Science, AskScience, and BadScience: On the Coexistence of Highly Related Communities

Jack Hessel and Chenhao Tan and Lillian Lee

Department of Computer Science
Cornell University
{jhessel, chenhao, llee}@cs.cornell.edu

Abstract

When large social-media platforms allow users to easily form and self-organize into interest groups, highly related communities can arise. For example, the Reddit site hosts not just a group called food, but also HealthyFood, foodhacks, foodporn, and cooking, among others. Are these highly related communities created for similar classes of reasons (e.g., to focus on a subtopic; to create a place for allegedly more “high-minded” discourse, etc.)? How do users allocate attention between such close alternatives when they are available or emerge over time? Are there different types of relations between close alternatives such as sharing many users vs. a new community drawing away members of an older one vs. a splinter group failing to cohere into a viable separate community? We investigate the interactions between highly related communities using data from reddit.com consisting of 975M posts and comments spanning an 8-year period. We identify a set of typical affixes that users adopt to create highly related communities and build a taxonomy of affixes. One interesting finding regarding users’ behavior is: after a newer community is created, for several types of highly-related community pairs, users that engage in a newer community tend to be more active in their original community than users that do not explore, even when controlling for previous level of engagement.

Table 1: The 10 most common Reddit group-name affixes.

| Affix | Example | # Pairs |
|-------|---------|---------|
| s     | auto, autos | 63      |
| porn  | space, spaceporn | 26      |
| circlejerk | hiphop, hiphocirclejerk | 23      |
| ask   | science, askscience | 21      |
| shitty | ideas, shityideas | 17      |
| music | running, runningmusic | 17      |
| help  | tech, techhelp | 11      |
| 2     | dota, dota2 | 9       |
| true  | atheism, trueatheism | 9       |
| learn | math, leaarnmath | 9       |

Introduction

Social networks are in constant flux, with new communities forming and old communities dying over time. On websites such as Facebook and Reddit, users have complete freedom to create communities at their own discretion. This has led to a very large number of communities arising organically from user initiative, for a variety of reasons. One reason is to create divisions that satisfy the need to better organize discussions; in fact, community design theory argues that “a growing Web community needs subdivisions which might be represented as towns, neighborhoods, topics, categories, conferences, or channels, depending on your metaphor” (Kim 2000; Jones and Rafaeli 2010). Or, new groups can develop because of religious, political, or other schisms; online examples include groups whose very names attempt to connote superiority to others, e.g., the subreddits trueatheism vs. atheism. Other reasons surely exist. The tremendous reach of modern social media provides researchers much greater data to examine these social processes at scale.

An interesting and frequently occurring version of the group creation process is that a new concept or culture may gain in popularity and, in a meme-like fashion, draw users to create a new community by using that concept as an affix1 of their community name. For example, on Facebook, after the creation of the OMG Confessions group, anonymous confession pages with names combining a college with the word confession or confessional proliferated to the degree that one can now find a confession page for almost every university campus. (Birnholtz, Merola, and Paul (2015) examine what kind of questions people ask on such pages.) Table 1 shows some examples from Reddit: the second column shows pairs of subcommunities where the name of one is a modified form of the other (ignore the third column for now).2

In this work, we investigate highly related communities that are based on affixes. An understanding of these highly related communities may help community organizers identify subtopics in a community and create an appropriate sub-

---

1An affix is either a prefix or a suffix.

2An additional, whimsical example from Reddit is random acts of, indicating people asking for or sending free things to others. Instantiations include random acts of pizza, random acts of amazon, and random acts of books. Althoff, Danescu-Niculescu-Mizil, and Jurafsky (2014) used random acts of pizza to study effective ways to ask for a favor.
division to cultivate focused discussions, or monitor sub-
groups that potentially feel marginalized or underserved,
and decide whether to change community norms or create
a dedicated community for that subgroup.

Despite the ubiquity of such affixes, and their appeal as
easily-identifiable (albeit sometimes imperfect) instances of
the important phenomenon of highly related communities,
little is known about canonical affixes and the activity in
the resultant highly related communities. For instance, are
eighborhoods, topics, and channels enough to capture all
possible affixes? Are there classes of affixes that are gen-
erally applicable? Perhaps different affixes behave in dif-
f erent ways. Moreover, once a highly related community is
created, how does it interact with the existing community?
Will it overtake it? Will the two share the same user base?
One of our goals is to analyze user behavior in the existing
community after they participate in the new community.

Organization and contributions. In this paper, we con-
struct a dataset from Reddit and present the first large-scale
study on the coexistence of highly related communities. De-
tails about the dataset are introduced in “Dataset Descrip-
tion”.

Our first contribution is to characterize the space of af-
fixes. We build a taxonomy of common affixes that users
adopt to create highly related communities. For instance,
we identify a category of “parody” affixes (circle jerk, shitty,
funny, lol, bad). This category generally shares the same
user base with its corresponding unaffixed community. On
the other hand, we identify a category of “derivative” affixes
(meta, anti, srs, post, ex) that likely attract different user
bases. Surprisingly, a non-trivial fraction of affixed com-
munities exist before the unaffixed ones. Also, an interesting
class of spinoff communities arises where early participants
in the new community come from the existing community.

Our second contribution is to introduce a framework for
analyzing users who try out spinoff communities (dubbed
“explorers”) and comparing them to “nonexplorers” who
never leave the original subreddit. We make the surpris-
ing observation that in multiple classes of affixes, users
who explore spinoff communities are more active in the origi-
nal communities after exploring when compared to similarly
active users who never tried the alternative. This resonates
with the findings in Tan and Lee (2015) that users who “wan-
der” to different (potentially completely unrelated) groups
tend to stay active longer on the site as a whole. Our ob-
servations may suggest that spinoff communities generally
serve a complementary rather than competitive role in multi-
community settings.

Finally, we summarize related work and offer some con-
cluding thoughts.

Dataset Description

Our starting point for understanding highly related com-
munities, affixes, spinoffs, nonexplorers, and explorers is an
examination of topically related communities. As such, we
compile a dataset from reddit.com, a site where users
are allowed to create communities called subreddits at their
discretion. Users can name the subreddits that they create
so that like-minded people can identify them effectively. As

| Data type   | count |
|-------------|-------|
| Subreddits  | 5,692 |
| Posts       | 88M   |
| Comments    | 887.5M |

a result of unmoderated creation and limitless naming pos-
sibilities, there are a wide variety of subreddits on Reddit,
e.g., funny, worldnews, politics, IAmA, todayilearned, etc. On
these subreddits, users submit link-based posts or text-based
posts, comment on others’ posts, and up/down vote posts and
comments. We construct a dataset that includes all ac-
tivities on Reddit from its inception until 2014, an 8-year
period, by combining two data sources: a post dataset that
was organized in Tan and Lee (2015), and all comments data
extracted by Jason Baumgartner. \(^3\) We focus on communities
that are active and that enjoy a reasonable number of users.
Specifically, we require all communities to include at least
300 unique users that made posts. This left us with just un-
der 5.7K communities. Table 2 presents basic statistics of
this dataset. \(^4\) The metadata for the Reddit conversation trees
that we used here is available for download. \(^5\)

As discussed in the introduction, user-defined subreddit
names are an important indicator of relationships between
highly related communities (e.g., food vs. HealthyFood).
We first retrieve all possible pairs of communities where one
community name is the other’s suffix or prefix, ignoring case
(food is the suffix of HealthyFood, ignoring case). We refer
to the difference between the names in a pair as the affix. For
instance, healthy is the affix in the pair food vs. HealthyFood.
There are around 4K such pairs over our dataset.

Using common affixes as a starting point allows us to
discuss the the space of possible highly related communi-
ties. For example, this framing allows us to make statistical
observations about all pairs with healthy or true as affixes.
Note that we omit some interesting highly related communi-
ties pairs by focusing on affixed pairs. One example is TwoX-
Chromosomes, a very popular “subreddit ... intended for
women’s perspectives,” and TrollXChromosomes, its satiri-
cal counterpart.

Identifying topically related communities. Unsurpris-
ingly, not all pairs of communities identified through affixes
are actually highly related communities. An example is “ru”
and “rum;” the first one is a Russian community while the
second one is about the liquor. In order to quantify sub-
reddit similarity, we compute the content similarity between

\(^3\) Information is available at https://pushshift.io. The
dataset in Tan and Lee (2015) was also originally extracted by Ja-
son Baumgartner.

\(^4\) The statistics reported here include posts and comments made
by users who deleted their accounts and banned accounts.

\(^5\) http://goo.gl/sHUfhC
pairs of communities. As suggested in Singer et al. (2014), subreddits can focus on text posts in addition to link-based posts. Therefore, we employ a method that can account for either link-dominant or text-dominant subreddits. Specifically, we use Jaccard similarity between the set of links to capture similarity based on links.\footnote{Jaccard similarity is defined as \( \frac{|A \cap B|}{|A \cup B|} \), where \( A \) and \( B \) are the set of links from two subreddits respectively.} and use Jensen-Shannon divergence between topic distributions (derived from a topic model trained on 6.6M text posts) to capture similarity based on text, following Hessel et al. (2015). Since these two metrics are not comparable by raw value, we compute the full background distribution based on all 1.62M possible pairs of the 5.7K communities in our dataset and compute the percentile of each affix pair in each distribution.

We consider a pair of communities to be topically related if either link similarity is above the 90th percentile or topical similarity based on text is above the 90th percentile. Accounting for our definition of topical similarity yields just over 1.7K pairs from our original set of 4K.

The last step of our preprocessing is to identify generalizable affixes that are commonly used in these highly related communities. We count the frequency of affixes and keep affixes that occur at least three times, so that all affixes in the final dataset carry a general meaning (it is not possible to make general statements about affixes that only occur once). This step brings us to 99 affixes and 572 pairs of highly related communities distributed between them.

### Characterizing affixes

The goal of this section is to explore the types of canonical affixes users on Reddit utilize. To accomplish this exploration, we first build a taxonomy of common affixes to better understand their basic properties and relationships. Next, we explore the temporal characteristics of the pairs. In general, we observe an accelerating culture of creating highly related communities, meaning that highly related communities are being created at increasing rates. We also observe that, in most cases, the affixed community in a pair was created after the unaffixed one, even though there is a non-trivial fraction that went the other way, e.g., ukpolitics and uspolitics both existed before politics. We further explore whether the newer community “overtakes” the older one in popularity. We then offer potential rationales that may help explain the surprising finding that a quarter of the newer communities are more active. The final characteristic that we examine is whether the newer community actually shares a user base with the older one, at least when the new one is forming. Despite the high similarity both in community name and in content, almost half of newer subreddits in pairs are not, in fact, born out of their older partners.

### The space of affixes

In order to achieve a basic understanding of what canonical affixes users adopt to create new communities, we first build a taxonomy of the 99 affixes from the dataset section in Table 3.

| Adjective-like | Verbatim-like | Noun-like | Minor |
|----------------|---------------|-----------|-------|
| “better”       | true, plus    | uk, reddit, chicago, us, dc, steam, canada, american, boston, android, online, web | s, al, ing, the, alternative |
| “parody”       | circlejerk, shitty, funny, lol, bad | porn, pics, music, memes, videos, vids, comics, apps, games, gaming, game | 2, 3, 4, 5 |
| “derivative”   | post, ex, meta, anti, srs | science, news, dev, servers, tech, tv, guns, recipes, city, u, college, man, girls |
| “genre”        | classic, fantasy, indie, folk, casual, dirty, classic, metal, academic, 90s, free, social | “equivalent, competition” |
| “nsfw”         | nsfw, nsfw, asian, trees, gonewild, gw, r4r, tree | “generation” |
|                |                | “subject” |
|                |                | “modifier” |

We start with a coarse structure based on part-of-speech. Among the adjective-like, the largest category is based on “genre”, e.g., rock vs. classicrock. Some other very interesting classes also arise: “better”, which indicates a certain level of superiority (e.g., atheism vs. trueatheism); communities dedicated to “parody” where users are likely aware of the culture in the unaffixed one (e.g., history vs. badhistory); and “derivative”, which probably attracts a very different audience (e.g., war vs. antiwar). In fact, anti and meta can be recursive, e.g., jokes, antijokes and antiantijokes.

Among the verb-like affixes, a class of self-improvement or learning communities exists, e.g., programming vs. learn-programming. In “actions”, there are many exchange related affixes, including trades (e.g., pokemon vs. pokemons), and swap (e.g., scotch vs. scotchswap). Altruistic behavior signified by random acts of... (e.g., pizza vs. random acts of pizza) has been studied specifically in Althoff, Danescu-Niculescu-Mizil, and Jurafsky (2014).

The noun-like affixes closely match the conceived metaphor of splitting space in community design theory (Kim 2000). Indeed, we see a group of affixes based on “place”, such as uk (e.g., politics vs. ukpolitics). “Medium”,
Temporal Relationships within Pairs

It is always possible to determine which community in a pair was created earlier. The first characteristic that we examine is the gap between the creation time of two communities in a pair. The overall average gap is 749 days since 2008, when users on Reddit were first allowed to create their own communities. If we compute the average gap grouped by the creation year of the older community, in Figure 2a we see a consistent trend that the newer community is created with a shorter and shorter gap over time. This suggests that there may be an accelerating culture of creating highly related communities over time, or that as there are more users on Reddit, affixed communities arise more quickly.

For most affixes, the community with the affixed was newer.

We further examine whether the newer community within a pair is the affixed one. This is indeed the case in 86% of our pairs. However, if we change our focus from pairs to affixes, we find that for 33% of the affixes, there was at least one instance where the affixed version was actually created before the original (see Figure 2b).

The four affixes for which the affixed version of the community more often exists first are ing, al, ism and s; these are mostly in the “equivalent/competition” class in Table 3. As a result, we observe phenomena like different communities focusing on exactly the same thing (e.g., wallpaper vs. wallpapers) or two communities eventually deciding to explicitly merge into one (e.g., wedding vs. weddings). Communities with different foci but similar names might also fall into this category, such as vegetarian vs. vegetarianism.

These four affixes do not cover all possible cases where the affixed was created earlier. For instance, twincitiessocial
was created before twincities.

Does the New Overtake the Old?

Another important characteristic is how active the newer community is compared to the older one after its inception. **Newer communities tend to be less active, but in a quarter of pairs, the newer one is more active.** We compute the log ratio in activity level (the total number of comments plus the total number of posts) between the newer community and the older community with add-one smoothing, only considering actions after the newer community was created so that we compare pairs during the same time period. According to this metric, a positive value means more activity in the newer community and a negative value means less activity in the newer community. Figure 3a demonstrates there is a trend that affixed versions of communities tend to be less active. The mean log ratio is -2.0, which suggests that new community is usually 13.5% as active as the older one. However, a nontrivial fraction of newer communities (25.7%) are more active.

A closer look at the more active newer communities. It’s somewhat surprising that 25.7% of newer communities overtake their established counterparts. Why does this occur? Figure 3 presents examples of possible reasons that the younger community might surpass its older counterpart.

The first reason is that the affix represents something that naturally appeals to more people. One example is writers vs. fantasywriters. As soon as fantasywriters was created, its activity level was more than twice as great as that in writers. Here are top 3 affixes that consistently lead to more activity: the (e.g., stopgirl vs. thestopgirl), ex (e.g., mormon vs. exmormon), steam (e.g., deals vs. steamdeals).

Second, the newer community may be “equivalent” to the older one and the newer one may win the competition. For example, in the case of auto vs. Autos, it took Autos a while to exceed the activity level in auto, but Autos is now much more popular (see Figure 3b).

Third, the newer community may actually be the non-modified one (14% of pairs have this property, as we earlier observed) and the newer one might achieve popularity because it is more general. For instance, politics is more popular than ukpolitics despite the later’s earlier founding. In this case, as soon as politics was created, its activity level exceeded ukpolitics.

The fourth and relatively rare reason is that the older one may have a large competitor, in other words, the newer one may originate from an even bigger community than the older one. An example is hiphop vs. makinghiphop. makinghiphop started at a similar size as hiphop but exceeded hiphop significantly later. Although hiphop and makinghiphop are both active, there is a much larger hiphop-related community on Reddit, hiphophead. makinghiphop might actually originate from hiphophead instead of hiphop.

Where are early participants in the new communities from?

The last reason in the above discussion leads to a natural question: where are the participants in the newer community from? Are they from the older one in a pair? This question is difficult to answer, as a subreddit may establish its own identity and unique audience over time, even if it was born out of an existing community. If we simply look at the overlap between two communities over all users, we may mistakenly believe that they have never shared the user base as a result of a large number of later users. We thus focus on the first \( n \) participants in the newer community (the *early participants*) and compute the fraction of them that were also members of the older community. A user is considered a member of the old community if they took any action in the old community within the last 30 days prior to interacting with the new community. We refer to this metric as “early participant fraction”. While we present results for \( n = 100 \), similar results hold for different \( n \).

Almost half of highly related communities do not really share early participants. As shown in Figure 4a, surprisingly, the majority of newer subreddits in highly related

![Figure 3:](a) Histogram of log activity level ratio between the newer community and the old one. (b) Case study on pairs in which the newer one has more activity, where activity is binned on a month-to-month basis.)

![Figure 4:](a) Histogram of the fraction of early participants in the new community that were from the old community over all pairs. (b) Average fraction of early participants in the new community that were from the old community sorted by categories.)
communities pairs are not “founded” by members of the older community. For example, only 7 of the first 100 participants in makinghiphop were members of hiphop.

Figure 4b presents the average early participant ratio for all categories in Table 3. It shows that “better”, “parody”, “action” and “learning” usually attract members from the older community. It also partly demonstrates why we obtain such a low average early participant ration. “equivalent” and “modifier” appear more than likely to attract completely different participants, e.g., vegetarian vs. vegetarianism. We also notice significant differences even within a single category. One notable example is meta (65.8% from the original community) vs. ex (1% from the original community).

From Highly Related Communities to Spinoffs

Thus far, we have explored the complex space of possible affixes, and the highly related communities that are created through them. We find that a non-trivial fraction of the new communities were not the affixed ones, or did not share the same user base of the older one. For these pairs, it is unclear whether the new community is a subdivision of the old one, or whether users in the existing community are affected by the new one’s presence. In order to better understand how users in the existing community may behave after exploring the new community, we will focus on a subset of highly related communities called spinoffs in the remainder of this work.

Spinoffs: Substitutions or Complements?

We now formally define spinoff communities. First: the newer of the two pairs in a highly related community is a spinoff if it satisfies the following properties: 1) more than 10% of the first 100 early participants in the newer community are members of the older community; and 2) the newer community is the affixed one, so that it is likely to represent a specialization or some other topic of interest. We will sometimes refer to a pair of highly related communities that contain a spinoff as a spinoff pair.

In this section, we investigate how a user’s behavior within the older subreddit is affected once they try out the newer spinoff: do such users get “distracted” by the new one, or does the new subcommunity complement the old one? Phrased differently, do users tend to decrease, increase, or not change their activity levels in the original community after trying the spinoff?

Surprisingly, we find that users who explore the spinoff generally become more active in the original community. Furthermore, with respect to the taxonomy we developed in Table 3, the magnitude of this trend depends on the type of affix: larger in “action”, “better”, and “parody”, smaller in “medium”, and negative in “nsfw”. Finally, it seems that this complementary effect is more prominent for users with lower activity level, although there is less data to compare users with different activity levels, and results may vary depending on specific pairs.

Disclaimer: we do not make any claims of causality given the observational nature of our dataset.

Figure 5: Schematic of the exploration experiment setup. TrueAtheism is a spinoff of Atheism, and the activity of two users is shown over time. Each box represents an interaction. With respect to the two subreddits shown, the dark user is an explorer, and the light user is a nonexplorer. Time $t$ is the time of the dark user’s first interaction with the spinoff subreddit. Here, the number of pre-interactions for both the dark and light users is 5. The dark user has 3 post-interactions, whereas the light user only has one.

Experiment setup

To understand user behavior in the original community after participating in the spinoff community, we propose an experiment framework in which we first pair “an explorer” and a “similar” “nonexplorer” in the original community. After identifying this pair of users, we compare their behavior pattern after the explorer first participated in the spinoff community, as illustrated in Figure 5.

Specifically, for each spinoff pair (e.g., Atheism vs. TrueAtheism in Figure 5), we define explorers as users who were active in the original community in a time window before their first participation in the spinoff community. The darker user in Figure 5 is an example. We denote the time of her first interactions in the spinoff community as $t$, and refer to her interaction in the original community from $t - w$ to $t$ as pre-interactions and her interaction in the original community from $t$ to $t + w$ as post-interactions. We consider users with at least 5 pre-interactions to ensure that they were indeed active in the original community.

A straightforward metric to compute is simply the ratio between the number of post-interactions and the number of pre-interactions for each user. However, this is problematic because we require users to have at least 5 pre-interactions but have no constraints on post-interactions. This causes our sample to be biased towards users with more pre-interactions than post-interactions.

To address this concern, for each exploring user $u_e$, we sample a similarly active user $u_{ne}$ in the original community who never interacts with the spinoff community. We call this user a “nonexplorer”. The rough idea is demonstrated by the light user in Figure 5, who had a similar number of pre-interactions and made a post in the original community around $t$ so that we know she was still active. The details of this sampling process are given in the appendix.

Metric: exploration effect. After we identify explorers and matching nonexplors, we compute the fraction of explorers who have more post-interactions than pre-interactions in...
the older community, and then compute the same fraction for nonexplorers. We take the difference between these two fractions and call it the “exploration effect” (see Equation 2, in the appendix). Higher values of this quantity indicate that user was more active in the original community than the non-explorer \( u_{ne} \). We use the macro average to aggregate results from different spinoff pairs because the number of explorers varies between pairs.

The only parameter in our framework is \( w \). Since our primary objective of interest in this work is the effect of the interaction with the spinoff community, we choose a relatively small window (30 days) to mitigate confounding factors that may affect user behavior over time and the dynamic nature of online communities (Danescu-Niculescu-Mizil et al. 2013; Backstrom et al. 2006; Ducheneaut et al. 2007; Kairam, Wang, and Leskovec 2012; Kumar, Novak, and Tomkins 2010). Our results are robust to reasonable changes in \( w \) (e.g., \( w = 20 \) days produces very similar results).

More active after exploring the spinoff community

We now apply this framework and examine how explorers behave in general. Surprisingly, we find that explorers are relatively more active compared to nonexplorers, i.e., the exploration effect is generally positive. We then further split explorers based on their activity level and study how our observation differ depending on activity level.

Comparisons across categories. Figure 6a presents exploration effect results for categories in our Table 3 taxonomy.\(^8\)

Somewhat counterintuitively, we find that for most spinoffs, users who explore become more active in the original subreddit after exploring, compared to similarly active users who never interacted in the new community (see Figure 6a).

Interestingly, the magnitude of this result varies based on the spinoff pair considered. We observe that “action” explorers are around 10% more likely to increase their activity after exploring, for example. “place” explorers, on the other hand, are roughly 2% less likely to increase their activity.

Our possible explanation for this observation is that users who explore “action” communities are often seeking to actively engage with a topic in a fashion above and beyond simple discussion. For example, the subreddit Bitcoin (which focuses on high-level discussions of the cryptocurrency) and its spinoff pair BitcoinMining (which focuses on lower-level issues, e.g., hardware useful for mining Bitcoins) exhibits a difference in interaction ratio of roughly 10%. If a user explores BitcoinMining from Bitcoin, this is likely a strong indication of their interest in digging deeper into the topic itself. It’s possible that viewing Bitcoin through the perspective of BitcoinMining increases overall engagement with the topic, at least in the short term.

In contrast, exploring “place” subreddits does not result in increased home activity nearly as often. For example, Bitcoin has another spinoff pair, BitcoinUK, that has an exploration effect of roughly zero. We have previously seen in Figure 4b that “place” spinoffs share relatively few early participants with their parent communities. Taken together,

---

\(^8\)Results are only reported for categories with more than 4

spinoff pairs.
these observations suggest that users seeking place-specific communities are not necessarily interested in engaging more deeply with the topic, so much as in who they discuss the topic with or how the topic affects them.

A closer look at the pairs. Figure 6b presents the top and bottom 5 pairs in terms of exploration effect. It further demonstrates how our results may vary across different pairs. All 5 bottom pairs present significantly negative exploration effect, which shows that it is not always the case that explorers with the highest pre-activity level present a smaller difference from nonexplorers.

Looking at the bottom 5, it partly supported our above discussion regarding places. Indeed, in “place” related pairs, gtaonline pulled people from gta and so did canadaguns for guns. Among the top 5, there is an even spread among several categories including “learning” (gaybros vs. askgaybros), “action” (airsoft vs. airsoftmarket), and “medium” (cringe vs. cringepics). The surprising affix is true. Although it seems to suggest superiority and separation, explorers actually become more active in the original community in this case, too.

Discussion. Our findings resonate with Tan and Lee’s (2015) results that users who continually explore new communities are, on average, more active than users who don’t. However, no causal relationship can be established that explains this result: exploration does not necessarily cause increased activity. However, in our dataset, exploration appears to be a strong signal of interest level.

Variations between explorers with different activity levels

We have established that users tend to be relatively more active in the “older” community after exploration, and have examined the variation across different categories and pairs. However, how does this effect differ for users with different pre-interaction levels? One could imagine that activity level prior to exploration affects whether or not users are more active after exploring. For example, upon discovering an alternative community, it’s possible very active users might remain more attached to their home community, whereas relatively inactive users might not have the same level of commitment.

To address this question, we split users into pre-interaction quartile levels within their spinoff pair, so that the users with the least number of pre-interactions are put in bin one, and users with the greatest number of pre-interactions are put in bin four. We then compute exploration effect for users in each quartile.\(^9\)

Figure 7 presents the fraction of users who had more post-interactions than pre-interactions for, respectively, explorers and nonexplorers in several popular subreddit pairs. In general, the relative effects of exploration appear to be different based on how active users are, but there are complex and varied relationships between user activity level and how much defection matters; these relationships differ based on which spinoff pair is considered. Since we split users further into quartiles, the amount of data is not sufficient to reach conclusions for all pairs.

One relatively consistent pattern across pairs is that explorers with the highest pre-activity level usually have a smaller difference from the nonexplorers compared to explorers in the lowest quartile, as shown in the left three figures in Figure 7, although this is not true for Figure 7d.

The trend of how the fraction or the difference changes with different pre-activity quartile is even more complex. Consider the case of Figure 7a; this figure illustrates that for users with low activity levels (first/second quartiles) exploring is much more indicative of increased future activity than nonexploration effect, but plot \(p_e\) and \(p_{ne}\) separately in these plots under the same name.
not exploring, and the difference is much less apparent for users with high activity levels – exploring and not exploring are associated with more similar levels of activity for users in the third/fourth quartiles.

Note that other pairs exhibit different patterns. For Science vs AskScience (Figure 7b) and Android vs androidapps (Figure 7c), the most active users (those in the 4\textsuperscript{th} quartile) appear to experience a slight “dip” in terms of the exploration effect.

**Related Work**

While there has been considerable interest in the topic in the social sciences (e.g., Hurtado; Berry (1997; 1997)), the study of situations wherein users engage with multiple, distinct communities represents a relatively new but increasingly relevant research area for computer scientists. Indeed, Kim (2000) argues that a growing Web needs subdivisions, while Jones and Rafaeli (2010) also argue that an effective community splitting strategy is essential for virtual communities and online discourse to thrive. Furthermore, Birnholtz et al.’s (2015) study of confession groups on Facebook could be viewed in the context of “place” style affixes. While our study has a different focus, our findings mirror those of Tan and Lee (2015) who found that users who tend to remain active on the Reddit platform (as an example of a meta-community) tend to continue to explore new sub-communities continuous throughout their “lifetime” on the site.

A number of studies have examined multi-community platforms in different contexts. Subcommunity survival (Turner et al. 2005; Iriberri and Leroy 2009; Kraut et al. 2012) is sometimes framed in the context of a meta-community. Also, Fisher et al. (2006) find that different newsgroups exhibit different conversation patterns, though they don’t examine if the same users behave differently across platforms (as in Vaisálescu, Filkov, and Serebrenik (2013)). Finally, Adamic et al. (2008) examine the quality of user answers across different categories of Yahoo Answers.

Despite exhibiting some undesirable upvoting patterns (Gilbert 2013), Reddit itself has been used as a data source in various contexts. For instance, the study of altruistic requests (Althoff, Danescu-Niculescu-Mizil, and Jurafsky 2014), the study of domestic abuse discourse (Schrading et al. 2015), and work about post titles (Lakkaraju, McAuley, and Leskovec 2013) demonstrate that useful information can be learned from Reddit comments and upvotes.

**Conclusion**

In this work, we use a dataset of all posts and comments from Reddit over an eight-year period to explore the space of naming affixes that lead to highly related communities on the platform. After building a taxonomy, we examine the early participants and other temporal aspects of the pairs, and introduce the idea of a spinoff community being “born out” of its unaffixed parent. Finally, we present the surprising result that users who explore in spinoff communities generally become relatively more active in their home communities instead of being “distracted”. We also find that the magnitude of this effect (and sometimes its sign) depend on the type of community pair and how active a user was prior to exploration.

There are several directions for possible future work. First, it would be interesting to examine more closely the origins of highly related communities. If a community is created because of a disagreement (e.g., Zachary’s Karate Club (1977)) one could potentially identify general characteristics of increasing unrest prior to a fission. Also, it would be interesting to delve deeper into differences between discourse on content in highly related communities pairs; how does discussion on TrueAtheism differ from discourse on Atheism, for example. It would be useful for community organizers if we can detect when a spinoff community is necessary or beneficial. Furthermore, it is an important direction to understand the mechanism behind our observation that users who explore in spinoff communities generally become relatively more active in their home communities. This could be potentially useful for community organizers to identify complementary communities.

Finally, we note that our consideration has presupposed a pairwise framing, i.e., we always assumed a pair of communities. In some cases, we noted more complex phenomena underlying community creation. For example, a number of srs communities were all created in a short period of time. Also, the world of pokemon subreddits may consist of multiple affixes that lead to different subdivisions. In general, one could generalize pairwise interactions to explore more complex relationships between communities.

**Acknowledgments.** We thank Cristian Danescu-Niculescu-Mizil, David Mimno, Mor Naaman, Skyler Seto, Tianze Shi, Justine Zhang and the anonymous reviewers for helpful comments and discussions. This work was supported in part by a Facebook fellowship and a Google Research Grant.

**Appendix: Sampling Method for Control Users**

The goal of this section is to describe how we sample a control user \( u_{ne} \) corresponding to each each exploring user \( u_e \). Ultimately, to compute the exploration effect, we need to find someone who never posts in the new subreddit, but takes a similar number of actions in the same time period. To choose this similarly active, nonexplorer user, we sample \( u_{ne} \) as follows:

1. From the set of all nonexplorer users, find the subset who have an interaction in the original community within 24 hours of \( u_e \)’s exploration time \( t \). Let these interactions occur at time \( t' \). If a nonexplorer user has more than one interaction between \( t - 24 \) hours and \( t + 24 \) hours, take the closest to \( t \).

2. Find the user \( u_{ne} \) in this candidate set that minimizes the difference between their own number of pre-interactions (re-centered at their \( t' \)) and \( u_e \)’s. Specifically, if we let \( p(u, t_a, t_b) \) be the number of interactions of user \( u \) in the original subreddit between \( t_a \) and \( t_b \) we find the loyal user

\[
\arg\min_{u_{ne}} |p(u_e, t - w, t) - p(u_{ne}, t' - w, t')|.
\]
3. If this difference is less than 5% of $u_e$’s pre-interactions, a similarly active user $u_{ne}$ has been successfully sampled.

Figure 5 demonstrates a pair of users that could be plausibly sampled in this manner. Both the light and dark users have the requisite 5 pre-interactions, and the light user makes a post within 24 hours of the dark user’s first exploration.

After sampling $k$ such user pairs $\{\langle u_{i,e}, u_{i,ne} \rangle \}_{i=1}^{k}$ for a given pair of subreddits$^{10}$, we first compute the proportion of exploring/nonexplorer users whose activity increased, i.e., have more post-interactions than pre-interactions. For instance, this fraction for exploring users is computed as

$$p_e := \frac{1}{k} \sum_{i=1}^{k} I[\text{post}(u_{i,e}) > \text{pre}(u_{i,e})] .$$ (1)

Finally, for each spinoff pair of communities, the quantity we are interested in is

$$p_e - p_{ne} .$$ (2)

We generally call the quantity given in Equation 2 the “exploration effect”. A larger exploration effect indicate that an explorer is more active in the original subreddit after posting to the splinter subreddit, when compared to a similarly active nonexplorer. In Figure 7, we plot $p_e$ and $p_{ne}$ separately, whereas in Figure 6 we plot $p_e - p_{ne}$.

References

Adamic, L. A.; Zhang, J.; Bakshy, E.; and Ackerman, M. S. 2008. Knowledge sharing and Yahoo answers: Everyone knows something. In Proceedings of WWW, 665–674.

Althoff, T.; Danescu-Niculescu-Mizil, C.; and Jurafsky, D. 2014. How to ask for a favor: A case study on the success of altruistic requests. In Proceedings of ICWSM.

Backstrom, L.; Huttenlocher, D.; Kleinberg, J.; and Lan, X. 2006. Group formation in large social networks: Membership, growth, and evolution. In Proceedings of KDD, 44–54.

Berry, J. W. 1997. Immigration, acculturation, and adaptation. Applied Psychology 46(1):5–34.

Birnholtz, J.; Merola, N. A. R.; and Paul, A. 2015. “Is it weird to still be a virgin”: Anonymous, locally targeted questions on Facebook confession boards. In Proceedings of CHI.

Danescu-Niculescu-Mizil, C.; West, R.; Jurafsky, D.; Leskovec, J.; and Potts, C. 2013. No country for old members: User lifecycle and linguistic change in online communities. In Proceedings of WWW, 307–318.

Ducheneaut, N.; Yee, N.; Nickell, E.; and Moore, R. J. 2007. The life and death of online gaming communities: A look at guilds in World of Warcraft. In Proceedings of CHI, 839–848.

Fisher, D.; Smith, M.; and Welser, H. T. 2006. You are who you talk to: Detecting roles in usenet newsgroups. In Proceedings of HICSS. IEEE.

Gilbert, E. 2013. Widespread underprovision on Reddit. In Proceedings of CSCW, 803–808.

Hessel, J.; Schofield, A.; Lee, L.; and Mimno, D. 2015. What do vegans do in their spare time? Latent interest detection in multi-community networks.

Hurtado, A. 1997. Understanding multiple group identities: Inserting women into cultural transformations. Journal of Social Issues 53(2):299–327.

Iriberri, A., and Leroy, G. 2009. A life-cycle perspective on online community success. ACM Computing Surveys (CSUR) 41(2):11.

Jones, Q., and Rafaeli, S. 2010. Time to split, virtually: ‘Discourse architecture’ and ‘community building’ create vibrant virtual publics. Electronic Markets 10(4):214–223.

Kairam, S. R.; Wang, D. J.; and Leskovec, J. 2012. The life and death of online groups: Predicting group growth and longevity. In Proceedings of WSDM, 673–682. ACM.

Kim, A. J. 2000. Community Building on the Web: Secret Strategies for Successful Online Communities. Addison-Wesley Longman Publishing Co., Inc., 1st edition.

Kraut, R. E.; Resnick, P.; Kiesler, S.; Burke, M.; Chen, Y.; Kittur, N.; Konstan, J.; Ren, Y.; and Riedl, J. 2012. Building successful online communities: Evidence-based social design. Mit Press.

Kumar, R.; Novak, J.; and Tomkins, A. 2010. Structure and evolution of online social networks. In Link mining: models, algorithms, and applications. Springer. 337–357.

Lakkaraju, H.; McAuley, J. J.; and Leskovec, J. 2013. What’s in a name?: Understanding the interplay between titles, content, and communities in social media.

Schrading, N.; Alm, C. O.; Putcha, R.; and Homan, C. 2015. An analysis of domestic abuse discourse on Reddit. 2577–2583.

Singer, P.; Flöck, F.; Meinhart, C.; Zeitfogel, E.; and Strohmaier, M. 2014. Evolution of Reddit: From the front page of the Internet to a self-referential community? In WWW (Companion).

Tan, C., and Lee, L. 2015. All who wander: On the prevalence and characteristics of multi-community engagement. In Proceedings of WWW, 1056–1066.

Turner, T. C.; Smith, M. A.; Fisher, D.; and Welser, H. T. 2005. Picturing Usenet: Mapping computer-mediated collective action. Journal of Computer-Mediated Communication 10(4):00–00.

Vasilescu, B.; Filkov, V.; and Sererbreinik, A. 2013. StackOverflow and GitHub: Associations between software development and crowdsourced knowledge. In Proceedings of SocialCom, 188–195. IEEE.

Zachary, W. W. 1977. An information flow model for conflict and fission in small groups. Journal of Anthropological Research 33(4):452–473.

$^{10}$We discard the pair of communities if $k < 100$. 