

Caitlin_Monaghan_Yammer

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1 Yammer: Investigating a Drop in User Engagement

1.1 Question:

1.1.1 What is behind the recent drop in weekly user engagement?

1.1.2 Engagement:

Making some type of server call by interacting with the product.

1.1.3 User Engagement:

Measured as the number of users who logged at least one engagement event during the week starting on that date.

Note: Engagement figure is based on login events specifically, though this always precedes any other engagement activities

Hypotheses:

- Does it differ by country?
 - Time range corresponds to common European holiday

```
In [1]: # Import packages
```

```
import os
import pandas as pd
from datetime import datetime
import seaborn as sns
```

```
In [2]: # Read in data
```

```
os.chdir('/Users/ckm/insight/Data Challenges/Week2')
df_emails = pd.read_csv('yammer_emails.csv')
df_events = pd.read_csv('yammer_events.csv')
df_users = pd.read_csv('yammer_users.csv')
df_rollup = pd.read_csv('dimension_rollup_periods.csv')
```

```
In [3]: # Functions to convert string timestamps to datetime, date, and time formats
```

```
def make_datetime(string):
    return datetime.strptime(string, '%Y-%m-%d %H:%M:%S')
def make_date(string):
    return datetime.strptime(string[:10], '%Y-%m-%d')
def make_time(string):
    return datetime.strptime(string[11:], '%H:%M:%S')

df_emails['datetime'] = df_emails['occurred_at'].apply(make_datetime)
df_emails['date'] = df_emails['occurred_at'].apply(make_date)
df_emails['time'] = df_emails['occurred_at'].apply(make_time)

df_users['datetime_c'] = df_users['created_at'].apply(make_datetime)
df_users['date_c'] = df_users['created_at'].apply(make_date)
df_users['time_c'] = df_users['created_at'].apply(make_time)

df_events['datetime'] = df_events['occurred_at'].apply(make_datetime)
df_events['date'] = df_events['occurred_at'].apply(make_date)
df_events['time'] = df_events['occurred_at'].apply(make_time)
```

```
In [4]: # Merge dataframes into one, add some other info.
```

```
df_emails = df_emails.rename({'action': 'event_type'}, axis='columns')
df_emails = df_emails.set_index('user_id', drop=False)
df_users = df_users.set_index('user_id', drop=False)
df_events = df_events.set_index('user_id', drop=False)
df_full = pd.concat([df_emails, df_events], sort=False)

def add_comp(row):
    return df_users['company_id'].loc[row]

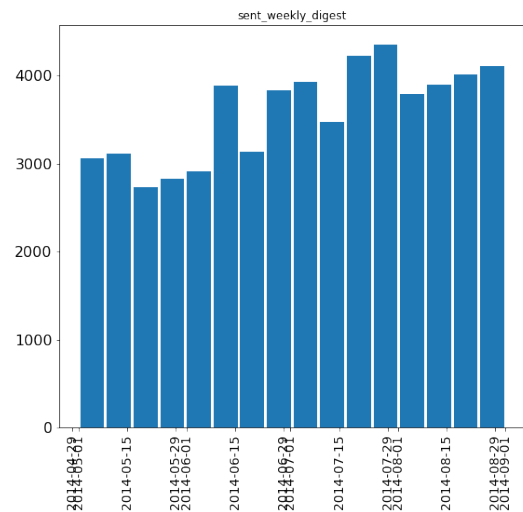
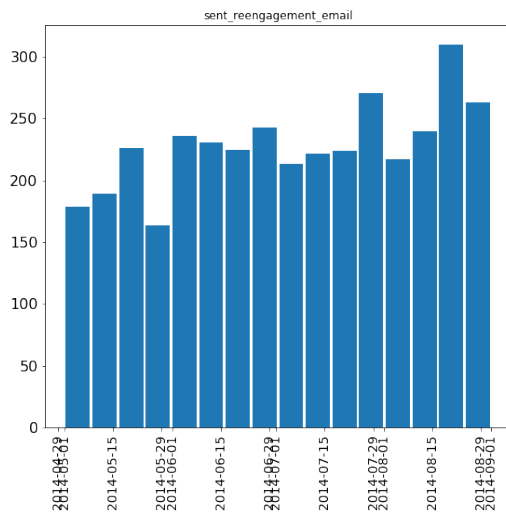
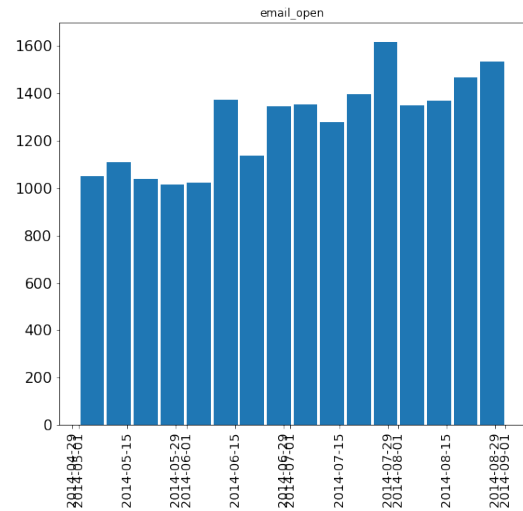
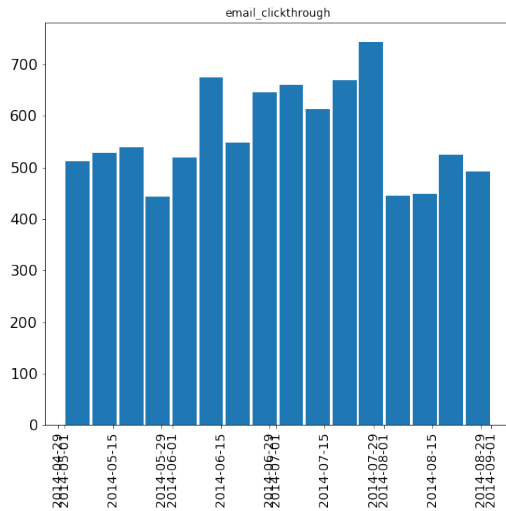
df_full['company_id'] = df_full['user_id'].apply(add_comp)
```

2 Yammer emails

2.1 Exploratory Data Analysis

2.2 Email clickthroughs are down, though they are still being opened:

```
In [5]: ax = df_emails.hist(column='date', by='event_type', figsize=(20,20), bins=16,
                             range=('2014-05-01', '2014-09-01'), rwidth=0.9,
                             xlabelsize=14, ylabelsize=16)
```

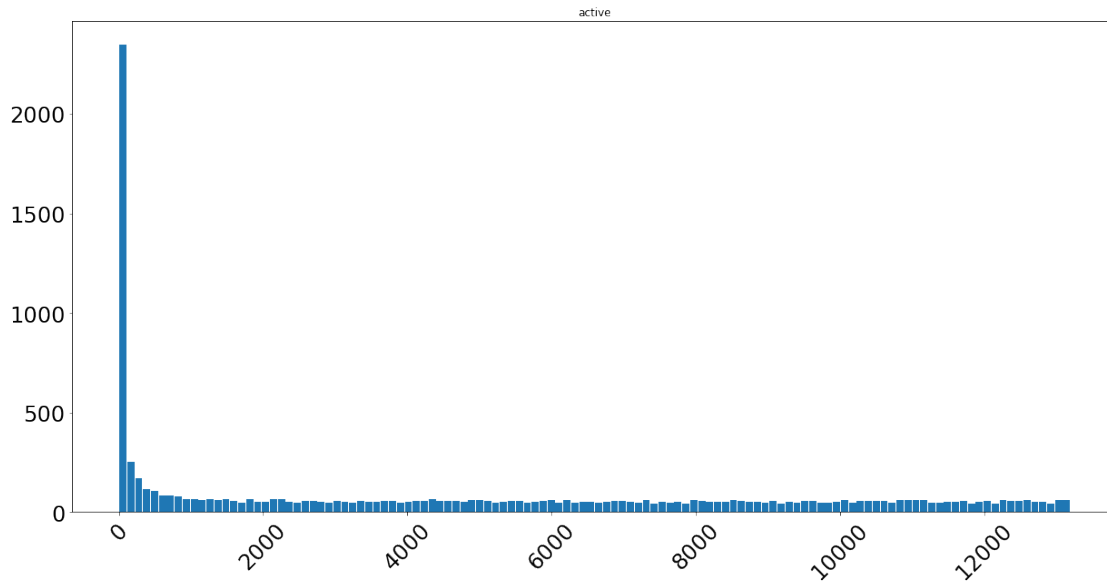


3 Yammer users

3.1 Exploratory Data Analysis

3.2 Number of active users in companies

```
In [6]: ax=df_users[df_users['state']=='active'].hist(column='company_id', by='state',
                                                    bins=120, figsize=(20,10), xrot=45,
                                                    rwidth=0.9,xlabelsize=24, ylabelsize=24)
```



3.3 Not many more unique users than companies:

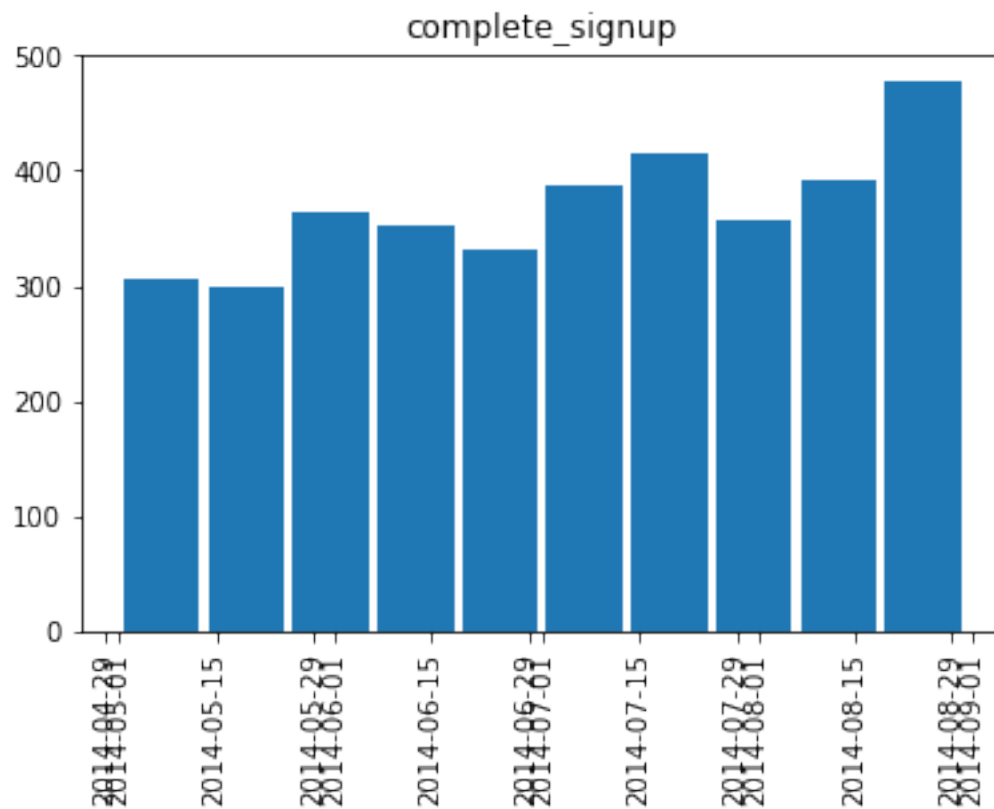
```
In [7]: companies = df_users['company_id'].unique()
        print('Unique companies: ' + str(len(companies)))
        print('Total users: ' + str(len(df_users)))
```

Unique companies: 13198

Total users: 19066

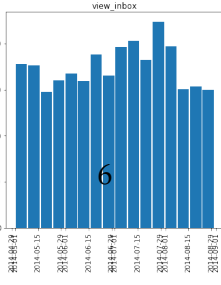
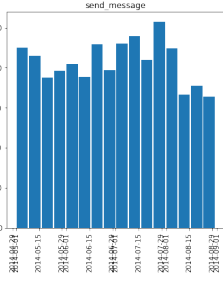
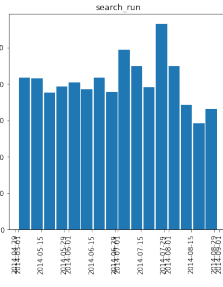
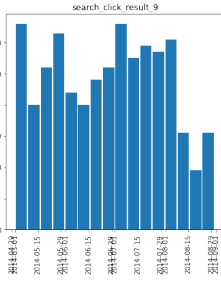
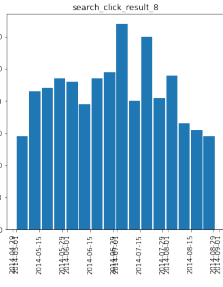
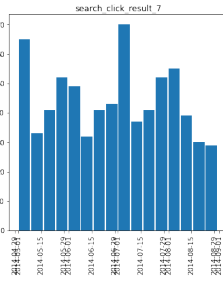
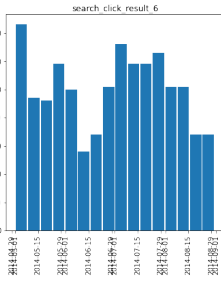
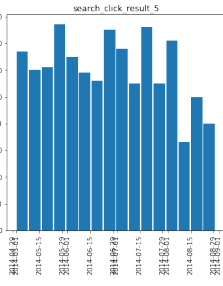
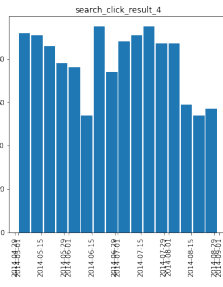
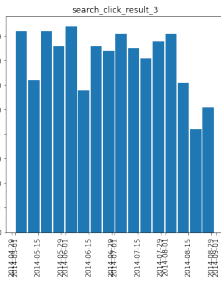
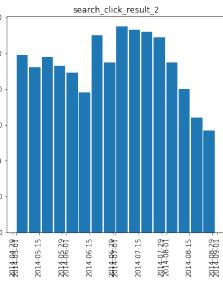
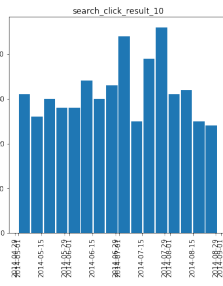
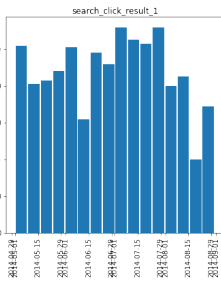
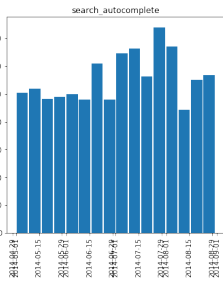
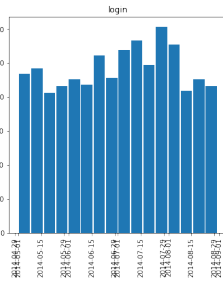
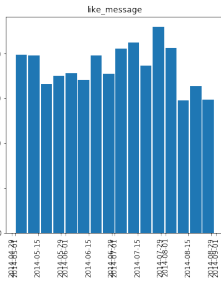
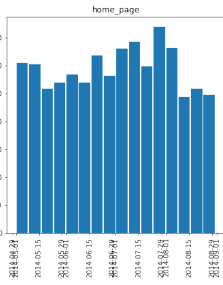
3.4 Signups are actually increasing

```
In [8]: ax=df_events[df_events['event_name']=='complete_signup'].hist(column='date',
                                by='event_name',
                                rwidth=0.9)
```



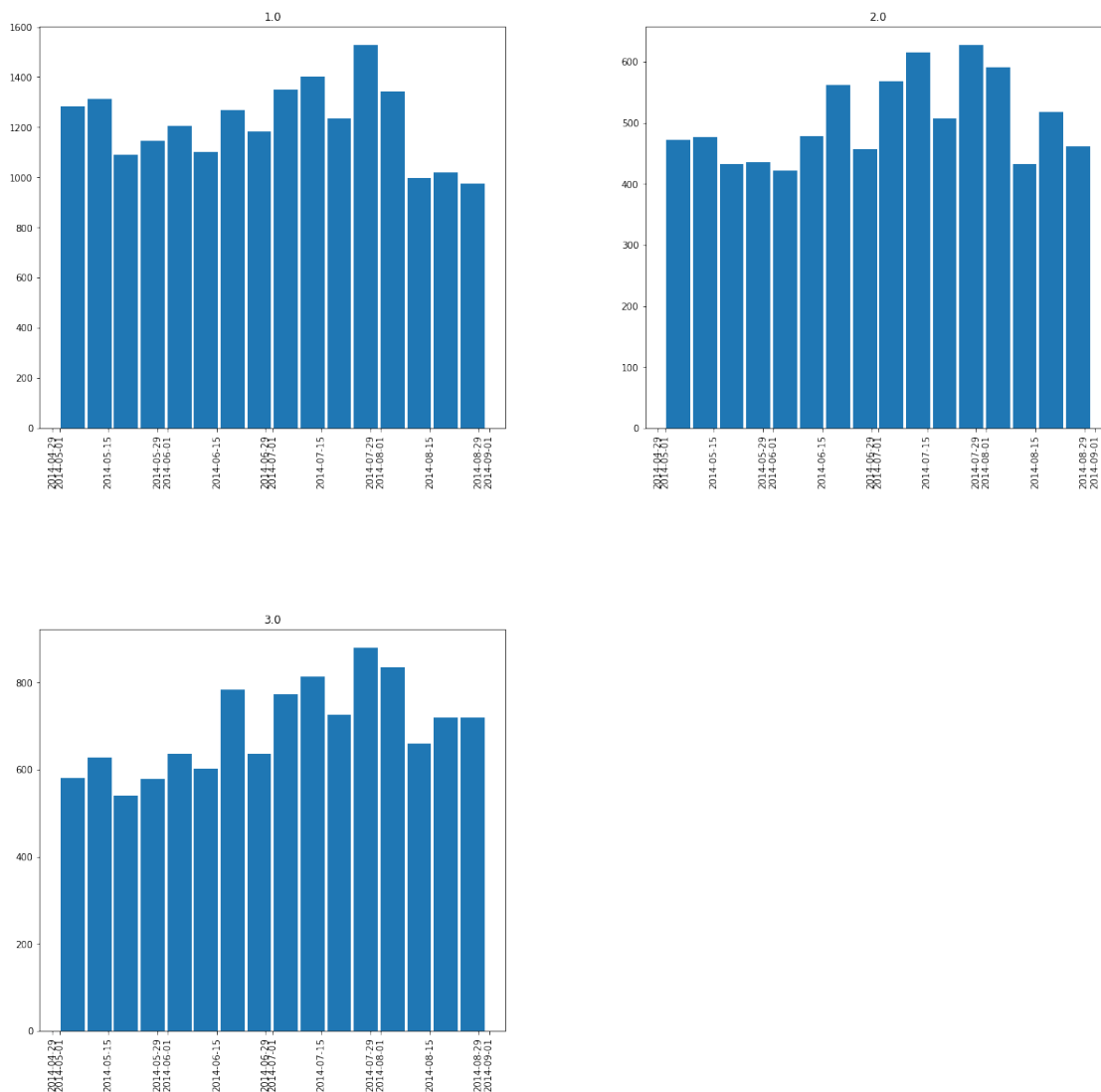
3.5 Decreasing trend consistent across event types:

```
In [9]: ax=df_events[df_events['event_type']=='engagement'].hist(column='date', by='event_name',
figsize=(20,50), layout=(6,3),
bins=16, rwidth=0.9)
```



3.6 Decrease most pronounced in user type 1, though present across all:

```
In [10]: ax=df_events[df_events['event_name']=='login'].hist(column='date', by='user_type',  
figsize=(20,20), bins=16,  
rwidth=0.9)
```



3.7 Decrease in logins appears more pronounced in many European countries:

```
In [11]: ax=df_events[df_events['event_name']=='login'].hist(column='date', by='location',  
figsize=(30,60), layout=(12,4),  
rwidth=0.9, bins=16)
```



3.8 Not due to complete failure on a device:

In [12]: `ax=df_events[df_events['event_name']=='login'].hist(column='date', by='device',
figsize=(30,40), layout=(7,4),
rwidth=0.9, bins=16)`



- 4 Ultimately, decrease in user engagement is likely due to the European holiday in August when fewer people are working.
- 5 Given the increase in user signups, engagement will likely rebound and return to previous growth.