Electronic Sources of Network Analysis

Data Collection Complexity

- social networks studied by observation
- Disadvantage requires close involvement for researcher
- Standardized surveys minimize influence of observer
- Disadvantage rely on active engagement of population to be studied
- Doubts whether responses are spontaneous and genuine
- Need multiple surveys to learn network dynamics
- Manual methods labor intensive 50% time spend on data collection
- Researchers needs to reanalyse to get accurate results
- So need to look for alternatives

- Solution reuse existing electronic records of social interaction
- Examples:
- publication or project databases showing collaborations among authors or institutes
- Corporate DB study networks of innovation
- News paper archives study of topics in politics
- But has significant price tag to access patent databases, media archives, legal and financial records
- Internet vast, diverse, dynamic and free for all resource
- Rely on electronic networks and online information sources data collection process automated
- Allows to exploit dynamics of electronic data to perform longitudinal analysis

Electronic Discussion Networks

- Firs study to prove versatility of electronic data series of works from Information Dynamics Labs of Hewlett-Packard
- Tyler et al studied communication among employees using corporate email archive
- Recreated network assigning tie between employee, if exchanged certain threshold of emails
- Also studied leadership role, formal and informal communities
- Adamic and Adar revisits local search find shortest path using local info
- Than finding most connected people use of additional info such as physical location and position results quick search
- Study using email electronic data limits by privacy issue

- Public forums and mailing lists has no privacy issues
- Ex. Analysis of USENET groups
- WWW mailing lists due to its openness within working group
- Gloor used headers of messages to automatically re-create the discussion networks of working group
- His dynamic visualization quickly identify key discussion

Blogs and Online Communities

- Blogs "personal publishing" or a "digital diary"
- Analysis of Blogs used for marketing trend analysis tools use content anlaysis
- Modern tool enables easy comment, react to comment, bloggers communications – leads to dynamic communities
- This manifests through aggregated blogs, blog rolls, Blog Walk series of meetings
- Hence blogs can be used to study network analysis.

Blogs and online communities

- Blogs enables researching due to structured data of RSS(Rich Site Summary)
- RSS aids dynamic analysis as contains metadata timestamp of the content
- Efimova and Anjewierden first to study blogs from a communication perspective
- Adar and Adamic offer visualization of communication in blogs

- Ex. US election 2004 exploited blogs to build networks among individual activists and supporters
- Blog analysis enables marketers to understand interested product choices of young demographics
- Blog Analysis becomes object of study lead to series of Sunbelt social networks conference

- Online community spaces and social networking services cater more socialization than blogs e.g. MySpace, LiveJournal
- Used by much younger demographic enables to study changes in youth culture
- LiveJournal data exposes data such as interest and social networks of users
- Backstrom et al. used this data to study influence of certain structural properties on community formation and community growth, how changes in the membership of communities relates to change in underlying discussion topics

- LiveJournal exposes data for research purposes in a semantic format (an exception, most don't do)
- Most online social networking services guard the data even from its user e.g. Friendster, Orkut, LinkedIn etc.
- A technological alternative to these centralized services is the FOAF network
- FOAF profiles stored on the website of users, linked together using hyperlinks
- Drawback of FOAF lack of tools to create, maintain profiles and exploit network

Web based Network

- Web vast, diverse and free to access nearly up to date
- Downside:
- quality of information varies significantly
- reusing for network analysis (web mining) requires efficient search provided by only commercial search engines
- Two features of web pages considered as basis of extracting
- social relations:
- links and co-occurrences
 - linking structure represents real world relationships
 - links are authoritative and relevant as it is chosen by author

- Drawback:
- Direct links between personal pages are very sparse as they use browsing of web as mode of navigation (than putting link, update link etc.)
- Automating this task for network analysis, results in home page search problem
- Hence studies are made on linking structure at higher level
- Ex.
- Heimeriks et al. studied communication and collaboration networks across different fields of research using a multi-layered approach

Co-occurences

- Co-occurrences of names in web pages serve as evidence of relationships and frequent phenomenon
- Extracting relationships based on co-occurrence of names requires web mining (as names embedded in text of web)
- This statistical methods is combined with analysis of the contents of web pages
- Web mining first tested for social network extraction by Kautz et al. on ReferralWeb project for referral chaining
- Referral chaining looking for experts with a given expertise close to the user of the system

- Referralweb extracted through co-occurrence analysis and page counts using the search engine, Altavista
- It collected page counts for individual names and number of pages where the names co-occurred
- Disadvantage: very shallow parsing of the web page as indirect references are not counted
- Ex. "the president of the United States" will not be associated with George Bush

Jaccard-coefficient

- Tie strength = number of co-occurrences / number of pages returned for the two names individually => Jaccard-coefficient
- The resulting value is tie strength 0 1 zero (no co-occurrences)
 and one (only co-occurrences)
- If this number exceeds certain fixed threshold was taken as evidence for the existence of a tie
- Jaccard-coefficient takes relative measure of co-occurrence and not absolute sizes of the sets
- Expertise of individual are extracted using proper name extraction,
 NLP technique, it is then used to extract new names (repeated 2 or 3 times) [snowballing technique]

- Kautz never evaluated his system for accuracy, he proved his recommendation system by, what it is based on and indicate the level of confidence in its decisions
- He proved it is better than official records, as personal pages are more up to date
- Different from Kautz approach, extraction of names and finding tie between names by SE is a quadratic problem, Matsuo et al. first extracted possible contacts from results of search engine for the individual names
- This significantly reduces the number of queries that need to be made to SE at a minimal loss.

- Jaccard-coefficient penalizes ties often occurs on the but less popular individuals
- To address this variant is used
- It divides the number of pages for the individual with the number of pages for both names
- if this number reaches a certain threshold, treated as tie

- Another approach to calculate the strength of association between the name of a given person and a certain topic
- Strength determined by the number of the pages where the name of an interest and the name of a person co-occur / total number of pages about the person
- Mutschke and Quan Haase, clustered keywords into themes based on the cooccurrences of keywords on publications in bibliography records
- assign documents to themes
- subsequently determines themes relevant for a person based on his or her publications

Problem of ambiguous names

- biggest technical challenge in social network mining is disambiguating person names
- Problem due to polysemy and synonymy
- Different variations of name, names with international characters –
 SE returns partial set of records
- Common names return all pages of all names
- coverage of the Web can be very skewed (over-represented) [(web pages are largely ranked by popularity]

Methods for disamguiation

- Bekkerman and McCallum deal ambiguity problem using limited background knowledge
- Instead of a single name they assume to have a list of names related to each other
- Disambiguate the appearances by clustering the combined results returned by the search engine (based on hyperlinks between the pages, common links or similarity in content) for the individual names

- Bollegala et al went on step further in mining the resulting clusters for key phrases
- i.e. adding key phrases to search query to reduce the set of results to target individual
- For example, when searching for *George Bush the beer brewer one* would add the term *beer to the query*
- Queries are too specific, results in lower recall

Average Precision

- When computing weight of directed link between two persons:
- Consider an ordered list of pages for the first person and a set of pages for the second (the relevant set)
- i.e. ask search engine for the top *N* pages for both persons but in the case of the second person the order is irrelevant for computation
- rel(n), the relevance at position n, where rel(n) is 1 if the document at position n is the relevant set and zero otherwise $(1 \le n \le N)$
- Let *P*(*n*) denote the precision at position *n* (*p* @*n*):

$$P(n) = \frac{\sum_{r=1}^{n} rel(r)}{n}$$

Average Precision

 Average precision is defined as the average of the precision at all relevant positions:

$$P_{ave} = \frac{\sum_{r=1}^{N} P(r) * rel(r)}{N}$$

