Hash Tables: Distributed Hash Tables

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Data Structures http://bit.ly/algospecialization

Outline

1 Online Storage Systems

2 Distributed Hash Tables

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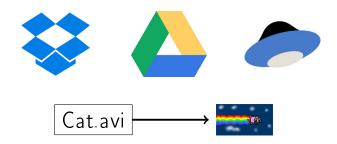
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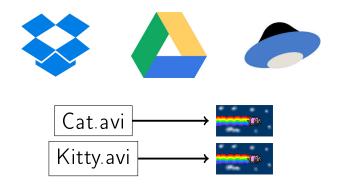
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- This lesson

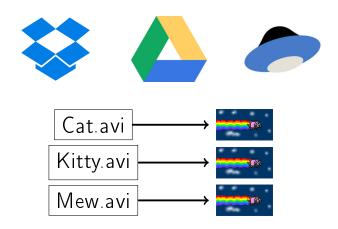


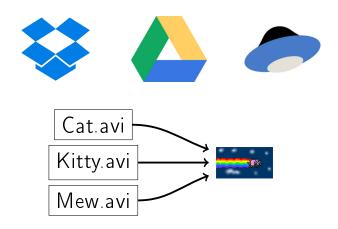


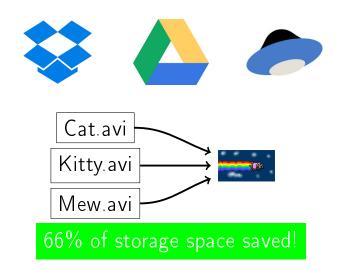












New File Upload



Need to determine whether there is already the same file in the system

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- If there's the same file, store a link to it instead of the new file

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- N grows, so total running time of uploads grows as $O(N^2)$

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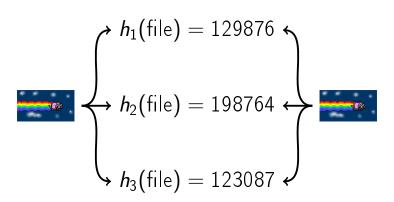
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- Compute hashes locally before upload

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- There are algorithms to find collisions for known hash functions
- However, even for one hash function collisions are extremely rare
- Using 3 or 5 hashes, you probably won't see a collision in a lifetime

Problem: O(N) Comparisons

Still have to compare with N already stored files

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- Only need the hashes to search in the table

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- Solution: distributed hash tables

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- Send request to that computer, search/modify in the local hash table

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- \bullet h(O) mod 1000 no longer works

Consistent Hashing

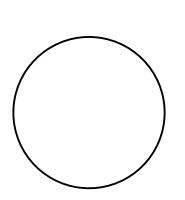
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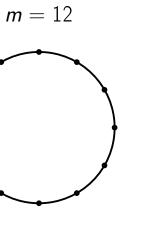
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- Map computer IDs to the same circle: $compID \rightarrow point number h(compID)$





$$m = 12$$

$$10$$

$$1$$

$$2$$

$$9$$

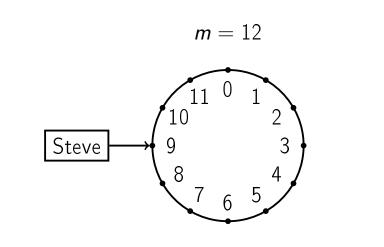
$$3$$

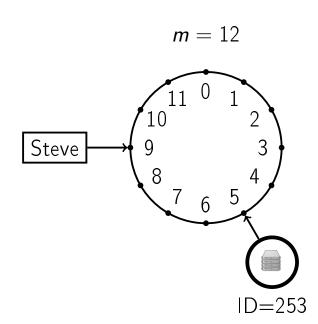
$$8$$

$$7$$

$$6$$

$$5$$



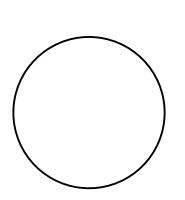


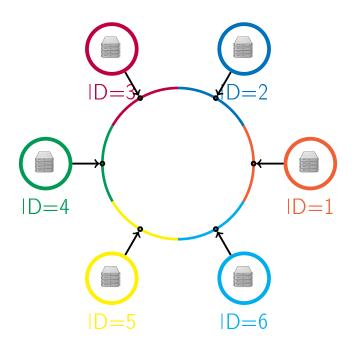
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- Each computer stores all objects falling on some arc of the circle



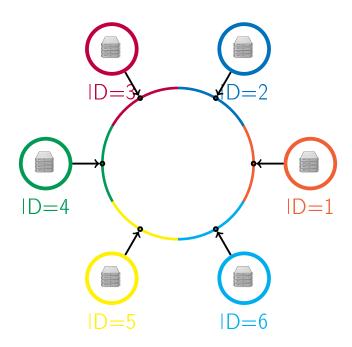


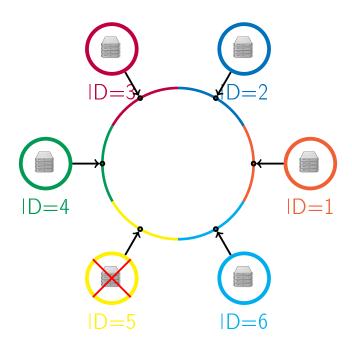
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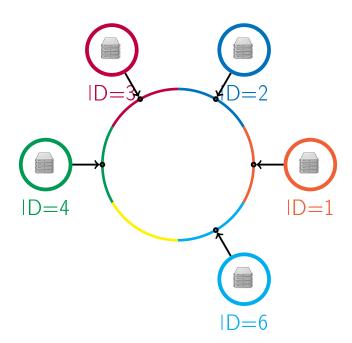
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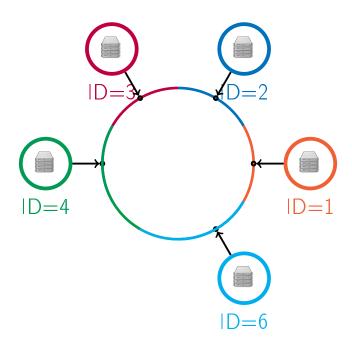
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- When a new computer is added, it takes data from the "neighbors"









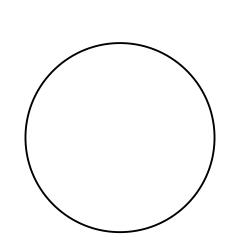
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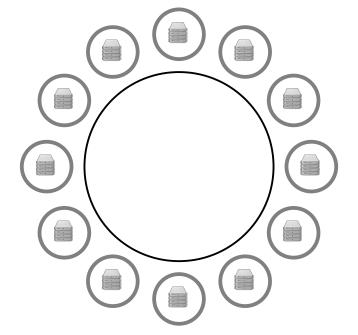
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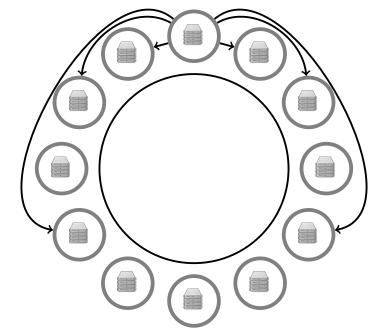
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- E.g., each node knows neighbors, $\pm 1, \pm 2, \pm 4, \pm 8, \ldots O(\log n)$ nodes, and can get/send any key in $O(\log n)$







Conclusion

- Distributed Hash Tables (DHT) store
 Big Data on many computers
- Consistent Hashing (CH) is one way to determine which computer stores which data
- CH uses mapping of keys and computer IDs on a circle
- Each computer stores a range of keys
- Overlay Network is used to route the data to/from the right computer