Evolution of Web Communities

Introduction

- Web community set of web pages created by individuals or associations with a common interest on a topic
- Link analysis techniques to extract web communities, but not for web community evolution
- Degree of web community evolution metrics growth rate, novelty, and stability
- Web graph helps to identify web community on a topic by extracting densely connected structure
- In a web graph nodes are web pages and edges are hyperlinks
- web community differs from a community of people Ex. web community may include competing companies

Application of Web Community Evolution Analysis

- Answering historical queries about topics on the Web
- Statistics on members (URLs) have appeared and disappeared
- Marketing Products
- Tracking social and cultural trends over time
- Emergence of quality web communities on a specific topic

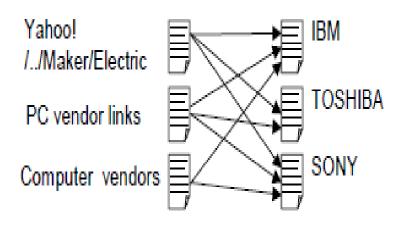
Extraction of Web Community Evolution

- Consists of two parts
- First part extracts communities and their relevance from web archive using web community chart
- Web community chart related communities are connected by weighted edges
- Allows Navigate through related communities, locate evolution around a particular community

- Second Part:
- Web community evolution viewer browsing communities evolution
- Allows to locate evolution of communities emergence and growing

Authorities and Hub

- Concept proposed by Kleinberg
- An authority page with good contents on a topic
- A hub page with a list of hyperlinks to valuable pages on the topic (points to many good authorities)
- HITS algorithm that extracts authorities and hubs from a given sub-graph



Improvement in HITS

- HITS improved exploiting anchor texts, edge weighting, document similarity, and Document Object Models
- Dean and Henzinger tailored HITS for finding related pages
- They proposed Companion, a related page algorithm
- Companion takes seed page as input, outputs related pages to the seed
- Its precision improved by considering the order of links in a page

Work done in Authorities and Hubs

- Kleinberg A set of authorities and hubs was regarded as a community
- Gibson et al. investigated the characteristic of communities derived by HITS
- Kumar et al. extracted > 100,000 cores of communities from web snapshot based on their graph evolution model
 - A core represents small complete bipartite graphs that consist of authorities and hubs
- Lempel and Moran proposed a random walk model for calculating authorities
- Flake et al. redefined a community given seed pages as a subgraph separated from the Web using a maximum flow/minimum cut framework

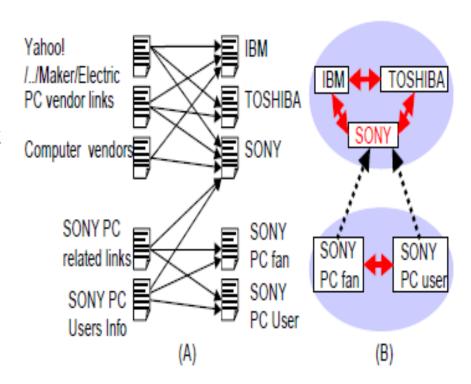
- Change frequency and lifetime of web pages has been studied
- Use this for web crawlers to determine timing for re-crawl
- Based on the page level analysis
- Site (or server) level analysis of web graph evolution is also studied
- Internet Archive to see past web pages stored in the Internet Archive's web archive
- User specify a single URL, see the past pages of that URL
- It is impossible to understand what topics are popular in the past

Overall Working of Web Community Chart Algorithm

- Algorithm builds web community chart from given set of seed pages
- Each seed page applied HITS and related page algorithm (RPA)
- Then, investigate how each seed derives other seeds as related pages
- RPA builds subgraph around the seed, extracts authorities and hubs from the graph using HITS
- Companion
 improves the precision of HITS and Companion
- It prunes error parts of subgraph used in HITS and Companion

Example

- In graph (B) two SONY PC fans, they derive each other and SONY as related pages
- SONY does not derive these fans, (no. of electric company lists > link lists of SONY PC)
- Symmetric derivation is a strong relationship – Ex. Community
- Asymmetric derivation is a weak relationship – Ex. between 2 communities

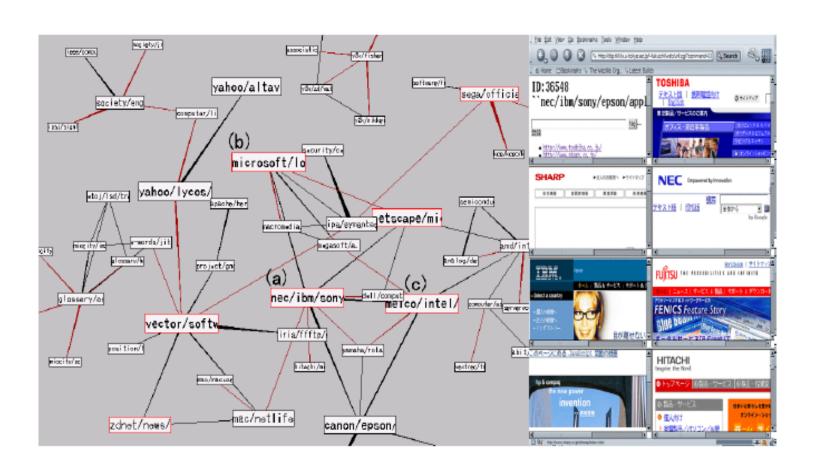


Algorithm for Constructing Web Community Charts

- Requisites:
- Web pages that have inlinks from *more* different servers (parameter to determine seed set)
- Build a directed graph that shows how each seed derives other seeds by Companion– (top *N authorities*, where *N is a parameter*) *called* authority derivation graph (ADG)
- extract a symmetric derivation graph (SDG) from ADG, to extract web communities from SDG
- extract densely connected subgraphs in SDG as cores of web communities
- form cores into communities by adding remaining nodes to these cores

- Core a core is defined as a set of triangles that share edges in SDG
- Extract all triangles of nodes from SDG. Two cores share some nodes, isolate them, and pass them to the next step
- Add remaining nodes in SDG to a neighboring core, if it has edges
 - If multiple candidates, select a core with most incoming edges.
 - Each core then becomes a community
- Remaining connected components that do not form triangles, extract such components as communities

Example of a Web Community Chart



EVOLUTION OF WEB COMMUNITIES

- <u>Types of Changes Emerge, Dissolve, Grow and shrink, Split, Merge</u>
- Notations used:
- t1, t2, ..., tn: Time when each archive crawled
- W(tk): The Web archive at time t_k .
- C(tk): The web community chart at time t_k .
- c(tk), d(tk), e(tk), ...: Communities in $C(t_k)$.

Types of Change

- Emerge: A community c(tk) emerges in C(tk), when c(tk) shares no URLs with any community in C(tk-1)
- Dissolve: A community c(tk-1) in C(tk-1) has dissolved, when c(tk-1) shares no URLs with any community in C(tk)
- Grow or Shrink: The community grows when new URLs appear in *c(tk)*, and shrinks when URLs disappeared from *c(tk-1)*
- Split: A community c(tk-1) may splits into smaller communities, then c(tk-1) shares URLs with multiple communities in C(tk)
- Merge: When multiple communities (c(tk-1), d(tk-1), ...) share URLs with a single community e(tk), these communities are merged into e(tk)

Evolution Metrics

- N(c(tk)): the number of URLs in the c(tk), similarly for others...
- Growth Rate, Rgrow(c(tk-1), c(tk)), represents the increase of URLs per unit time
 - allows us to find most growing or shrinking communities
 - when c(tk-1) does not exist, use zero as N(c(tk-1))

$$R_{grow}(c(t_{k-1}), c(t_k)) = \frac{N(c(t_k)) - N(c(t_{k-1}))}{t_k - t_{k-1}}$$

- Rstability(c(tk-1), c(t)), amount of disappeared, appeared, merged and split URLs per unit time
- No change of URLs, the stability becomes zero
- A stable community is desirable

$$R_{stability}(c(t_{k-1}), c(t_k)) =$$

$$\frac{N(c(t_{k-1})) + N(c(t_k)) - 2N_{sh}(c(t_{k-1}), c(t_k))}{t_k - t_{k-1}}$$

- Rnovelty(c(tk-1), c(tk)), represents the number of newly appeared URLs per unit time
- c(tk) has grown mainly by newly appeared URLs
- Most emerged communities at time tk can be found by sorting communities by the novelty

$$R_{novelty}(c(t_{k-1}), c(t_k)) = \frac{N_{ap}(c(t_k))}{t_k - t_{k-1}}$$

 disappearance rate, Rdisappear(c(tk-1), c(tk)), is the number of disappeared URLs from c(tk-1) per unit time

$$R_{disappear}(c(t_{k-1}), c(t_k)) = \frac{N_{dis}(c(t_{k-1}))}{t_k - t_{k-1}}$$

• merge rate, Rmerge(c(tk-1), c(tk)), is the number of absorbed URLs from other communities by merging per unit time

$$R_{merge}(c(t_{k-1}), c(t_k)) = \frac{N_{mg}(c(t_{k-1}), c(t_k))}{t_k - t_{k-1}}$$

 split rate, Rsplit(c(tk-1, c(tk)), is the number of split URLs from c(tk-1) per unit time

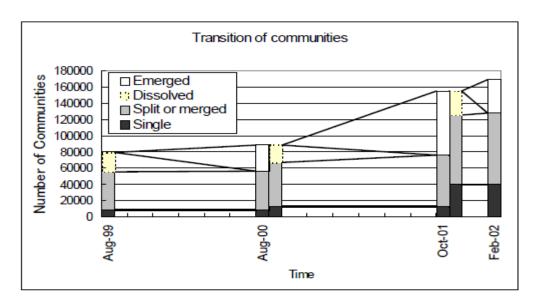
$$R_{split}(c(t_{k-1}), c(t_k)) = \frac{N_{sp}(c(t_{k-1}), c(t_k))}{t_k - t_{k-1}}$$

ANALYSIS

- For experiments, 4 web archives of Japanese web pages (in .jp domain) crawled in 1999, 2000, 2001, and 2002 used
- From 1999 and 2000, 17 million pages in each year collected
- In 2001, it doubled due to improved crawling rate

web graph build with URLs and links by extracting anchors from all pages in the

archive



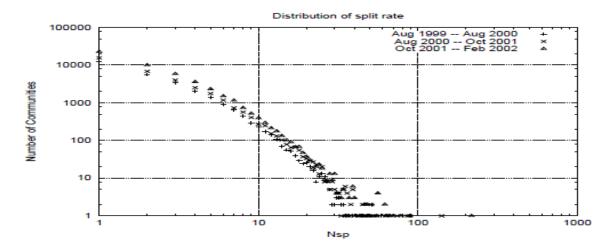


Figure 7: Distribution of split rate

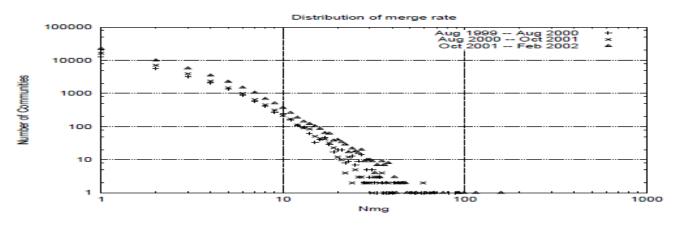


Figure 8: Distribution of merge rate

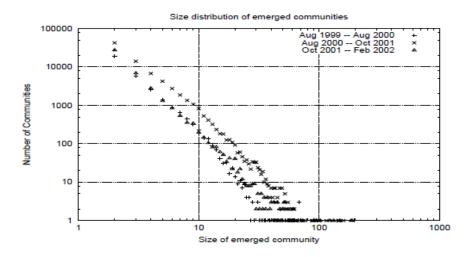


Figure 9: Size distribution of emerged communities

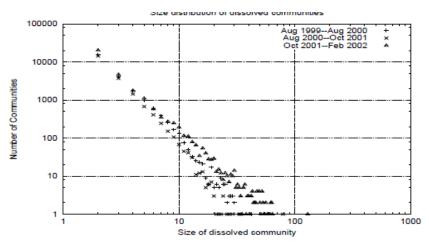


Figure 10: Size distribution of dissolved communities

Findings

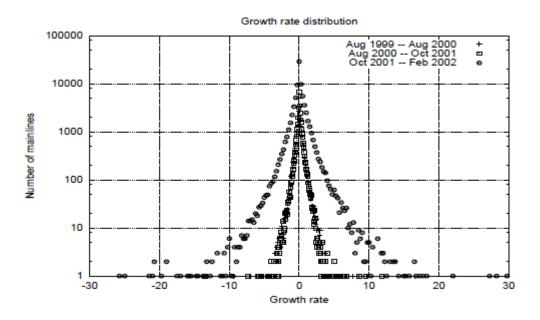


Figure 11: Distribution of growth rate

Findings

- All 4 curves follow power law distribution
- size distribution of communities follows the power law and its exponent (2.9 to 3.0)did not change so much over time
- split or merge rate is small in most cases (almost the same number of communities are merged at *tk with the same rate as the* split rate)
- Emerged and Dissolved Communities small communities are easy to emerge and dissolve
- growth rate is small for most of communities, and the graph has clear y-axis symmetry

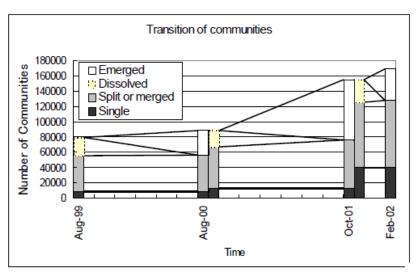


Figure 6: Transition of communities

	t_{k-1} : Aug. 99 t_k : Aug. 00		
# Branch lines at t_{k-1}	28,467	32,490	41,501
# Main lines	26,723	34,396	83,771
# Branch lines at t_k	29,722	41,752	44,305

Table 3: Number of main lines in split or merged communities

Findings

- Types of Changes: from tk-1 to tk:
- Each bar represents the number of communities in charts at the time
- Bars with split vertically, shows it has previous and the next charts to be compared
- more than half of communities are involved in split and merge
- number of single communities is small
- About half of survived communities in 1999 and 2000 are included in main lines for one year
- About 66% of survived communities in 2001 are included in main lines for four months

EVOLUTION VIEWER

- Viewer provides various means to extract evolving communities as follows:
- Searching communities by keywords or a URL
- Sorting and filtering communities by the evolution metrics defined in the viewer
- User can extract various kinds of evolving communities, most growing communities, the most emerging communities related to a specific topic

Example - find a community of Islam in 2001 by the keyword "Islam"

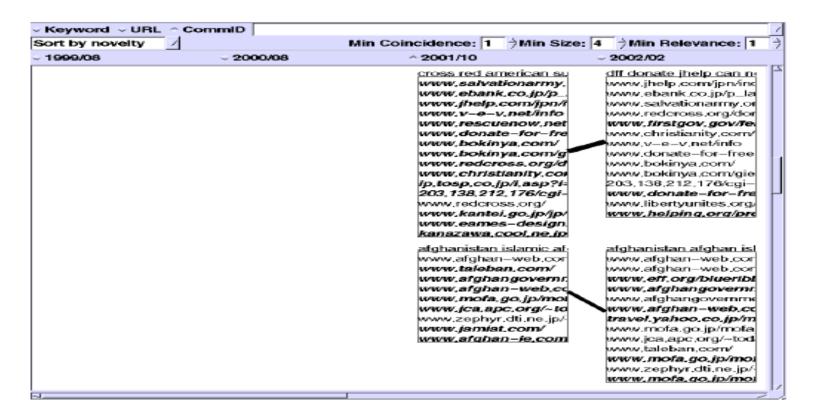


Figure 13: Emerged communities around a pacifist community