

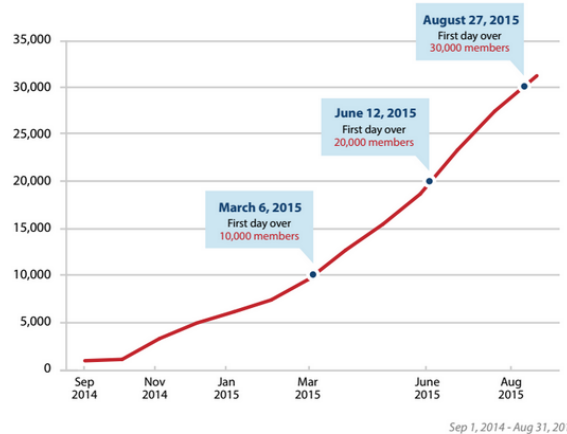
The Problem:

True Adoption of Enterprise Collaboration Software

- Enterprise Collaboration Software is only valuable if people actually use it.
- Achieving true adoption of Collaboration Software is a lot harder than you might guess.
- Significant investment to get adoption: software investment, training, internal marketing, etc.
- Gamification presents an opportunity but thus far is based on intuition rather than data

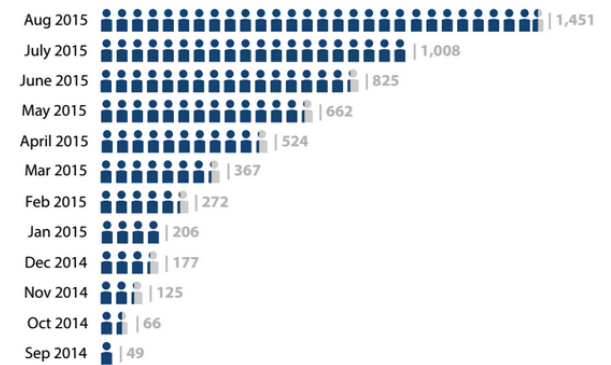
Total Members

Total members as of August 31, 2015: **30,487**



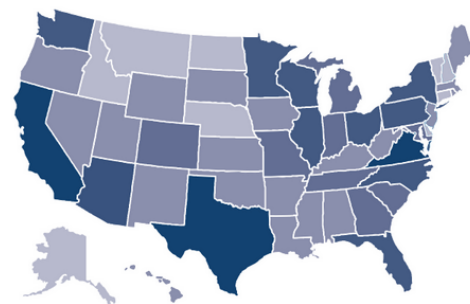
Total Groups

Total as of August 31, 2015: **1,451**



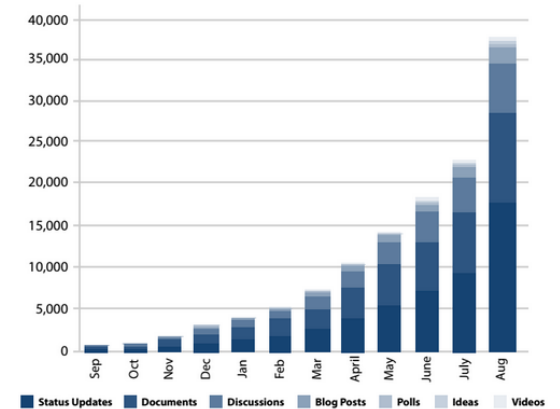
Total Page Views By State

Total page views as of August 31, 2015: **2,299,018**



Total Content Posted

Total content posted as of August 31, 2015: **38,236**



The Question(s)

- Can we predict which users will “adopt” based on behavior in the first month?
- If so, can we use the model to inform our process for on-boarding new users? Interpretability of model would be an excellent bonus.

Sub Question 1: which rollout to use as sample?

- Needs sufficient data to model
- Prefer on that is still “in progress” to inform ongoing work.
- Answer: large, slow-rolling deployment that just hit 1-year birthday with just over 10% of total potential audience on platform and a bit over 7,500 users on platform at least 6 months

Sub Question 2: what is definition of “adopt”?

- Lapsed: logs in < 4 times a month or perform less than 20 actions/month
- Collaborator: at least 25 weighted usages (5 points / major content + 1 point / minor content)
- Consumer: viewing but not creating; not Lapsed but not Collaborator

The Data

- Underlying SQL database captures a ton of data about users and the activity they perform in the system (think google analytics level data without blinding).
- User profile data is voluntary and thus data completeness / quality is limited; ongoing side investigation for better profile data (e.g. ip address giving location).
- Other activity in Excel logs (training participation, demo participation).

Data Cleaning

Normalizing data to show the following per user (big effort):

- UserId
- Creation Date
- Location
- Organization
- Per Month Data
 - # Logins
 - # “Small” Activities
 - # “Large” Activities
 - # Group Memberships
 - # Social Relationships
 - Most Active Group # “Small” Activities
 - Most Active Group # “Large” Activities

Data Exploration

- ~48% of users drop to 0 activity after 6 months
 - This includes users who leave the organization completely so this “end-state” is over-reported; trying to find a way to clean these users out of data.
- ~21% of users are “Contributors” after 6 months
- Intuitions based on playing with data
 - Thus far, no clear pattern in first month personal activity; some users do a lot and then go quiet; others start with little activity and then go nuts. This lead to re-definition of activity into “Small/Large”; haven’t had a chance to really understand this.
 - Membership in an “Active” group that adopts tool seems to correlate with continued participation; do we need to further engineer this feature?
 - Anecdotal evidence / theory that number of relationships built with active users leads to activity (looking to engineer a feature to represent this).

Future Questions (Sandbox)

- Predicting whether a user is going to drop-out based on current rolling months worth of activity; enables team to re-target users before
- Predicting whether an employee is going to quit based on activity (or changes in activity) could enable focused interventions from managers.