University of Massachusetts at Amherst March 11, 1996

Speculative Data Dissemination and Service

to Reduce Server Load, Network Traffic and Service Time in Distributed Information Systems

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Monday, March 11^{th} , 1996

IEEE ICDE'96, New Orleans, Louisiana February 28^{th} , 1996

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Wednesday, February 28^{th} , 1996

[†] This work is supported partially by a grant from the NSF.

Locality of Reference in a Client-Server Environment

Locality of Reference Flavors

◊ Temporal:

A document accessed frequently in the past is likely to be accessed again in the future.

◊ Spatial:

A document "neighboring" a recently accessed document is likely to be accessed in the future.

♦ Geographical:

A document accessed by a client is likely to be accessed in the future by "neighboring" clients.

How to Capitalize on it?

- On the client side, use "caching" and "prefetching" (e.g. Distributed file systems, Sun NFS, AFS, [Standberg 1985, Morriss 1986, Howard 1988], Proxy caching [Danzig 1993, Acharya 1993, Papadimitriou 1994], Cooperative client caching [Blaze 1993, Dahlin 1994]).
- On the server side, use "information dissemination" [Bestavros 1994], "geographical caching" [Braunh and Claffyh 1994], "speculative service" [Bestavros 1995], "geographical push caching" [Gwertzman and Seltzer 1995].

Information Caching versus Information Dissemination

Motive

♦ The scalability of Internet services hinges on efficient distribution and partitioning of system resources to reduce the amount of data that must be moved.

Information Caching

- $\diamond\,$ Initiated by a client or a group of clients.
- ◇ Geared towards reducing service time.
- ◇ Relies on temporal locality of client reference patterns.
- ♦ Ensuring consistency is expensive.

Information Dissemination

- \diamond Initiated by servers.
- ♦ Geared towards balancing load and reducing traffic.
- ♦ Relies on temporal/geographical popularity of documents.
- ♦ Ensuring consistency is cheap.
- $\diamond\,$ Requires collaboration of "server proxies".

Client-initiated Caching Study

Experiment Description

- \diamond We instrumented Mosaic to log all user accesses on our site [BCC:95].
- \diamond We studied cache performance at various levels:
 - Session Caching: One cache per session
 - Host Caching: One cache per host
 - LAN Caching: One cache per LAN
- ♦ We used the logs obtained from Mosaic to perform trace simulations for various protocols [BCCCHM:95].

Sessions	4,700
Users	591
URLs Requested	575,775
Files Transferred	130,140
Unique Files Requested	46,830
Bytes Requested	$2713 \mathrm{MB}$
Bytes Transferred	1849 MB
Unique Bytes Requested	1088 MB

Summary Statistics for Trace Data Used in This Study

Client-initiated Caching Effectiveness

Experiments Results

- $\diamond\,$ Poor Byte Hit Rate $<\,40\%$ with infinite cache.
- ♦ Sharing amongst multiple clients is limited too!



Cache Expansion Index for Remote documents

The Server's Perspective

Server Log Analysis

- ◊ We collected the logs of our departmental HTTP server and those of the Rolling Stones Multimedia server.
- ♦ We used the logs to analyze the popularity of various documents and to drive trace simulations of various server-initiated protocols.

	cs-www.bu.edu	www.stones.edu
Period	56 days	110 days
URL requests	$172,\!635$	4,068,432
Bytes transferred	$1,447 { m MB}$	112,015 MB
Average daily transfer	26 MB	1,018 MB
Files on system	2,018	N/A
Files accessed (remotely)	974~(656)	N/A (1,151)
Size of (accessed) file system	50 MB (37 MB)	N/A (402 MB)
Unique clients (10+ requests)	8,123	60,461

Summary Statistics for Log Data Used in This Study

The Server's Perspective

Log Analysis of http://cs-www.bu.edu

- \diamond Popular documents are *very* popular!
- \diamond Only 10% of all blocks accounted for 91% of all requests!



Popularity of data blocks and bandwidth reduction from their dissemination http://cs-www.bu.edu

The Server's Perspective

Log Analysis of http://www.stones.com

- ♦ Same conclusions as before.
- ♦ Making 25MB of data available to clients at a proxy one-hop closer to them would save more than 900MB/day of network bandwidth.



http://www.stones.com

Information Dissemination Protocol

Underlying Model

- \diamond A set of *service proxies* act as information "outlets" on the Internet.
- ♦ These service proxies offer space/bandwidth "for-rent" to other servers or proxies that constitute its Cluster.
- ♦ A server may belong to several clusters, thus allowing some of its files to be dessiminated to multiple service proxies.
- \diamond Service proxies are themselves servers who may be members of other clusters.



Underlying Model for Information Dissemination

Information Dissemination Protocol

Questions to be answered

- Given the access pattern at a server, which clusters should the server choose to join?
- ◊ Given the access pattern at a server, which files should the server disseminate? and where?
- Given the popularity profile of all servers in a cluster, how should the
 resources (space/bandwidth) at the service proxy be allocated?

Assumptions

- ♦ The dissemination protocol should not require any "special" features/capabilities from other protocols.
- ◇ File popularity is a "universal" phenomenon (i.e. the probability of accessing a file is independent of who is accessing it). This is a *conservative* simplifying assumption.
- ◇ File popularity does not change drastically in a short period of time. This assumption has been verified.

Optimal Allocation of Storage at the Proxy

Notation

- ♦ $C = S_0, S_1, S_2, ..., S_n$ is the set of servers in a cluster. S_0 is the service proxy of C.
- $\diamond R_i$ is the total number of bytes per unit time serviced by server S_i to clients outside C.
- $\diamond H_i(b)$ is the probability that a request to S_i will be to the most popular b bytes disseminated to S_0 .
- ♦ B_i is the number of bytes that S_0 duplicates from S_i . $B_0 = B_1 + B_2 + \dots + B_n$ is the total storage space available at S_0 .

Goal

 \diamond Choose B_i to maximize the percentage of traffic serviced at \mathcal{S}_0 .

$$\alpha_{\mathcal{C}} = \frac{\sum_{i=1}^{n} R_i \times H_i(B_i)}{\sum_{i=1}^{n} R_i}$$

Which Proxies Should be Contracted?

Characterizing the Client Tree and Choosing Proxies

- Using the record route option of TCP/IP, it is possible to build a complete tree originating at the server with clients at the leaves. For http://cs-www.bu.edu, this tree consisted of 18,000 nodes.
- \diamond The most popular files are disseminated down the tree and stored at proxies closer to the clients.
- ◇ The location of such proxies depends on the demand from the various parts of the tree.
- ♦ Analysis of http://cs-www.bu.edu logs for a consecutive 26-week period suggests that the shape of the tree (especially internal nodes) and the distribution of load is quite static over time.



Clients

How Much Bandwidth is Saved?

How far could we "push" information towards clients?

- $\diamond\,$ At least 8-9 hops!
- Replicating the most popular 25 MB from http://www.stones.com on *few* proxies yields a whoping saving of > 8 GB of network bandwidth per day.



How far away are clients?

How Much Bandwidth is Saved?



Expected reduction in bandwidth as a result of dissemination

The Notion of Speculative Service

Could the next client request be predicted?

- \diamond In many cases, the answer is *yes*.
- ♦ Servers could "speculatively" service documents *before* they are requested (a.k.a. server-initiated prefetching).

Two kinds of dependencies:

- \diamond Embedding dependencies: Document \mathcal{D}_j is embedded in \mathcal{D}_i .
- \diamond Traversal dependencies: Document \mathcal{D}_j is often requested as a result of an access to \mathcal{D}_i .

Traversal



How far away are clients?

Document Access Interdependency Matrix

Notation

- ♦ Let p[i, j] denote the conditional probability that document \mathcal{D}_j will be requested, within T_w units of time after the request for \mathcal{D}_i .
- ♦ Let P denote the square matrix representing p[i, j], for all possible documents $0 \ge i, j \le N$. Let P^{*} denote the transitive closure of P.
- ♦ Thus, $p^*[i, j]$ is the probability that there will be a sequence of requests (inter-request time ; T_w) starting with \mathcal{D}_i and ending with \mathcal{D}_j .

Server log analysis

 \diamond Using server logs, the P and P^{*} matrices could be easily constructed.



Simulation Model

- ♦ Successive requests separated by less than StrideTimeout units of time belong to the same "stride".
- ♦ Clients maintain a session cache. The session cache is purged if the time between successive requests exceeds SessionTimeout.
- Service Time ratio, and Miss rate ratio.

Parameter	Meaning	Base Value
CommCost	Cost of communicating 1 Byte	1 unit
ServCost	Setup cost for a service request	10,000 unit
StrideTimeout	Value of time window T_w	$5.0 \mathrm{secs}$
SessionTimeout	Cache invalidation timeout	∞ secs
MaxSize	Maximum size to prefetch	∞ (no limit)
Policy	Speculative service algorithm	$p^*[i,j] \ge T_p$
HistoryLength	Length of the logs used for P	60 days
UpdateCycle	Frequency of recomputing P	1 day

Baseline Results

- ♦ Significant improvement in performance (above what is achievable by client caching) could be achieved for a miniscule increase in traffic.
- ♦ 5% extra bandwidth results in a whopping 30% reduction in server load, a 23% reduction in service time, and a 18% reduction in client miss-rate.
- $\diamond\,$ Beyond some point, speculation does not seem to pay off.



Stability of the P and P^* Relations

- \diamond We varied the **UpdateCycle** from 1 to 7 days, while keeping the **HistoryLength** at 60 days. This change resulted in a 3% degradation in all measured metrics, suggesting that P and P^* do change (albeit very slowly) with time.
- ◇ Also, we varied the HistoryLength from 60 to 30 days, while keeping the UpdateCycle at 1 day. This change resulted in a 5% improvement in all measured metrics, suggesting that an aging mechanism must be used to phase-out dependencies exhibited in on older server traces, in favor of dependencies exhibited in more recent ones.



Effect of Client Caching

- ♦ We compared simulations with SessionTimeout equal to 3,600 seconds (large cache) and to 120 seconds (small cache).
- ♦ The presence of client caching (even if modest) is likely to further improve the performance of speculative service.
- ◇ In order to reap all the benefit from speculative service, client must cache "prefetched" documents long enough.



Cooperative Clients

- ◇ Performance could be further improved if documents "already cached at the client" are not speculatively serviced!
- ♦ Our simulations showed that speculative service with cooperative clients results in better bandwidth utilization, especially when the client performs "some" caching.



Effect of Document Size

 The benefits of speculation are most pronounced when documents serviced speculatively are small. We studied this by varying the MaxSize parameter.



Variations on Speculative Service

Server-assisted Prefetching

- ◇ Servers could pass the list of "probable future documents" to the client (instead of passing along the document themselves).
- ♦ Prefetching could be done at the discretion of clients.

Client-initiated Prefetching

- ◊ Using user traces, it is possible for the client software to perform "prefetching" [Bestavros and Cunha: 1995].
- ◇ Client-initiated prefetching is *very* effective for "frequently traversed documents" but ineffective for "newly/rarely traversed documents".
- ♦ Client-initiated prefetching and server-initiated speculative service are "complementary".

Conclusion

In a Client-Server model, servers are in a much better position to discover and utilize information about locality of reference, whether *temporal*, *spatial*, or *geographical*.

- ◇ Temporal locality of reference could be exploited to disseminate information closer to clients, to complement client-initiated caching.
- ♦ Spatial locality of reference could be exploited to initiate service speculatively, to complement client-initiated prefetching.
- ♦ Geographical locality of reference could be exploited to optimize the placement of replicas, to match the paterns of demand from clients.

For current and future projects, visit our Research Group Home Page at

http://cs-www.bu.edu/groups/oceans