Peer-to-Peer Systems and Distributed Hash Tables



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CS 240: Computing Systems and Concurrency Lecture 8

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Credits: Michael Freedman and Kyle Jamieson developed much of the original material. Selected content adapted from B. Karp, R. Morris.

Today

- 1. Peer-to-Peer Systems
 - Napster, Gnutella, BitTorrent, challenges
- 2. Distributed Hash Tables
- 3. The Chord Lookup Service
- 4. Concluding thoughts on DHTs, P2P

What is a Peer-to-Peer (P2P) system?



- A **distributed** system architecture:
 - No centralized control
 - Nodes are roughly symmetric in function
- Large number of unreliable nodes

Advantages of P2P systems

- High capacity for services through resource pooling:
 - Many disks
 - Many network connections
 - Many CPUs
- No centralized server or servers may mean:
 - Less chance of service overload as load increases
 - A single failure won't wreck the whole system
 - System as a whole is harder to attack

P2P adoption

- Successful adoption in some niche areas -
- Client-to-client (legal, illegal) file sharing

 Popular data but owning organization has no money
- 2. Digital currency: no natural single owner (Bitcoin)
- 3. Voice/video telephony
 - Skype used to do this...

Example: Classic BitTorrent

- 1. User clicks on download link
 - Gets torrent file with content hash, IP addr of tracker
- 2. User's BitTorrent (BT) client talks to tracker
 - Tracker tells it list of peers who have file
- 3. User's BT client downloads file from one or more peers
- 4. User's BT client tells tracker it has a copy now, too
- 5. User's BT client serves the file to others for a while

Provides huge download bandwidth, without expensive server or network links

The lookup problem



Centralized lookup (Napster)



Flooded queries (original Gnutella)



Routed DHT queries (Chord)



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What is a DHT (and why)?

- Local hash table:
 key = Hash(name)
 put(key, value)
 get(key) → value
- Service: Constant-time insertion and lookup

Distributed Hash Table (DHT): Do (roughly) this across millions of hosts on the Internet!

What is a DHT (and why)?

- Distributed Hash Table: key = hash(data) lookup(key) → IP addr (Chord lookup service) send-RPC(IP address, put, key, data) send-RPC(IP address, get, key) → data
- Partitioning data in large-scale distributed systems
 - Tuples in a global database engine
 - Data blocks in a global file system
 - Files in a P2P file-sharing system

Cooperative storage with a DHT



- App may be **distributed** over many nodes
- DHT distributes data storage over many nodes

BitTorrent over DHT

- BitTorrent can use DHT instead of (or with) a tracker
- BT clients use DHT:
 - Key = **?**
 - Value = ?

BitTorrent over DHT

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 - Key = file content hash ("infohash")
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BitTorrent over DHT

- BitTorrent can use DHT instead of (or with) a tracker
- BT clients use DHT:
 - Key = file content hash ("infohash")
 - Value = IP address of peer willing to serve file
 - Can store multiple values (*i.e.* IP addresses) for a key
- Client does:
 - get(infohash) to find other clients willing to serve
 - put(infohash, my-ipaddr) to identify itself as willing

Why DHT for BitTorrent?

- The DHT comprises a single giant tracker, less fragmented than many trackers
 - So peers more likely to find each other

- Classic BitTorrent tracker is a single point of failure
 - Legal attacks
 - Denial-of-Service (DoS) attacks

What is hard in DHT design?

- **Decentralized:** no central authority
- Scalable: low network traffic overhead
- Efficient: find items quickly (latency)
- Dynamic: nodes fail, new nodes join

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Chord lookup algorithm properties

Interface: lookup(key) \rightarrow IP address

- Efficient: O(log *N*) messages per lookup *N* is the total number of nodes
- Scalable: O(log *N*) state per node
- Robust: survives massive failures
- Simple to analyze

Chord identifiers

- **Key identifier** = SHA-1(key)
- **Node identifier** = SHA-1(IP address)
- SHA-1 distributes both uniformly

How does Chord partition data?
 – *i.e.*, map key IDs to node IDs

Consistent hashing



Key is stored at its **successor:** node with next-higher ID

Chord: Successor pointers



Basic lookup



Simple lookup algorithm

Correctness depends only on successors

Improving performance

- **Problem:** Forwarding through successor is slow
- Data structure is a linked list: O(n)
- Idea: Can we make it more like a binary search?
 Need to be able to halve distance at each step

"Finger table" allows log N-time lookups



Finger *i* Points to Successor of *n*+2^{*i*}



Implication of finger tables

- A binary lookup tree rooted at every node
 Threaded through other nodes' finger tables
- This is **better** than simply arranging the nodes in a single tree
 - -Every node acts as a root
 - So there's no root hotspot
 - No single point of failure
 - But a lot more state in total

Lookup with finger table

Lookup(key-id)

look in local finger table for

- highest n: my-id < n < key-id
- if n exists

call Lookup(key-id) on node n *//nexthop* else

return my successor //done

Lookups Take O(log N) Hops



An aside: Is log(n) fast or slow?

- For a million nodes, it's 20 hops
- If each hop takes 50 milliseconds, lookups take a second
- If each hop has 10% chance of failure, it's a couple of timeouts
- So in practice log(n) is better than O(n) but **not great**

Failures may cause incorrect lookup



Successor lists

- Each node stores a **list** of its *r* immediate successors
 - After failure, will know first live successor
 - Correct successors guarantee correct lookups
 - Guarantee is with some probability
 - -r is often O(log N)
 - E.g., 20 for 1 million nodes

Lookup with fault tolerance

Lookup(key-id) look in local finger table and successor-list for highest n: my-id < n < key-idif n exists call Lookup(key-id) on node n //nexthop if call failed, remove n from finger table and/or successor list return Lookup(key-id) else

return my successor //done

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The DHash DHT

- Builds key/value storage on Chord
- **Replicates** blocks for availability
 - Stores k replicas at the k successors after the block on the Chord ring

DHash replicates blocks at *r* successors



- Replicas are easy to find if successor fails
- Hashed node IDs ensure independent failure

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DHTs: Impact

- Original DHTs (CAN, Chord, Kademlia, Pastry, Tapestry) proposed in 2001-02
- Following 5-6 years saw proliferation of DHT-based applications:
 - Filesystems (e.g., CFS, Ivy, OceanStore, Pond, PAST)
 - Naming systems (e.g., SFR, Beehive)
 - Content distribution systems (e.g., Coral)
 - Distributed databases (e.g., PIER)
- Chord is one of the most cited papers in CS!

Why don't all services use P2P?

- 1. High latency and limited bandwidth between peers (*vs* servers in a datacenter)
- 2. User computers are **less reliable** than managed servers
- **3. Lack of trust** in peers' correct behavior — Securing DHT routing hard, unsolved in practice

DHTs in retrospective

- Seem promising for finding data in large P2P systems
- Decentralization seems good for load, fault tolerance
- But: the security problems are difficult
- But: churn is a problem, particularly if log(n) is big
- So DHTs have not had the impact that many hoped for

Take-away ideas: What DHTs got right

Consistent hashing

- Elegant way to divide a workload across machines
- Very useful in clusters: actively used today in Amazon Dynamo, Apache Cassandra and other systems
- Replication for high availability, efficient recovery after node failure
- Incremental scalability: "add nodes, capacity increases"
- Self-management: minimal configuration
- Unique trait: no single server to shut down/monitor

Next topic: Scaling out Key-Value Storage: Amazon Dynamo