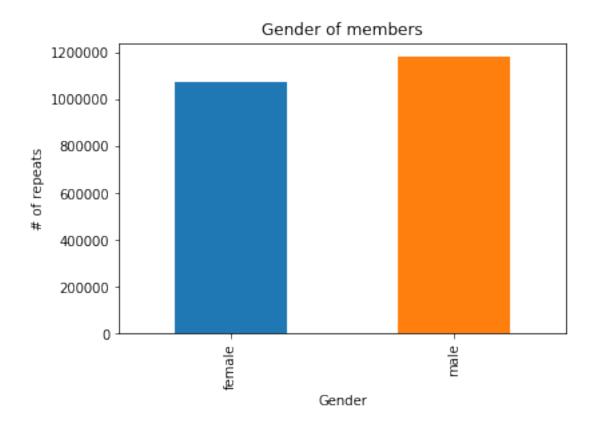
### 0.3 members

```
In [10]: members['membership_days'] = members['expiration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].subtract(members['registration_date'].
                                         members['membership_days'].describe()
Out[10]: count
                                                                                   34403.000000
                                                                                        1091.632474
                                        mean
                                         std
                                                                                        1151.438441
                                                                              -16191.000000
                                         min
                                         25%
                                                                                                72.000000
                                         50%
                                                                                            701.000000
                                         75%
                                                                                        1769.000000
                                                                                        5149.000000
                                         Name: membership_days, dtype: float64
In [11]: repeats=train[train.target==1]
                                         member_repeats=repeats.groupby('msno',as_index=False).song_id.count()
                                         member_repeats.columns=['msno','count']
                                         members=members.merge(pd.DataFrame(member_repeats), on='msno', how='left')
In [12]: members.head()
```

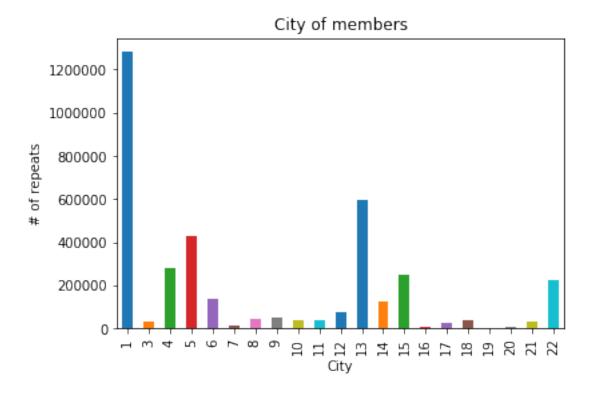
```
Out[12]:
                                                                  bd gender \
                                                            city
                                                      msno
         0 XQxgAYj3k1VKjR3oxPPXYYFp4soD4TuBghkhMTD4oTw=
                                                                        NaN
                                                               1
         1 UizsfmJb9mV54qE9hCYyU07Va97c0lCRLEQX3ae+ztM=
                                                               1
                                                                        NaN
         2 D8nEhsIOBSoE6VthTaqDX8U6lqjJ7dLdr72mOyLya2A=
                                                               1
                                                                   0
                                                                        \mathtt{NaN}
         3 mCuD+tZ1hERA/o5GPqk38e041J8ZsBaLcu7nGoIIvhI=
                                                                   0
                                                                        NaN
                                                               1
         4 q4HRBfVSssAFS9iRfxWrohxuk9kCYMKjH0EagUMV6rQ=
                                                                        NaN
            registered_via registration_init_time expiration_date membership_days \
         0
                         7
                                        2011-08-20
                                                         2017-09-20
                                                                                 2223
                         7
         1
                                        2015-06-28
                                                         2017-06-22
                                                                                  725
         2
                                                                                  457
                         4
                                        2016-04-11
                                                         2017-07-12
         3
                         9
                                        2015-09-06
                                                         2015-09-07
                                                                                    1
         4
                          4
                                        2017-01-26
                                                         2017-06-13
                                                                                  138
            count
           560.0
         1
              {\tt NaN}
         2
             11.0
         3
              {\tt NaN}
              NaN
Gender
In [13]: df=pd.DataFrame(members.sort_values(by='count',ascending=False))
         gender=df.groupby(['gender'])['count'].sum()
         gender.plot(kind='bar')
         plt.title('Gender of members')
         plt.ylabel('# of repeats')
```

plt.xlabel('Gender')

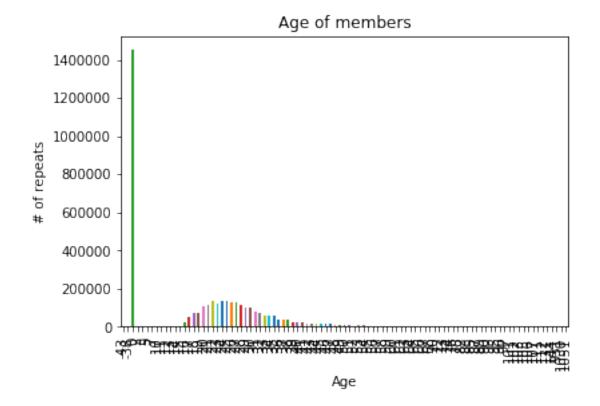
Out[13]: Text(0.5, 0, 'Gender')



# City

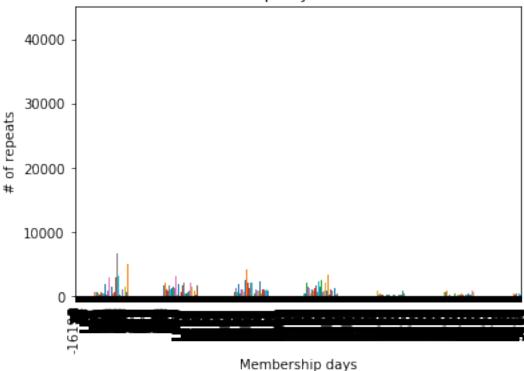


#### bd



## membership days

# Membership days of members



#### In [17]: train.info()

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 7377418 entries, 0 to 7377417
Data columns (total 8 columns):
msno
                          object
                          object
song_id
source_system_tab
                          object
                           object
source_screen_name
source_type
                          object
target
                           int64
song_year
                           float64
2017_song_year_percent
                          float64
dtypes: float64(2), int64(1), object(5)
memory usage: 506.6+ MB
```

#### 0.4 songs

```
In [19]: #songs = songs.drop(['count'], axis=1)
         songs.head()
Out[19]:
                                                 song_id song_length genre_ids \
         O CXoTN1eb7AI+DntdU1vbcwGRV4SCIDxZu+YD8JP8r4E=
                                                               247640
                                                                             465
         1 o0kFgae9QtnYgRkVPqLJwa05zIhRlUjfF701tDw0ZDU=
                                                               197328
                                                                             444
         2 DwVvVurfpuz+XPuFvucclVQEyPqcpUkHROne1RQzPsO=
                                                                             465
                                                               231781
         3 dKMBWoZyScdxSkihKG+Vf47nc18N9q4m58+b4e7dSSE=
                                                               273554
                                                                             465
         4 W3bqWd3T+VeHFzHAUfARgW9AvVRaF4N5Yzm4Mr6Eo/o=
                                                               140329
                                                                             726
                 artist name
                                                        composer
                                                                      lyricist \
         0
             (Jeff Chang)
                   BLACKPINK
         1
                              TEDDY | FUTURE BOUNCE |
                                                      Bekuh BOOM
                                                                        TEDDY
         2
                SUPER JUNIOR
                                                             NaN
                                                                           NaN
         3
                       S.H.E
         4
                                                 Traditional Traditional
            language
                       count
                         5.0
         0
                 3.0
         1
                31.0 1873.0
         2
                31.0
                      147.0
         3
                 3.0
                       161.0
                        35.0
                52.0
In [20]: df=pd.DataFrame(songs.sort_values(by='count',ascending=False))
         lang=df.groupby(['language'])['count'].sum()
         lang.plot(kind='bar')
         plt.title('Language of songs')
         plt.ylabel('# of repeats')
         plt.xlabel('Language')
Out[20]: Text(0.5, 0, 'Language')
```

