Example topic 1

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□ Peer-to-peer

P2P file sharing

Notes based on notes by K.W. Ross, J. Kurose, D. Rubenstein, and others

P2P: centralized directory

Original "Napster" design

- 1) When peer connects, it informs central server:
 - IP address
 - content
- 2) Alice queries for "Hey Jude"
- 3) Alice requests file from Bob





 File list and IP address is uploaded



napster.com centralized directory







2. User requests search at server.



napster.com centralized directory









napster.com



4. User chooses server

napster.com centralized directory





Napster's centralized server farm had difficult time keeping up with traffic



P2P: problems with centralized directory

- single point of failure
- performance bottleneck
- copyright infringement: "target" of lawsuit is obvious

file transfer is decentralized, but locating content is highly centralized

Unstructured P2P: Gnutella

Focus: decentralized method searching for files

- central directory server no longer the bottleneck
- o more difficult to "pull plug"



Each application instance serves to:

- store selected files
- o route queries from and to its neighboring peers
- respond to queries if file stored locally
- ⁹ serve files

Gnutella: protocol



Distributed Search/Flooding



Distributed Search/Flooding



Hierarchical Overlay

- Between centralized index, query flooding approaches
- Each peer is either a group leader or assigned to a group leader
 - TCP connection between peer and its group leader
 - TCP connections between some pairs of group leaders
- Group leader tracks content in its children



Example: KaZaA Architecture (2)

supernodes

- Nodes that have more connection bandwidth and are more available are designated as "supernodes"
- Each supernode acts as a mini-Napster hub, tracking the content and IP addresses of its descendants

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Parallel Downloading; Recovery

If file is found in multiple nodes, user can select parallel downloading

Most likely HTTP byte-range header used to request different portions of the file from different nodes

Automatic recovery when server peer stops sending file

KaZaA Corporate Structure

- Software developed by FastTrack in Amsterdam
- FastTrack also deploys KaZaA service
- FastTrack licenses software to Music City (Morpheus) and Grokster
- Later, FastTrack terminates license, leaves only KaZaA with killer service

- Summer 2001, Sharman networks, founded in Vanuatu (small island in Pacific), acquires FastTrack
 - Board of directors, investors: secret
- Employees spread around, hard to locate
- Code in Estonia

Lessons learned from KaZaA

KaZaA provides powerful file search and transfer service <u>without</u> server infrastructure

- Exploit heterogeneity
- Provide automatic recovery for interrupted downloads
- Powerful, intuitive user interface

<u>Copyright infringement</u>

- International cat-andmouse game
- With distributed, serverless architecture, can the plug be pulled?
- Prosecute users?
- Launch DoS attack on supernodes?
- Pollute?

P2P Case study: Skype

- Inherently P2P: pairs of users communicate.
- Proprietary application-layer protocol (inferred via reverse engineering)
- Hierarchical overlay with Supernodes (SNs)
- Index maps usernames to IP addresses; distributed over SNs



Peers as relays

- Problem when both Alice and Bob are behind "NATs".
 - NAT prevents an outside peer from initiating a call to insider peer



Peers as relays

- Problem when both Alice and Bob are behind "NATs".
 - NAT prevents an outside peer from initiating a call to insider peer
- Solution:
 - Using Alice's and Bob's SNs, Relay is chosen
 - Each peer initiates session with relay.
 - Peers can now communicate through NATs via relay



Structured p2p systems

<u>Distributed Hash Table (DHT)</u>

DHT = distributed P2P database

Database has (key, value) pairs;
key: ss number; value: human name
key: content type; value: IP address

Key	Value
00	
01	
10	
11	

Peers query DB with key
DB returns values that match the key

Peers can also insert (key, value) pairs

DHT Identifiers

Key	Value
000000	
000001	
000002	
ffffff	

Assign integer identifier to each peer in range [0,2ⁿ-1]
Each identifier can be represented by n bits.

- Require each key to be an integer in same range.
- To get integer keys, hash original key.
 - E.g., key = h("Led Zeppelin IV")
 - This is why they call it a distributed "hash" table

How to assign keys to peers?

Central issue:

• Assigning (key, value) pairs to peers.

- Rule: Assign key to the peer that has the closest ID.
- Convention in lecture: closest is the closest successor of the key.
- **Ex:** n=4; peers: 1,3,4,5,8,10,12,14;
 - \bigcirc key = 13, then successor peer = 14

○ key = 15, then successor peer = 1



Each peer only aware of immediate successor and predecessor.

"Overlay network"

Circle DHT (2)



<u>Circular DHT with Shortcuts</u>



Each peer keeps track of IP addresses of predecessor, successor, short cuts.
E.g., Example above reduced from 6 to 2 messages.
Possible to design shortcuts so O(log N) neighbors, O(log N) messages in query

Example: Chord Routing [see paper]

- A node s's ith neighbor has the ID that is equal to s+2ⁱ or is the next largest ID (mod ID space), i≥0
- To reach the node handling ID t, send the message to neighbor #log₂(t-s)
- Requirement: each node s must know about the next node that exists clockwise on the Chord (0th neighbor)
- Set of known neighbors called a finger table





each data item (e.g., file or metadata pointing to file copies) has a key



DHT Layered Architecture



BitTorrent-like systems

- File split into many smaller pieces
- Pieces are downloaded from both seeds and downloaders
- Distribution paths are dynamically determined
 - Based on data availability



File distribution: BitTorrent

□ P2P file distribution



Download using BitTorrent

Background: Incentive mechanism

Establish connections to large set of peers

- At each time, only upload to a small (changing) set of peers
- Rate-based tit-for-tat policy
 - Downloaders give upload preference to the downloaders that provide the highest download rates



Download using BitTorrent Background: Piece selection



- Rarest first piece selection policy
 - Achieves high piece diversity
- Request pieces that
 - the uploader has;
 - the downloader is interested (wants); and
 - is the rarest among this set of pieces











□ MapReduce

<u>Motivation</u>

Process lots of data

• Google processed about 24 petabytes of data per day in 2009.

□ A single machine cannot serve all the data

• You need a distributed system to store and process in parallel

MapReduce

□ MapReduce [OSDI'04] provides

- Automatic parallelization, distribution
- I/O scheduling
 - Load balancing
 - Network and data transfer optimization
- Fault tolerance
 - Handling of machine failures

□ Need more power: Scale out, not up!

 Large number of commodity servers as opposed to some high-end specialized servers

> **Apache Hadoop:** Open source implementation of MapReduce

MapReduce workflow



MapReduce



Failure in MapReduce

- □ Failures are norm in commodity hardware
- Worker failure
 - Detect failure via periodic heartbeats
 - Re-execute in-progress map/reduce tasks
- 🗖 Master failure
 - Single point of failure; Resume from Execution Log
- Data stored on multiple nodes

🗆 Robust

 Google's experience: lost 1600 of 1800 machines once!, but finished fine

Example: Word Count

Input Files

Apple Orange Mango Orange Grapes Plum

Apple Plum Mango Apple Apple Plum

http://kickstarthadoop.blogspot.ca/2011/04/word-count-hadoop-map-reduce-example.html

MapReduce: map, shuffle, reduce

The overall MapReduce word count process Shuffling Final result Input Splitting Mapping Reducing Bear. 1 Bear, 2 Deer. 1 Bear, 1 **Deer Bear River** Bear, 1 River, 1 Car. 1 Car, 1 Car, 3 Bear. 2 **Deer Bear River** Car, 1 Car. 3 Car, 1 Car Car River Car Car River Car. 1 Deer, 2 Deer Car Bear River, 1 River, 2 Deer. 1 Deer, 2 Deer, 1 Deer. 1 Deer Car Bear Car, 1 River, 1 River, 2 Bear, 1 River, 1

Map: Each worker applies the map function to local data + writes the output to a temporary storage. Master ensures only one copy of the redundant input data is processed.

Shuffle: Workers redistribute data based on the output keys (produced by the map function) such that all data belonging to one key is located on the same worker node

Reduce: Workers process each group of output data, per key, in parallel.

MapReduce: map, shuffle, reduce



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□ MapReduce

- Programming paradigm for data-intensive computing
- Distributed & parallel execution model
- Simple to program