Churn Model review of Semantic Overlay Networks in P2P

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Abstract—In this paper, we study the behaviour of a semantic overlay network that supports full-fledged information retrieval in the presence of peer churn. In a peer-to-peer (P2P) system, nodes typically connect to a small set of random nodes (their neighbours), and queries are propagated along these connections Such query flooding tends to be very expensive. We adopt a model for peer churn, and study the effect of network dynamics on peer organisation and retrieval performance. We propose that node connections be influenced by content, so files will connect to other similar nodes. Our work is the first to demonstrate the issues involved in the design of semantic overlay networks when introducing peer churn. The over-lay network is evaluated on a realistic peerto-peer environment using real-world data and queries, and taking into account the dynamics of user-driven peer participation. (SON). Queries are routed to the appropriate SONs, increasing the chances that matching files will be found quickly, and reducing the search load on nodes that have unrelated content. Using this evaluation, we draw conclusions on the performance of the system in terms of clustering efficiency, communication load and retrieval accuracy in such a realistic setting. Thus, semantically related nodes form a Semantic Overlay Network

Keywords-P2P, Semantic Overlay Networks, Query, churn.

1. INTRODUCTION :

grown Peer-to-peer systems (P2P) have dramatically in recent years. They offer the potential for low cost sharing of information, autonomy, and privacy. The main idea behind P2P is that instead of relying on central components, functionality is provided through decentralised overlay architectures. In overlay networks, peers typically connect to a small set of other peers. Queries are then, propagated to the network searching for information qualifying query criteria by utilising existing connections and following a predetermined query forwarding strategy. However, query processing in current P2P systems is very inefficient and does not scale well. The inefficiency arises because most P2P systems create a random overlay network where queries are blindly forwarded from node to node. As an alternative, there have been proposals for "rigid" P2P systems that place content at nodes based on hash functions, thus making it easier to locate content later on (e.g., [5, 6]). Although such schemes provide good performance for point queries (where the search key is known exactly), they are not as effective for

approximate, range, or text queries. Furthermore, in general, nodes may not be willing to accept arbitrary content nor arbitrary connections from others.

Semantic Overlay Networks (SONs) is a flexible network organization that improves query performance while maintaining a high degree of node autonomy. With Semantic Overlay Networks (SONs), nodes with semantically similar content are "clustered" together. To illustrate, consider Fig. 1 which shows eight nodes, A to H, connected by the solid lines.

When using SONs, nodes connect to other nodes that have semantically similar content. For example, nodes A, B, and C all have "Rock" songs, so they establish connections among them.



Fig.1: Semantic overlay Networks.

In Semantic Overlay Networks (SONs), peers that are semantically, thematically or socially similar are organised into groups. SONs, while being highly flexible, improve query performance and guarantee high degree of peer autonomy. This technology has proven useful not only for information sharing in distributed environments, but also as a natural distributed alternative to Web 2.0 application domains, such as decentralised social networking in the spirit of Flickr3. Contrary to structured overlays that focus on providing accurate location mechanisms (e.g., [7]), SONs are better suited for loose P2P architectures, which assume neither a specific network structure nor total control over the location of the data. Additionally, SONs offer better support of semantics due to their ability to provide mechanisms for approximate, range, or text queries, and emphasise peer autonomy.

The management of large volumes of data in P2P networks has generated additional interest in methods for effective network organisation based on peer contents and consequently, in methods supporting information retrieval (IR) (e.g., [4]). Most of these research proposals, while exploiting certain architectural [4] or modelling [1] aspects of peer organisation, assume for their experimental evaluation an ideal scenario where peers never leave or join the network. However, studies of P2P content-sharing systems have concluded that peers are typically dynamic (e.g., [13]). A peer joins the network when a user starts the application. While being connected, the peer can contribute resources to the network and search for resources provided by other peers. The independent arrival and departure of peers creates the collective effect called churn. These userdriven dynamics of peer participation is a critical issue, since churn affects the overlay structure [2], the resiliency of the overlay [14], the selection of key design parameters [11], and the content availability which in turn, affects retrieval effectiveness.

To the best of our knowledge, the work presented in this paper is the first to address the issues involved in the design and the evaluation of the SONs when introducing peer churn. We adopt a model for peer churn proposed by Yao et al. [10], and study the effect of network dynamics on peer organisation and retrieval performance. The overlay network is evaluated on a realistic P2P environment using real-world data and queries. Based on the results of this evaluation, we draw conclusions on the performance of the system in terms of clustering efficiency, communication load and retrieval accuracy.

The remainder of the paper is organised as follows. SON-like structures supporting IR functionality and also research on modeling peer dynamics are reviewed in Sec. 2. Section 3 presents the model used to describe peer churn, while Sec. 4 presents a SON architecture and the related rewiring protocol. The experimental evaluation of the system is presented in Sec. 5, followed by conclusions and issues for further research in Sec. 6.

2 RELATED WORK AND BACKGROUND:

This section provides a brief survey of the work related to data organisation and retrieval in SONs, along with research on modelling the dynamics of P2P networks.

2.1 Semantic Overlay Networks

Initial IR approaches implementing SON-like structures supporting content search in a distributed collection of peers, where a two-tier architecture is proposed. In this architecture, a peer provides content-based information about neighbouring peers and determines how to route queries in the network. Along the same lines, Klampanos et al. [4] propose an architecture for IR-based clustering of peers. In this architecture, a representative peer (hub) maintains information about all other hubs and is responsible for query routing.

Additional work on peer organisation using SONs is based on the idea of "small-world networks by proposing a creative self-organising semantic small world (SSW) network based on the semantics of data objects stored locally to peers. Voulgaris et al. [16] propose an epidemic protocol that implicitly clusters peers with similar content. peers share concepts from a common ontology and this information is used for organising peers into communities with similar contents. Most of these works assume a static peer network for their evaluation.

2.2 Churn Models

Characterising churn requires fine-grained and unbiased information about the arrival and departure of peers in a network. This task is rather challenging due to the large size and highly dynamic nature of P2P systems. Several studies (e.g., [17]) present a high level view of churn by analysing its characteristics (e.g., median session length) in large scale P2P systems. Gummadi et al. [17] measure session lengths by monitoring a router at the University of Washington. Sen et al. [18] analyse P2P traffic by monitoring flows in FastTrack5, All studies infer that session lengths follow some known probability distribution ranging from Poison to heavy-tailed (or Pareto) distributions. To identify the key challenges in characterising churn, determine common pitfalls in measuring churn such as biased peer selection, which they believe are the main factors for the conflicting results, and develop techniques



Fig. 2. On-line and off-line behaviour for peers pi and pj

to address these difficulties. Along the same lines, Leonard et al. [14] present a realistic model for peer lifetimes in P2P networks and investigate the resilience of random graphs to lifetime-based peer failure. Yao et al. [10] introduce a generic model that captures the heterogenous behaviour of peers. They view each peer as an alternating renewal process and consider that on-line/on-line durations are independent and unique for each peer.

3 CHURN MODEL:

We present a model of user behaviour characterising peer arrivals and departures in a P2P system. The model takes into account heterogeneous browsing habits, formalises recurring user participation in P2P systems and explains the relationship between the various lifetime distributions observable in P2P networks. Churn Model. We consider a P2P network with N participating peers. Each peer pi is either alive (i.e., present in the system) at a specific moment or dead (i.e., logged-off). This behaviour is modelled by a renewal process {Zi (t) } for each peer pi.:

$$Z_i(t) = \begin{cases} 1, \, p_i \text{ is alive at time } t; \\ 0, \text{ otherwise.} \end{cases}, \quad 0 \le i \le N.$$
(1)

This framework is illustrated in Fig. 1 that presents the renewal processes { Zi(t) } and { Zj(t) } for peers pi and pj

respectively during time period t. The random variables $L_{i,t} > 0$ and $D_{i,t} > 0$ represent the on-line and off-line durations for peer pi respectively.

The following assumptions are made:

1.To capture the independent nature of peers, we assume that peers behave independently of each other and processes $\{ Zi(t) \}$ and $\{ Zi(t) \}$, for any $i \neq j$,

are independent. This means that peers do not synchronise their arrivals or departures and generally exhibit uncorrelated lifetime characteristics.

2.Although the model is generic enough to allow dependencies between cycle lengths, without loss of generality we treat all lifetime and off-time processes as independent and we use identical distributed sets of variables. Thus, for each process { Zi (t) } its on-line durations { $L_{i,t}$ } are described by some joint distribution $F_i(x)$ and its off-line durations { $D_{i,t}$ } are described by another joint distribution $G_i(x)$. This means that for each peer its on-line and o®-line durations are independent.

Lifetime Distribution: A Pareto distribution [10] is used to represent the on-line durations:

$$F(x) = 1 - (x/x_m)^{-k}; x_m > 0; k > 1$$
(2)

Off-line durations are respectively represented by an alike Pareto distribution G(x). Pareto distribution is parameterised by the quantities xm and k, which stand for the scale and the shape of the distribution respectively. The scale parameter sets the position of the left edge of the probability density. The shape parameter determines the skewness of the distribution.

Peer Availability: Yao et al. [10] define the average online duration (or life-time) of a peer pi as li = E[Li] and its average off-line duration as di = E[Di].

The availability ai of peer pi (i.e., the probability that pi is in the system at a random moment) is calculated as:

$$a_i = \lim_{t \to \infty} P(Z_i(t) = 1) = \frac{l_i}{l_i + d_i}.$$
(3)

According to this model, the only parameters that control a peer's availability are the on-line li and the off-line di durations. Note that these parameters are independent and unique for each peer. Parameters li and di are drawn indepen- dently from two Pareto distributions. Once pair (li; di) is generated for each peer pi, it remains constant for the entire evolution of the system.

4 A SEMANTIC OVERLAY NETWORK FOR INFORMATION RETRIEVAL:

The present work builds on iCluster P2P network [3], which extends the idea of peer organisation in small-world networks by allowing peers to have multiple and dynamic interests. iCluster peers are responsible for serving both users searching for information and users contributing information to the network. Each iCluster peer is characterised by its information content (i.e., its document

collection), which may be either automatically (by text analysis) or manually assigned to each document (e.g., tags or index terms). To identify its interests, a peer categorises its documents by using an external reference system (i.e., an ontology as in [9] or a taxonomy such as the ACM categorisation system) or by clustering. Thereupon, a peer may be assigned more than one interests.

Interests are created and deleted dynamically to reflect a peer's variety in the documents it contributes to the network. Each peer maintains a routing index (RI) holding information for short- and long-range links to other peers:

short-range links correspond to intra-cluster information (i.e., links to peers with similar interests)

long-range links correspond to inter-cluster information (i.e., links to peers having different interests and thus belonging to different clusters) Entries in the routing index contain the IP addresses of the peers the links point to and the corresponding interests of these peers. The idea is to let peers self-organise into clusters with similar content. Peer organisation is achieved through a rewiring protocol that is (periodically) executed independently by each peer. The purpose of this protocol is to establish connections among peers with similar interests. Eventually, by creating new connections to peers and by discarding connections that are outdated, dynamic clusters of peers emerge. Overall, rewiring is a highly dynamic procedure that involves simultaneous operations by many peers inducting to peer clusters. Queries can then be resolved by routing the query towards clusters based on their likelihood to match the query. Once reaching a cluster, the peer receiving the query is responsible for forwarding it to other peers within the same cluster.

The basic protocols that determine the way peers join the overlay network, connect to and disconnect from the network, and the way queries are processed are thoroughly presented in [3]. Below, we present the protocol that specifies the way peers dynamically self-organise into clusters with similar content. Rewiring Protocol. Peer organisation proceeds by establishing new connections to similar peers and by discarding old ones. Each peer pi periodically (e.g., when joining the network or when its interests have changed) initiates a rewiring procedure (independently for each interest) by computing the intracluster similarity (or neighbourhood similarity)

$$NS_i = \frac{1}{s} \cdot \sum_{\forall p_j \in RI_i} sim(I_i, I_j),$$
(4)

where s is the number of short-range links of p_i according to interest I_i , p_j is a peer contained in RI_i that is on-line, I_j is the interest of p_j , and *sim()* can be any appropriate similarity function (e.g., the cosine similarity between the term vector representations [12]). The neighbourhood similarity NS_i is used here as a measure of *cluster cohesion*. If NS_i is greater than a threshold Θ , then p_i does not need to take any further action, since it is surrounded by peers with similar interests. Otherwise, p_i issues a FINDPEERS(ip(p_i), I_i , L, *t*R) message, where L is a list and *t*R is the time-tolive (TTL) of the message. List L is initially empty and will be used to store tuples of the form < ip(p_i); I_i >, containing the IP address and interest of peers discovered while the message traverses the network.

System parameters θ and tR need to be known upon bootstrapping. A peer pi receiving the FindPeers() message appends its IP address $ip(p_i)$ and its interest I_i to L (or the interest most similar to Ii if pi has multiple interests), reduces *t*R by one, and forwards the message to the m neighbouring peers (m \leq s) with interests most similar to I_i. This message forwarding technique is referred to in the literature as gradient walk (GW) [15]. When t = 0,

Procedure Rewiring $(p_t, I_t, \tau_R, \theta, m)$ Initiated by p_t when neighborhood similarity NS_t drops below θ .

input: peer p_i with interest I_i and routing index RI_i output: updated routing index RI₄

1:	compute $NS_i = \frac{1}{g} \cdot \sum_{\forall p_i \in RI_i} sim(I_i, I_j)$
2:	if $NS_i < \theta$ then
3:	$L \leftarrow \{ \}$
4:	create FINDPEERS()
5:	$p_k \leftarrow p_i$
6:	repeat
7:	$L \leftarrow L :: \langle ip(p_k), I_k \rangle$
8:	send FINDPEERS() to
	m neighbours of p_k with interests most similar to I_i
9:	$p_k \leftarrow \text{every peer receiving FINDPEERS}()$
10:	$\tau_R \leftarrow \tau_R - 1$
11:	until $\tau_R = 0$
12:	return list L to p_i
13:	update RI_i with information from L

Fig. 3. The rewiring protocol

the FINDPEERS() message is sent back to the message initiator p_i. Figure 3 illustrates the above rewiring procedure in algorithmic steps. When the message initiator p_i receives the FINDPEERS() message back, it utilises the information contained in L to update its routing index RI_i by re- placing old short-range links corresponding to peers with less similar interests with new links corresponding to peers with more similar interests. Peers also store longrange links in their routing indexes which stand as short paths to dissimilar clusters. For the update of the long-range links, peer pi uses a random walk⁷ in the network with TTL /R.

5 EVALUATION:

In this section, we evaluate the rewiring protocol of iCluster under churn in a realistic setting using realistic data and queries.

5.1 Performance Measures

As it is typical in the evaluation of P2P information retrieval systems, performance is measured in terms of network traffic and retrieval effectiveness. The network traffic is measured by the number of rewiring (respectively search) messages sent over the network during rewiring (respectively querying). The retrieval effectiveness is evaluated using recall (i.e., the number of relevant documents retrieved over the total number of relevant documents in the network). Notice that in our setting precision is always 100% since only relevant documents are



Fig. 4. Lifetime distributions

retrieved. To describe the effect of peer clustering, we use clustering efficiency [8], a measure that gives information about the underlying network structure.

Clustering efficiency takes values in the interval [0, 1]. The higher its value is the better the organisation of the network is considered.

5.2 Experimental Test

Dataset: The dataset contains over 556,000 web documents from the TREC-68 collection belonging in 100 categories, and has been previously used to evaluate information retrieval algorithms over distributed document collections. The queries employed in the evaluation of the corpus are strong representatives of document categories (i.e., the topics of the categories).

Setup: We consider N loosely-connected peers, each of which contributes documents in the network from a single category. The base unit for time used in the experiments is the period t. The start of the rewiring procedure for each peer is randomly chosen from the interval [0, 4000] and its periodicity is randomly selected from a normal distribution of 2000, in the spirit of [15]. Therefore, each peer starts (and goes over again) independently the rewiring process. We start recording the network activity at time 4000.t, when all peers have initiated the rewiring procedure at least once.

We experimented with different values of similarity threshold Θ , message forwarding TTL *t*R and query forwarding TTLs *t*b,*t*f. We consider that a given parameter value is better than another if it results in better clustering and retrieval for less communication load.

The simulator used to evaluate the rewiring protocol was implemented in C/C++ and all experiments were run on a Linux machine. Our results were averaged over 25 runs (5 random initial network topologies and 5 runs for each topology).

We experimented with different lifetime distributions. Figure 4(a) illustrates two different on-line session lengths distributions. By definition, the most skewed the distribution is, the smaller the lifetimes of the most peers are. The first case in Fig. 4(a) corresponds to a difficult scenario compared to the second case, since peers are on-line for shorter time periods and leave the network more often. Figure 4(b) presents the percentage of off-line peers as a function of time. In the first scenario the percentage of the peers that are logged-off reaches up to 42% (for t = 11:5K), while in the second scenario the percentage of the peers that are logged-off every moment is kept under 25%.

6 CONCLUSIONS:

In this work, we studied the performance of P2P networks under churn. Building upon established peer churn models describing user behaviour in a realistic setting (where peers connect and disconnect from the network), we evaluated the effects of churn on network organisation and the sufficiency of rewiring to retain a clustered organisation of peers. For this we relied on iCluster [3], an approach for organising peers into highly dynamic networks and supporting efficient information retrieval. This is the first work to address the issues involved in the design and the evaluation of protocol when introducing peer churn. Incorporating replication and caching (to ensure consistency between redundant resources, to improve reliability, fault-tolerance, or accessibility) in selforganising dynamic P2P networks, and studying more elaborate methods for load balancing (in terms of processing and communication activity) are important issues for future research.

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