In [28]: df_logs.describe()

Out[28]:

	date	num_25	num_50	num_75	num_985	nu
count	1.839636e+07	1.839636e+07	1.839636e+07	1.839636e+07	1.839636e+07	1.83963
mean	2.017032e+07	6.191401e+00	1.508789e+00	9.413759e-01	1.079905e+00	3.02824
std	8.916720e+00	1.342827e+01	3.908539e+00	1.924840e+00	3.518409e+00	4.20364
min	2.017030e+07	0.000000e+00	0.000000e+00	0.000000e+00	0.000000e+00	0.00000
25%	2.017031e+07	0.000000e+00	0.000000e+00	0.000000e+00	0.000000e+00	7.00000
50%	2.017032e+07	2.000000e+00	1.000000e+00	0.000000e+00	0.000000e+00	1.70000
75%	2.017032e+07	7.000000e+00	2.000000e+00	1.000000e+00	1.000000e+00	3.70000
max	2.017033e+07	5.639000e+03	9.120000e+02	5.080000e+02	1.561000e+03	4.11070

In [30]: df_logs.isnull().values.any()

Out[30]: False

The logs table doesn't have any null values, and the ranges for the features seems sensible.

2. Data Storytelling

Q1: Count something

The mean number of unique songs listened to daily is about 29.

In [32]: df_logs.num_unq.mean()

Out[32]: 29.036145516162382

Looking at the mean of is_auto_renew, we see that the majority of users, 78%, have their plans set to automatically renew.

In [26]: df_trans.is_auto_renew.mean()

Out[26]: 0.7853025382789347

Q2: Find some trends

The registration of new members is increasing over time.

```
In [46]:
          pd.to datetime(df members['registration init time'], format='%Y%m%d')
          .dt.year.value_counts()
Out[46]: 2016
                  2246761
         2015
                  1620525
         2014
                   975776
         2013
                   524722
         2017
                   481684
         2012
                   283190
         2011
                   179051
         2010
                   115075
         2007
                    89830
         2008
                    67690
         2009
                    63633
         2006
                    53953
                    41349
         2005
         2004
                    26234
         Name: registration_init_time, dtype: int64
```

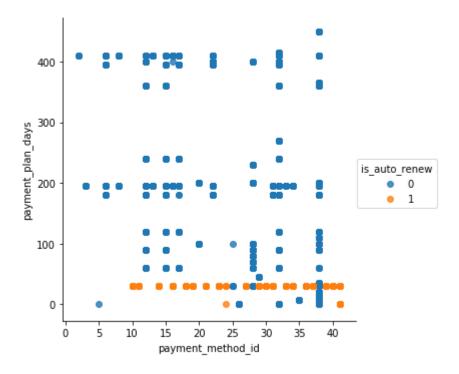
The month with the most signups is January, and the month with the fewest signups is May (TODO: START AND STOP SERIES AT THE SAME MONTH).

```
pd.to_datetime(df_members['registration_init_time'], format='%Y%m%d')
         .dt.month.value counts()
Out[6]: 1
               688712
         12
               652783
         2
               647056
         10
               645650
         11
               643998
         3
               591504
        8
               540394
         7
               492109
        9
               489340
        4
               477881
        6
               465483
        5
               434563
        Name: registration init time, dtype: int64
```

Q5. Make scatterplots

All of the auto renew plans have only 2 lengths (payment plan days).

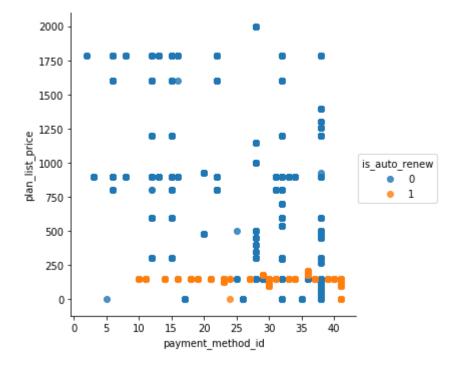
Out[22]: <seaborn.axisgrid.FacetGrid at 0x7f70c6f5bd30>



These two auto_renew plans have a few different prices.

In [28]: sns.lmplot(x="payment_method_id", y="plan_list_price", hue="is_auto_r
enew", data=df_trans,fit_reg=False)

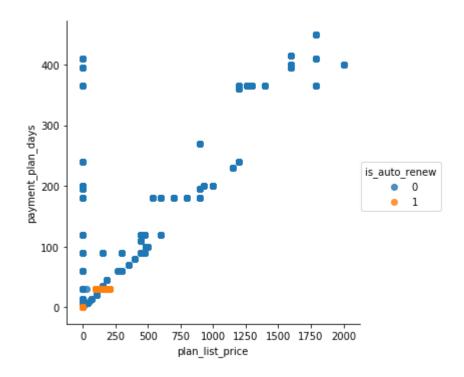
Out[28]: <seaborn.axisgrid.FacetGrid at 0x7f70c6ec49e8>



From the scatterplot of payment_plan_days vs plan_list_price it looks like there is a max plan price for each plan duration.

In [27]: sns.lmplot(x="plan_list_price", y="payment_plan_days", hue="is_auto_r
enew", data=df_trans,fit_reg=False)

Out[27]: <seaborn.axisgrid.FacetGrid at 0x7f70c2c166d8>



Auto renew is never false when is_cancel is true.

In [30]: sns.lmplot(x="is_auto_renew", y="is_cancel", data=df_trans,fit_reg=Fa
lse)

Out[30]: <seaborn.axisgrid.FacetGrid at 0x7f70c2baecc0>

