Simple Scenario

Suppose you're building a big web cache that holds copies of web pages your users have downloaded:



How do you allocate pages/images to the cache servers?

Static Partitioning

- Items A–C go to this server/bucket/bin, D–Fd-f go to that server/bucket/bin, ...
- Requires **planning**
 - If you used the server name, what if "cowpatties.com" had 1'000'000 pages, but "zebras.com" had only 10?
 - This may cause load imbalance
- Could fill up the bins as they arrive
 - Requires tracking the location of every object at the frontend.
 - May be reasonable design for huge objects

Conventional Hashing

- Recall that a hash function maps elements of a (usually super-big) universe U, like URLs, to "buckets", such as 32-bit values
 - A "good" hash function is **easy** to remember and evaluate.
 - For all practical purposes, a "good" hash function behaves like a totally random function.
- Given a "good" hash function, we can set

bucket = hash(x) mod num_buckets

- Now the server we use is a deterministic function of the item
 - e.g., sha1(URL) \rightarrow 160 bit ID % 20 \rightarrow a server ID

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- Now the server we use is a deterministic function of the item
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- But what happens if we want to **add or remove a server**?

Consistent Hashing

 The key idea is: in addition to hashing the names of all objects (URLs) x, like before, we also hash the names of all the servers s. The object and server names need to be hashed to the same range, such as 32-bit values.



- Given an object x that hashes to the bucket h(x), we scan buckets to the right of h(x) until we find a bucket h(s) to which the name of some server s hashes.
- We wrap around the array, if necessary.

Consistent Hashing



- Hash of object = closest clockwise bucket ("*successor*")
- N servers partition the circle into N segments, with each server responsible for all objects in one of these segments.

Properties

- Balance: assuming reasonable hash functions, by symmetry, the expected load on each of the N servers is exactly a 1/N fraction of the objects.
- Smoothness: suppose we add a new server s which objects have to move? Only the objects stored at s. When a server is added, the expected number of items that move to the newly added server is (#items)/(1+#servers).
- Complexity: to implement Lookup and Insert we can use a hash table, a heap, a balanced binary search tree, with O(log(n)) lookup and insert implementations.

Implementation

• To insert an item x:

(a) Find the successor of $h_i(x)$ in the BST (if it has no successor in the BST then return the machine with the smallest h_m value)

(b) Store x in the returned machine.

• To delete an item x:

(a) Find the successor of $h_i(x)$ in the BST (if it has no successor in the BST then return the machine with the smallest h_m value)

(b) Delete x in the returned machine.

- BST: a Binary Search Tree whose keys are the values assigned to the machines.
- h_{i:} the function hashing items to the interval [0, 1].
- h_m: the function hashing machines to the interval [0, 1].

Implementation

• To insert a new machine Y:

There may be some existing items that should be stored in the new machine Y, but these items now are all stored in the successor of $h_m(Y)$ (or the machine with the smallest h_m if $h_m(Y)$ is the largest value).

(a) Find the successor of $h_m(Y)$ in the BST (if it has no successor in the BST then return the machine with the smallest h_m value)

(b) Move all items whose h_i value is less than $h_m(Y)$ to the newly inserted machine Y .

• To delete an existing machine Y:

(a) Find the successor of $h_m(Y)$ in the BST (if it has no successor in the BST then return the machine with the smallest h_m value)

(b) Move all items in Y to the returned machine.

- BST: a Binary Search Tree whose keys are the values assigned to the machines.
- $h_{i:}$ the function hashing items to the interval [0, 1].
- h_m: the function hashing machines to the interval [0, 1].

Virtual Nodes

- While the **expected load** of each server is a 1/N fraction of the N objects, the **actual load** of each server will vary.
 - If you pick N random points on the circle, you're very unlikely to get a perfect partition of the circle into equalsized segments.
- To reduce imbalance, systems often represent each physical node as k different buckets, sometimes called "virtual nodes" (but really, it's just multiple buckets).
 - For example, we can hash a server with K different hash functions on the same co-domain.
- Objects are assigned as before.

Virtual Nodes



- With N servers and K virtual nodes per server, by symmetry, each virtual node still expects to get a 1/(KN) fraction of the objects.
- This replication increases the number of keys stored in the balanced binary search by a factor of K, but it reduces the variance in load across servers significantly.
- Choosing K ≈ log₂(N) is large enough to obtain reasonably balanced loads.

Use of consistent hashing

- The implementation of consistent hashing first appeared in a research paper in 1997 (STOC).
- In 1999, the trailer "Star Wars: The Phantom Menace" release put <u>apple.com</u> servers offline, while <u>akamai.com</u>, implementing consistent hashing, was able to serve a unauthorised copy.
- Consistent hashing is re-purposed in 2001 to address technical challenges that arise in peer-to-peer (P2P) networks (e.g., Chord and BitTorrent).
- In 2006 Amazon implements its internal Dynamo system using consistent hashing.