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**Data Mining Analysis of Subject Priorities
Among Prominent News Corporations**

by

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**Approved by
Supervising Committee:**

Dedication

This report is dedicated to my father Alan Glasser, whose guidance and encouragement

has always inspired me to seek new heights of learning;

To my son Ben, who may not have always understood what I was doing, but will

understand the extra time that I spend with him from now on;

Most of all to my wife Virginia, who cared for me and put up with me during these

difficult years.

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Finally, I wish to applaud the efforts of *The Daily Show* and the staff of Air America Radio. Collectively, they helped me understand that although entertainment does not constitute news, the two are not mutually exclusive.

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Abstract

Data Mining Analysis of Subject Priorities Among Prominent News Corporations

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In recent years, major news corporations seem to dedicate an increasing amount of time and space to “fluff,” reporting on celebrities, entertainment and crime stories, rather than more essential national and international news. As such news content is increasingly gathered online, it has become feasible to aggregate large amounts of data from a wide range of sites. This report proposes a model for collecting information from news agencies, then applying the techniques of Data Mining to organize this reporting in a way that identifies the priorities of individual organizations.

In addition, the rise of user-based taxonomies has made it possible broadly to evaluate the interests of people who actively read and recommend news. In the final analysis, data collected from users of Digg.com are compared with data collected from media sites. This provides a benchmark for determining whether the delivery of “fluff” news is delivered is a fair response to popular demand, or whether typical news readers are dissatisfied with the level of serious event coverage found in the media.

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INTRODUCTION

Chapter 1: The Nature of Modern Media

It is Monday evening, and Jon Stewart is on a roll. On the satirical news program, *The Daily Show*, Stewart briefly notes the important news of the day on June 11, 2007: Peter Pace has been ousted from his job as Chairman of the Joint Chiefs of Staff, America's highest-ranking military officer. This information is immediately followed by a clip of a CNN anchor saying: “We’ll talk about that more at the top of the hour; we do have some live pictures out of Paris Hilton that we want to talk about.”

Stewart watches in mock disbelief as clip after clip shows major news networks giving lengthy coverage to the supposedly important story of Paris Hilton being put in a car following her arrest. He notes: “All networks covered it gleefully... but know this: CNN didn’t want to!”



Illustration 1: Jon Stewart reacts to coverage of Paris Hilton

There follows a rapid montage of prominent CNN anchors scoffing at the insignificance of the story, even as they continue to talk about it around the clock.

“We’ve always avoided even mentioning the name of the hotel heiress, because we can never figure out what she was famous for.”

“What is the obsession with the woman – and I’m talking about Paris Hilton – who does absolutely nothing?”

“We’re not sure what upsets you [viewers who sent feedback to CNN by mail] more: Paris going to jail, or the fact that we’re even **covering** this story...”

“Of course it is the case of Paris Hilton and – I know! But hear me out...”

“Are we just so pathetic and so lonely that we have to live through people like Paris Hilton?”

Naturally, Jon Stewart is ready with the punch line. “If by ‘we’ you mean ‘CNN’, and if by ‘lonely’ you mean ‘nobody’s watching you,’ then, uh... yes. Poor CNN. ‘Why are they making us do this?’”

So reluctant were they to cover the story, he adds, that they covered it all day. (The Daily Show, 2007.)

DO MEDIA OUTLETS COVER ISSUES THAT DON’T INTEREST PEOPLE?

Though *The Daily Show* is a comedy program, it raises a serious point. If CNN anchors really do not wish to cover stories like the incarceration of Paris Hilton, then who forces them to? In theory, the for-profit corporation is answerable to its viewers, whose eyeballs translate directly into advertising dollars. But do cable subscribers really want wall-to-wall discussion of Paris Hilton? Or is it really the case, as one correspondent seemed to say, that the feedback was overwhelmingly negative? And if the latter is true, then why does the story receive so much attention?

These are questions which are presumably well explored by the marketing experts and business leaders who manage CNN. For ordinary consumers of news, however, their reasoning is a bit more opaque. Presumably a news channel has two goals, which may

sometimes come into conflict. The first is to present serious news which is intended to inform viewers of important current events. The second is to entertain people and make a profit.

This paper will investigate the following research question: *Do media sources place an overly high emphasis on sensationalism over legitimate news, even beyond that which is called for based on pandering to the public's taste for entertainment?*

For this purpose, we can broadly divide the news into two categories: “relevant” news and “sensationalist” (or “fluff”) news. Suppose also that it were possible to track these stories along two axes: first, the amount of attention given to each type of story by media outlets; and second, the level of interest in each type of story among readers. If this level of categorization could be accomplished, then we might discover one of three possible outcomes:

1. The coverage of sensationalist news is less than the public demand for sensationalist news. In this case, we might explain the data by claiming that media managers view their mission as one of presenting serious news, and to a certain extent manage to disregard their function as entertainment and avoid catering to the lowest common denominator.
2. The coverage of sensationalist news is about equal to the public demand for it. In this case, one could explain the data by saying that managers view their primary mission as one of catering to public demand, providing entertainment to the widest possible consumer base, in order to maximize profit regardless of journalistic concerns.
3. The coverage of sensationalist news is significantly greater than the public demand for it. This may be the most interesting outcome, although perhaps the hardest to explain. If the media are intentionally delivering

fluff that the public does not want to know about, then there is some other motive besides either journalistic integrity or broad appeal to an audience. Although the true motives cannot be determined by this study, some speculations will be discussed at the end of the paper.

Of course, “the media” is a plural term, which does not describe a single monolithic entity. Many different corporations exist, with highly variable motives and missions. For this reason, it is necessary to analyze media outlets on a case by case basis, rather than as a uniform group.

With this in mind, the purpose of this project is to collect reporting patterns from a wide range of media sources and compare their priorities with that of the public interest. Sources covered will include respected mainstream newspapers, cable news, and some (perhaps) deliberately biased organizations.

In the final analysis, five questions will be investigated:

1. Taken as a whole, do media outlets present a level of fluff news that is congruent to public demand?
2. Taken individually, which media outlets most closely resemble the public demand for specific types of stories?
3. In general, what patterns can we identify about how the public reads the news?
4. What level of coverage do individual politicians receive by media?
5. How does this level of coverage compare to that which is warranted based on reader interest?

STUDIES IN JOURNALISM: THE AGENDA-SETTING THEORY

Maxwell McCombs is a professor of journalism at the University of Texas. Along with Donald Shaw, McCombs is widely known for conducting the first systematic study of the agenda-setting hypothesis of media in 1972. This developed into a popular theory of journalism which states that “the news media, by their display of news, come to determine the issues the public thinks about and talks about.” (Severin and Tankard, 2000, p 207.)

The agenda-setting theory stands in contrast to the notion that popular media merely reacts to public sentiment. It is also in contrast to the “magic bullet theory” of journalism, which holds that the media directly plants opinions into people minds. Agenda-setting theory states that the media does not necessarily have the power to directly change people’s minds, but it does have the power to change the issues on which people focus their attention.

For example, early in the 20th century, a reporter named Lincoln Steffens wrote about how he got in a competition with a reporter another newspaper, where each tried to find crime stories in order to outsell each other. Soon, many other New York papers worked to find crimes and keep up with the others. As a result, there was widespread public perception of a crime wave, which even drew the attention of Teddy Roosevelt. Yet there was no crime wave; merely eager reporters shifting the public’s attention so that they came to see crime as the most important issue of the day. (Severin and Tankard, 2000, p. 207.)

The 1972 story by McCombs and Shaw focused on polling undecided voters in Chapel Hill, North Carolina. A similar study was performed five years later in which they demonstrated that over time, public perception of the relative importance of an issue

is more likely to be influenced by the number of stories appearing on that topic, rather than the other way around. (Severin and Tankard, 2000, pp. 209-211.)

This theory of journalism provides a context for this study. One of the important goals for this research is to use the World Wide Web as an alternative method for gathering information on both the media agenda and the public agenda. Since most major news services now duplicate all their printed material in online documents, a large number of stories can be gathered to represent the media agenda, and this can be done far more quickly than any human reader can by sorting through printed publications. In addition, public submissions to a popular news sharing service will be taken as an approximate representation of the public's agenda.

SOFTWARE DEVELOPMENT

Chapter 2: Modeling the Problem with Web 2.0

One problem with conducting studies via survey is that surveys are expensive, requiring significant time and manpower to collect a representative sample of the general public. Another problem is that people who respond to opinion polling may be a self-selecting and non-representative set of the general public.

As the last decade of the 20th century saw the rise of the World Wide Web as a serious media phenomenon, the early 21st century has witnessed the rise of a suite of software technology collectively described by the buzzword “Web 2.0.” Web 2.0 sites focus on distributed content creation. No longer is web content generated exclusively by technically adept gurus; instead, a growing arsenal of server-side tools allows people with essentially no programming experience to effortlessly contribute to a vast information network.

Examples of Web 2.0 sites include Wikipedia¹, a user-created encyclopedia that can be edited by anyone with an account; Blogspot² and other “web logging” sites (or “blogging” for short), where individuals can publish textual posts in which they present news content or personal stories to anyone who is interested in keeping track of their blog; and Digg³, a social content sharing website at which some users submit the URL of existing web pages and then other users rate the contents of those pages.

With all this information being spontaneously generated by a user base numbering in the millions, the time seems right to ask whether there is a cheaper and more efficient

¹ <http://www.wikipedia.org>

² <http://www.blogspot.com>

³ <http://www.digg.com>

method for identifying the focus of public opinion. Rather than hiring people to laboriously pore through physical newspaper clippings and watch hours of television, we can take advantage of convenient sites like the Google News archive⁴, where dedicated software engineers have already done the hard work of collecting searchable text from many thousands of online media outlets. Rather than calling people on the phone to learn their opinions, we can use content-sharing sites such as Digg to learn about broad patterns of people's reading habits. That information has already been made freely available; we need only figure out how to assemble it in meaningful ways.

To reiterate the bearing that these sites have on journalism issues, the data collected from Google News represents the media agenda, while the data collected from Digg represents the public agenda.

⁴ <http://news.google.com/archivesearch>

Chapter 3: Planning the Analysis

The technique for gathering news went through several iterations. Some of the revisions will be described in the remainder of this paper, but there follows a brief summary of the final method applied. Justification for these decisions will be described in detail later.

1. The analysis will be restricted to a list of 12 topics, clearly identified as either “serious” or “fluff.”
2. The overall trends in the news will be drawn from all sources that are catalogued on the Google News archive. However, for in-depth analysis of specific media outlets, only 12 sources will be studied.
3. News stories will be collected by a web crawler, from a series of months ranging from August 2006 to February 2007.
4. To gather information from Google News, a separate web search is performed for each month in the study. To determine the extent of major coverage, the program will keep a count of how many stories are listed on the first search page returned for each topic. This provides a metric for how important the story is consider in the context of media agenda.
5. Because the search page only shows the total number of stories found, additional web searches must be performed in order to expand the list and find which news outlets covered each story.
6. Another web search is performed on Digg, covering the same time period. The stories about each topic will be noted in each month, along with the average scores assigned to those stories by Digg readers.

7. The ultimate goal is to compare the priorities of the media agenda (Google News sources) to the public agenda (scores submitted to Digg).

Submitting specific monthly searches to Google News gets around a fundamental problem with computer analysis. While existing software can very quickly scan all news over a period of time, computer programs have a difficult time identifying what a particular story is about. A human reader who looked at a news article might skim the first paragraph and immediately conclude that the article is about, for instance, presidential candidate Barack Obama. Individuals who are well informed about the news might even learn most of what they need from an article merely by reading through a few paragraphs. However, despite efforts that researchers have made towards artificial intelligence since the invention of the computer, a simple software package that reads the content of a story and identifies its significance is currently beyond easy access.

Reading and organizing news is a task which has already been solved (to some extent) by Google News. Developing an “intelligent” news reader to interpret these stories would needlessly duplicate this work, and therefore it is more efficient to just submit topics to Google and assume that the returned stories have been correctly identified as falling under that topic..

THE SEARCH FOR NEWS TAGS

Web tagging is a subject that has gained considerable attention lately. On many sites that are geared toward user content creation, contributors are allowed to assign “tags” to each resource that they generate. On Wikipedia, for example, a given article may be placed in multiple categories, which are arranged by users in a hierarchy to

enable later visitors to easily browse items with similar tags. Tags are searchable, offering keywords that can be used by a database. Unlike reading the full text of an article, the use of tags allows an article to be briefly identified by subject.

Therefore, my first hope was that Google News would itself assign tags to articles and make the job of collecting data easy. Unfortunately, it turned out that this was not the case. Failing to find this easy solution, I then searched the web for a site that, like Digg.com, might post contemporary articles and encourage others to assign manual tags. However, I could not find any such site.

On further reflection, it is obvious why such a search must be futile. To give an idea of the scope of the categorization problem, consider these figures. When you visit news.google.com, you might typically be greeted with a number of stories resembling the following:

State Dept. steps up watch on Blackwater

Chicago Tribune - 23 hours ago

By Aamer Madhani and Bay Fang | Washington Bureau October 6, 2007

WASHINGTON - With Congress poised to expand laws overseeing private security contractors in Iraq, the State Department announced a new set of procedures Friday that will allow for closer ...

Iraqis Claim Jurisdiction, But US Also May Oversee Incident ... ABC News

Iraqi-US commission holds first meet on Blackwater AFP

New York Times - The Courier News - Worcester Telegram - JURIST

all 1,352 news articles »

In turn, clicking on the link labeled “all 1,352 news articles »” takes you to a “cluster” of similar articles, all of which are dated within the last 5 days or so, and many being roughly within a single day.

Furthermore, the front page of Google News alone contains links to some 30-40 clusters, each cluster representing its own individual group of related news articles. The above cluster contained 1,352 articles, which makes it unusually large. A more typical

front page cluster may contain something closer to 100-500 stories. A quick calculation reveals that if the front page of Google News displays 35 clusters which each contain perhaps 200 stories that were posted within a single day, this means that we would have to rely on users to accurately tag 7,000 new stories per day on the front page alone. This does not even include all other stories, which are not selected by the search algorithm to appear on the front page.

By way of comparison, Wikipedia (arguably the most successful web 2.0 site) has a statistics page⁵ showing that a total of just over two million English language articles have been produced within its six year history. Dividing this up, we can verify that somewhere around 913 new articles have been generated per day. Users cannot reasonably be expected to tag all the stories on the front page of Google News by hand, when this accomplishment would require a level of daily activity more than seven times as much as is now applied on Wikipedia.

To take a brief tangent from the topic of the report, it would be very valuable if major news corporations could be persuaded to apply a standard tagging format to their own articles as they are generated. This would greatly improve the browsing experience for readers of media web sites. However, this is not available at the present time, and therefore it is an empty wish for the purposes of this project.

INFERRING TAGGED NEWS FROM SEARCHES

The Google Corporation is known first and foremost for their innovative approach to web searching. Instead of looking for tags within a story, we can use Google News searches on specified topics, implicitly “tagging” these stories as pertaining to the search topic that retrieved them.

⁵ <http://en.wikipedia.org/wiki/Special:Statistics>

The first programming task was to write a web page reader which could identify common topics. Google News supplies a valuable tool to meet this need. Each day, the front page shows a list of eight common terms, under the heading “In The News.” The first iteration of the program was a simple page reader that could run each day. The program would read the major topics and transfer all terms into a “Topic” table in my database. After this had run for a couple of weeks, I collected a list of some 100 unique topics to choose from. From this list, I selected 12 topics on which to focus, based on the following criteria:

1. Each topic must have a significant and regular presence in the news.
2. Taken all together, the topics should represent a reasonable spread of categories that can be obviously identified as serious and fluff.

The categories that I had planned were: 1. United States News (serious); 2. Politicians (serious); 3. Celebrities (fluff); 4. Entertainment (fluff).

As the news currently contains a significant amount of coverage on the upcoming 2008 presidential primary campaign, all politicians are United States presidential candidates.

The list of topics appears in Table 1, along with their categorization.

Topic	Category	Description
Abu Ghraib	US News	Prison in Iraq, currently under control of United States military. Abu Ghraib has been a prominent news subject since May 2004, when U.S. soldiers were discovered torturing and abusing prisoners.
Anna Nicole Smith	Celebrity	Former Playboy model who married elderly billionaire J. Howard Marshall. Briefly starred in a reality TV show. Died of a drug overdose on February 8, 2007, a few months after her 20 year old son also died
Barack Obama	Politician	Currently the second highest polling candidate in the field of Democratic candidates.
Britney Spears	Celebrity	Young female pop singer and sex symbol.
Giuliani	Politician	Current Republican front runner. Since first name is alternately reported as “Rudolph” or “Rudy,” only his distinctive last name is used here, in order to catch the most possible stories.
Harry Potter	Entertainment	Fictional protagonist of a popular series of books and movies.
Hillary Clinton	Politician	Current Democratic front runner.
John Edwards	Politician	Currently the third highest polling candidate in the field of Democratic candidates.
Mitt Romney	Politician	Currently the second highest polling candidate in the field of Republican candidates.
New Orleans	US News	Major United States city, mostly decimated by Hurricane Katrina in August 2005. Reconstruction of the city has continued to be slow and many destroyed homes have not been rebuilt.
Paris Hilton	Celebrity	Daughter of Richard Hilton, and heiress to the Hilton Hotel fortune. Hilton is regarded as a somewhat vacuous individual, yet receives frequent media attention. Former star of a reality television show called “The Simple Life.” Her arrest in 9/06 generated brief, intense coverage.
Tiger Woods	Entertainment	Successful African-American professional golfer; this topic is intended to represent sports interests.

Table 1: List of collected topics

Once the topics were chosen, it was apparent what would be the correct way to proceed in collecting data. First, do a web search on one of the above news topics. Then, find all links on the page labeled “all N news articles,” note the number, and save the

link. This link would represent a single news “cluster”, an event that is covered by one or more media sources. (Stories that appear by themselves, and not in a cluster with other stories, are ignored.) Finally, read the contents of the cluster by following the appropriate link and reading off-site links.

Each link can be parsed fairly simply to identify what site it comes from. For instance, a link to

< <http://www.washingtonpost.com/wp-dyn/content/article/2006/04/13/AR2006041302159.html> >

can be identified as coming from the Washington Post by extracting the first string after “http://” and before the next “slash” character: “www.washingtonpost.com”. To make later searching easier, a number is immediately associated with all web pages as soon as they are added to the database, uniquely identifying the source site, which then does not need to be extracted from the full URL again.

Naturally, the list had to be altered a bit throughout the course of the project. The list originally included the topics “Rupert Murdoch,” “Gulf Coast,” and “Blackwater,” which came up as popular subjects during my initial scans. The “Rupert Murdoch” topic was eliminated because it consistently resulted in a very low story count, which made the sample difficult to work with. “Gulf Coast” was eliminated because the subject was somewhat redundant with “New Orleans.” Although interesting and topical, “Blackwater” (a privately owned security organization which receives military funding for operations in Iraq) generated very little news until early 2007, so it lost value as a story when the program began scanning older stories. The “Abu Ghraib” topic was not sampled initially, but it was added near the end in order to include more subjects in the “News” category.

SELECTION OF MEDIA SOURCES

The first attempt to collect data simply involved doing topic searches several times over a few days, then recording the news that appeared in every cluster retrieved from these searches. After doing this data collection, I analyzed the database to determine which sites were responsible for most of the content. Based on this list, I tagged a list of 12 web sites as worth tracking. The criteria employed were:

1. Each site must have a relatively high percentage of total stories scanned, in order to provide a statistically significant data source.
2. Each site must be clearly identifiable as a separate news corporation; hence stories received from aggregators or news wires are not considered.
3. Overall, the sites should represent a reasonable spread of different types of media, from established newspapers to cable news.

I have listed my online sources below. Although I have tried to avoid coloring the data with personal opinions, many of the sources have a clearly stated mission, or a reputation for biases, whether they are political or fluff. The descriptions of the sites provided in table 2 cannot be purely objective, but as much as possible they are summarized from a combination of their own mission pages and other external descriptions, such as Wikipedia references.

Source	Web Address	Perceived characteristics
The New York Times	http://www.nytimes.com	Online location of large, established newspaper
The Washington Post	http://www.washingtonpost.com	Online location of large, established newspaper
Cable Network News (CNN)	http://www.cnn.com	Founded in 1980 by Ted Turner, CNN was the first major news channel to feature 24-hour news coverage.
ABC News	http://www.abcnews.com	ABC News is a division of the ABC network. They distribute news to television, radio, and the internet.
USA Today	http://www.usatoday.com	Daily national newspaper; widest circulation of any in the US. Known for simplifying stories for a broad audience and using colorful charts and tables.
Fox News	http://www.foxnews.com	Cable news network owned by Rupert Murdoch. Despite the slogan "Fair and Balanced," Fox News has a reputation as a highly partisan right wing network.
New York Post	http://www.nypost.com	Tabloid-style newspaper, also owned by Rupert Murdoch.
Wall Street Journal	http://online.wsj.com	Financial newspaper that covers business news. Conservative op-ed, but neutral reporting. The paper was also recently bought by Rupert Murdoch.
Washington Times	http://www.washingtontimes.com	Politically conservative daily newspaper, founded in 1992 by Sun Myung Moon, leader of the Unification Church.
BBC	http://www.bbc.co.uk	British Broadcasting Corporation, primary television news coverage in Great Britain.
The Guardian	http://www.guardian.co.uk	Somewhat left-leaning British newspaper
Salon	http://www.salon.com	Online-only magazine, focusing on American politics with a distinctly liberal point of view.

Table 2: List of media sites used

In order to determine the priorities of the various news sites, I chose to look at all the clusters identified by Google which ran over a minimum size of 20 stories and see which news sites were featured stories within that cluster. Then I planned to identify each combination of web site and topic based on the percentage in which they were present in the relevant clusters.

IDENTIFYING POPULAR OPINION WITH DIGG.COM

Searching Digg proved to be much easier than searching Google News. In contrast to the 7,000 daily front page stories on Google News, Digg has a much smaller base of user-submitted stories to read. In fact, it turned out that I could collect every story relating to each of my topics, back through the beginning of Digg's existence.

Unlike Google News, Digg provides a fairly obvious way to "score" stories based on overall interest or disinterest in a topic. The process of visiting Digg as an ordinary user works like this:

1. Find an interesting web link.
2. Log in to an account on Digg, and click "submit new."
3. Enter a headline and a brief description for the link.

Other users who encounter a previously submitted article can click on a direct link to view the original story. They can also click a second link to arrive at a page within the Digg site, where they can comment on the significance of the story or engage in discussion with other users. Finally, and most importantly to this study, a link on one side says "Digg it." Clicking this link will assign one point value to the story, and the overall score will then be displayed next to each story. At most, one point may be assigned by each reader.

In this way, all stories can be viewed and their scores can be tallied in just one page load. Once this is done, we can determine the total number of stories submitted for each topic, and the average score assigned to stories within that topic.

Chapter 4: Database Design

The database consists of six tables. Each table name is prefixed with “NM_”, indicating that it is part of the News Miner group of tables.

The table schema is as follows:

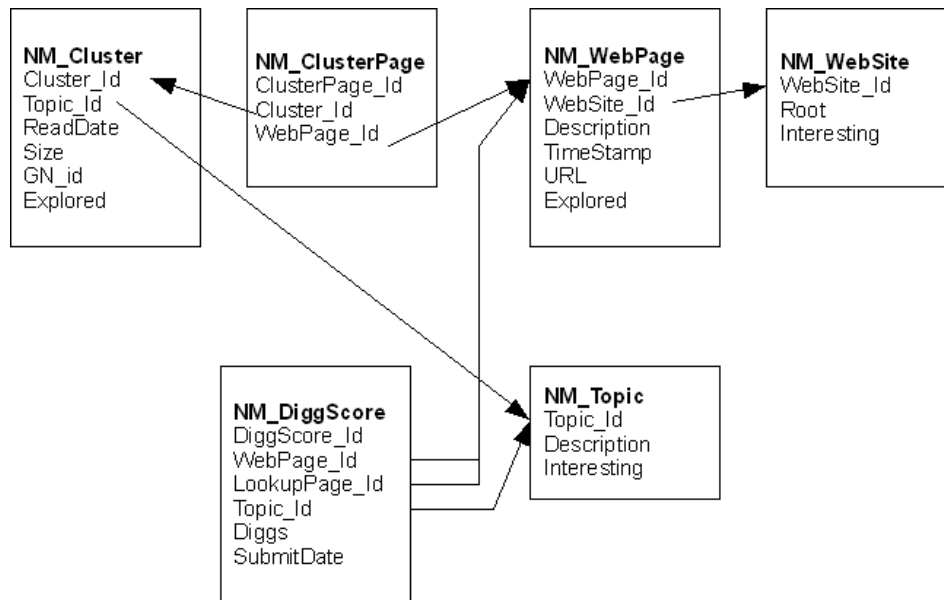


Figure 1: Entity-Relation Diagram

The description of the tables is listed below.

- **NM_Topic** is a descriptor of one topic in the news. Hundreds of topics were entered, but only the topics listed in Table 1 were scanned.
- **NM_WebSite** represents an individual site, such as www.nytimes.com or www.foxnews.com. The “Interesting” field is a binary value indicating whether it is being treated as an important news source in the analysis.

- **NM_WebPage** represents a single web page. It is associated with one **NM_WebSite** entry. The URL is the Uniform Resource Locator that identifies the site on the web. The TimeStamp represents the date and time at which the page was added to the database. The Description is a string that is usually pulled directly from the link text from which the story was obtained. However, the program is also designed in such a way that the link can be retrieved from the title field after directly visiting the page.
- **NM_Cluster** represents a cluster of Google News stories. Each cluster is associated with exactly one topic. The ReadDate is a date and time which is provided by Google News to indicate on what date the stories within the cluster occurred. GN_id is a unique identifying number that can be used to retrieve the search page from Google News. The Explored field is a boolean value which indicates whether or not stories have been collected from within the cluster yet. Stories are assigned to Clusters within the database via the **NM_ClusterPage** table.
- **NM_ClusterPage** is a table that associated WebPages with Clusters. Each ClusterPage entry links exactly one cluster id with exactly one web page id. The relation between WebPages and Clusters is intended to be one-to-many, so each cluster will have multiple stories but a web page can be associated with at most one cluster.
- **NM_DiggScore** represents one story collected from a Digg search. A Digg score has two web page associated with it: the original story, and a URL for the discussion page. Like Cluster, the Digg story has a topic based on the search term. IT also has a score (the “Diggs” field).

Chapter 5: Program Architecture

The News Miner program is written in Java. It uses the Java Database Connectivity (JDBC) library to access an online MySQL database at my website, <http://mysql.apollowebworks.com>. It also uses the JTidy library, a package which interprets HTML text which may not be well-formed, and converts each document into a Document Object Model (DOM) tree. JTidy can be found at <http://jtidy.sourceforge.net/>.

CLASSES

The program consists of 23 classes and approximately 3500 lines of code. Many of the Java classes directly correspond to tables in the database. Instantiations of classes such as **Cluster**, **WebPage**, and **DiggScore**, represent single table entries. Methods in these classes communicate with a common **MiningDB** class to read entries straight out of the database, and save new entries to corresponding rows in the table.

The **WebPage** class also has methods for scanning a page and collecting links, or viewing lists of other selected HTML tags in order to facilitate collecting new information.

There is also a **Digg** class and a **GoogleNews** class, both inherited from a common **AggregatorSite** class. The classes for these aggregator sites contain static methods for retrieving the front page and individual search pages. There is also a **GoogleNewsPage** class and a **DiggPage** class, both of which extend the **WebPage** class. These classes are responsible for reading HTML patterns specific to those sites.

For instance, the **GoogleNews** class has a “getMonthlyClusters” method for exploring stories within clusters that have not yet been associated with an entries in the

ClusterPage table. Calling “getMonthlyClusters” will generate a series of **GoogleNewsPage** instances. These pages are then read using the method **GoogleNewsPage.getNewsStories**, which in turn populates the database with new **WebPage** and **ClusterPage** entries. While a cluster is being explored, the autocommit property of the database is turned off. New web pages are committed only after all pages in a cluster have been retrieved. The **NM_Cluster** row is marked with the “Explored” field set to true in the same operation. Thus, a cluster in the database is always either completely explored, or completely unexplored.

I designed a simple interface on the Java application, offering a menu displaying options for various data-gathering tasks. The tasks are: 1. Get monthly clusters (one topic); 2. Get monthly clusters (all topics); 3. Explore current clusters; 4. Get new Digg scores (one topic); 5. Get new Digg scores (all topics); 6. Generate results table; 7. Analyze results.

Throughout the months when I was collecting data, I would execute one or more instances of the program each day to fill in gaps in the current data.

Chapter 6: Web Interface

A simple web interface was written in Perl/CGI, using the DBI library to access the SQL database. The resulting web pages are primarily intended to provide a way to conveniently drill down into the data, which is organized by topic, then month, then cluster (or Digg Score); and finally links to the stories on the original source pages.

Newsminer clusters

Month	Abu Ghraib	Anna Nicole Smith	Barack Obama	Britney Spears	Giuliani	Harry Potter	Hillary Clinton	John Edwards	Mitt Romney	New Orleans	Paris Hilton	Tiger Woods
3/2006	524	97	199	108	83	211	478	403	148	883	161	1055
4/2006	975	176	56	137	300	117	335	42	296	1025	83	1879
5/2006	906	113	532	246	86	196	215	440	209	894	298	953
6/2006	837	131	54	224	334	145	419	191	72	321	189	1221
7/2006	55	146	54	78	1725	242	640	96	424	1239	63	2096
8/2006	10	531	225	207	49	124	598	766	469	1446	152	2064
9/2006	254	422	288	144	420	96	753	66	163	710	196	2544
10/2006	462	198	185	68	632	306	420	115	150	520	264	375
11/2006	319	127	382	481	590	341	1431	231	188	528	252	1104
12/2006	622	89	189	544	203	632	358	338	414	982	263	798
1/2007	454	117	202	260	288	234	1008	397	1359	1199	372	1256
2/2007	49	698	261	306	397	245	717	142	548	314	58	550

Figure 2: Screenshot of the web interface

The above figure is a screenshot of the top level Google News browsing interface. This page displays a grid with topics on the horizontal axis, and months scanned on the vertical axis. Each cell in the grid displays a number representing the total number of prominent (front page) stories collected on one topic in one month. (See the appendix for the URL where this data may be reviewed.)

Clicking on any cell takes you to a page where a number of clusters (up to 20 per monthly topic) are displayed. Each cluster is identified by the headline of the first within that group of articles. The clusters range in size from two up to several hundred stories. Each cluster can be clicked in order to visit the original Google News page, and you can then browse the stories in that cluster. A similar interface also exists for exploring the submitted stories in Digg at their original locations.

Chapter 7: Revising the Approach

As every programmer knows, even the best laid plans are always subject to change based on new information discovered in the process of developing software. This section lists some of the issues that I had to deal with in generating my data.

SWITCHING FROM DAILY NEWS TO THE HISTORICAL ARCHIVE

My initial plan was to simply scan the news each day and record stories as they appeared. After a few days of collecting data in this manner, it became clear that this would not generate a sufficient amount of data to create an acceptable sample size. Therefore I chose to use the archive, which records older data.

The archive has an unfortunate limitation, in that it only contains stories up to mid-February of 2007. This explains why the range of topic searches only extends through February, rather than up to the beginning of the project in September.

GETTING AROUND GOOGLE'S MALWARE DETECTION FEATURE

Once the program began gathering data from the archive, a new problem unexpectedly materialized. Topic searches suddenly began to return no clusters at all. The results of one search seemed to indicate that there had been no news at all about Anna Nicole Smith during the entire month of March. A quick check with a browser made it clear that this was not true.

Going through the program in a debugger led to a visited URL which contained this message:

We're sorry...

... but your query looks similar to automated requests from a computer virus or spyware application. To protect our users, we can't process your request right now.

We'll restore your access as quickly as possible, so try again soon. In the meantime, if you suspect that your computer or network has been infected, you might want to run a virus checker or spyware remover to make sure that your systems are free of viruses and other spurious software.

We apologize for the inconvenience, and hope we'll see you again on Google.

To continue searching, please type the characters you see below:

This was followed by a typical set of “CAPTCHA” characters to type. (CAPTCHA is an acronym that stands for “Completely Automated Public Turing test to tell Computers and Humans Apart.”⁶)

Evidently, Google’s software was smart enough to recognize a computer program repeatedly accessing similar searches within a very short period of time. A post on Google’s blog explains the intent behind this feature. According to Niels Provos from Google’s Anti-Malware Team:

“we have seen self-propagating worms that use Google search to identify vulnerable web servers on the Internet and then exploit them. The exploited systems in turn then search Google for more vulnerable web servers and so on. This can lead to a noticeable increase in search queries and sorry is one of our mechanisms to deal with this.” (Provos, 2007)

Certainly no one could fault Google for acting on their legitimate interest in protecting their servers from being overtaxed by viruses and possibly denial of service attacks. However, the automated approach they had chosen rendered it very difficult to convince them that I was a real person with an important task to complete. I attempted to get through to a human respondent in the hopes of obtaining special permission to carry on this research, but no one responded to my email and newsgroup postings.

⁶ More information can be found at <http://www.captcha.net/>.

Since the authentication was done on a per-browser basis, typing the CAPTCHA text in a browser such as Firefox did not solve the problem within the program. The project had moved beyond strict data mining issues and into the realm of internet security. I was reluctant to invest time in getting the program to answer the CAPTCHA challenge. So instead, I dealt with the problem first by physically switching IP addresses. For a few days I found myself transporting a laptop computer to various coffee shops and classrooms in order to grab data in brief bursts before being interrupted. In order to be considerate to Google's resources, I added a "sleep" routine so as to avoid hitting their servers too hard. This was only a temporary solution, however.

I am greatly indebted to a contact who uses the internet handle "nephlm," since he suggested the eventual solution to my problems. The Electronic Frontier Foundation sponsors a program called "Tor," which is described in this way:

Tor is a toolset for a wide range of organizations and people that want to improve their safety and security on the Internet. Using Tor can help you anonymize web browsing and publishing, instant messaging, IRC, SSH, and other applications that use the TCP protocol. Tor also provides a platform on which software developers can build new applications with built-in anonymity, safety, and privacy features.

Tor aims to defend against traffic analysis, a form of network surveillance that threatens personal anonymity and privacy, confidential business activities and relationships, and state security. Communications are bounced around a distributed network of servers called onion routers, protecting you from websites that build profiles of your interests, local eavesdroppers that read your data or learn what sites you visit, and even the onion routers themselves.

This proved to be an ideal solution to my problem. After installing Tor and configuring my program to access web sites through a local proxy port, I was able to make it appear as though my IP address changed periodically. Google lost the ability to track my activity, and I was able to continue collecting data uninterrupted.

This approach may perhaps raise some ethical issues. It is possible, though unlikely, that virus designers could learn from this approach. However, in order to use Tor, special software has to be knowingly installed on a client computer, and web traffic has to be redirected through a running proxy server. It seems unlikely that a virus could do all this work undetected. In any case, throughout the project I continued to instruct the program to sleep for a few seconds between each page access. This was no longer necessary, but I felt it was a considerate compromise to ease the load on Google's servers.

REDUCING THE SCOPE OF CLUSTERS SEARCHED

The Java database interface is designed to be as simple as possible on the development end, but the execution was not extensively optimized. The program uses an online database rather than a local file; this makes it possible to easily access the same data from any location without transferring files other than source code. The down side of this approach is that transferring data takes a noticeable amount of time. Even setting up the data in batch files and transferring several rows at once does not save very much time.

The program can identify new web page entries and associate them with clusters at the rate of about two per second. At this rate, scanning a single cluster of size 1,000 would take 8 minutes. The program was wasting far too much time scanning a few clusters with hundreds of stories in them; and it was also scanning a very large number of clusters that had only two or three stories each – most of which would not contain stories by any of the target sites.

In order to speed up the process, I chose not to scan all available clusters in depth. Instead, I only looked at details within clusters that have at least 25 stories and no more than 100. The total number and size of all clusters may be analyzed without this detail, but for identifying the agendas of individual news organizations, this size limitation still provides a good representative sample of all stories.

FOCUSING ON MORE RELEVANT CLUSTERS

Initially, I did not limit the data to clusters that appeared on the first page of each search. In a search which returned multiple pages of resulting clusters, I treated all clusters as equally relevant. Once the web interface was implemented (as described in chapter 6), I uncovered serious flaws in this method. Many of the clusters were not genuinely related to the topic with which they were tagged. This was due to a high presence of minor stories in which the search term appeared somewhere deep within the text of one article.

To pick one example, consider the search for “Paris Hilton” in March 2006 (scanned, though not included in the final results due to date constraints). The largest cluster in the database contained 105 stories. Readers can currently view this cluster at < <http://news.google.com/archivesearch?q=Paris-Hilton&cid=8616030813290510> >

The first story in this cluster is titled “Morrison or motivation?” It is a story in a blog associated with a paper in Louisville, KY. It begins with this sentence: “While I would argue IU point guard Earl Calloway has the mustache of all staches, it's a **Paris Hilton** no-brainer that Adam Morrison is one of, if not THE, best players in the country.” Throughout the cluster, we find similar stories about professional basketball player Adam Morrison.

These articles have nothing to do with Paris Hilton at all. The only relation is a thrown away remark at the beginning of one article which used the name as a colloquialism. Such was the case in a distressingly large number of the stories that had been scanned.

The reason for this misclassification is that I had a Google News search which returned 89 pages of results, with 20 results per page. Initially the program recorded stories from all 89 pages. Since Google sorts these clusters by relevance, the likelihood quickly falls off that a given cluster after page 1 will truly be about the appropriate subject. Celebrity names are used as easy pop culture references. A politician's name may appear in stories about his or her home state, and only briefly mention the name itself; other stories are about the entire field of presidential candidates rather than one politician in particular. Many online news publications also have side bars containing a selection of randomly chosen popular stories, so a key word may appear on a page from a story that is not even tangentially related.

As a result, as stated initially, I restricted the crawler to interpret only results on the first page as relevant.

DATA ANALYSIS

In this section, I will outline the analysis that was done on the final data. Some of this analysis was generated in the form of graphs on Microsoft Excel, while other data are analyzed using a combination of Java and the WEKA data mining package.

Chapter 8: Viewing Coverage Over Time

As shown in Figure 2, the web interface displays an abridged table showing the number of stories contained within collected clusters on the various topics each month. Once this component was working, I collected all the Google News Stories in an abridged table, and all the Digg stories in another table. Here are the results from Google News, broken down by total number of clustered stories appearing in the first 20 clusters in each month.

<i>Site</i>	<i>Abu Ghraib</i>	<i>Anna Nicole Smith</i>	<i>Barack Obama</i>	<i>Britney Spears</i>	<i>Giuliani</i>	<i>Harry Potter</i>	<i>Hillary Clinton</i>	<i>John Edwards</i>	<i>Mitt Romney</i>	<i>New Orleans</i>	<i>Paris Hilton</i>	<i>Tiger Woods</i>
Aug-06	10	531	225	207	49	124	598	766	469	1446	152	2064
Sep-06	254	422	288	144	420	96	753	66	163	710	196	2544
Oct-06	462	198	185	68	632	306	420	115	150	520	264	375
Nov-06	319	127	382	481	590	341	1431	231	188	528	252	1104
Dec-06	622	89	189	544	203	632	358	338	414	982	263	798
Jan-07	454	117	202	260	288	234	1008	397	1359	1199	372	1256
Feb-07	49	698	261	306	397	245	717	142	548	314	58	550

Table 3: Google News clustered stories by month

The data from Digg is shown in a different way. While we can make a case that the total number of top search stories indicates the prominence of a news topic in a given month, this does not indicate the genuine popularity of a story on Digg. For Digg stories,

our table will display two numbers: The total number of stories found, followed by the average reader-assigned score of each story.

<i>Site</i>	<i>Abu Ghraib</i>	<i>Anna Nicole Smith</i>	<i>Barack Obama</i>	<i>Britney Spears</i>	<i>Giuliani</i>	<i>Harry Potter</i>	<i>Hillary Clinton</i>	<i>John Edwards</i>	<i>Mitt Romney</i>	<i>New Orleans</i>	<i>Paris Hilton</i>	<i>Tiger Woods</i>
Aug-06	5, 6.80	0	1, 19.00	43, 4.49	8, 8.12	24, 6.08	22, 6.55	0	0	47, 6.79	103, 4.76	24, 4.83
Sep-06	9, 9.33	26, 5.31	8, 7.62	18, 4.17	9, 5.33	29, 4.52	10, 4.80	0	1, 5.00	24, 4.08	88, 4.62	19, 4.47
Oct-06	6, 6.67	7, 6.43	19, 6.26	16, 2.25	0	37, 5.11	18, 7.78	1, 1.00	1, 5.00	9, 3.00	52, 3.44	9, 3.44
Nov-06	7, 7.14	6, 1.67	15, 16.73	151, 4.50	20, 6.15	45, 4.24	9, 6.00	7, 11.57	8, 7.25	17, 5.24	87, 3.23	14, 4.29
Dec-06	2, 4.00	7, 2.14	31, 9.06	226, 3.28	9, 4.44	76, 5.38	13, 3.54	35, 6.94	0	32, 4.38	121, 3.37	16, 4.94
Jan-07	9, 11.33	10, 2.30	77, 6.99	158, 3.13	20, 8.25	38, 5.21	79, 7.82	12, 6.33	10, 4.50	91, 4.98	168, 4.61	29, 3.52
Feb-07	5, 9.00	451, 3.82	124, 5.78	373, 3.21	77, 8.47	116, 4.34	51, 6.06	41, 10.80	24, 5.21	63, 4.71	131, 2.63	19, 4.58

Table 4: Digg submissions by month

Digg is a relatively new site, and the number of stories submitted has increased as the site increased in popularity. Searches that extend back before August tend to have very sparse data associated with them. Although early searches went all the way back to January 2006, these rows were eliminated in order to consider consistent levels of Digg submissions.

Now there are two “scores” for each topic in each month: a magnitude of “relevant” stories from Google, and an average score from Digg. The data can be reduced by grouping all topics within the categories defined in Table 1.

We will then further normalized the monthly data by taking the maximum value from each column, and dividing the entire column by that number. For instance, in August 2006, the “Entertainment” column from the Google data showed an average of 1094 stories per subject (124 for “Harry Potter” and 2064 for “Tiger Woods”). This is the highest category score for that month, so we divide the entire row of four topics by this value to get the weight. This reduces all cells to a single value with a magnitude

between 0 and 1, which makes it possible to make meaningful comparisons between months. A similar weighted score is obtained by normalizing the average score for each category in Digg.

This approach yields a time sequence for both Google News and Digg, which are graphed on the next page. On the following page, you will find several charts comparing coverage on two of the four topics, celebrities and politicians.

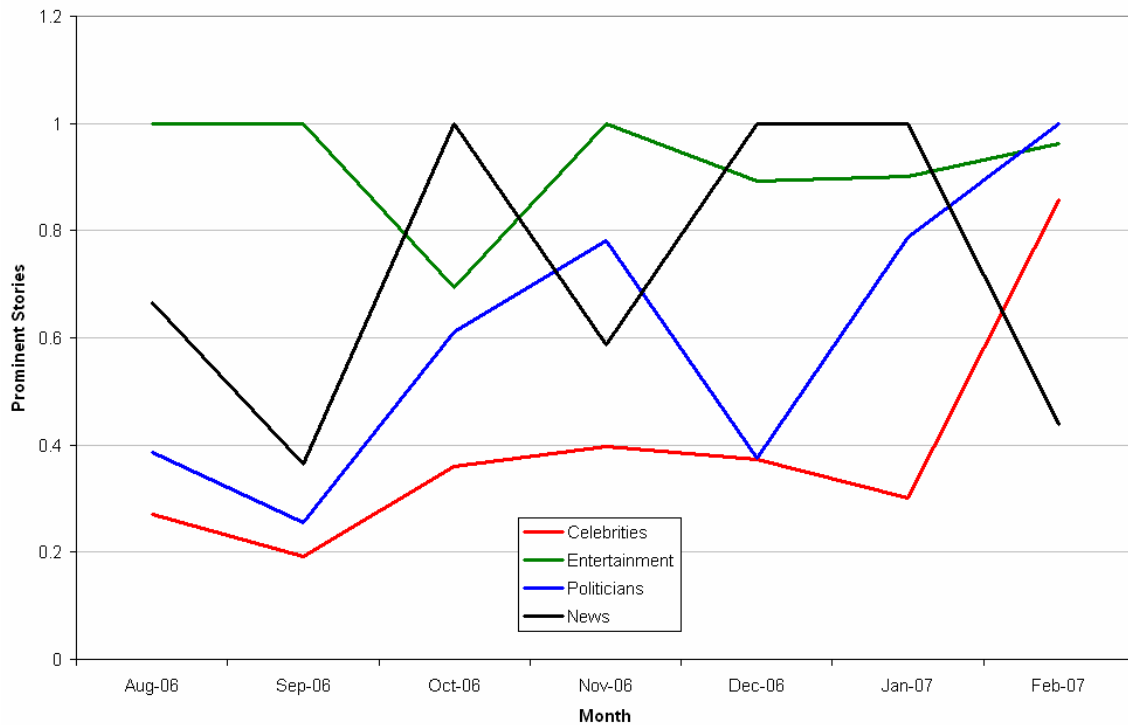


Figure 3: Monthly story priorities from Google News

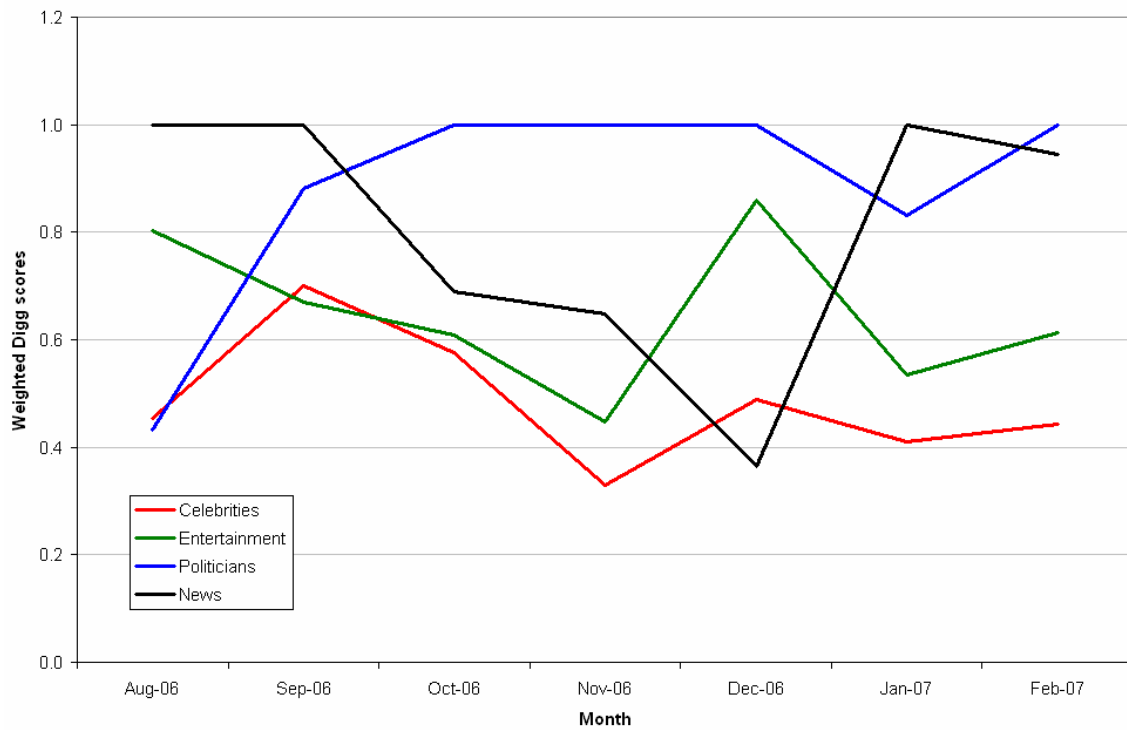


Figure 4: Monthly story priorities from Digg

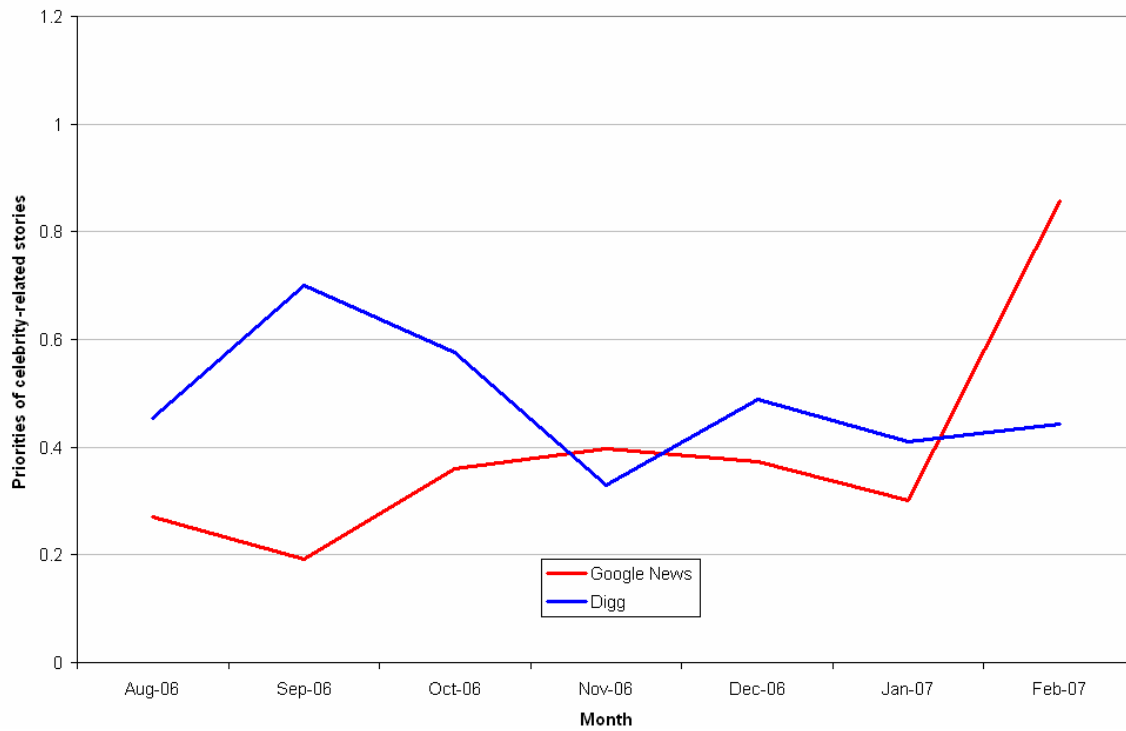


Figure 5: Comparison of focus given to celebrities

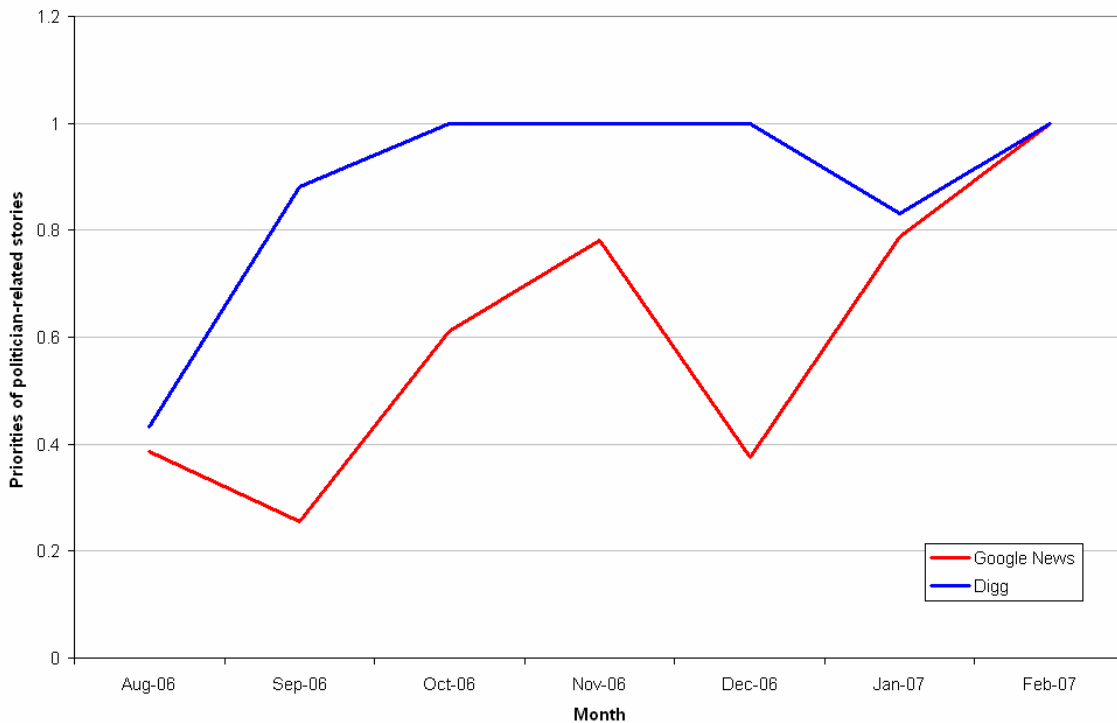


Figure 6: Comparison of focus given to politicians

For the most part, figure 5 does not indicate an overly strong focus on celebrities that is out of line with public opinion. Both Digg and Google maintain consistently low interest in the celebrities listed. The spike in February indicates a fairly high amount of coverage on the death of Anna Nicole Smith. Interestingly, this was seen as a significant event based on available news stories, but Digg readers regarded it as relatively uninteresting.

On the other hand, Digg readers maintained a consistently high level of interest in political stories, which does not appear to be shared on the news graph until the beginning of 2007, when media outlets started to interest themselves with the primary season.

Chapter 9: Topic-specific results in Digg

The Digg data is crucial for testing hypotheses about what topics people find interesting. All search terms are associated with two numbers: the total number of stories identified, and the average score (number of Diggings) per story. The results are indicated below. They cover a period from August 2006 through February 2007, as in the results above. The topics are sorted by average score. “Count” indicates the total number of stories submitted within the designated time period.

Topic	Score	Count
Giuliani	10.01	243
Abu Ghraib	8.22	49
John Edwards	7.67	189
Barack Obama	7.24	348
Hillary Clinton	6.29	291
Mitt Romney	5.36	72
New Orleans	4.95	340
Harry Potter	4.84	511
Tiger Woods	4.13	150
Paris Hilton	3.71	891
Anna Nicole Smith	3.62	632
Britney Spears	3.41	1238

Table 5: Priorities found in Digg topics

Several facts are immediately clear when we view the data in this format. Our three “celebrity” subjects all have the highest story counts and the lowest average scores. The large “Count” fields indicate that many people found stories about those people that they considered were worth submitting. However, the scores tell a different story. It appears that, while many people submit stories about Hilton, Smith, and Spears, the general Digg population does not regard these stories favorably. This supports the

hypothesis that news readers do not prefer to read stories about fluff topics; whereas they do tend to favor stories about politicians and other more weighty news subjects.

Rudy Giuliani stories received the highest average score, 10.01. This includes only data through February. When expanding the data range to include stories in all times, the preference for the “Giuliani” topic becomes even more pronounced, with Giuliani stories gaining an average score of over 18.

Thus, we might be tempted to assume that Giuliani must be the most popular candidate. However, if we look at the Digg page with the all-time highest rated stories on Giuliani⁷, we see a very different story.

- Mr Giuliani Please Stop Mentioning 9/11
- Rudy Giuliani Constitutionally Ineligible To Be President
- Anger at Giuliani 9/11 fundraiser "\$9.11 for Rudy" in poor taste
- America's Worst Nightmare: President Giuliani
- Giuliani: "For Me Every Day Is An Anniversary Of Sept. 11" GET OFF IT!
- Rudy Giuliani: "Freedom is Slavery"
- Rudy Giuliani's daughter is supporting Barack Obama
- DIGG this! Soldier to Giuliani: Have you done your foreign policy homework?
- Reporter Arrested on Orders of Giuliani Press Secretary
- Giuliani Closed Off Streets to Avoid 9/11 Victims' Families

Notably, these stories are all negative. There is not a single friendly news story in the first page of Giuliani links. (It bears mentioning that these stories did not all occur in or before February; they are taken from the set of all Digg data and include stories up to early October 2007.) One interpretation of this result is that what people really want from the news is more stories that speak badly of Rudy Giuliani. Many top stories regarding other politicians are similarly negative, but no other candidates generate nearly as much negative interest as Giuliani does.

⁷ <http://digg.com/search?s=giuliani&submit=Search§ion=all&type=title&area=all&sort=most>
Visited 10/02/2007

Chapter 10: Correlating Results with Individual Media Sources

I used the Java API from the WEKA package to analyze the final results. My Java program has a class called FinalResults which is responsible for reading significant information out of the database and converting it to an arff file, which is the format used by WEKA.

The format of the arff file for clusters is expressed in this way:

```
@RELATION newscluster
@ATTRIBUTE Topic {'Anna Nicole Smith', 'Barack Obama', 'Britney Spears', 'Giuliani', 'Gulf Coast', 'Harry Potter', 'Hillary Clinton', 'John Edwards', 'Mitt Romney', 'New Orleans', 'Paris Hilton', 'Rupert Murdoch', 'Tiger Woods'}
@ATTRIBUTE Month Numeric
@ATTRIBUTE Timestamp DATE "yyyy-MM-dd HH:mm:ss"
@ATTRIBUTE abcnews {0,1}
@ATTRIBUTE bbc {0,1}
@ATTRIBUTE wsj {0,1}
@ATTRIBUTE washingtontimes {0,1}
@ATTRIBUTE cnn {0,1}
@ATTRIBUTE foxnews {0,1}
@ATTRIBUTE guardian {0,1}
@ATTRIBUTE nypost {0,1}
@ATTRIBUTE nytimes {0,1}
@ATTRIBUTE salon {0,1}
@ATTRIBUTE usatoday {0,1}
@ATTRIBUTE washingtonpost {0,1}
@ATTRIBUTE digg {0,1}
```

A row in this table takes this form:

```
"Hillary Clinton",3,"2006-03-02 00:00:00",1,0,0,0,0,0,0,1,0,0,0,0
```

In this example, we are looking at a cluster about Hillary Clinton, from month 3 (counting up from January 2006, this means March of 2006), which contains stories from both ABC News and the New York Times.

WEKA allows visualization of data attributes, which makes it immediately possible to inspect the data results and see graphs such as this one:

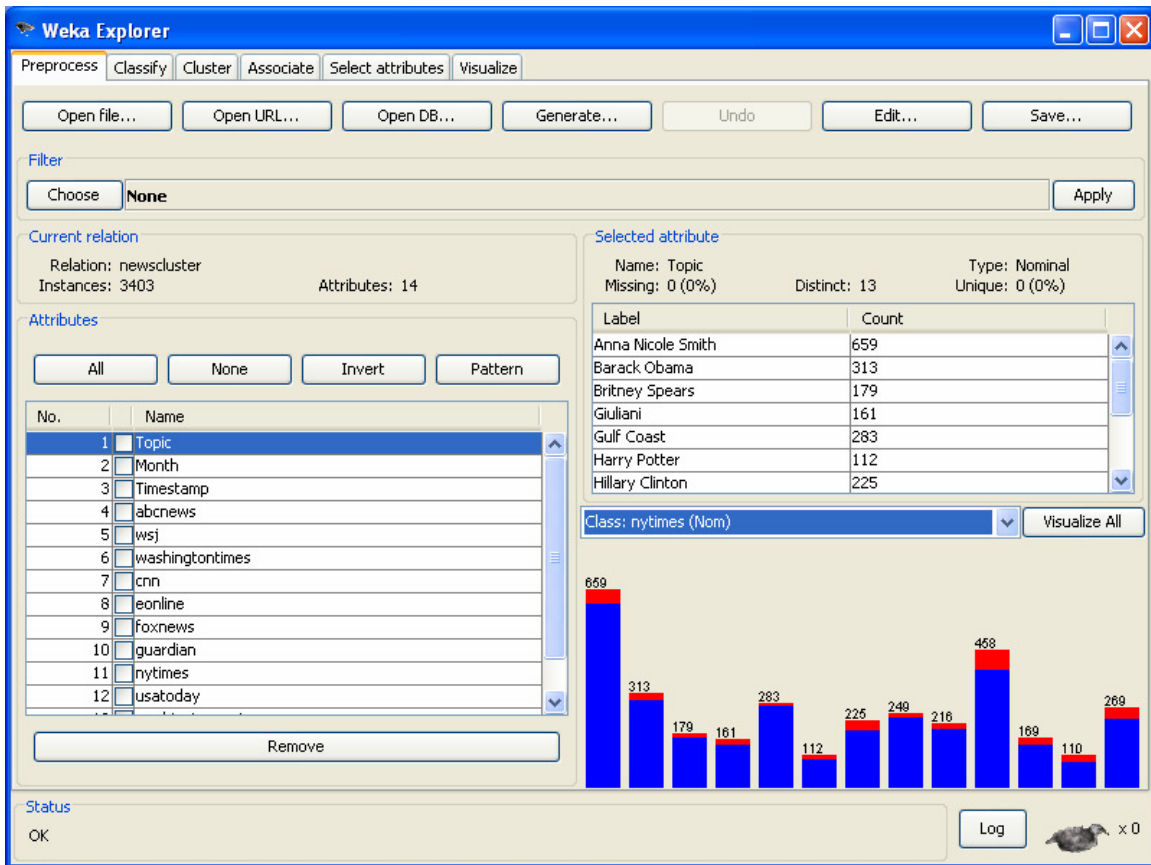


Figure 3: Data visualization

By selecting the “Topic” line on the left it is possible to see bars representing the number of clusters covering each topic. After selecting “nytimes” as the class, the stories are highlighted in red (indicating clusters that include a story by the New York Times) and blue (indicating a cluster in which the New York Times did not participate.)

In the above figure you can see some general features of the data. An individual news source such as nytimes has a small overall presence in the clusters, but higher red bars indicate greater coverage. For instance, the cluster on the left represents stories about Anna Nicole Smith, and the red portion fills about 7% of the bar, indicating New York Times stories about Anna Nicole Smith. In this example, one of the highest proportions is given to the topic of New Orleans, the topic with 458 clusters. The New

York Times has a presence of 15% in stories about New Orleans. It is important to note that this graph comes from an early iteration of the data scan, and may not reflect the current final data.

Once the data is placed in this format, it is then transformed into a series of new .arff files, each of which contains the results from a single news source. Since, as mentioned above, the New York Times has a story in 7% of 648 clusters, about Anna Nicole Smith, the intermediate table will contain a row value of “nytimes”, “Anna Nicole Smith”, 7%, 648. In this way, a data file is generated for each site, and each file contains one row for every topic, revealing what percent of clusters have stories from the selected source.

In the final step, the data was normalized again. The reason this was done is that we are not merely interested in the overall presence or absence of a source in a topic; we are interested instead in the priority. If the New York Times is present in 15% of all clusters about New Orleans, but only 10% of clusters overall, then obviously New Orleans has a higher priority for nytimes than most clusters. Therefore, I normalized the values of all results from each site so that they fit into a range from 0 (lowest priority) to 1 (highest priority). These are inclusive values: Every row in the final table contains at least one “1” and at least one “0”.

Like the Google News sources, the Digg topics were also normalized to bring them within a range from 0 to 1, though these values were based on average scores rather than story counts. As seen in Table 5, Britney Spears was the lowest scoring Digg topic and Giuliani was the highest. The final results of normalized topics across all sites are tabulated below.

Source	Abu Ghraib	Anna Nicole Smith	Barack Obama	Britney Spears	Giuliani	Harry Potter	Hillary Clinton	John Edwards	Mitt Romney	New Orleans	Paris Hilton	Tiger Woods
abcnews	0.42	0.00	0.35	0.00	0.74	0.40	0.51	1.00	0.51	0.33	0.17	0.03
bbc	0.42	0.00	0.22	0.68	0.31	0.71	0.01	0.02	0.55	0.18	0.23	1.00
cnn	0.23	0.00	0.43	0.30	0.66	0.00	0.72	0.72	1.00	0.02	0.09	0.07
foxnews	0.51	1.00	0.37	0.36	0.71	0.16	0.71	0.47	0.63	0.24	0.45	0.00
guardian	0.40	0.34	0.29	0.09	0.61	0.10	0.51	0.76	1.00	0.30	0.00	0.02
nypost	0.15	0.97	0.00	0.47	1.00	0.50	0.44	0.40	0.23	0.15	0.74	0.10
nytimes	0.63	0.00	0.55	0.09	1.00	0.00	0.98	0.10	0.30	0.30	0.16	0.27
salon	0.69	0.71	0.72	0.12	0.85	0.20	0.97	1.00	0.60	0.18	0.53	0.00
usatoday	0.00	1.00	0.50	0.15	0.25	0.16	0.29	0.10	0.71	0.93	0.27	0.22
washpost	0.42	0.13	0.94	0.19	0.87	0.20	1.00	0.72	0.99	0.38	0.00	0.11
washtimes	0.65	0.02	0.96	0.45	0.60	0.24	0.74	1.00	0.85	0.30	0.00	0.01
wsj	0.70	0.00	0.52	0.90	0.99	0.68	0.62	0.33	1.00	0.20	0.00	0.11
digg	0.73	0.03	0.58	0.00	1.00	0.22	0.44	0.65	0.30	0.23	0.04	0.11

Table 6: Detailed priorities of all media sources

How do the various news sites stack up in their coverage of fluff news? As noted before, the three celebrity topics – Paris Hilton, Britney Spears, and Anna Nicole Smith – were the lowest scoring Digg topics, so media sources that devote a great deal of coverage to those subjects are generally out of line with public opinion. Anna Nicole Smith received extensive coverage from Fox News, New York Post, and USA Today. Paris Hilton tended to be a generally low scoring topic, aside from receiving a priority of 0.74 from New York Post. Britney Spears also tended to score low, except in the Wall Street Journal (0.90) and the BBC (0.68).

As for presidential candidates, Giuliani does seem to compel some of the highest levels of coverage, receiving priorities of { 1, 1, .99, .87, .85, .74, .71 } from 7 of the 12 media sites. Hillary Clinton and John Edwards are also clear favorites of the media, each receiving a priority above 0.7 from 6 sites, and above 0.9 from 3 of the sites. However, Clinton’s best coverage comes from “mainstream” news sites – Washington Post and New York Times; whereas John Edwards appears to do better with TV news and the right

wing Washington Times. (This may have something to do with played-up stories about his expensive haircuts.)

As a final analytical step, I used WEKA's clustering tool to see which sources are most aligned with each other, and also to see which news sites are most closely aligned with Digg. I picked three as the number of clusters, and simple K-Means as the clustering algorithm. By this measurement, sites would be considered similar to each other based on the least mean sum of the squares for all priorities.

The clusters I got as a result were:

1. ABC News, Wall Street Journal, Washington Times, CNN, Guardian, Washington Post

Key topics: John Edwards (0.76), Giuliani (0.75)

2. BBC, Fox News, New York Post, Salon, USA Today

Key topics: Anna Nicole Smith (0.74), Giuliani (0.62)

3. Digg, New York Times

Key topics: Giuliani (1.0), Hillary Clinton (0.71)

Tentatively, this indicates that the New York Times is closest to Digg in terms of delivering content that people are interested in reading. Of course, it is worth bearing in mind that Giuliani and Clinton are both New York politicians, so it is natural that a New York paper would cover them; it may just be a coincidence that those happen to be topics of relatively high interest to Digg readers. The second cluster, containing both of the Murdoch run publications and the notoriously fluffy USA today, seem to take interest in Giuliani stories, but spend too much time on a topic like Anna Nicole Smith. Of course, these numbers do not indicate whether the coverage on a given topic is generally positive or negative, which is obviously an important consideration.

On the other hand, this clustering scheme should be taken with a grain of salt. For instance, the BBC is in a cluster with a generally high Anna Nicole Smith rating, even though the BBC itself has the lowest priority on Anna Nicole Smith. Although the relation between BBC and Fox News within a cluster may not be immediately apparent, when we inspect table 6 we can get a better sense of the similarities. BBC and Fox are similar not in their highly covered stories, but in their less frequently covered stories. For instance, both sites have similarly low-to-middling coverage of Abu Ghraib, New Orleans, and Barack Obama.

POTENTIAL OBJECTIONS TO METHODOLOGY

This study would not be complete without acknowledging and addressing some of the shortcomings of the methods used. I would like to present a few possible objections to the methodology, and consider future avenues for studying this topic.

Digg users do not represent the general public.

Even though I earlier raised the point that people responding to media surveys are a self-selected group, this may be still truer of Digg users. First of all, all Digg users have internet access, which implies that they are more likely be people of means than the entire population. Second, Digg users are active participants in an online community, which implies a higher level of engagement than the average TV watcher.

Although these are valid concerns, at worst we can say that the Digg community recommends a distinct demographic of the population. Considering that this is a demographic which is particularly involved in reading and discussing the news, it seems to me that these results would be significant to media marketers, even though they cannot be universally applied. In order to gather better information, it might help to collect web traffic statistics directly from major news sites, rather than relying on a voluntary scoring system.

It is not appropriate to normalize the data. The total number of news clusters / Digg submissions is a significant factor, yet the data selection disregards this factor.

Clearly, mine are not the only acceptable methods for breaking down the data. My reasoning in focusing on the presence of a news source in each news cluster is this: A news cluster indicates that *something has happened* with regard to a particular topic. The

sample space I used, clusters sized between 25 and 100 stories, insures that the event is relatively noteworthy but not universally covered. The extent to which a media source chooses to cover or not cover this event definitely indicates their priorities.

As for the number of total stories per topic: We might suggest that the New York Times should have covered more stories about Blackwater in 2006, because Blackwater eventually proved to be a newsworthy topic. However, since almost no sources covered Blackwater in 2006, the fact that the New York Times did not have this foresight should not necessarily be considered an issue with their reporting.

In the future, it would be a relatively simple matter to come up with and apply more techniques for breaking down the data in various ways. The program development and data collection were the most time consuming phases; now that this groundwork is complete, many other analyses could be applied.

The filtering process relies too much on Google's algorithms. This assumes that Google has done its work correctly.

This is a fair criticism. Part of the point of this report was to investigate a method for gathering existing data, which can be applied in other work. Google functions as a "black box" in this experiment, and it is assumed that the search engine prioritizes stories in an appropriate way. However, Google may have hidden biases that color the data. We can probably leave the analysis of Google's bias up to future studies.

To address this concern, we might use Lexis-Nexis instead of Google. Or, future iterations of this project might require implementing a program that directly monitors news sites to obtain new stories. An advantage of this would be that a story's prominence within the site could be scanned and recorded, so front page stories could be distinguished from minor stories.

CONCLUSION

The introduction to this paper raised the possibility that media sources might give an undue amount of attention to sensationalist news. The analysis provided in this report does not indicate that this is universally the case. However, in Figure 5, there was a clear spike in celebrity news following a major celebrity event – the death of the relatively minor celebrity Anna Nicole Smith – which was not accompanied by a spike in interest by Digg readers. The data shows that news readers in general, as represented by Digg users, are generally uninterested in stories about celebrities and far more interested in coverage of more serious topics, particularly politics. Some media sources deliver this information, while others appear to place too high an emphasis on sensationalism. It is this type of difference that is most interesting to investigate.

What are some possible explanations for the discrepancy? We can speculate on various reasons. Here are a few ideas.

1. **Reporting on celebrities is cheaper.** Serious journalism requires man power, travel expenses, and educated research. By contrast, celebrity news only requires a small number of dedicated photographers and interviewers, many of whom are paid for the material they produce but not salaried. Therefore, despite a desire for hard-hitting news among readers, perhaps it is more cost-effective to produce a lot of stories that generate minor interest.
2. **Serious journalism generates serious enemies.** Reporting serious news often involves coming into conflict with people who have power and money. Journalists often criticize multinational corporations which employ large legal

teams, or politicians who have the power to pass punitive legislation. Writing about celebrities is less risky, even if they are wealthy celebrities.

3. **Fluff reporting is presented as a deliberate distraction.** This is the “bread and circuses” model of the media. Perhaps media conglomeration has put a large number of news corporations in the hands of a small group of owners with a specific political agenda, and this agenda runs contrary to responsible journalism. Under this theory, the media has hidden incentives not to report serious stories, and therefore misdirects the public with stories about celebrities and similar irrelevancies.
4. **Due to limitations of this study, it may simply be that fluff sells better than the data implies.** As the Methodology Objections section notes, Digg users are not necessarily representative of the general public, and fluff may be more popular among general readers than among Digg readers. Even if this were not the case, a niche market obviously exists for fluff journalism. The results of Table 6 can be further studied for an interesting perspective on which news sources cater to which niche market. It could be instructive to compare their discovered priorities with their stated missions and see if the two are at odds.

I am not a student of journalism, and so I cannot do much more than guess which answer is correct, if any. Nevertheless, I believe this is an interesting topic to explore, as it may shed light on which media outlets are really in tune with the pulse of the public, and which are merely pushing their own agenda.

It is my hope that the program developed for this report will ultimately be a first step in a larger contribution to the study of journalism. The concept is tremendously

expandable, and further work could provide more comprehensive content analysis on individual media sources. Depending on the needs of the journalism community, the program could be modified to perform in-depth monitoring on one or more individual newspapers over a period of time. This might serve as an interactive online research aid.

As an example of how this might be put to use, stories could be stored in association with key words. An interested researcher could ask to retrieve all stories about, for instance, Giuliani, and then add additional tags or mark the stories based on the tone of the coverage, labeling each story as positive, negative, or neutral. Essentially this is the same sort of thing that researchers already do when reading a newspaper, but an extended version of this project might act as a way to massively streamline the amount of work that needs to be done.

The potential of the web as a research tool is vast, and developers are still discovering novel applications for this power every year. Combining the power of programmatic text readers with the dedication of a large, independent user base will continue to open up new and surprising avenues of study for a long time to come.

Appendix

Readers may be interested in exploring the data on which this report was based. I invite you to view the web application that was described in Chapter 6. At the current time, my data is publicly available at the following address:

<http://www.apollowebworks.com/newsminer>

Glossary

HTML

HyperText Markup Language. It is the predominant markup language used in web pages.

DOM

Document-Object Model. Most programs that analyze HTML and XML documents rely on the creation of a well-formed tree of document elements, which can be analyzed for content and presentation layouts.

URL

Uniform Resource Locator. This is a unique identifier assigned to every page on the web, which can be used to locate and retrieve its contents.

SQL

Standard Query Language. Executing SQL statements is the most common method for searching and updating tables in a database.

WEKA

Waikato Environment for Knowledge Analysis. WEKA is a free software package used for analyzing data.

API

Application programming interface. An API is a source code interface that an operating system or library, such as WEKA, provides to support requests by computer programs.

CGI

Common Gateway Interface. CGI is a standard protocol for interfacing external application software with a web server.

References

- The Daily Show. (June 11, 2007). "Paris Hilton Gets In a Car." Video reproduction retrieved 11/1/2007, from http://www.spikedhumor.com/articles/110359/The_Daily_Show_Paris_Hilton_Gets_in_a_Car.html
- Severin, W. and Tankard, J. 2000. Communication Theories: Origins, Methods and Uses in the Mass Media. New York: Allyn & Bacon
- Tan, P., Steinbach, M., and Kumar, V. 2005. Introduction to Data Mining. Boston: Pearson Education, Inc.
- Witten, I. and Frank, E. 2005. Data Mining: Practical Machine Learning Tools and Techniques. San Francisco: Morgan Kaufmann Publishers
- Niels Provos. (July 9, 2007) "The reason behind the 'We're sorry...' message." Retrieved September 29, 2007, from <http://googleonlinesecurity.blogspot.com/2007/07/reason-behind-were-sorry-message.html>
- Tor: Anonymity Online. <http://tor.eff.org/>
- Weinberger, D. 2007. Everything Is Miscellaneous: The Power of the New Digital Disorder. New York: Times Books
- Mindich, D. 2005. Tuned Out" Why Americans Under 40 Don't Follow the News. Oxford University Press USA

Vita

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