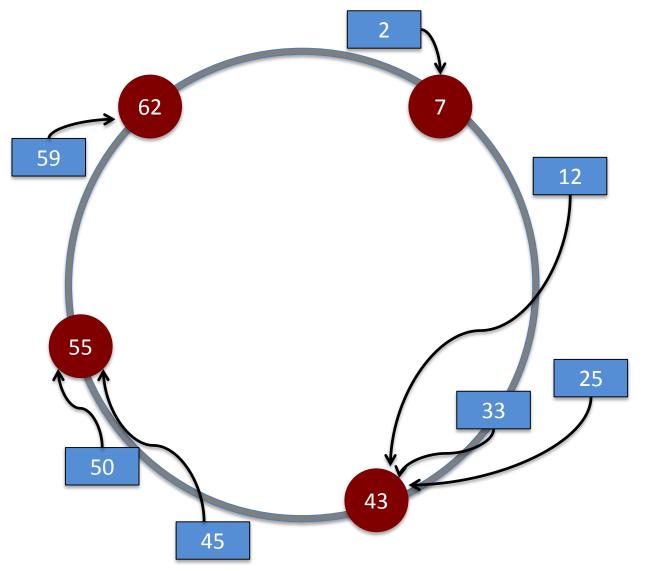
Distributed Data Management Summer Semester 2015 TU Kaiserslautern

Prof. Dr.-Ing. Sebastian Michel
Databases and Information Systems
Group (AG DBIS)

http://dbis.informatik.uni-kl.de/

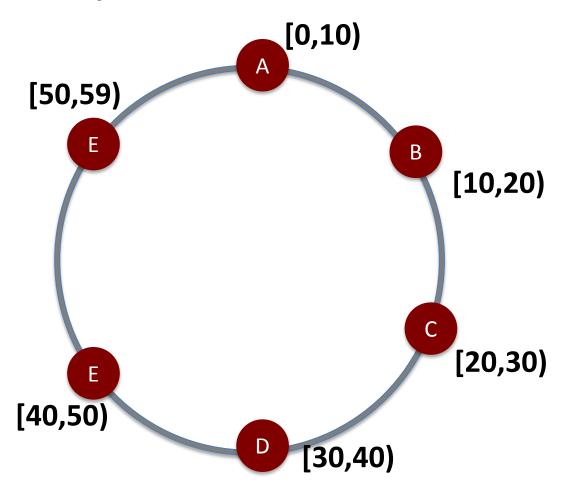
Recap: Consistent Hashing



"Consistent Hashing" in Amazon Dynamo

- Global view of partitioning following the principles of consistent hashing
- No routing tables, no multi-hop routing (reason, network #roundtrips is too expensive for low latency) (check SLA=Service Level Agreements, e.g., 300ms for 99.9%)
- Instead: dissemination of full network information, using gossiping as information dissemination (will see later) => then O(1) lookup cost

Replication in Amazon's Dynamo



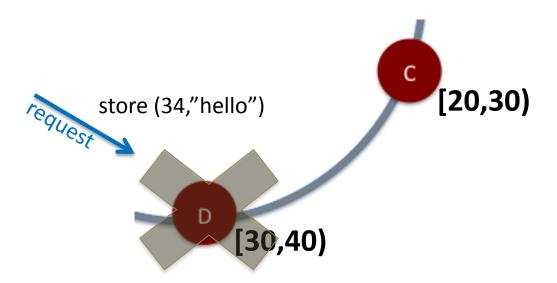
Key	Node	Replica
3	Α	В, С
12	В	C, D
19	В	C, D
20	С	D, E
37	D	E, F
40	E	F, A
54	F	A, B

Replicas are stored at X (here 2) successors of node that "owns" the key.

Replica holders are physically distinct nodes (because of virtual nodes).

Hinted Handoff

What if a node for a key is not available?



 Store data at other node, coordinator, or neighbor. With hint that it is for the (currently) unavailable node.

Hinted Handoff (Cont'd)

- Problem: Hinted Handoff information can get lost if holding node is unavailable.
- Requires protocol that fixes such inconsistencies.

- Each node stores a set of entries of the form
 <key, value, version>
- According to, here, ranges on the "ring", but protocols we see now are independent on that.

Synchronization Process

- Given N nodes (replicas)
- Each of them might or might not have the recent value of an object

 Communication between nodes has to ensure consistent view on data (replicas)

Deterministic Solution

 Node that gets new information sends the information to the N-1 other nodes. (Also called direct mail).

Pros and Cons?

Deterministic Solution

 Node that gets new information sends the information to the N-1 other nodes. (Also called direct mail).

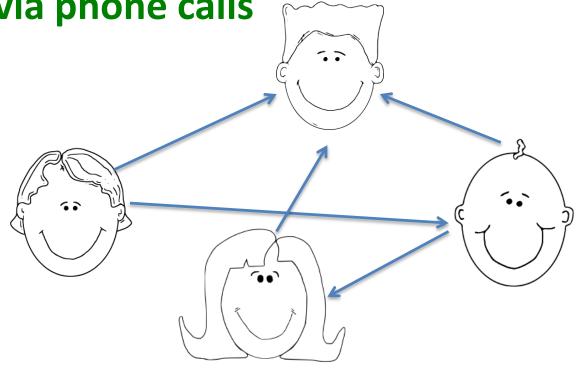
- Pros and Cons?
- Very efficient, no duplicate messages that waste network bandwidth or CPU time.
- But what if a nodes fails?

Epidemic Algorithms

- Anti-entropy: Information is constantly exchanged with randomly selected node. Items to be exchanged are always the current versions items stored in the nodes. Do that continuously.
- Rumor spreading: Information is exchanged with randomly chosen nodes, multiple rounds, then stop. With high probability, data is consistently replicated afterwards.

Rumor Spreading

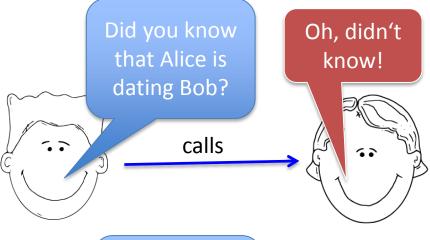
 Think: Spreading rumors between people, say, via phone calls



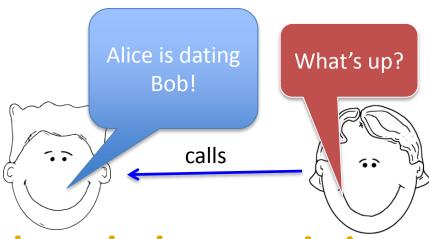
• Two issues: Understanding how rumor spreads (in social networks, e.g.,) or how to devise algorithms that behave similarly (we, here, will look at algorithms)

Variants of Gossiping

 Push: Holder of new information actively distributes it.

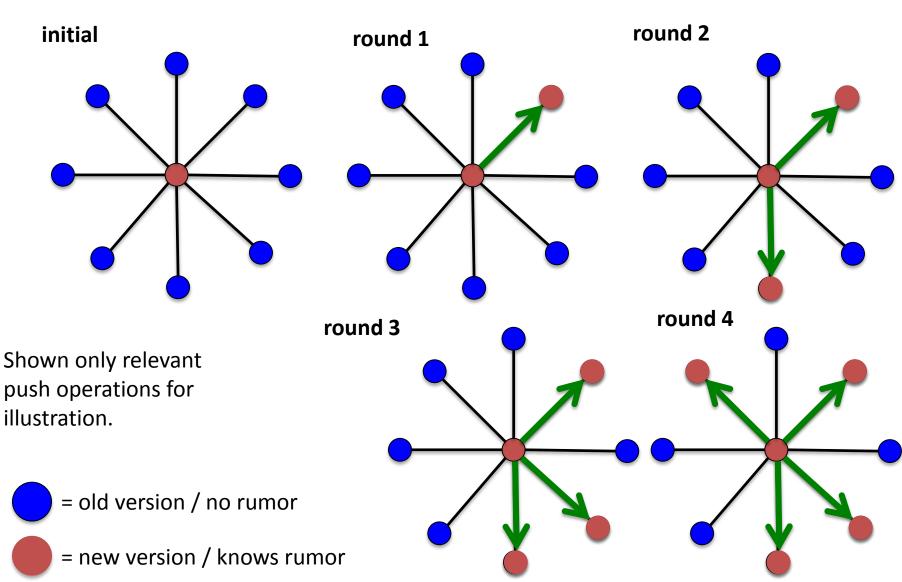


 Pull: People actively call to obtain news.



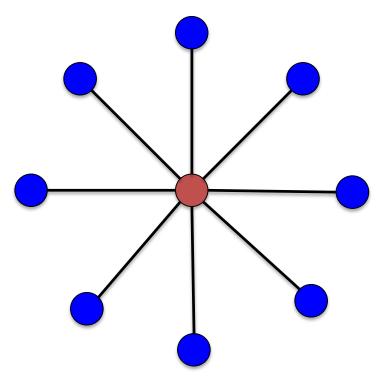
What are the strong and weak characteristics of both strategies?

Push

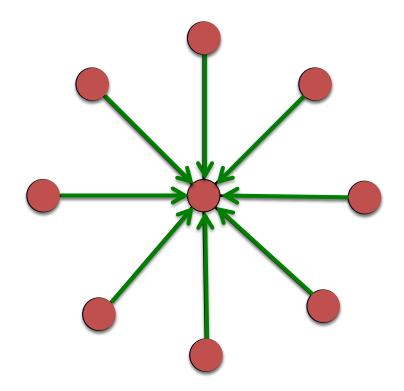


Pull

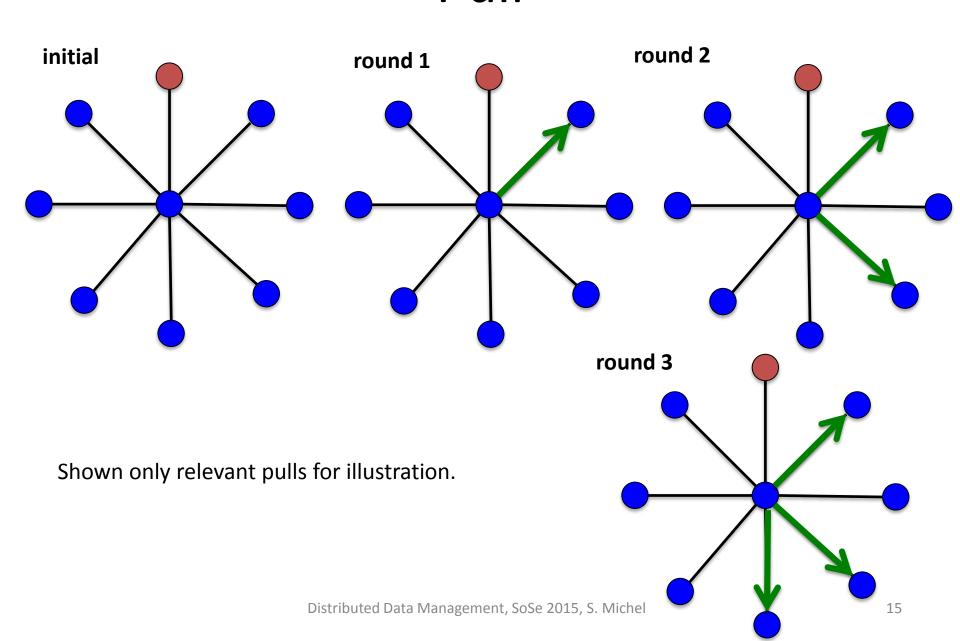
initial



round 1



Pull



Push-Pull

- Combination of push and pull
- Also works in rounds.

- In each round:
 - each node contacts a random neighbor
 - if one of the two has the rumor it tells the other
 - push: caller sends rumor
 - pull: caller receives (learns) rumor

Behavior

 Rumor spreading in case of complete graphs, random graphs or hypercube graphs:

in O(log n) rounds all nodes know the rumor with high probability (w.h.p.)

Also **robust to failures:** if communication links fail with certain probability f<1 then, e.g., O(1/(1-f)) more time needed

Robert Elsässer, Thomas Sauerwald: On the runtime and robustness of randomized broadcasting. Theor. Comput. Sci. 410(36): 3414-3427 (2009)

Anti-Entropy as Secondary Protocol

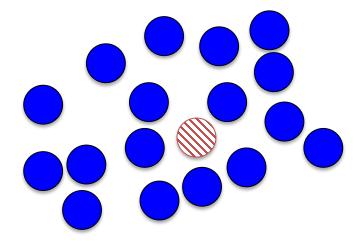
- Demers et al.* put Anti-Entropy in the role of being used after "direct mail" or rumor spreading protocols.
- To fix missing information due to unavailable nodes or
- in case rumor spreading did not receive 100% of all nodes (as it comes with only a "with high probability" guarantee)

^{*}Alan J. Demers et al.: Epidemic Algorithms for Replicated Database Maintenance. PODC 1987: 1-12

Anti-Entropy as Secondary Protocol (2)

 Assume case of having majority of nodes that are in sync already and have the same latest version

What is the method of choice, push or pull?



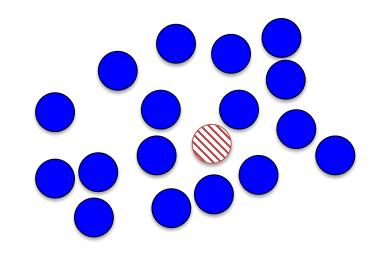
Latest version (="knows rumor")



Not latest version (="does not know rumor")

Anti-Entropy as Secondary Protocol (3)

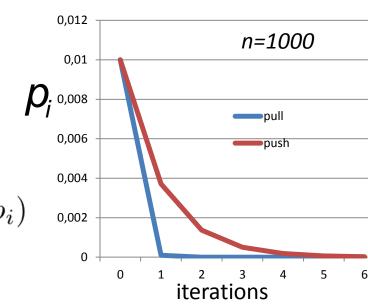
 Then: Pull or pull-push is much better suitable then push only.



Say p_i is probability that a node is not informed, then in next round

for pull:
$$p_{i+1} = (p_i)^2$$

for push:
$$p_{i+1}=p_i imes (1-rac{1}{n})^{n(1-p_i)}$$

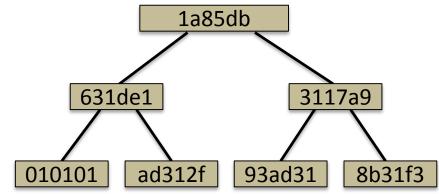


Optimizing Data Exchange/Comparison

- Points before addresses the protocols for data exchange between two nodes.
- In each such process, potentially lots of data is required to be sent/processed.
- Large potential for optimization through compression (signatures).
- First shot: use checksums (e.g., MD5 or SHA-1) of data
- If checksum is the same, data is precisely the same (almost certainly)

Merkle Trees

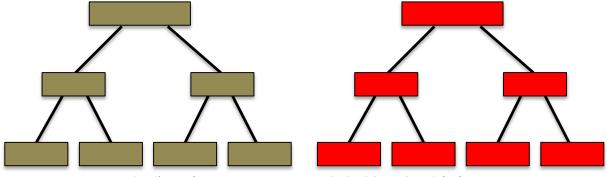
- Hash Trees (invented 1979 by R. Merkle)
- Parent node is hash of its children



- Used in distributed systems for checking consistency of data
- Allows hierarchical checking

Comparing Merkle Trees

- Start at root.
- If the same hash, then stop.
- Otherwise: compare corresponding nodes in levels. For nodes with different hash: go down to children, etc.
- Eventually: Found different data (leaves)
 => exchange them



Merkle Trees in Dynamo

 Each node maintains a separate Merkle tree for each key range (as we have multiple due to virtual nodes!)

• Two nodes compare Merkle trees of the ranges they have in common, as described before.

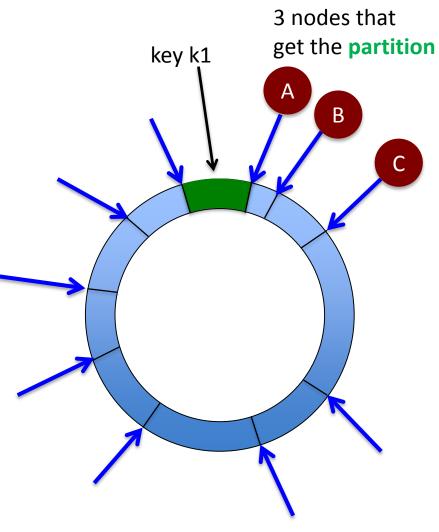
Partitioning / Replication & Dynamics

- Have seen consistent hashing
- Now, slight variations for (said) better performance (again, in Dynamo)

- Dynamics: new nodes cause key ranges of nodes to change.
- Merkle trees need to be recomputed
- Data for "moving" ranges gathered and transferred.

Traditional Consistent Hashing

- S*T nodes are placed randomly (S=number of real nodes, T=virtual instances per node, called also Tokens in*)
- Range between them defines partitions
- N (here =3) copies of partitions in N-1 successors of node that hashing tells to be responsible



not possible to add nodes without affecting data partitioning

Random Placement with Equal-Sized

• Have Q equal sized partitions (Q >> T*S), where

- Nodes are (as before) placed randomly.
- Partition is assigned to N nodes that follow (successors) the end of the partition.

Decoupling of partitioning and partition placement Partition bounds don't change. Efficient 27

key k1

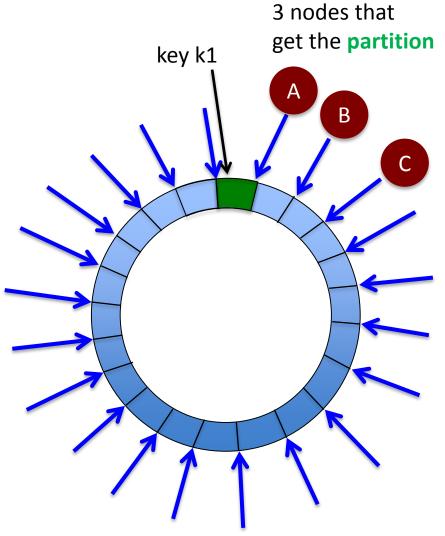
maintenance.

3 nodes that

get the partition

Q/S Virtual Nodes for each Node

- Q/S virtual nodes per node (S=number of nodes in system)
- i.e., one partition per virtual node + replication
- When node enters: steals positions from existing ones
- At leave: gives back
- Such that property remains (means: extra work to do!)



Best load balancing among discussed schemes.

Literature

- Ion Stoica, Robert Morris, David R. Karger, M. Frans Kaashoek, Hari Balakrishnan: Chord: A scalable peer-to-peer lookup service for internet applications. SIGCOMM 2001: 149-160
- Alan J. Demers, Daniel H. Greene, Carl Hauser, Wes Irish, John Larson, Scott Shenker, Howard E. Sturgis, Daniel C. Swinehart, Douglas B. Terry: Epidemic Algorithms for Replicated Database Maintenance. PODC 1987: 1-12
- Robert Elsässer, Thomas Sauerwald: On the runtime and robustness of randomized broadcasting. Theor. Comput. Sci. 410(36): 3414-3427 (2009)
- Richard M. Karp, Christian Schindelhauer, Scott Shenker, Berthold Vöcking: Randomized Rumor Spreading. FOCS 2000: 565-574
- Ralph C. Merkle: A Digital Signature Based on a Conventional Encryption Function. CRYPTO 1987: 369-378
- http://www.allthingsdistributed.com/2007/10/amazons_dynamo.html

Literature

- http://www.cs.berkeley.edu/~brewer/cs262b-2004/PODC-keynote.pdf
- Seth Gilbert, Nancy A. Lynch: Brewer's conjecture and the feasibility of consistent, available, partition-tolerant web services. SIGACT News 33(2): 51-59 (2002)
- Peter Bailis, Shivaram Venkataraman, Michael J. Franklin, Joseph M. Hellerstein, Ion Stoica: Probabilistically Bounded Staleness for Practical Partial Quorums. PVLDB 5(8): 776-787 (2012)
- Leslie Lamport: Time, clocks, and the ordering of events in a distributed system. Communications of the ACM. 21, Nr. 7, July 1978.
- Colin Fidge: Timestamps in Message-Passing Systems That Preserve the Partial Ordering. Australian Computer Science Communications, Vol. 10, No. 1, pp. 56-66, February 1988.
- Philip A. Bernstein, Nathan Goodman: Concurrency Control in Distributed Database Systems. ACM Comput. Surv. 13(2): 185-221 (1981)
- Gerhard Weikum, Gottfried Vossen (2001): Transactional Information Systems, Elsevier, ISBN 1-55860-508-8
- http://highlyscalable.wordpress.com/2012/09/18/distributed-algorithms-in-nosqldatabases/
- David R. Karger, Eric Lehman, Frank Thomson Leighton, Rina Panigrahy, Matthew S. Levine, Daniel Lewin: Consistent Hashing and Random Trees: Distributed Caching Protocols for Relieving Hot Spots on the World Wide Web. STOC 1997: 654-663

Summary NoSQL Part

- Walked through core characteristics of faulttolerant (replicated) distributed data stores.
- Started with simple replica management and state machine replication. t-fault tolerance. Different failure models.
- Paxos for consensus. Logical clocks and vector clocks for bringing order to "events" in a distributed system.
- CAP theorem, BASE, consistency models.
- Data placement (consistent hashing) and synchronization methods (rumor spreading).