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A social network caught in the Web

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Abstract

We present an analysis of Club Nexus, an online community at Stanford University. Through the Nexus site we were able to study a reflection of the real world community structure within the student body. We observed and measured social network phenomena such as the small world effect, clustering, and the strength of weak ties. Using the rich profile data provided by the users we were able to deduce the attributes contributing to the formation of friendships, and to determine how the similarity of users decays as the distance between them in the network increases. In addition, we found correlations between a user's personality and their other attributes, as well as interesting correspondences between how users perceive themselves and how they are perceived by others.

Introduction

Community Web sites are becoming increasingly popular – allowing users to chat, organize events, share opinions and photographs, make announcements, and meet new friends. Several prior studies have focused on characterizing these online interactions (Curtis, 1992; Yee 2001), and others have attempted to measure the effect of the Internet on real life social interactions (Wellman et al., 2002a and 2002b). Our study has a somewhat different focus: while we can learn much about the online community itself, we are more interested in gleaning from it insights about the underlying real world social networks.

The community we chose for our study is Club Nexus. Club Nexus was introduced at Stanford in the fall of 2001. It is a system devised by students to serve the communication needs of the Stanford online community. Students can use Club Nexus to send e-mail and invitations, chat, post events, buy and sell used goods, search for people with similar interests, place personals, display their artwork or post editorial columns. Within a few months of its introduction, Club Nexus attracted over 2,000 undergraduates and graduates, together comprising more than 10 percent of the total student population.

The electronic nature of online community participation presents an opportunity to study human behavior and interactions with great detail and on an unprecedented scale. Traditional methods of gathering information on social networks require researchers to conduct time consuming and expensive mail, phone, or live surveys. This limits the size of the data sets and requires additional time and effort on the part of the participants. When studying an online community, our ability to learn more about the social network is simply a side-effect of users transmitting information digitally.

Previously we were able to analyze a portion of the Stanford social network reflected in the homepages of Stanford students and the hyperlinks between them (Adamic and Adar, 2003). Our finding that personal homepages can create a large social network was an inspiration for Club Nexus. Because users are explicitly asked to name their friends, Club Nexus is more

densely connected than the homepage network where users link to their friends of their own accord. The structured format of the profiles lends itself to easier statistical analysis than the free-form text of personal homepages. The data presents an opportunity to study, among other things, the online community's structure, social interactions and how factors such as personality and interests influence one's choice of friends. In this paper we take the first step of analyzing the community as a social network, and compare profiles supplied by the users to characterize the connections.

User Registration and Data

Upon registering, users were required to supply their names, e-mail addresses, birthdays (for birthday reminder notifications to their friends), major, graduate or undergraduate status, year in school, residence, and home country and state. They could also optionally list the high school (and college if they are graduate students) that they attended, as well as their phone number, hometown, homepage and picture. The data that we used in all of our analysis was anonymized, with user names replaced by unique ID's and only year, graduate or undergraduate status, and department retained from the above information. All results of our study are presented in aggregate to further ensure the users' privacy.

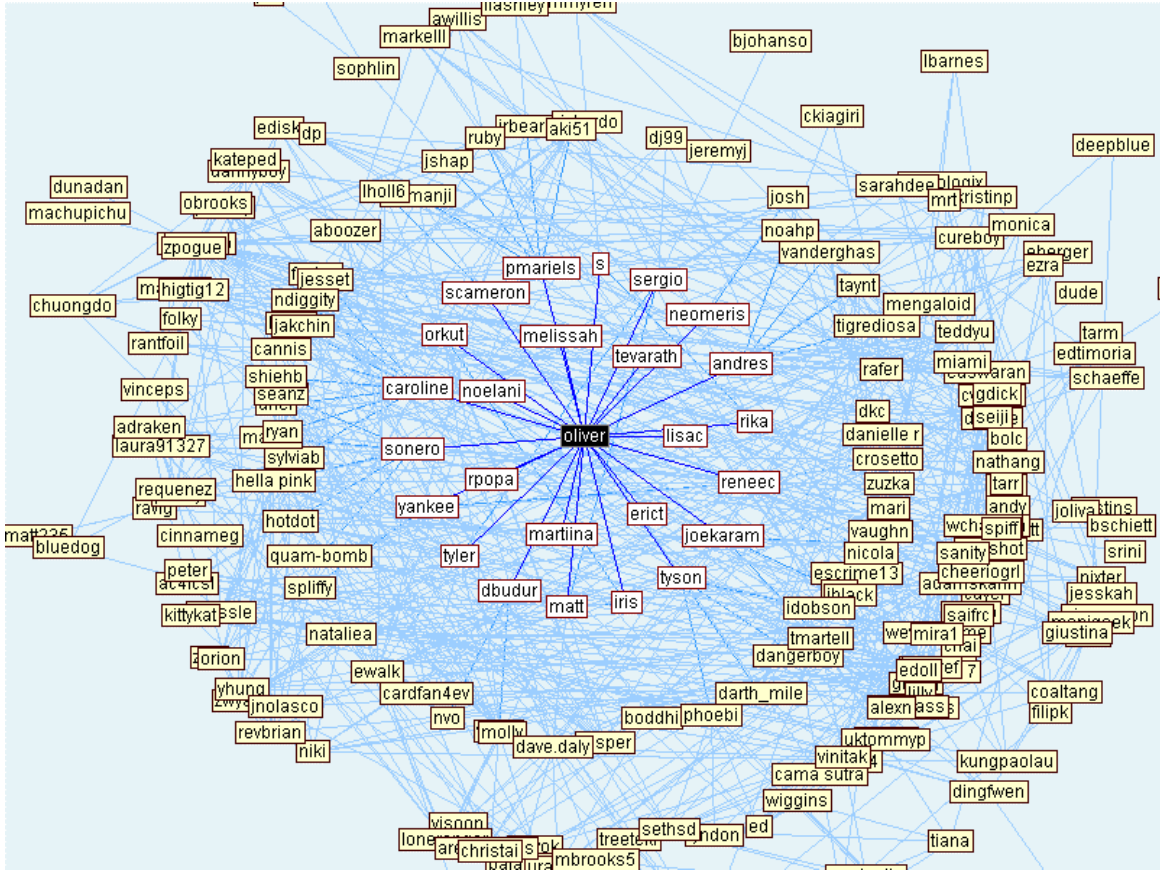


Figure 1. Nexus Net as seen from a single user perspective. Clicking on any of the nodes re-centers the graph around that user.

In the second registration step, users were asked to list their friends and acquaintances at Stanford. In ‘Nexus-speak’ these people are called ‘buddies’. Users identified their buddies by searching for them in the Stanford directory or by entering their names manually. If a user adds a buddy who is already registered, the buddy will get a notification that the user has requested to be their buddy and can accept or decline the request. If the ‘buddy’ is not yet registered, they will get an invitation to join Club Nexus. This viral sign-up strategy resulted in a rapid build-up of the user base. Several months after Club Nexus was introduced, users were given the opportunity to rank how ‘trustworthy’, ‘nice’, ‘cool’, and ‘sexy’ their buddies were. This added a new dimension to the interaction data.

In addition to basic demographic information users were asked to add a list of interests and hobbies to their profile by checking off as many choices as they liked from listings of social activities, sports, movies, music, and books. These choices could then be used by Club Nexus to match up users with similar preferences. In a final step, users were asked to select 3 items from lists of adjectives to describe their personalities, the kinds of people they turn to for support, the ways they like to spend their free time, and what they look for in friendship and romance.

The resulting dataset was a social network with rich profiles for each of the members. In the following sections we first analyze Club Nexus from a network perspective and then look at the relationship between the user attributes and their choices in contacts.

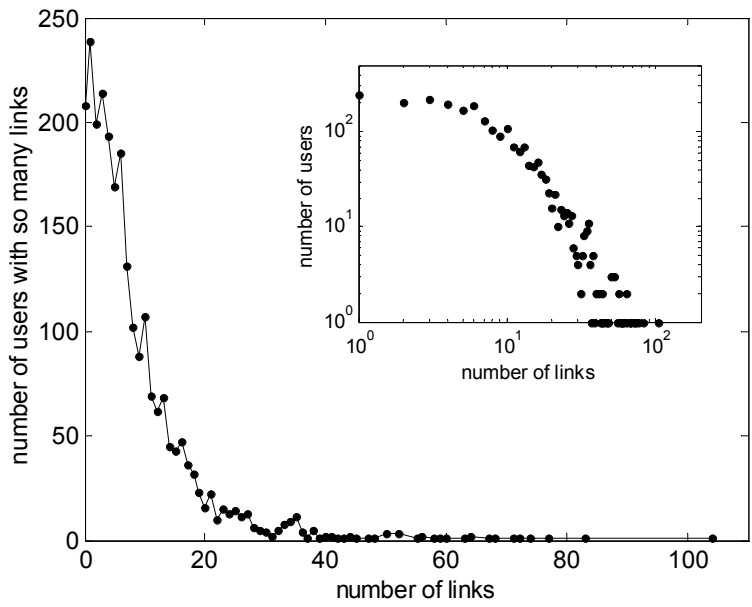


Figure 2. The number of connections users have; a single buddy being the most common case. The inset shows the same distribution on a log-log scale.

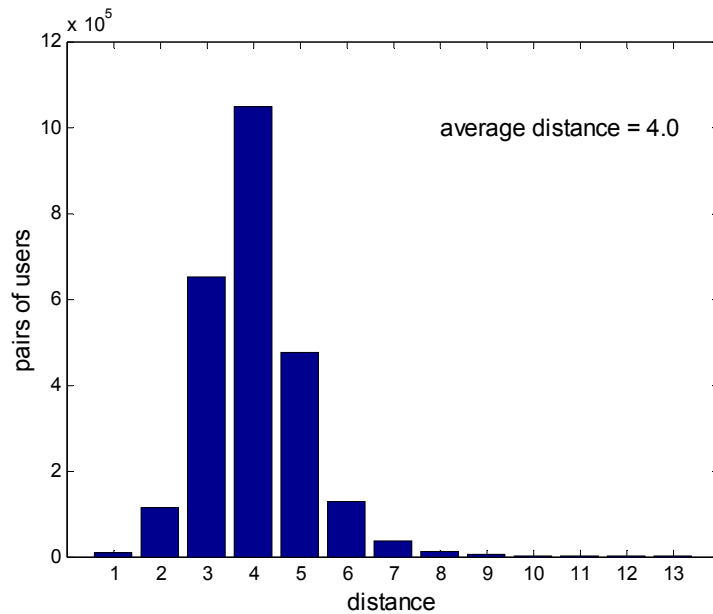


Figure 3. Distribution of user to user distances, with a mode at 4 hops.

Network Analysis

The ‘Nexus Net’, a large social network, consists of 2469 Nexus users and 10119 links between them, two individuals being linked if they include each other on their buddy lists. Users can browse the network using the visual interface shown in Figure 1 and can automatically contact their neighbors out to some radius. For example, to organize an event, they can invite just their friends or the friends’ friends.

As is typical of both social networks in general and online communities in particular, the number of buddies a user has is distributed highly unevenly. Figure 2 shows that users most frequently listed just one buddy (over 200 listed no buddies), but some individual users had dozens of connections, and one had even exceeded a hundred. Part of the skewness in the connectivity distribution is due to the fact that some people are naturally more social than others, but it also reflects a varying eagerness on the part of users to enter their social contacts into an online service. In general, we expect that most Club Nexus users have more friends offline than just those that they list as their buddies with the service.

In analyzing the social network we observed a small world effect (Migram 1967; Watts and Strogatz 1998), where the distance between any two users, measured in the number of hops along the Nexus Net, is only 4 on average (the full distribution is shown in Figure 3). This average might at first seem low in view of the fact that Club Nexus represents a diverse group of users, both undergraduates and graduates at various stages in their studies representing many departments. This is the counterintuitive aspect of the small world phenomenon: individuals tend to socialize in smaller cliques, often determined by factors such as year in school, department or dorm, yet any two users are separated by only a small number of hops. One can determine to what degree cliques are present by measuring the amount of clustering, also sometimes referred to as transitivity (Newman 2001). The clustering coefficient tells us how many of a user’s friends’ friends were friends of the user themselves. The clustering coefficient of Club Nexus is 0.17, 40 times higher than it would be for a random network with the same number of users and

connections. The clustering coefficient tells us that there is a significant amount of structure in the social interactions reported in Club Nexus. The apparent conflict between clustering and short paths was resolved by Watts and Strogatz (1998). They used a simple model of social networks to show that as long as there is a small fraction of ‘random’ connections between cliques, social networks could display both high clustering and small average shortest paths.

While the above analysis of the network topology is insightful, things become even more interesting when user profiles are taken into account. We will explore these profile features in the next section and will later return to their impact on network properties.

Properties of Individual Profiles

Profile Data and Statistical Tools

In the process of registering users were asked to list aspects of their personality by selecting 3 words out of a choice of 10 to 15 describing their personalities, what they look for in friendship and romance, how they spend their free time and what kind of people they turn to for support. All users completed this section as it was required for initial registration. Users were also asked to optionally express their preferences about book and movie genres, indoor, outdoor and water sports, and other activities. The 418 (of the 2469) users who did not make a selection in any category were omitted from the analysis regarding preferences.

We used Z-scores to characterize the relationships between different attributes the users chose. Z-scores indicate how likely it is to find a connection between two attributes by chance. Specifically, the probability that a Z-score falls above 2 or below -2 is 5 percent. So we can say that any correlation with an absolute Z-score greater than 2 is significant at the $p=.05$ level. For example, if we are interested in whether people who consider themselves funny enjoy watching comedies, we count the number of people (518 in all) who selected ‘funny’ as one of the 3 descriptive words for themselves. We then count the number of users (1511 out of 2051 that specified their interests) who selected comedies as a movie genre they liked. This gives a probability $p = 0.74$ that a randomly chosen user likes comedies. It then follows that of the 518 ‘funny’ users, an expected $p*518=382$ would enjoy comedies with a standard deviation of 10. We observe that in actuality, 416 users who think they are ‘funny’ also enjoy comedies. This gives us a Z score of $((\text{number observed})-(\text{number expected})) / (\text{standard deviation}) = 3.43$. The probability that this occurs by chance, that is, that there is no connection between whether users consider themselves funny and whether they like comedies is 0.0003. Hence, although the difference is slight (about 10 percent more funny users like comedies than one would expect from a random sample), the effect is significant. From here on, when we write that ‘users possessing quality A tend to like B’, we simply mean that the proportion of users having A and liking B is significantly different than the proportion of users overall who like B. In no way do we mean to say that all users having A are a certain way. Wherever practical, we’ve included the difference between observed and expected quantities in the tabulated results in the appendices. We would also like to remind the reader that the results pertain only to the Stanford community, which is not necessarily representative of the population overall.

Personality and Preferences

We used this kind of analysis to find correlations between users’ personalities and their preferences. Due to the large number of pairings of personality and preference, a few of the

relationships may be found statistically significant by chance. But since so many pairings were found to be statistically significant, a majority of them represent true tendencies that paint reasonable portraits of personality types.

Using this technique we found that users tended to be consistent in how they described themselves and what they looked for in others. For example, those who described themselves as sexy were more likely to look for sex in both friendship and romance. Those who described themselves as attractive thought appearance and looks were important. Those who thought themselves to be funny sought laughter both in friendship and romance.

Statistical correlations between personalities and preferences aligned for the most part with stereotypes pertaining to those personalities. Individuals labeling themselves as ‘weird’ tended to have ‘weird’ friends and were more likely to prefer spending their free time alone and staying at home, not ‘doing anything exciting’ or ‘doing physically challenging activities’. They are more likely to enjoy science fiction and fantasy books and movies, heavy metal, and computer gaming. They don’t especially value looks in relationships and don’t tend to describe themselves as fun, attractive or successful.

On the other hand, those who described themselves as “successful” spent their free time fulfilling commitments and catching up on chores. They also placed an emphasis on appearance and sex in romantic relationships and friendships and liked to spend their time doing physically challenging activities, including weightlifting, tennis, boating, jet and water skiing. They are also three times more likely to read business books.

For a complete list of all significant relationships between personality and preferences the reader may consult Appendix A. The appendix also lists some interesting correlations that appear between an absence of a characteristic and the person’s choices. For example, those users who did not select the word ‘responsible’ to describe themselves include individuals who enjoy books on sex, erotic, gay and lesbian, and independent movies, listen to funk, jungle, reggae, and trance, and enjoy skateboarding and raving.

Academic Major and Personality

We also examined the relationship between a person’s academic major or department and what adjectives (three from a list of sixteen) they selected to describe themselves. Because there are many different majors, the data were spread out thinly. We were still able to glean a few statistically significant trends, shown in Table 1. Physics, math, and electrical engineering majors stayed true to a “nerdy” stereotype, being approximately twice as likely to spend their free time learning and to describe themselves as ‘weird’.

Thirteen of the 29 Public Policy majors (double the average proportion) described themselves as kind, while a high number of the 62 Political Science majors thought they were attractive (29 vs. 16 percent) and lovable (24 vs. 12 percent). Those who had not yet declared a major (presumably freshmen) were most amiable to ‘doing anything exciting’ (209 out of 337). On the other hand, the 46 history majors were three times as likely to enjoy spending their free time at home. Unsurprisingly, the 74 English majors were twice as likely to enjoy spending their free time reading and to consider themselves creative. They were also twice as likely to describe

themselves as sexy (18 percent), while on the other hand, only 3 of the 136 Electrical Engineering majors chose to describe themselves in that way.

	personality (% of total)	major
free time	learning (17%)	Physics (46%), Philosophy (37%), Math (31%), EE (26%), CS (24%)
	reading (26%)	English (55%)
	staying at home (8%)	History (24%)
	doing anything exciting (52%)	undecided/undeclared (62%)
	fulfilling commitments (16%)	
	watching TV (17%)	
you	intelligent (32%)	Philosophy (59%), CS (42%)
	successful (4%)	CS (7%)
	socially adaptable (14%)	STS (46%)
	attractive (16%)	Political Science (29%), International Relations (25%)
	lovable (12%)	Political Science (24%)
	kind (25%)	Public Policy (45%)
	weird (12%)	Physics (34%), Math (28%), EE (18%)
	fun (26%)	Human Biology (38%)
	creative (22%)	Product Design (62%), English (42%)
	sexy (8%)	English (18%)

Table 1. Personality traits and positive correlations to majors.

Gender Differences

We next examined how gender influences personality and preferences. While most differences were slight (as shown in Appendix B), typically in the range of 5-10 percent, some were quite marked such as the fact that twice as many women as men liked to read romance novels. Although one cannot say that *all* women or *all* men are a certain way, for the most part these slight tendencies conformed to existing stereotypes of gender differences.

More men than women enjoy computer, science fiction, professional, technical, science, and business books. More women than men enjoy romance novels, fiction, books about health, mind and body, cooking and art and photography. More men favor football, frisbee golf, table tennis, and golf, while more women prefer gymnastics, field hockey, and softball. More men enjoy science fiction, war, and action movies, as opposed to the romance, family and drama movies women like to watch. More men indicated that they like to spend their free time learning and doing physically challenging activities, while more women said that they like to catch up on chores and socialize. Men preferred friends with mutual acquaintances and common interests, while women valued laughter, honesty and trust. Women looked for the same characteristics in romantic partners, but men were more likely than women to appreciate appearance, sex, and physical attraction. When turning to someone for support, some men gravitated to extremes, turning to ‘eternal optimists’ or the ‘give-it-to-you-straight’ people. Women sought support of a more emotional kind and turned to the ‘unconditional accepters’ and the ‘listeners’. Finally, more men than women described themselves as intelligent, while more women than men thought they were fun, lovable and friendly. This may be more indicative of the men’s propensity to boast than true intelligence, because there is no confirmed relationship between overall intelligence and gender (Halpern, 2000).

Association by Similarity

Many studies have confirmed the tendency of people to share common interests with their social contacts (Lazarsfeld and Merton, 1954; Touhey, 1974; Feld, 1981). We took advantage of the richness of the Club Nexus dataset to see what common interests or traits most influenced friendship. To this end, we used a quantity we termed ‘association ratio’ to measure network homophily. For a given trait, the association ratio is the proportion of contacts made between people sharing a trait to the proportion of individuals in the population possessing the trait. For example, 329 or 16 percent of the users indicated that they liked ballroom dancing and they had 2727 buddy links. If one’s selection of friends were independent of their enjoyment of ballroom dancing, then 16 percent or 437 of the links would be to other ballroom dancers. However, a full 704 of the links stay within the group of ballroom dancers. This gives us a ratio of 1.61 as the strength of association between ballroom dancers. We also calculate a Z score to confirm that the ratio is not likely to have occurred by chance.

Nearly all interests showed a statistically significant tendency of those individuals sharing them to associate with one another (for detailed results see Appendix C). We found further that, in general, activities or interests that are shared by a smaller subset of people showed stronger association ratios than very generic activities or interests that could be enjoyed by many. For example, raving (1.64), ballroom dancing (1.61), and Latin dancing (1.49) showed stronger association in the social activity category than barbecuing (1.20), partying (1.18), or camping (1.11), although all had very high Z-scores. In sports in particular, multi-player team or niche sports were better predictors of social contacts than sports that could be pursued individually or casually. Among water sports, synchronized swimming, diving, crew, and wake boarding were better predictors than boating, fishing, swimming or windsurfing. In the land sports category, team sports, in particular women’s team sports such as lacrosse and field hockey were better predictors than soccer (often played casually as opposed to in a competitive college team), tennis, or racquetball. In the ‘other sport’ category, niche or extreme sports such as freestyle biking, skateboarding, freestyle frisbee, ultimate frisbee, and sky diving are more predictive than sports that have wider appeal such as backpacking, weightlifting, aerobics, jogging, hiking, snow skiing, martial arts, or bicycling.

We observed that niche book, movie, and music genres were more predictive of friendship than generic ones. Gay and lesbian books, read by 63 users, had a ratio of 4.37, followed by professional and technical, teen, and computer books. In contrast, the general category of ‘fiction & literature’ had a ratio of 1.09.

Specific movie genres such as gay and lesbian, performing arts, religion and erotic & softcore had higher scores than genres that appeal to a wider audience such as action, drama, mystery, documentary or comedy. Non-mainstream music genres like gospel, jungle, bluegrass/rural and heavy metal were more predictive than jazz, pop, classical and rock.

We also checked for homophily in the users’ self-described personalities (see Appendix D). Users who described themselves as ‘sexy’, ‘talented’, ‘fun’, ‘weird’, or ‘lovable’ liked to associate with those who described themselves likewise. We know from the previous analysis that those who describe themselves as ‘sexy’ are more likely to value sex in friendships and romance. It makes sense therefore that they would like to associate with other sexy people. We

did not, however, observe homophily for individuals who described themselves as ‘intelligent’, ‘responsible’, ‘kind’, ‘competent’ and ‘successful’. Unsurprisingly, those who like to spend their free time fulfilling commitments and socializing preferentially associate with others who like to do the same. In contrast, users who like to stay at home or be alone do not preferentially associate with other loners.

One observation we made concerning the relationship between a user’s profile and their social network is that listing more preferences and interests correlates slightly ($\rho=0.2$) to the number of buddies listed with Club Nexus. There are two possible explanations: 1) Users who invested the time to enter their friends into the database would also take the time to list more of their interests and activities. 2) More active users also maintain more social contacts. Unfortunately, the manner in which the data were collected does not allow us to differentiate between the two.

Finally, we were able to use the user’s profiles and their positions in the network to test the weak link hypothesis (Granovetter 1973). It states that connections between dissimilar individuals are important in creating cross-community links. We calculated the betweenness of an edge: how many shortest paths pass through it (Freeman, 1977; Wasserman and Faust, 1994). We then compared the betweenness of the edge to how similar the two individuals sharing the edge were, based on the overlap of their profiles. We found a negative correlation coefficient $\rho = -0.2$, meaning that interactions between dissimilar people play a role in making the average distance between any two users in the community shorter. One should also not underestimate the role of highly connected individuals. There is a very strong correlation ($\rho = 0.77$) between the betweenness of an individual and the number of buddies they have. Users with many friends naturally serve as a social bridge, and their friends are less likely to all form one social clique, which is indicated by a negative correlation ($\rho = -0.12$) between an individual’s betweenness score and the clustering coefficient for their friends.

Similarity and Distance

So far we have established that people who share interests or characteristics are more likely to be friends than those who don’t. We take this a step further by examining how similar people are on average to each other as a function of their separation in the Nexus Net. In Figure 4 we compare what fraction of an individual’s first, second, third, fourth etc. neighbors share the same attribute such as department and year in school as the individual. We find that the similarity drops off rapidly for most categories, that is, there is a much higher likelihood that we share a characteristic with a friend or a friend’s friend than that we share it with someone 4 steps removed.

Specifically, we find that the year of study is much more important for undergraduate students than for graduate students, but that the department is more important for graduates than a major is for undergraduates. This can be explained by the observation that undergraduate students take many required classes with others in their class, but graduate students usually spend most of their time interacting with individuals in their research group and sometimes collaborate with others in their department. The courses that they take tend to be more specialized and will usually expose them primarily to other graduate students in their own field. Finally, we find that attributes such as tastes in books and movies also show a decay in similarity with increasing distance in the

network. The effect is much reduced, possibly because these variables do not influence to the same extent how and with whom students spend their time.

Nexus Karma

Several months after Club Nexus was launched, Nexus Karma was announced by e-mail as a new feature. Users were given the opportunity to rank how ‘trusty’, ‘nice’, ‘cool’, and ‘sexy’ their buddies were on a scale of 1 to 4. One could not pick and choose which buddies to rank, but rather had to rank all of them at once. After a week, users who had been ranked by at least 3 buddies were themselves sent an e-mail asking them to rank their buddies in turn. There was a tremendous response to this, with 446 users ranking 1735 different friends. This data allowed us to step beyond users’ self-perceptions and allowed us to integrate user’s perceptions of each other into the network data.

We found that users had a tendency to give a similar score to a buddy across all categories. That is, users tended to rank their friends as ‘3, 3, 3, 3’ as opposed to ‘1, 4, 2, 3’. This resulted in a high correlation coefficient between the different attributes. There were still, however, perceptible differences in the scores given. Users on average received the highest scores for niceness (3.37) and trustiness (3.22), followed by coolness (3.13) and sexiness (2.83). While pairs of dissimilar attributes such as ‘trusty--sexy’ or ‘nice--sexy’ had a lower correlation coefficient of 0.4, the pairs of attributes ‘trusty-nice’ and ‘cool-sexy’ had higher correlation coefficients of 0.7. This indicates that although users had an overall opinion about their buddies, they tended to associate trustiness with niceness and coolness with sexiness.

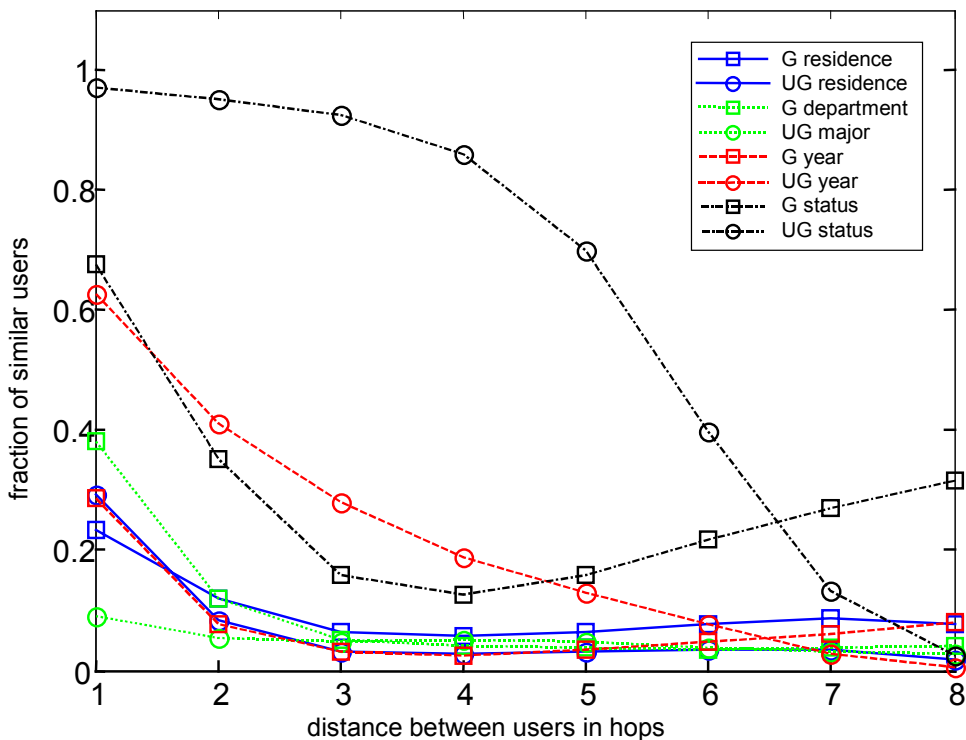


Figure 4. Average fraction of users with a common trait (year, undergraduate or graduate status, etc.) as a function of the distance from a user having that trait. The plot is truncated at 8 hops because less than .03% of the pairs are separated by more than 8 hops.

We found mild or negligible correlation between a person's average ranking in each category and the number of buddies that they have. This negates the hypothesis that people perceived as cool or nice have more friends. Interestingly, we found a slight negative relationship ($\rho \sim -0.1$) between the number of buddies a person has and the average 'trusty', 'nice', and 'cool' scores that they gave them. A simple interpretation is that those who list only a few of their friends with Club Nexus tend to list their closest ones, those they would rate most highly. Users who list a large number of friends are more likely to include those that they don't have the highest opinion of.

We did find interesting correlations between the ratings users received from others and the adjectives that they chose to describe themselves. We used a t test for two sample means to see if the average ranking in a category differed at the 1 percent significance level between those who did and did not choose a particular adjective to describe themselves. A few adjectives displayed a slight, but significant, difference. For example, those who described themselves as responsible received higher (3.36 on average vs. 3.23 for those not describing themselves as responsible) 'trusty' scores on average, but scored slightly lower in the 'cool' (3.02 vs. 3.13) and 'sexy' (2.67 vs. 2.85) categories. The reverse was true of those who described themselves as 'attractive' or 'sexy'. They were ranked more highly on average in the 'sexy' category, but fared worse in the 'trusty' and 'nice' categories. As one would expect, 'friendly' and 'kind' users received higher scores in the 'nice' category, while 'kind' people were also ranked as more 'trusty'. Users who described themselves as 'weird' received lower 'sexy' scores, while 'funny' people were perceived as less 'nice'. This not only demonstrates a clear correspondence between the way that individuals perceive themselves and the way that they are perceived by others, but also an interesting dichotomy between desirable qualities such as being funny or attractive and whether people possessing those qualities are perceived as nice.

We were also interested in the reasons why individuals gave the rankings that they did. One might expect that nicer people are more generous with their judgments. Indeed, the higher a user's 'nice' score, the higher the 'trusty', 'nice', and 'cool' scores ($\rho=0.14-0.17$) they give to their friends. Similarly, the higher a user's 'trusty' score, the higher the 'trusty', 'nice', 'cool', and 'sexy' scores that user gives to others ($\rho=0.14-0.20$). We also found evidence that some friendships are closer than others. For example, users who share friends (and hence belong to the same clique) are more likely to give each other high scores ($\rho=0.10-0.13$). We further found that users tend to reciprocate their 'trusty' and 'nice' scores, meaning that if user A gives user B a higher than average score, then user B is somewhat more likely to do the same for user A. Note that users' ratings of one another are independent because they are not told, except in aggregate, what score their friends have given them. Users did not however seem to reciprocate on their 'cool' and 'sexy' opinions.

These are only some of the insights that can be gleaned from the Nexus Karma data set. We hope to study it in greater detail in future work.

Conclusions and future work

We have presented a preliminary social network analysis of the Club Nexus online community. The online community in many respects appears to reflect the underlying community structure at

Stanford University. The size of the network allowed us to study phenomena such as the small world effect and the strength of weak ties, while the richness of the profiles allowed us to characterize social ties and identify what factors influence friendships.

Our analysis was able to detect many expected trends (e.g. English majors liking to spend their free time reading or people sharing a narrow or unusual interest becoming friends), while at the same time finding non-obvious relationships (e.g. 'responsible' people being perceived as slightly less 'cool'). What makes Club Nexus special is that one is able to observe these patterns on a large scale with many different variables. The richness of this information can be used to model dynamics such as the spread of ideas on a network or the way that people can find each other through their contacts. The ranking data from Nexus Karma can help us better understand reputation mechanisms now used by online retailers (Resnick and Zeckhauser: 2002).

As the Club Nexus community evolves, there will be opportunity to study the changes in the network over time, as well as to analyze social dynamics such as the adoption of a new feature introduced at the web site. Whereas tracking social networks over time by traditional methods such as telephone or live interviews is very expensive and time consuming, studying online communities is relatively effortless but may provide new and valuable insights.

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Appendix A

Table A1. Correlations between a user's personality and their preferences		
personality	preference/ activity	item (Z score, number of individuals who selected both the personality trait and the item, the number expected if random)
attractive	book	business (4.34,48,26.5)
	movie	erotic & softcore (3.09,48,31.5)
	music	disco (3.04,51,34.1)
	other	weightlifting (4.64,102,67.8)
	social	bar-hopping (5.43,154,107.4), clubbing (5.83,190,137.3), hip-hop dancing (4.20,121,87.2), hot tubbing (5.14,157,112.4), partying (5.55,246,195.4)
	watersport	crew (3.99,30,14.9), diving (3.32,20,9.8), jet skiing (3.41,68,46.4), scuba diving (3.18,62,42.6)
creative	book	art & photography (6.47,143,88.7), philosophy (3.42,120,91.0), fiction & literature (3.26,282,248.6), classics (2.75,164,137.4)
	music	folk (4.80,83,50.9), bluegrass/rural (3.26,57,37.8), jazz(3.16,164,133.6)
	movie	art (6.95,141,83.8), documentary (2.79,129,104.2), independent (5.04,206,155.7)
not* friendly	book	philosophy (3.19,297,252.1), sex (3.22,169,133.9),
	movie	erotic & softcore (2.83,139,110.6), independent (2.93,480,431.4)
	music	funk (2.93,194,159.5)
	social	hot tubbing (3.03,444,394.7), raving (2.80,181,149.0)
fun	book	entertainment (3.04,137,108.8)
	landsport	beach volleyball (2.87,127,101.1), football (2.86,123,97.5)
	movie	adventure (3.16,305,268.8), drama (3.45,325,285.7), horror (2.53,108,86.5), romance (2.98,200,168.2)
	music	rap/hip hop (4.20,282,234.2), soul/R&B (2.60,193,165.4),
	other	ice skating (2.69,100,78.1)
	social	partying (5.93,369,301.8), clubbing(4.99,268,211.9), hip-hop dancing (4.43,179,134.6), bar-hopping (3.02,198,165.9), hot tubbing (2.92,205,173.5)
	watersport	surfing (2.78,68,49.4), wake boarding (3.31,54,35.1), water skiing (2.91,89,66.8)
funny	landsport	table tennis (2.51,153,128.3)
	movie	comedy (3.43,416,381.6)
	music	rap/hip hop (2.73,262,231.1), rock (3.05,377,344.2)
	other	bowling (3.04,179,147.7), couch potatoing (3.16,204,170.2)
intelligent	book	philosophy (2.90,169,138.7), politics (2.81,141,113.7), science (3.99,151,112.4), science fiction (2.88,213,180.0)
	other	computer gaming (2.74,134,108.0)
kind	movie	science fiction (2.67,201,172.3)
lovable	book	cooking (2.53,38,25.8), entertainment (3.27,73,52.0), romance (2.89,35,22.0)
	movie	adventure (2.59,149,128.5), mystery (2.92,93,72.1), romance (5.63,122,80.4)
	music	easy listening (3.13,48,31.6), latin (2.81,71,52.9), rap/hip hop (2.80,134,112.0), soul/R&B (4.20,110,79.1), trip-hop (2.52,40,27.5)
	other	aerobics (2.58,36,24.0)
	social	hip-hop dancing (4.43,95,64.4)
	watersport	swimming (2.82,121,99.1)
not* responsible	book	sex (3.88,149,110.6)
	movie	erotic & softcore (3.15,120,91.3), gay & lesbian (3.11,55,36.5), independent (3.70,412,356.2)
	music	funk (3.11,165,131.7), jungle (3.15,99,73.1), reggae (2.87,199,165.4), trance (2.86,231,195.2)
	other	skateboarding (2.54,63,46.2)
	social	raving (4.14,166,123.1)
sexy	book	sex (7.71,51,19.2), teen (5.26,19,6.2), health mind & body (3.06,29,17.0)
	landsport	wrestling (3.05,14,6.4)
	movie	erotic & softcore (9.80,53,15.8), western (4.40,28,12.8), gay & lesbian (4.32,17,6.3), horror (2.64,41,28.2)
	music	funk (4.98,45,22.8), house (3.88,47,28.2), disco (3.52,31,17.2), jungle (3.31,24,12.7), trip-hop (2.75,30,18.8), reggae (2.73,42,28.7)

	other	weightlifting (4.19,56,34.1), bungee jumping (3.26,20,10.0), skateboarding (2.89,16,8.0)
	social	hot tubbing (6.58,97,56.5), bar-hopping (4.60,82,54.0), raving (4.78,42,21.3), partying (4.29,126,98.3), hip-hop dancing (3.88,66,43.9), folk dancing (3.62,15,6.2), clubbing (3.58,92,69.0)
	watersport	jet skiing (3.26,38,23.3), surfing (2.86,27,16.1)
<i>not*</i> sexy	book	science fiction (2.61,305,268.5)
socially adaptable	book	sociology (3.41,39,23.2)
	music	house (2.75,65,47.6), rap/hip hop (2.97,154,128.9)
	other	snowboarding (2.55,65,48.8)
	social	bar-hopping (3.63,120,91.3), clubbing (2.92,141,116.7), raving (2.83,52,36.1)
	watersport	water polo (3.56,23,11.3)
<i>not*</i> socially adaptable	movie	art (3.01,185,151.7), fantasy (2.59,197,167.3), performing arts (2.64,112,88.6)
	other	laser gaming (2.67,99,76.8)
successful	book	business (5.88,21,6.6)
	landsport	tennis (3.41,43,28.2)
	other	weightlifting (4.16,32,16.8)
	social	barbecuing (3.05,41,27.8)
	watersport	boating (2.85,22,12.7), jet skiing (4.30,25,11.5), water skiing (3.05,20,10.7)
<i>not*</i> successful	book	fantasy (3.03,222,186.0), science fiction (2.66,246,212.9)
	movie	art (3.43,189,151.2), fantasy (2.74,198,166.7), gay & lesbian (2.70,43,28.8), performing arts (3.02,115,88.3)
talented	book	professional & technical (3.17,22,11.6)
	movie	performing arts (4.20,37,19.5)
	other	skateboarding (2.87,16,8.1)
<i>not*</i> talented	book	fantasy (3.37,184,148.2), science fiction (3.01,203,169.6)
	movie	art (2.80,148,120.4), fantasy (2.96,163,132.8), independent (2.79,257,223.6)
	other	laser gaming (2.97,83,61.0)
unique	landsport	track & field (2.58,74,55.9)
	movie	independent (2.52,191,165.1)
<i>not*</i> unique	book	science fiction (4.59,82,53.4), fantasy (3.75,69,46.7)
	landsport	fencing (2.90,14,6.7)
	movie	fantasy (3.34,61,41.8), art (3.27,56,37.9), science fiction (3.05,84,64.0)
weird	book	fantasy (3.32,78,56.3), science fiction (3.75,90,64.4)
	movie	art (3.01,64,45.7), fantasy (3.27,71,50.4), science fiction (2.90,98,77.1)
	music	heavy metal (2.56,39,26.6)
	other	computer gaming (3.24,57,38.6)

***Note: Personality traits preceded by “not” (for example, “not friendly”) do not mean that individuals described themselves as having that trait. Rather, they elected not to select a certain characteristic (e.g. friendly). “Not” simply means the absence of a self-described characteristic.**

Appendix B

TableB1. Preferences of male users	
preference/activity	item (Z score, number observed, number expected)
book	computers (5.74,172,113.3), science fiction (5.65,430,338.6), professional & technical (4.72,125,83.1), science (4.49,272,211.5), business (3.85,133,96.4), politics (3.33,259,213.9), philosophy (3.08,306,260.9), sports (3.07,179,144.0), adventure (2.35,337,300.7)
landsport	football (5.92,312,229.6), frisbee golfing (5.16,195,137.4), table tennis (5.00,384,306.1), golf (4.72,258,196.4), baseball (4.70,199,145.2), basketball (4.02,442,374.8), cricket (2.88,54,36.8), fencing (2.32,57,42.2), racquetball (3.67,94,65.1), squash (2.32,82,63.9), tennis (2.94,466,415.1), soccer (2.10,382,347.6), wrestling (2.03,60,46.4)
movie	science fiction (7.42,533,405.5), war (6.98,395,288.6), action (4.03,771,693.5), spy film (3.59,450,389.2), erotic & softcore (3.26,148,114.5), adventure (2.69,684,632.6), anime (2.88,202,166.9), sports (2.98,262,221.1), western (3.02,121,92.8)
music	heavy metal (4.55,191,139.8)
other	computer gaming (7.02,296,203.0), weightlifting (5.55,326,246.4), billiards (4.57,432,356.7), ultimate frisbee (4.53,246,188.0), mountain biking (4.51,175,126.5), paintballing (4.35,241,185.6), laser gaming (2.30,146,121.7), bicycling (2.15,257,227.1)
social	barbecuing (3.04,462,409.7), raving (2.45,183,154.2), hot tubbing (2.06,444,408.5)
watersport	fishing (2.23,183,156.7), sailing (2.03,205,179.5)
personality	trait (Z score, observed, expected)
freetime	learning (4.21,314,253.1), doing physical challenging activities (4.07,414,347.6)
friendship	mutual friends (3.51,217,173.5), common interests (3.33,875,811.0), appearance/look (3.05,92,67.5), sex (2.65,72,53.0)
romance	appearance/look (5.09,293,222.9), sex (3.41,239,194.6), physical attraction (2.93,686,630.2)
support	the eternal optimists (3.88,325,267.5), the give-it-to-you-straight people (3.08,872,812.8), i've-been-down-and-dirty-a-few-times-myself people (2.12,414,378.4)
you	intelligent (2.99,523,469.4)

Table B2 Preferences of female users	
preference/activity	item (Z score, number observed, number expected)
book	romance (8.28,139,71.5), fiction & literature (5.55,557,470.2), health mind & body (4.86,123,81.1), cooking (4.35,122,83.9), art & photography (4.26,218,167.7), entertainment (3.05,205,168.9), mystery & thriller (2.99,209,173.3), psychology (2.52,145,119.2), classics (2.17,290,260.0)
landsport	gymnastics (4.31,53,29.8), field hockey (4.08,35,17.9), softball (2.70,84,63.2)
movie	romance (11.48,420,261.1), family (5.61,135,85.5), drama (5.16,524,443.6), musical (5.09,230,169.7), performing arts (3.54,125,92.6), comedy (2.38,637,600.6), independent (2.12,325,294.5)
music	soul/R&B (5.39,331,256.8), pop (4.49,442,373.6), country/western (4.08,121,85.1), rap/hip hop (3.06,410,363.7), folk (2.34,118,96.2), latin (2.46,201,171.7)
other	aerobics (9.69,160,77.9), ice skating (4.93,172,121.2), jogging (3.92,262,211.5)
social	hip-hop dancing (6.62,294,209.1), lating dancing (3.94,165,124.0), clubbing (3.44,380,329.1)
watersport	swimming (2.79,363,322.0)
personality	trait (Z score, observed, expected)
freetime	catching up on chores and things (3.80,244,196.4), socializing (3.75,715,659.8)
friendship	laughter (6.66,791,696.4), honesty/trust (3.95,736,678.9), communication (2.24,514,479.0)
romance	laughter (7.18,579,466.7), honesty/trust (2.92,685,641.5)
support	unconditional accepters (5.99,355,271.1), the listeners (3.43,596,542.6), chicken-soup people (2.95,156,125.2)
you	fun (4.05,307,251.6), lovable (2.56,142,116.1), friendly (2.53,446,407.0)

Appendix C: Individual preferences and association ratios

Table C1. Book genres and association ratios					
genre	association ratio	Z score	# users	# connections	# expected
gay & lesbian	4.37	15.35	63	88	20
professional & technical	1.75	6.61	138	128	73
computers	1.65	8.52	188	256	154
teen	1.64	3.04	74	36	22
sex	1.41	6.82	230	340	240
sports	1.39	5.91	239	288	207
business	1.37	4.20	160	162	118
romance	1.32	3.63	180	158	120
religion & spirituality	1.31	5.63	258	376	286
politics	1.31	7.80	355	700	535
art & photography	1.29	9.29	422	1056	819
sociology	1.28	3.24	165	156	121
fantasy	1.26	9.80	491	1356	1075
entertainment	1.26	8.46	425	1064	845
health, mind & body	1.23	3.11	204	202	164
psychology	1.21	4.69	300	496	408
science	1.21	4.91	351	572	474
cooking	1.20	3.03	211	236	195
science fiction	1.20	8.54	562	1610	1343
biography	1.19	4.89	337	630	527
travel	1.17	3.71	306	450	382
nonfiction	1.16	4.79	419	868	750
philosophy	1.15	4.56	433	882	769
mystery & thriller	1.14	4.63	436	968	848
adventure	1.14	5.18	499	1198	1051
horror	1.13	1.32	144	102	89
classics	1.13	6.88	654	2096	1851
history	1.10	3.62	483	1068	969
fiction & literature	1.09	11.17	1183	6568	6004
outdoor & nature	0.88	-1.13	140	68	77

Table C2. Movie genres and association ratios					
genre	association ratio	Z score	# users	# connections	# expected
gay & lesbian	5.65	24.75	76	154	27
performing arts	1.76	13.22	233	472	268
religion	1.46	2.89	92	54	36
erotic & softcore	1.44	5.57	190	208	144
sports	1.38	9.95	367	760	548
anime	1.37	6.82	277	408	298
musical	1.36	11.66	427	1154	851
western	1.32	3.33	154	136	103
family	1.26	3.85	215	252	200
fantasy	1.25	8.40	440	1078	859
art	1.25	7.48	399	898	718
crime	1.24	7.58	421	952	765
independent	1.24	14.70	741	3056	2471
biography	1.18	3.12	245	304	257
war	1.17	6.12	479	1132	965
thriller	1.16	9.82	744	2850	2461
romance	1.15	7.49	657	1984	1727
history	1.15	4.21	398	754	657
science fiction	1.14	7.52	673	2192	1921

horror	1.12	3.08	338	576	512
adventure	1.11	11.20	1050	5372	4828
spy film	1.11	5.62	646	1974	1777
action	1.11	12.39	1151	6250	5633
drama	1.10	11.39	1116	5996	5429
mystery	1.08	3.63	589	1554	1437
documentary	1.06	2.20	496	1060	999
comedy	1.05	9.34	1511	10002	9533

Table C3. Music genres and association ratios

genre	association ratio	Z score	# users	# connections	# expected
gospel	2.06	6.76	105	80	38
jungle	1.78	8.67	152	202	113
bluegrass/rural	1.48	5.70	180	188	126
heavy metal	1.48	7.83	232	354	239
trance	1.44	13.92	406	1158	804
funk	1.42	8.14	274	454	318
latin	1.42	13.71	432	1212	855
house	1.40	10.08	338	758	543
folk	1.38	6.27	242	332	240
trip-hop	1.33	5.18	225	298	224
soul/R&B	1.31	16.43	646	2498	1904
techno	1.30	14.56	588	2152	1652
rap/hip hop	1.30	24.99	915	5004	3850
new age	1.30	3.28	157	146	112
easy listening	1.29	5.05	258	344	266
reggae	1.25	6.26	344	640	510
blues	1.23	5.93	348	664	538
country/western	1.23	3.16	214	212	172
disco	1.22	3.15	206	234	192
jazz	1.19	9.70	636	2124	1783
world music	1.18	5.01	384	724	612
pop	1.18	15.50	940	4668	3951
classical	1.12	6.87	716	2372	2116
rock	1.10	15.54	1363	8670	7871

Table C4. Land sports and association ratios

sport	association ratio	Z score	# users	# connections	# expected
touch rugby	33.08	N/A	4	2	0
lacrosse	3.12	7.09	54	34	10
field hockey	2.64	5.00	45	24	9
wrestling	2.29	6.73	77	60	26
cricket	2.24	4.44	61	28	12
fencing	2.14	4.77	70	36	16
frisbee golfing	1.99	16.59	228	494	247
squash	1.79	5.21	106	74	41
track & field	1.72	12.93	251	482	279
gymnastics	1.66	3.55	75	46	27
softball	1.64	6.87	159	176	107
baseball	1.59	9.98	241	400	251
badminton	1.56	8.76	221	336	214
football	1.56	15.50	381	970	621
golf	1.33	7.43	326	582	439
volleyball	1.22	6.18	388	764	624
beach volleyball	1.20	5.71	395	804	670
basketball	1.15	6.95	622	1758	1530
table tennis	1.14	5.29	508	1232	1081
soccer	1.13	5.53	577	1506	1334
tennis	1.05	2.52	689	1924	1835

racquetball	0.80	-1.34	108	34	42
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Table C5. Water sports and association ratios					
sport	association ratio	Z score	# users	# connections	# expected
synchronized swimming	3.25	5.91	45	22	6
diving	2.42	4.72	59	26	10
crew	2.24	6.97	90	68	30
wake boarding	1.64	6.01	137	136	83
jet skiing	1.38	7.28	280	442	320
surfing	1.33	4.16	193	190	142
scuba diving	1.33	5.93	257	376	282
water skiing	1.29	5.10	261	354	274
canoeing	1.24	5.36	309	538	434
water polo	1.18	0.97	80	32	27
sailing	1.13	2.72	298	406	358
kayaking	1.13	2.93	309	472	416
boating	1.10	2.11	309	418	380
swimming	1.08	5.30	810	2968	2751
fishing	1.08	1.36	260	294	273
windsurfing	0.87	-1.12	135	56	64

Table C6. Other sports and association ratios					
sport	association ratio	Z score	# users	# connections	# expected
freestyle biking	2.11	3.46	48	20	9
skateboarding	1.60	4.15	96	74	46
freestyle frisbee	1.58	4.06	96	74	46
ultimate frisbee	1.46	10.66	312	662	453
ski diving	1.45	5.18	165	174	119
miniature golfing	1.41	14.01	426	1296	918
computer gaming	1.40	9.78	337	702	501
laser gaming	1.31	4.59	202	264	202
mountain biking	1.30	4.13	210	220	169
bowling	1.28	13.45	585	2060	1604
rock climbing	1.28	6.22	302	554	434
road biking	1.27	2.15	124	76	59
couch potatoing	1.26	14.64	674	2770	2206
paintballing	1.25	5.55	308	538	431
rollerblading	1.24	3.89	228	280	224
billiards	1.23	10.79	592	1908	1549
triathlon	1.23	0.78	54	14	11
snowboarding	1.22	5.34	346	594	486
bungee jumping	1.19	1.54	120	76	64
ice skating	1.19	4.06	305	476	400
weightlifting	1.16	4.49	409	758	655
backpacking	1.16	5.40	477	1062	918
aerobics	1.12	1.62	196	172	152
jogging	1.10	3.83	532	1284	1171
hiking	1.08	4.30	690	2094	1939
snow skiing	1.08	2.91	517	1196	1112
martial arts	1.05	0.65	211	182	173
bicycling	1.04	0.97	377	564	543

Table C7. Social activities and association ratios					
activity	association ratio	Z score	# users	# connections	# expected
raving	1.64	12.00	256	502	305
ballroom dancing	1.61	13.91	329	704	437
lating dancing	1.49	10.80	312	620	416

bar-hopping	1.34	17.24	648	2312	1720
folk dancing	1.34	1.51	74	26	19
hip-hop dancing	1.33	13.62	526	1652	1238
hot tubbing	1.32	17.74	678	2790	2121
clubbing	1.24	17.27	828	3814	3074
barbecuing	1.20	10.93	680	2364	1967
partying	1.18	22.31	1179	7372	6224
camping	1.11	6.83	745	2618	2353

Appendix D: Personalities and association ratios

Table D1. How users describe themselves and what kind of people seek out others like them					
personality	association ratio	Z score	# users	# connections	# expected
sexy	1.46	5.47	204	192	131
talented	1.40	5.17	213	210	149
fun	1.25	11.22	633	1852	1479
weird	1.25	4.32	286	332	265
lovable	1.22	4.20	292	406	333
unique	1.11	4.15	547	1194	1074
funny	1.10	4.06	619	1474	1345
friendly	1.10	7.55	1024	4024	3674
socially adaptable	1.09	2.12	342	482	440
attractive	1.07	1.76	406	522	486
creative	1.04	1.48	541	982	941
intelligent	1.01	0.42	779	1848	1833
responsible	0.99	-0.28	500	686	692
kind	0.99	-0.44	625	1226	1239
competent	0.92	-1.40	294	226	246
successful	0.70	-1.57	99	18	25

Table D2. How users spend their free time and whether those who spend their free time in the same way are more likely to be friends.					
free time activity	association ratio	Z score	# users	# connections	# expected
fulfilling commitments	1.34	9.30	398	826	614
socializing	1.12	21.12	1660	11374	10156
catching up on chores and things	1.09	2.71	494	850	782
learning	1.07	1.82	420	536	498
doing anything exciting	1.07	8.05	1280	6278	5850
watching TV	1.07	1.85	415	602	561
reading	1.02	0.66	631	1186	1166
getting outside	1.01	0.97	940	2882	2841
staying at home	0.97	-0.32	209	126	129
alone	0.96	-0.93	380	398	415
doing physical challenging activities	0.96	-1.46	577	878	916