## PROJECT 5: TEXT MANIPULATION ON E-MAIL MESSAGES

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## Project Overview

In this assignment, we focus on manipulating text data into a desirable form and analyzing the structured data through the application of regular expressions. We will be using two large data sets for this project. The first data set is composed of all e-mails from Enron before the company folded. The second contains the archive of the R-help mailing list from April 1998 to February 2011. The data sets have several aspects in common, so we can use these characteristics to create general functions that should be compatible with all e-mail files.

### R-Help mailing list

0.1. Format of the data. The R-help mailing list archive is stored as 167 .tar.gz files corresponding to each month from April 1998 to February 2011. Each of the .tar.gz files is a compressed text file consisting of all e-mail messages send through the mailing list during that month. E-mail volume per month varies from 36 to 8674 messages, and there is an increasing trend through time. A plot of the e-mail messages sent per month may be seen in Figure 1. Upon closer inspection, the spike in e-mail volume seen in Figure 1 occurs in March 2006. As a diagnostic check, we can look at the file size corresponding to the compressed file of that month's e-mail messages, and sure enough, the March 2006 file is much larger than the other files. We can take a closer look at e-mail volume by inspecting the data set by day rather than by month. A histogram of e-mail volume by day is shown in Figure 2.

Somewhat unsurprisingly at this point, the 49 days with the most e-mails sent through the R-help mailing list all occurred in 2006. All of these 49 days fell in March, April, or May of that year.

0.2. **Functions of interest.** We can use regular expressions in order to search the subjects and bodies of the R-help e-mails for function calls. First, let's look through the subject lines. A search with regular expressions found 19150 possible uses of functions in subject lines, the most common of which are tabulated in the first two columns of Table 1.

The functions appearing in Table 1 are all quite common, as expected. Several of them (also including *lmer*, not listed in Table 1) refer to various types of linear models, indicating that many people are using the R-help mailing list for advice or troubleshooting when performing data analysis. Further, graphical and summarizing functions appear on the list, which are used to visually or numerically summarize a data set.

Now, let's look at the functions used in the bodies of the e-mail messages. The most commonly used functions in e-mail bodies are displayed in the third and fourth columns of Table 1. It should come as no surprise that the functions used in e-mail bodies are even more generic than those in subject lines. Users of the R-help mailing list are likely to include their code in the bodies of their e-mails, so it makes sense that these generic functions appear most often since they have wide applications. Many of these functions deal with data creation, possibly to set up test data for more complex function calls.

We can also explore what packages the users of the R-help mailing list are utilizing. The most commonly loaded packages are shown in Table 4.

Date: March 3, 2011.

#### E-mails sent through R-help by month

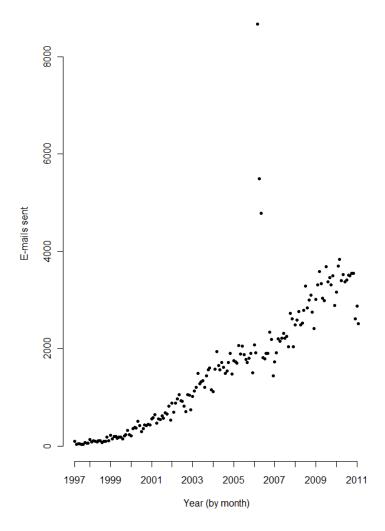


FIGURE 1. From 1998 to 2011, the popularity of the R-help mailing list appears to grow exponentially with some unexpectedly high traffic during 2006.

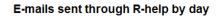
# 1. Enron Email List

The main purpose of this part of the project is to find all the email recipients and senders from the Enron data set, and summarize and graphically represent the results. The first task was to extract the header of the emails, and from this extract all of the information about who sent the email, and who it was sent to. This included looking for Bcc (Blind Carbon Copy) and Cc (Carbon Copy) fields.

This was accomplished using regular expressions, and we found that if we do not include Bcc and Cc, we see that the person who emailed anyone the most was Jeff Dasovich. If we include Bcc and Ccs, the list remains the same. We may now want to investigate who Jeff was emailing, and we have two ways of doing that.

The first is to build another frequency table, as follows: In which we can see the number of times Jeff emailed his top 5 recipients. We may also look at a network graph;

Which gives us the same information, but visually, and without information about the weights. However, with network graphs it is easy to see who people Jeff emailed were communicating with.



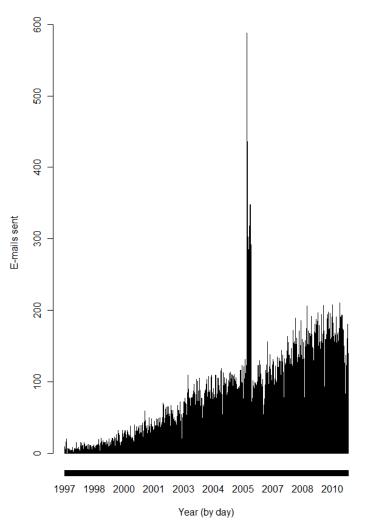


FIGURE 2. The e-mail volume appears to grow when we look at e-mails sent per day. The peaks occur in March 2006, when around 500 e-mails were sent on multiple days.

Subject Line		E-Mail Body	
Function	Frequency	Function	Frequency
lm	527	С	169356
plot	465	function	55049
par	326	list	39738
library	242	library	31071
apply	213	plot	30538
optim	194	length	29409
$_{ m glm}$	191	rep	28783
source	191	rnorm	24379
c	183	data.frame	22974
summary	181	paste	21357
lme	178	matrix	20023
image	166	lm	16186

TABLE 1. The twelve most commonly referenced functions in subject lines and e-mail bodies from the R-help mailing list.

Subject Line		E-Mail Body	
Library	Frequency	Library	Frequency
survival	13	lattice	2392
car	10	ZOO	1384
MASS	10	MASS	1345
gplots	9	nlme	1005
Matrix	9	ggplot2	824
	8	Hmisc	604
tcltk	8	chron	555
fCalendar	6	lme4	497
SSPA	6	tcltk	477
"package"	5	RODBC	471
svIDE	5	gsubfn	393
convert	4	survival	375

TABLE 2. The twelve most commonly referenced libraries in subject lines and e-mail bodies from the R-help mailing list.

Rank	Sender	Number of Emails
1	jeff.dasovich@enron.com	131302
2	veronica.espinoza@enron.com	110601
3	rhonda.denton@enron.com	90730
4	cheryl.johnson@enron.com	74632
5	jae.black@enron.com	59196

Table 3. The top 5 people who sent email from the Enron email list.

Rank	Sender	Recipient	Frequency
1	jeff.dasovich@enron.com	richard.shapiro@enron.com	2909
2	jeff.dasovich@enron.com	paul.kaufman@enron.com	2768
3	jeff.dasovich@enron.com	susan.mara@enron.com	2746
4	jeff.dasovich@enron.com	james.steffes@enron.com	2725
5	jeff.dasovich@enron.com	karen.denne@enron.com	2489

Table 4. The top five people that Jeff Dasovich emailed.

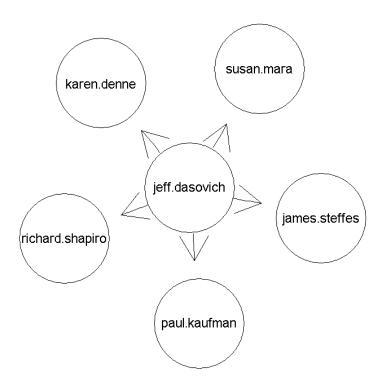


FIGURE 3. A small network of who Jeff emails

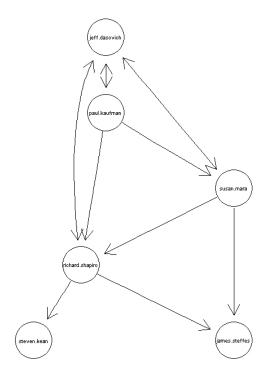


FIGURE 4. A small network of who Jeff emails, and who they email.