John Stevens

University of Essex

Abstract

Social network analysis largely concentrates on one-point-in-time network data. I demonstrate the suitability of social network techniques (developed in Stevens (2010)) to compare education-based social network data. This paper uses university email networks, and concentrates on structures arising when users choose to contact, reply to, or copy other members of the network into communications, through the cc email attachment. Staff and students use e-mail differently. While social connection features in the e-mails of most users, staff devote a higher proportion of their e-mail to research, teaching and administrative functions. Social contacts and learning feature more prominently in the e-mails of undergraduates, and postgraduate student e-mails have characteristics lying between these groups. The email activity between staff and students mainly relates to teaching and learning.

Keywords: Social network; email; teaching.

Introduction

An important feature of this paper is the exploration of the belief that students are characterized by their social nature and ability to make new friendships. The article compares two email networks and concentrates on their network structure of friendships.

Other Social Network Studies

Social network studies tend to be case studies of a single group or setting. Relatively less attention has been paid to comparisons using networks from multiple datasets and longitudinal comparisons. Studies employing multiple datasets focus on one of two distinct general questions. The first asks whether a network of a specific relational content, in aggregate, exhibits common structural

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tendencies. The second enquires as to what structural features are distinguished among different kinds of social relations. In approaching the first sort of question, some studies have examined the same relation measured in multiple settings. Empirical examples include friendships in schools or classrooms (Bearman, Jones, and Udry 1997; Hallinan 1989; and Leinhardt 1972).

Paul Holland and Samuel Leinhardt (1976) used triples or triad census counts to compare local structural similarities among a collection of fifty-one networks of different relational contents, measured on different species. Stanley Wasserman and Philippa Pattison (1996) compare the direction and magnitude of parameters (size, density, and connectedness) in networks using graphs, allowing the calculation of measures of dissimilarity between graphs for a variety of social networks. Both studies found differences in the "structural signatures" of different kinds of relations, notably antagonistic relations such as fighting and dominance on the one hand, and relations of affection (friendship, liking) and affiliation on the other.

Social networking techniques have been used equally effectively to compare the networks formed by other species, as well as to compare human networks with the networks of other species. John Skvoretz and Katherine Faust (2002) documented that humans show tendencies toward mutuality and in-stars and away from transitivity, compared to non-human primates, which showed tendencies in the opposite direction on these properties. Later the triad census approach was applied (2007) to study forty-two networks among four types of species, human, nonhuman primates, non-primate mammals, and birds. Faust and Skvoretz also introduced the technique of applying network analysis to study change across time. Their aim was to assess whether two, three... or many differing networks were similarly structured despite their surface differences. A recent study by Faust (2006) uses a triad census to compare different social networks of, amongst other species, chimpanzees; she comes to the conclusion that "caution should be taken in interpreting higher order structural properties when they are explained by local network features pg. 12".

Research Question

This paper will investigate the following research questions.

- To determine how email activity between staff within a university is separated between workflow and administration-based emails, compared with e-mail sent outside the university.
- To determine if the email activity between staff and students mainly relates to teaching and learning.
- To determine if emails between students on a university campus are mainly within year group and of a social nature by using qualitative interviews.

Research Methods and Data sets

This section describes the research methods and two data sets used in this paper and their sample sizes.

Data Analysis Methods

This study implements the "two-way" network comparison technique detailed in Stevens (2010). It details the output of these implementations by performing a "2 way" cross-sectional analysis on the email social network data.

The raw email contact data used in this paper was collected and anomized by Computer Service at the University of Essex. A random user identifier was assigned to all email users¹. The processing of this raw data into useable data was performed using PERL computer programs. This data is then imported into UCINET and Netdraw computer packages for analysis. By its nature this is a complete sample of 50177 email sent by 2801 users.

The data collection technique is recoded in anomized log files, by means of which an individual's emails that were recorded in a given one-week period. The computer service only provided me with the following information:

- The department and user type assigned to each user identifier
- The source user identifier and destination user identifier of all emails sent within the university in a given week

¹ This research was completed in keeping with the Data Protection Act (1988) and no content of emails was provided to me at any point in this research.

• The number of users with email redirects.

For obvious legal and ethical reasons computer service only provided the following information for the research reported in this paper.

Staff email network

The first email friendship network I investigate is a University based staff network. By its nature this is a complete sample. The set has information on 1910 different members of staff who sent, or received email to other members of staff in a one-week period. In comparison with the student email network that I describe next, the email volume of the staff email network is less affected by term dates. This is shown below in Figure 1.



Figure 1:- Internal staff email volume by week (2582 staff with 110 redirects) (Source: Data collected by University of Essex computer service and detailed in X (19-20 February 2009))

The data points are limited as a result of SSPS graphs. The basic descriptive statistics of the staff email data set for a given week is shown below in table 1.

No. Staff email originators	1910
No. Emails	40071
Average	20.97

Table 1:Staff emails response rates

Student email network

The second email friendship network I investigate is a university based student email network. The set has information on 819 different students who sent email to other students or received email from other students, in a one-week period. Unlike the staff email network which I described previously, student's attendance and the email volume of the student email network are affected by term dates. This is shown in Figure 2.



Figure 2:Internal student email volume by week (8591 students with 254 redirects) (Source: Data collected by University of Essex computer service)

The basic descriptive statistics of the student email data set for a given week are shown below in Table 2.

No. Student email originators	891
No. Emails	10106
Average	11.34

A descriptive analysis of a staff to staff email network

In this section I investigate the staff to staff email network in further detail. The results detailed in the following sections were produced using the UCINET network analysis package and the graphs using Netdraw social network drawing package. The graph below depicts the staff email network within the University and illustrates that it is far denser than the student email network described in the next section. Table 4 Highlights the very considerable variation the number of email contacts by the department in which a member of staff works.



Figure 3: Figure 3: Staff email network (Source: Data collected by University of computer service)

Departments

The volume of emails sent (out degree) by staff in each academic department within the University is shown in Table 4. (N = 40071)

Department	No of Emails	No of Staff	Average	
Department	Sent		0-	
Art history and theory	403	25	16.12	
Biological science	2068	147	14.06803	
Computing and electronics	3059	147	20.80952	
East 15	712	81	8.790123	
Economics	2057	41	50.17073	
Business school	1755	101	17.37624	
Government	1525	74	20.60811	
Health and human sciences	1013	111	9.126126	
History	738	31	23.80645	
International academy	528	92	5.73913	
Language and linguistics	1485	60	24.75	
Law	1606	80	20.075	
Literature	1386	52	26.65385	
Mathematics	514	23	22.34783	
Philosophy	104	19	5.473684	

Table 3:No of staff emails sent per department

Psychology	2138	52	41.11538
Sociology	1189	65	18.29231

I found that members of staff in science and engineering departments, and the business school, sent the most emails to each other, staff in the social sciences sent a moderate number of emails, and staff in the arts and humanities sent the least. It could be argued that the volume of emails and density of a network or subnet are one good indicator of the level of interconnection within that network, although of course they could be using other methods of communication.

Staff type

The volume of emails sent (out degree) by staff type within the University is shown in the following table. (N=40071)

Employee Type	No of Emails Sent	No of Staff	Average
Academic	22280	760	29.31579
Administration / Support	17791	892	19.94507

Table 4:No of staff emails sent per employee type

It can be seen from Table 5 that the largest number of internal emails were sent by academic staff, but administrative and support staff also sent a large number of emails within the university. Academic staff sent only slightly more emails than support staff within the department despite there being a ratio of approximately one member of support staff in a department, to 8 members of academic staff. I surmise and have anecdotal evidence that academic staff use email to contact students for teaching and external to the University to liaise with other users about research, but these points are not accessible given the confidentiality constraints in this research.

A descriptive analysis of student-to-student email network

In this section I propose to examine the student-to-student email network in further detail. Figure 4 shows a plot of the email network for a given week which illustrates it to be far less dense than the staff email network. The cluster of email respondents within the network that relates to the student's department is clearly shown.



Figure 4:Student email network (Source: Data collected by University of Essex computer service)

Department

The volume of emails sent (out degree) by students in each academic department within the University is shown in Table 6. (N=10106)

Department	No of Emails	No of Students	Average	
	Sent			
Art History and	30	133	0.225564	
Theory				
Biological Sciences	286	1007	0.284012	
Business School	128	1638	0.078144	
CCFEA	400	60	6.666667	
Computing and	119	616	0.193182	
Electronic Systems				
Economics	92	690	0.133333	
Government	383	554	0.691336	
Health and Human	258	1020	0.252941	
Sciences				
History	288	442	0.651584	
Humanities	115	950	0.121053	
ISER	49	27	1.814815	
Language and	75	678	0.110619	
Linguistics				
Law	6408	899	7.12792	
Literature	220	507	0.433925	
Mathematics	24	167	0.143713	

Table 5:No of student emails sent (according to student's department)

Psychoanalytical	47	151	0.311258
Studies			
Psychology	75	547	0.137112
Sociology	1111	361	3.077562

It can be seen from Table 6 that by far the highest number of internal emails were sent by students within the law department. Students in the arts department sent far fewer emails. It is surprising that there is very little variation in the number of emails sent by students from different departments, and little variation in the density of the network within departments. The density of a network or subnet is a very good indicator of the level of interconnection within that network.

Student type

The volume of emails sent (out degree) by student degree level within the University is shown in Table 7.

Student Type	No of Emails Sent	No of Students	Average
1 st year Undergraduate	13	2995	0.004341
2 nd year Undergraduate	763	2532	0.301343
3 rd year Undergraduate	1167	1824	0.639803
Post graduate (taught)	6445	1986	3.245217
Post graduate (Research)	1718	1285	1.336965

Table 6:No of student emails sent per student type (Source: Calculated using bespoke Pearl computer programs)

Qualitatively I have found by interviewing five students that emails between students on site are mainly within year group and mainly of a social nature. I also found that student emails off site are mainly regarding contacts back home and keeping in touch with past friendships. I have no quantitative data to support this supposition.

The student email network is a less dense and interconnected acquaintanceship network than the staff email network with markedly fewer emails sent by younger students (i.e., undergraduates) because they use email accounts outside the university, a lot of which they will have been using since there childhood. In contrast, post graduate students use email accounts within the university.

Comparing staff and student email networks

In this section I compare the staff and student email networks using multidimensional scaling of the network data. Firstly I briefly compare the number of staff-to- student and student-to- staff email networks for completeness.

It is a widely held view that an educational establishment such as university is a small community of its own. This study found that, in terms of email activity, this university appears to have two separate communities of staff and students. This result is not surprising because the division between staff and students was assumed at the start in the research question and in the method. The study does however, show that these two groups use emails in different ways. Investigating staff-tostudent and student-to-staff e-mails, this study found that these forms of email traffic with the given sender and recipient pairs, and I assume this is virtually exclusive for teaching and learning for research and taught courses alike. The volume of emails sent differentiated by email sender type is shown in Table 2.

Direction of email traffic	Student	Staff
Student	10106	1056
Staff	66840	40071

 Table 7:Number of emails sent, by direction (student-to-staff and staff-to-student)

As can be seen from the table the pattern is very biased toward staff sending emails, which is mainly for the purpose of sending teaching materials to students and only a minimal number of emails are sent in the reverse direction from student to staff. This email activity is mainly within the students own department.

Metadata Table

Table 9 compares these two email networks, which have a vastly different number of nodes. The staff network has grown in an organic fashion over 25 years of email use in the university. In contrast, the student network has a turnover of students, on average, every two years. For this reason the staff network is rather denser and more centralised than the student network, and has a smaller path length than the student network. The student social network is more scattered in layout than the staff social network. In the student network, an Individual's friends are rather more local and in a more isolated student situation rather than outside of the university, but are still within the United Kingdom.

Data set	No. of nodes	Density the of social network, Bott (1957)	Average Shortest Path, Scott (1991)	Network % centralization Freeman 1992)	Node Betweeness, Freeman (1996)
Staff email network	324	0.0118	4.442	18.06	1027.264
Student email network	833	0.0051	3.618	4.88	221.775

Table 8:Comparing University staff and student email networks (Source: Calculated using UCINET using staff and student email datasets)

The main thing that this network comparison highlights is that there are two different networks or communities, staff and student, co-located in the same location of a University.

Results

Elaborating the results of the Prefscal procedure shown in Figure 5 by highlighting the nodes produced a graphical output to compare the social networks when the outputs are combined together. This procedure has the notable advantage that the user can select to output rows or columns in the model space and use a rectangular input data matrix. However, this output is unsatisfactory, providing no additional information except that the networks are different. The source data of the Prefscal output (Table 9) is more revealing, showing that Network 1 is more developed than Network 2.



Figure 5:Combined output (Initial Stress: 0.3610) (Source: Scaling output of Prefscal as implemented in SPSS when computing metadata Table 9)

section I found that a multi data set comparison cannot be successfully used to compare crosssectional social network data sets. This new method of analysis detailed in Stevens (2010) only shows that the two networks are different. It can be concluded from this analysis that this new method of analysis should not be used for comparing two networks.

Summary

I assume that the emails used by staff within a university are predominantly for work, with staff performing either teaching or administration based tasks by email. Email is more of a social network and learning orientated tool for undergraduate students and the same is true to a lesser extent for postgraduate students. The email activity between staff within the university is mainly workflow and administration based and also to a lesser extent research emails, as research contacts tend to be outside the University. The email activity between staff and student is mainly regarding teaching and learning. Emails between students on site are mainly within year group and largely social in nature. Student emails off site are mainly regarding contacts back home and keeping in touch past friendships.

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