CS5412: TIER 2 OVERLAYS

Lecture VI

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Recap

 A week ago we discussed RON and Chord: typical examples of P2P network tools popular in the cloud

Then we shifted attention and peeked into the data center itself. It has tiers (tier 1, 2, backend) and a wide range of technologies

Many of those use a DHT "concept" and would be build on a DHT. But we can't use Chord here!

Today's focus

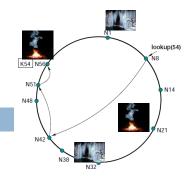
How can we create distributed hash tables optimized for use in cloud computing settings?

If you look deeply into systems like the ones we discussed last time, you'll find DHT technology at the base. So with a DHT you can layer fancier things on top... but the DHT determines the speed!

First problem with Chord: Cost

- Internal to a cloud data center a DHT needs to be blindingly fast
 - Put operation should have cost no higher than 1 RPC directly to the nodes where the data will live
 - Get operation could have a cost of 1 RPC
- In Chord with as few as 1000 participants, costs can include 9 routing hops. So this is unacceptable

Another problem: Hot spots



- As conditions in a network change
 - Some items may become far more popular than others and be referenced often; others rarely: hot/cold spots
 - Members may join that are close to the place a finger pointer should point... but not exactly at the right spot
 - Churn could cause many of the pointers to point to nodes that are no longer in the network, or behind firewalls where they can't be reached
- This has stimulated work on "adaptive" overlays

Today look at three examples

 Beehive: A way of extending Chord so that average delay for finding an item drops to a constant: O(1)

 Pastry: A different way of designing the overlay so that nodes have a choice of where a finger pointer should point, enabling big speedups

 Kelips: A simple way of creating an O(1) overlay that trades extra memory for faster performance

File systems on overlays

 If time permits, we'll also look at ways that overlays can "host" true file systems

- CFS and PAST: Two projects that used Chord and Pastry, respectively, to store blocks
- OceanStore: An archival storage system for libraries and other long-term storage needs

Insight into adaptation

Many "things" in computer networks exhbit Pareto popularity distributions

This one graphs
 frequency by category
 for problems with
 cardboard shipping
 cartons

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Notice that a small subset
 of issues account for most problems

Beehive insight

- Small subset of keys will get the majority of Put and Get operations
 - Intuition is simply that everything is Pareto!
- By replicating data, we can make the search path shorter for a Chord operation
- ... so by replicating in a way proportional to the popularity of an item, we can speed access to popular items!

Beehive: Item replicated on N/2 nodes

If an item isn't on "my side" of the Chord ring it must be on the "other side"

In this example, by replicating a (key,value) tuple over half the ring, Beehive is able to guarantee that it will always be found in at most 1 hop. The system generalizes this idea, matching the level of replication to the popularity of the item.

Beehive strategy

- Replicate an item on N nodes to ensure O(0) lookup
- \square Replicate on N/2 nodes to ensure O(1) lookup

. . .

Replicate on just a single node (the "home" node)
 and worst case lookup will be the original O(log n)

 So use popularity of the item to select replication level

Tracking popularity

- Each key has a home node (the one Chord would pick)
- Put (key,value) to the home node
- □ Get by finding any copy. Increment access counter
 - Periodically, aggregate the counters for a key at the home node, thus learning the access rate over time
 - A leader aggregates all access counters over all keys, then broadcasts the total access rate
 - ... enabling Beehive home nodes to learn <u>relative</u> rankings of items they host
 - ... and to compute the optimal replication factor for any target O(c) cost!

Notice interplay of ideas here

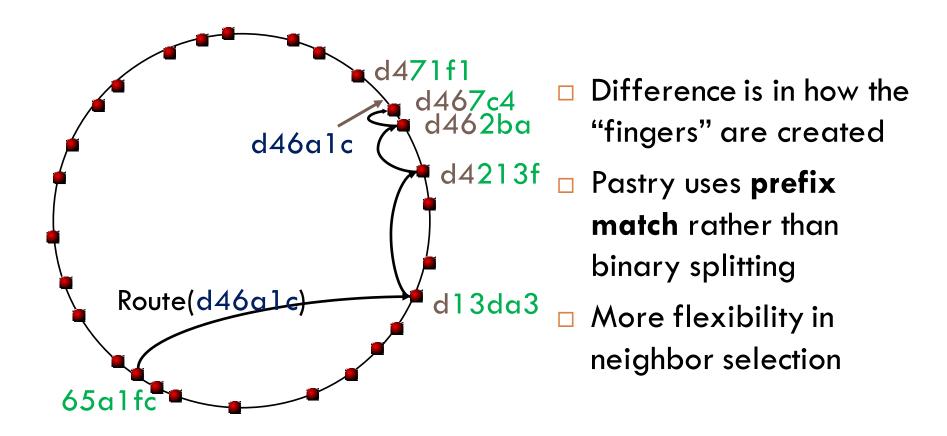
- Beehive wouldn't work if every item was equally popular: we would need to replicate everything very aggressively. Pareto assumption addresses this
- Tradeoffs between parallel aspects (counting, creating replicas) and leader-driven aspects (aggregating counts, computing replication factors)
- We'll see ideas like these in many systems throughout CS5412

Pastry

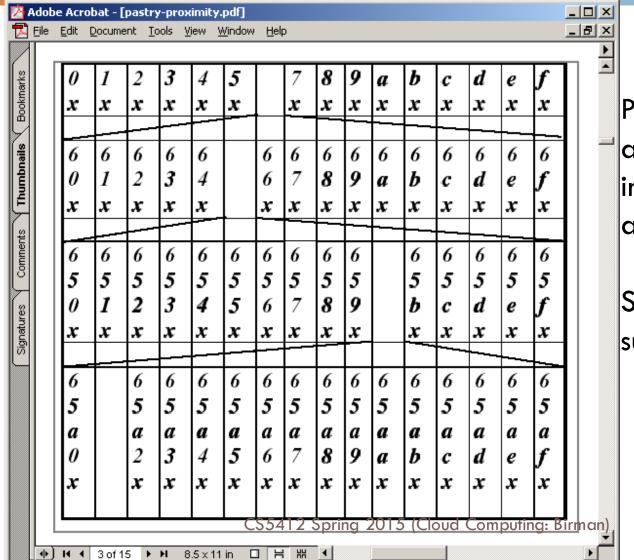
□ A DHT much like Chord or Beehive

- But the goal here is to have more flexibility in picking finger links
 - □ In Chord, the node with hashed key H must look for the nodes with keys H/2, H/4, etc....
 - In Pastry, there are a set of possible target nodes and this allows Pastry flexibility to pick one with good network connectivity, RTT (latency), load, etc

Pastry also uses a circular number space



Pastry routing table (for node 65a1fc)



Pastry nodes also have a "leaf set" of immediate neighbors up and down the ring

Similar to Chord's list of successors

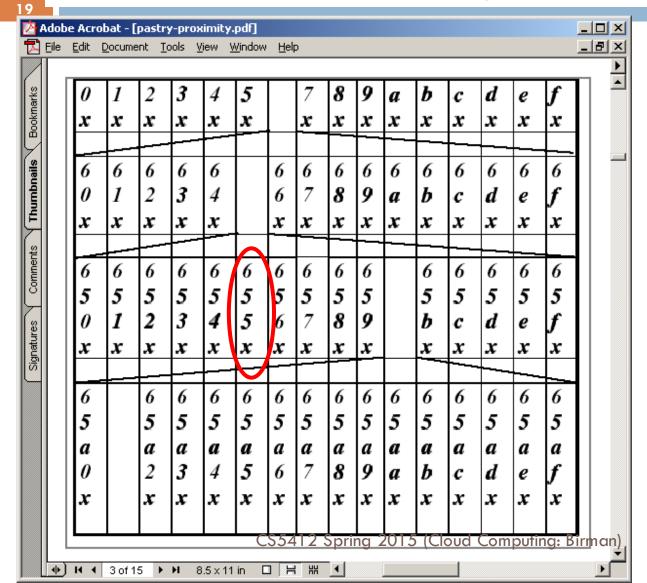
Pastry join

- \square X = new node, A = bootstrap, Z = nearest node
- A finds Z for X
- In process, A, Z, and all nodes in path send state tables to X
- X settles on own table
 - Possibly after contacting other nodes
- X tells everyone who needs to know about itself
- Pastry paper doesn't give enough information to understand how concurrent joins work
 - 18th IFIP/ACM, Nov 2001

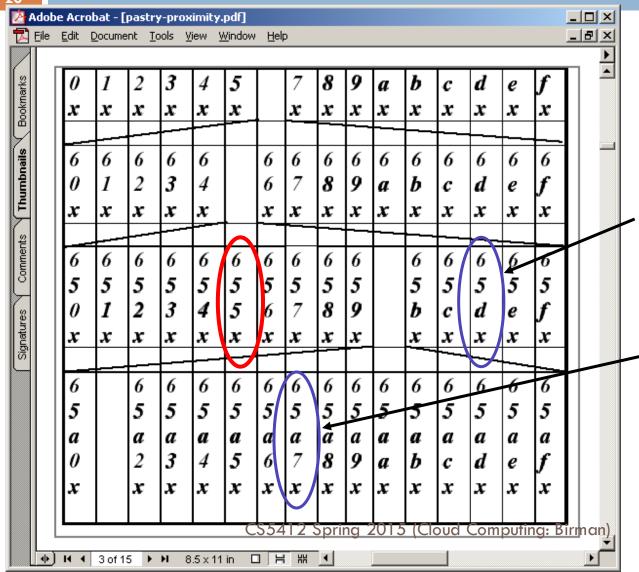
Pastry leave

- Noticed by leaf set neighbors when leaving node doesn't respond
 - Neighbors ask highest and lowest nodes in leaf set for new leaf set
- Noticed by routing neighbors when message forward fails
 - Immediately can route to another neighbor
 - Fix entry by asking another neighbor in the same "row" for its neighbor
 - □ If this fails, ask somebody a level up

For instance, this neighbor fails



Ask other neighbors



Try asking some neighbor in the same row for its 655x entry

If it doesn't have one, try asking some neighbor in the row below, etc.

CAN, Chord, Pastry differences

- CAN, Chord, and Pastry have deep similarities
- □ Some (important???) differences exist
 - CAN nodes tend to know of multiple nodes that allow equal progress
 - Can therefore use additional criteria (RTT) to pick next hop
 - Pastry allows greater choice of neighbor
 - Can thus use additional criteria (RTT) to pick neighbor
 - In contrast, Chord has more determinism
 - How might an attacker try to manipulate system?

Security issues

- □ In many P2P systems, members may be malicious
- If peers untrusted, all content must be signed to detect forged content
 - Requires certificate authority
 - Like we discussed in secure web services talk
 - This is not hard, so can assume at least this level of security

Security issues: Sybil attack

- Attacker pretends to be multiple system
 - If surrounds a node on the circle, can potentially arrange to capture all traffic
 - Or if not this, at least cause a lot of trouble by being many nodes
- Chord requires node ID to be an SHA-1 hash of its IP address
 - But to deal with load balance issues, Chord variant allows nodes to replicate themselves
- A central authority must hand out node IDs and certificates to go with them
 - Not P2P in the Gnutella sense

General security rules

- Check things that can be checked
 - Invariants, such as successor list in Chord
- Minimize invariants, maximize randomness
 - Hard for an attacker to exploit randomness
- Avoid any single dependencies
 - Allow multiple paths through the network
 - Allow content to be placed at multiple nodes
- □ But all this is expensive...

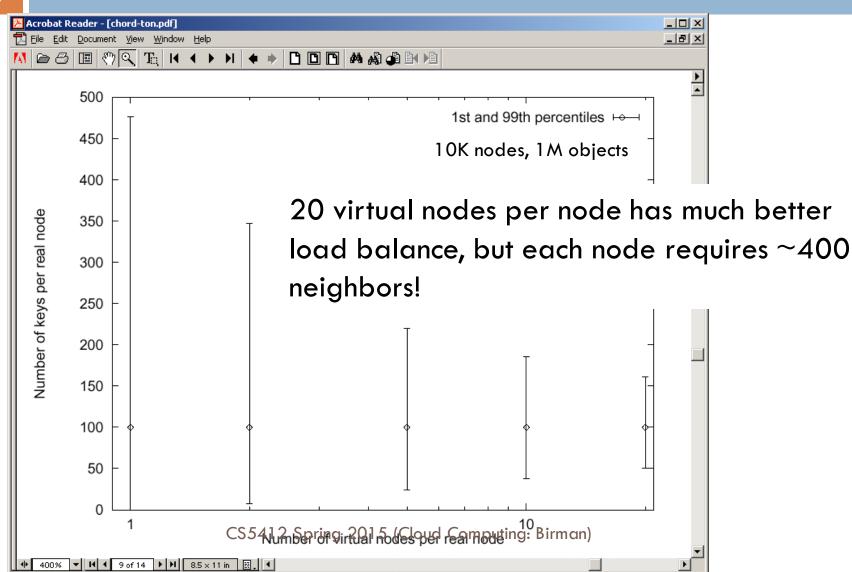
Load balancing

- Query hotspots: given object is popular
 - Cache at neighbors of hotspot, neighbors of neighbors, etc.
 - Classic caching issues
- Routing hotspot: node is on many paths
 - Of the three, Pastry seems most likely to have this problem, because neighbor selection more flexible (and based on proximity)
 - This doesn't seem adequately studied

Load balancing

- Heterogeneity (variance in bandwidth or node capacity
- Poor distribution in entries due to hash function inaccuracies
- One class of solution is to allow each node to be multiple virtual nodes
 - Higher capacity nodes virtualize more often
 - But security makes this harder to do

Chord node virtualization



Fireflies

- □ Van Renesse uses this same trick (virtual nodes)
- In his version a form of attack-tolerant agreement is used so that the virtual nodes can repell many kinds of disruptive attacks
- We won't have time to look at the details today

Another major concern: churn

- Churn: nodes joining and leaving frequently
- Join or leave requires a change in some number of links
- Those changes depend on correct routing tables in other nodes
 - Cost of a change is higher if routing tables not correct
 - In chord, ~6% of lookups fail if three failures per stabilization
- But as more changes occur, probability of incorrect routing tables increases

Control traffic load generated by churn

- Chord and Pastry appear to deal with churn differently
- Chord join involves some immediate work, but repair is done periodically
 - Extra load only due to join messages
- Pastry join and leave involves immediate repair of all effected nodes' tables
 - Routing tables repaired more quickly, but cost of each join/leave goes up with frequency of joins/leaves
 - Scales quadratically with number of changes???
 - Can result in network meltdown???

Kelips takes a different approach

- \square Network partitioned into \sqrt{N} "affinity groups"
- Hash of node ID determines which affinity group a node is in
- Each node knows:
 - One or more nodes in each group
 - All objects and nodes in own group
- But this knowledge is soft-state, spread through peer-topeer "gossip" (epidemic multicast)!

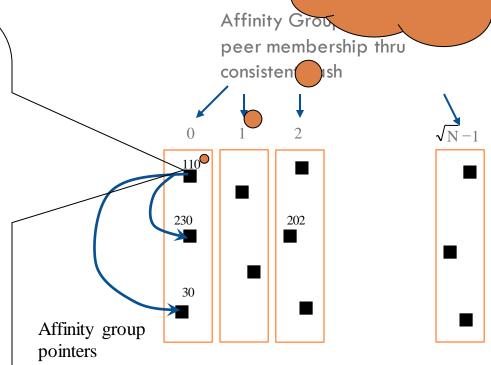
Rationale?

- Kelips has a completely predictable behavior under worst-case conditions
 - It may do "better" but won't do "worse"
 - Bounded message sizes and rates that never exceed what the administrator picks no matter how much churn occurs
 - Main impact of disruption: Kelips may need longer before Get is guaranteed to return value from prior Put with the same key

110 knows about other members – 230, 30...

Affinity group view

id	hbeat	rtt
30	234	90ms
230	322	30ms



 \sqrt{N} members per affinity group

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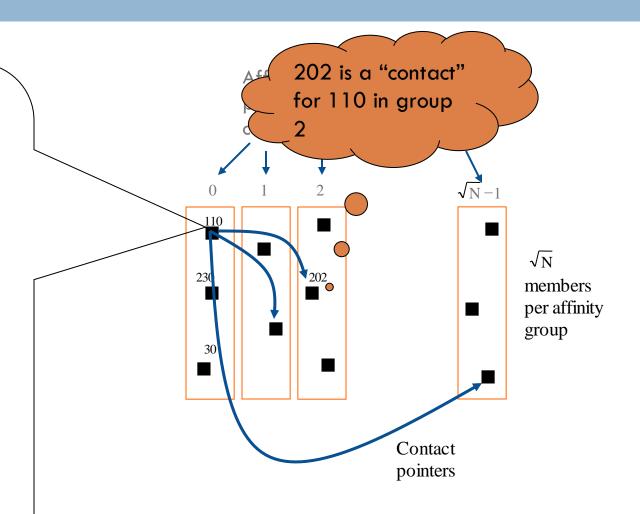
Kelips

Affinity group view

id	hbeat	rtt
30	234	90ms
230	322	30ms

Contacts

group	contactNode
2	202



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Kelips

"cnn.com" maps to group 2. So 110 tells group 2 to "route" inquiries about cnn.com to it.

Affinity group view

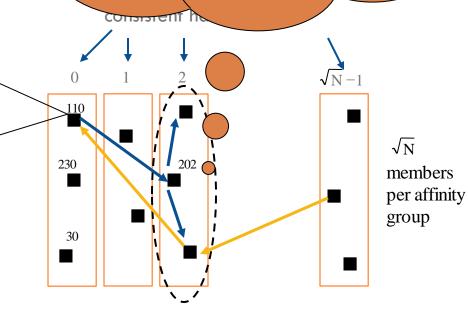
id	hbeat	rtt
30	234	90ms
230	322	30ms

Contacts

group	contactNode
•••	•••
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Resource Tuples

resource	info
•••	•••
cnn.com	110 C



Gossip protocol replicates data cheaply

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How it works

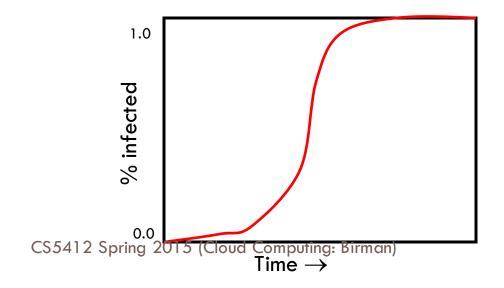
- Kelips is entirely gossip based!
 - Gossip about membership
 - Gossip to replicate and repair data
 - Gossip about "last heard from" time used to discard failed nodes
- Gossip "channel" uses fixed bandwidth
 - ... fixed rate, packets of limited size

Gossip 101

- Suppose that I know something
- I'm sitting next to Fred, and I tell him
 - Now 2 of us "know"
- Later, he tells Mimi and I tell Anne
 - Now 4
- □ This is an example of a *push* epidemic
- Push-pull occurs if we exchange data

Gossip scales very nicely

- Participants' loads independent of size
- Network load linear in system size
- Information spreads in log(system size) time



Gossip in distributed systems

- We can gossip about membership
 - Need a bootstrap mechanism, but then discuss failures, new members
- Gossip to repair faults in replicated data
 - "I have 6 updates from Charlie"
- □ If we aren't in a hurry, gossip to replicate data too

Gossip about membership

- Start with a bootstrap protocol
 - For example, processes go to some web site and it lists a dozen nodes where the system has been stable for a long time
 - Pick one at random
- Then track "processes I've heard from recently" and "processes other people have heard from recently"
- Use push gossip to spread the word

Gossip about membership

- Until messages get full, everyone will known when everyone else last sent a message
 - With delay of log(N) gossip rounds...
- But messages will have bounded size
 - Perhaps 8K bytes
 - Then use some form of "prioritization" to decide what to omit – but never send more, or larger messages
 - Thus: load has a fixed, constant upper bound except on the network itself, which usually has infinite capacity

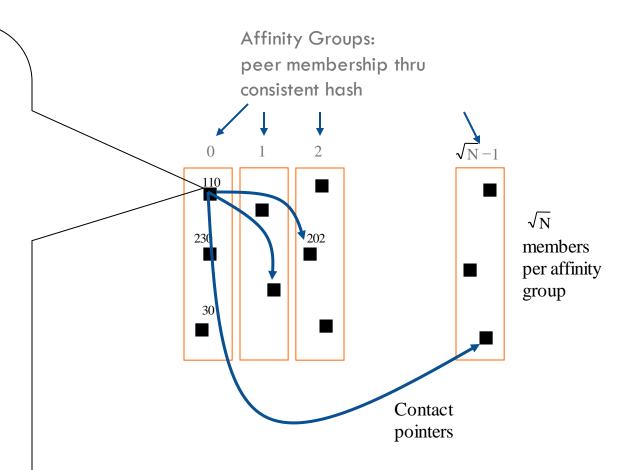
Back to Kelips: Quick reminder

Affinity group view

id	hbeat	rtt
30	234	90ms
230	322	30ms

Contacts

group	contactNode
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How Kelips works Node 175 is a contact for Node 102 in some affinity Hmm...N group much be affinity group 2 Node 102 RTT: 6 ms



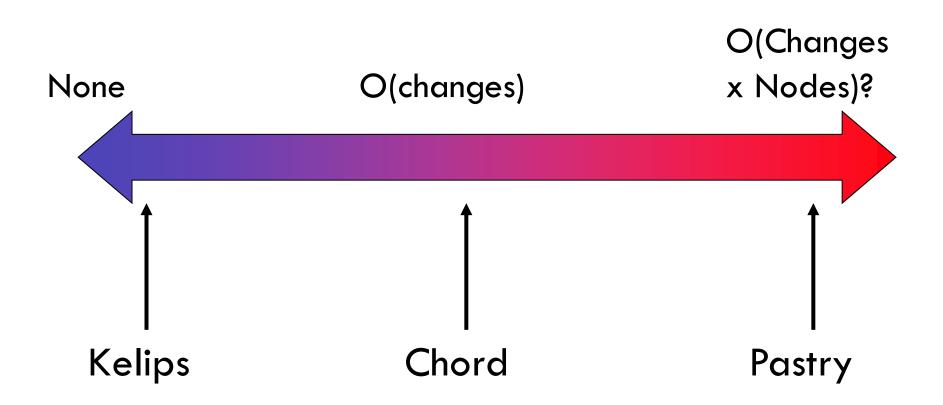
Gossip data stream

- Gossip about everything
- Heuristic to pick contacts: periodically ping contacts to check liveness, RTT... swap so-so ones for better ones.

Replication makes it robust

- Kelips should work even during disruptive episodes
 - $lue{}$ After all, tuples are replicated to \sqrt{N} nodes
 - Query k nodes concurrently to overcome isolated crashes, also reduces risk that very recent data could be missed
- we often overlook importance of showing that systems work while recovering from a disruption

Control traffic load generated by churn



Summary

- Adaptive behaviors can improve overlays
 - Reduce costs for inserting or looking up information
 - Improve robustness to churn or serious disruption

 As we move from CAN to Chord to Beehive or Pastry one could argue that complexity increases

Kelips gets to a similar place and yet is very simple,
 but pays a higher storage cost than Chord/Pastry