Principles of Software Construction: Objects, Design, and Concurrency

Part 6: Concurrency and distributed systems

Distributed systems

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Administrivia

- Homework 5b due Thursday, 11:59 p.m.
 - Finish by Friday (10 Apr) 10 a.m. if you want to be considered as a "Best Framework" for Homework 5c
 - Our evaluation considers:
 - Novelty
 - Functional correctness
 - Documentation

— ...

Key concepts from last Thursday

Concurrency at the language level

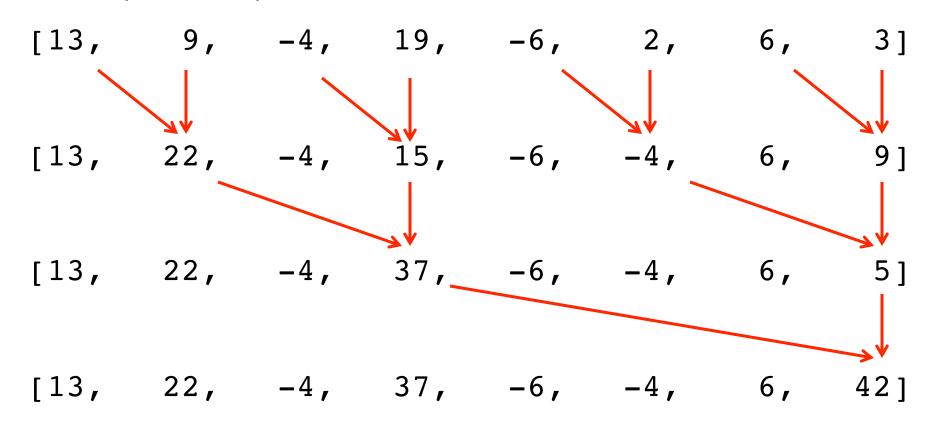
Consider: int sum = 0; Iterator i = coll.iterator(); while (i.hasNext()) { sum += i.next();

In python:

```
sum = 0;
for item in coll:
    sum += item
```

Parallel prefix sums algorithm, winding

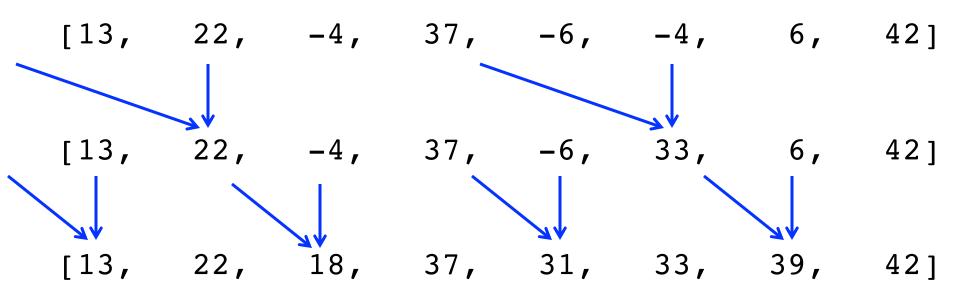
Computes the partial sums in a more useful manner





Parallel prefix sums algorithm, unwinding

Now unwinds to calculate the other sums



Recall, we started with:

$$[13, 9, -4, 19, -6, 2, 6, 3]$$

A framework for asynchronous computation

The java.util.concurrent.Future<V> interface

```
V get();
V get(long timeout, TimeUnit unit);
boolean isDone();
boolean cancel(boolean mayInterruptIfRunning);
boolean isCancelled();
```

• The java.util.concurrent.ExecutorService interface

```
Future
Future<V> submit(Runnable task);
Future<V> submit(Callable<V> task);
List<Future<V> invokeAll(Collection<Callable<V>> tasks);
Future<V> invokeAny(Collection<Callable<V>> tasks);
```

Fork/Join: another common computational pattern

- In a long computation:
 - Fork a thread (or more) to do some work
 - Join the thread(s) to obtain the result of the work
- The java.util.concurrent.ForkJoinPool class
 - Implements ExecutorService
 - Executes java.util.concurrent.ForkJoinTask<V> or java.util.concurrent.RecursiveTask<V> or java.util.concurrent.RecursiveAction



Parallel prefix sums algorithm

- How good is this?
 - Work: O(n)
 - Depth: O(lg n)
- See PrefixSumsSequentialImpl.java
 - n-1 additions
 - Memory access is sequential
- For PrefixSumsNonsequentialImpl.java
 - About 2n useful additions, plus extra additions for the loop indexes
 - Memory access is non-sequential
- The punchline: Constants matter.

Today: Distributed system design

- Java networking fundamentals
- Introduction to distributed systems
 - Motivation: reliability and scalability
 - Failure models
 - Techniques for:
 - Reliability (availability)
 - Scalability
 - Consistency

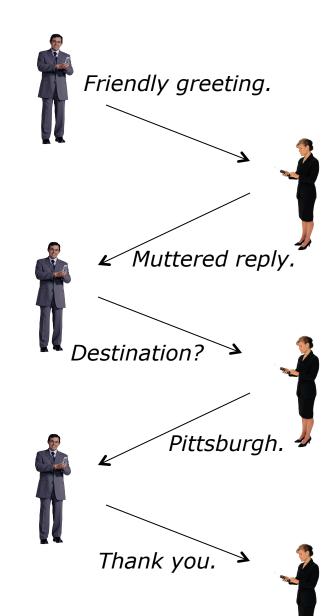
Our destination: Distributed systems

- Multiple system components (computers) communicating via some medium (the network)
- Challenges:
 - Heterogeneity
 - Scale
 - Geography
 - Security
 - Concurrency
 - Failures

(courtesy of http://www.cs.cmu.edu/~dga/15-440/F12/lectures/02-internet1.pdf

Communication protocols

 Agreement between parties for how communication should take place

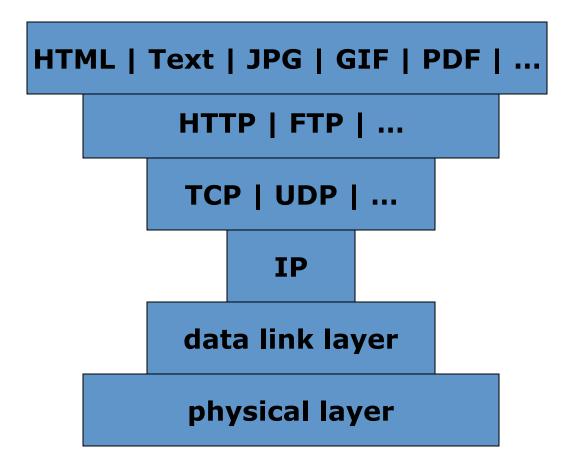


(courtesy of http://www.cs.cmu.edu/~dga/15-440/F12/lectures/02-internet1.pdf

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Abstractions of a network connection



Packet-oriented and stream-oriented connections

- UDP: User Datagram Protocol
 - Unreliable, discrete packets of data
- TCP: Transmission Control Protocol
 - Reliable data stream

Internet addresses and sockets

- For IP version 4 (IPv4) host address is a 4-byte number
 - e.g. 127.0.0.1
 - Hostnames mapped to host IP addresses via DNS
 - ~4 billion distinct addresses
- Port is a 16-bit number (0-65535)
 - Assigned conventionally
 - e.g., port 80 is the standard port for web servers

Networking in Java

The java.net.InetAddress: static InetAddress getByName(String host); static InetAddress getByAddress(byte[] b); static InetAddress getLocalHost(); The java.net.Socket: Socket(InetAddress addr, int port); boolean isConnected(); boolean isClosed(); close(); void InputStream getInputStream(); OutputStream getOutputStream(); The java.net.ServerSocket: ServerSocket(int port); Socket accept(); void close();

Simple sockets demos

- NetworkServer.java
- A basic chat system:
 - TransferThread.java
 - TextSocketClient.java
 - TextSocketServer.java

Higher levels of abstraction

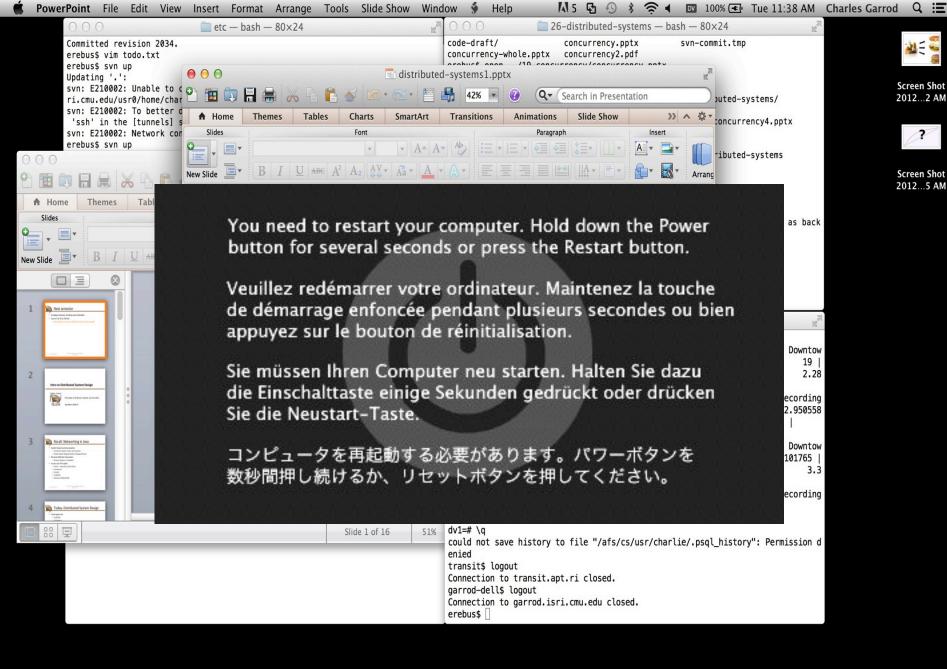
- Application-level communication protocols
- Frameworks for simple distributed computation
 - Remote Procedure Call (RPC)
 - Java Remote Method Invocation (RMI)
- Common patterns of distributed system design
- Complex computational frameworks
 - e.g., distributed map-reduce



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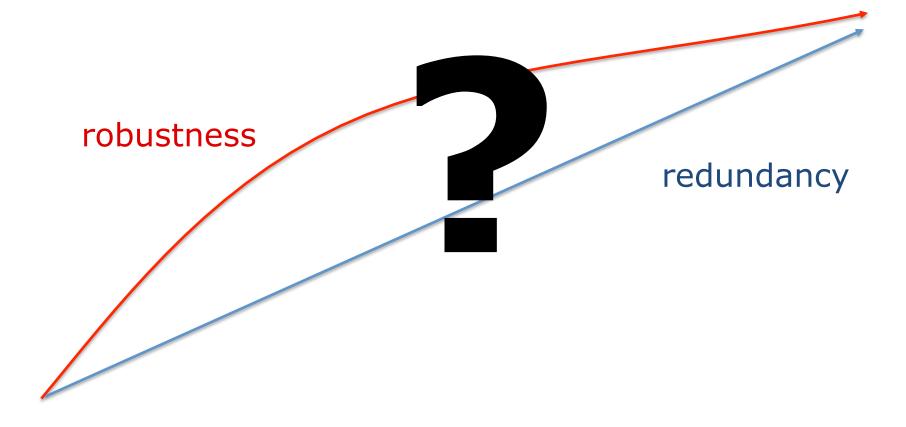


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Aside: The robustness vs. redundancy curve

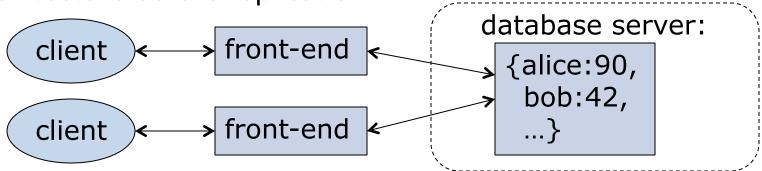


Metrics of success

- Reliability
 - Often in terms of availability: fraction of time system is working
 - 99.999% available is "5 nines of availability"
- Scalability
 - Ability to handle workload growth

A case study: Passive primary-backup replication

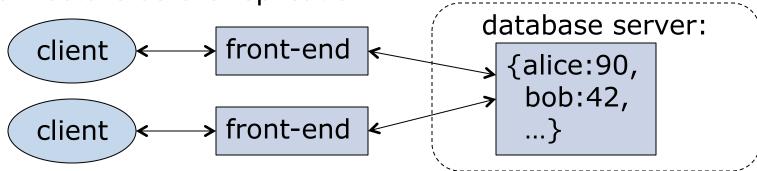
• Architecture before replication:



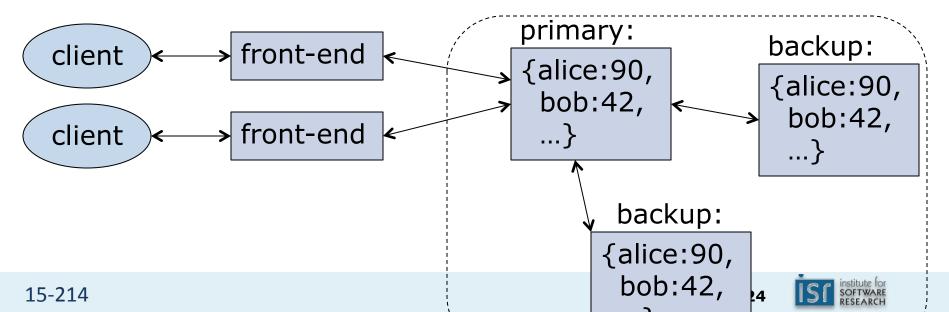
Problem: Database server might fail

A case study: Passive primary-backup replication

• Architecture before replication:



- Problem: Database server might fail
- Solution: Replicate data onto multiple servers



Passive primary-backup replication protocol

- 1. Front-end issues request with unique ID to primary DB
- Primary checks request ID
 - If already executed request, re-send response and exit protocol
- 3. Primary executes request and stores response
- 4. If request is an update, primary DB sends updated state, ID, and response to all backups
 - Each backup sends an acknowledgement
- 5. After receiving all acknowledgements, primary DB sends response to front-end

Issues with passive primary-backup replication

- If primary DB crashes, front-ends need to agree upon which unique backup is new primary DB
 - Primary failure vs. network failure?
- If backup DB becomes new primary, surviving replicas must agree on current DB state
- If backup DB crashes, primary must detect failure to remove the backup from the cluster
 - Backup failure vs. network failure?
- If replica fails* and recovers, it must detect that it previously failed
- Many subtle issues with partial failures

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More issues...

- Concurrency problems?
 - Out of order message delivery?
 - Time...
- Performance problems?
 - 2n messages for n replicas
 - Failure of any replica can delay response
 - Routine network problems can delay response
- Scalability problems?
 - All replicas are written for each update
 - Primary DB responds to every request



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Types of failure behaviors

- Fail-stop
- Other halting failures
- Communication failures
 - Send/receive omissions
 - Network partitions
 - Message corruption
- Data corruption
- Performance failures
 - High packet loss rate
 - Low throughput
 - High latency
- Byzantine failures

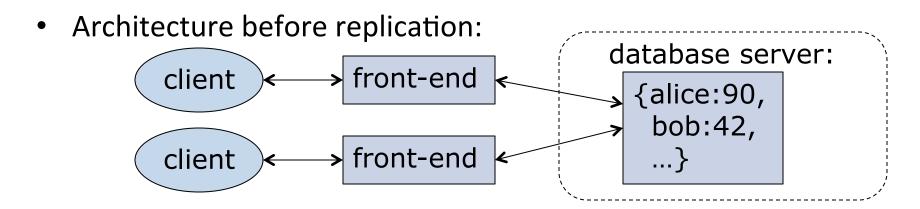
Common assumptions about failures

- Behavior of others is fail-stop (ugh)
- Network is reliable (ugh)
- Network is semi-reliable but asynchronous
- Network is lossy but messages are not corrupt
- Network failures are transitive
- Failures are independent
- Local data is not corrupt
- Failures are reliably detectable
- Failures are unreliably detectable

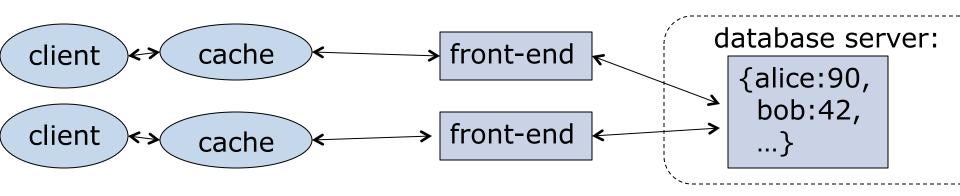
Some distributed system design goals

- The end-to-end principle
 - When possible, implement functionality at the end nodes (rather than the middle nodes) of a distributed system
- The robustness principle
 - Be strict in what you send, but be liberal in what you accept from others
 - Protocols
 - Failure behaviors
- Benefit from incremental changes
- Be redundant
 - Data replication
 - Checks for correctness

Replication for scalability: Client-side caching

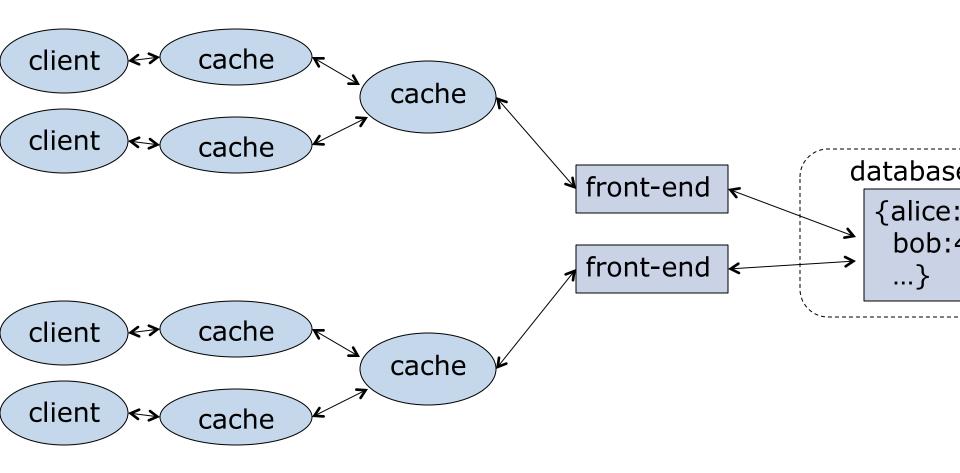


- Problem: Server throughput is too low
- Solution: Cache responses at (or near) the client
 - Cache can respond to repeated read requests



Replication for scalability: Client-side caching

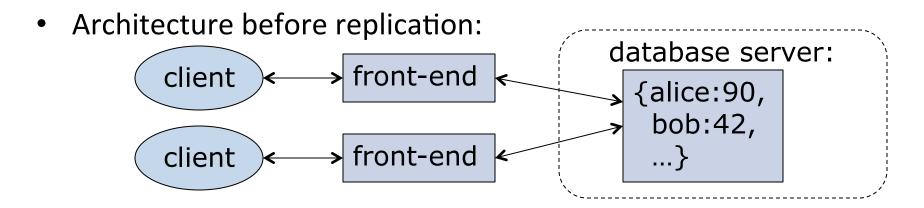
Hierarchical client-side caches:



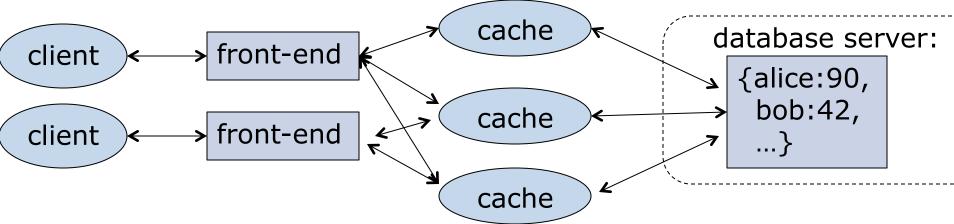
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Replication for scalability: Server-side caching



- Problem: Database server throughput is too low
- Solution: Cache responses on multiple servers
 - Cache can respond to repeated read requests



Cache invalidation

- Time-based invalidation (a.k.a. expiration)
 - Read-any, write-one
 - Old cache entries automatically discarded
 - No expiration date needed for read-only data
- Update-based invalidation
 - Read-any, write-all
 - DB server broadcasts invalidation message to all caches when the DB is updated



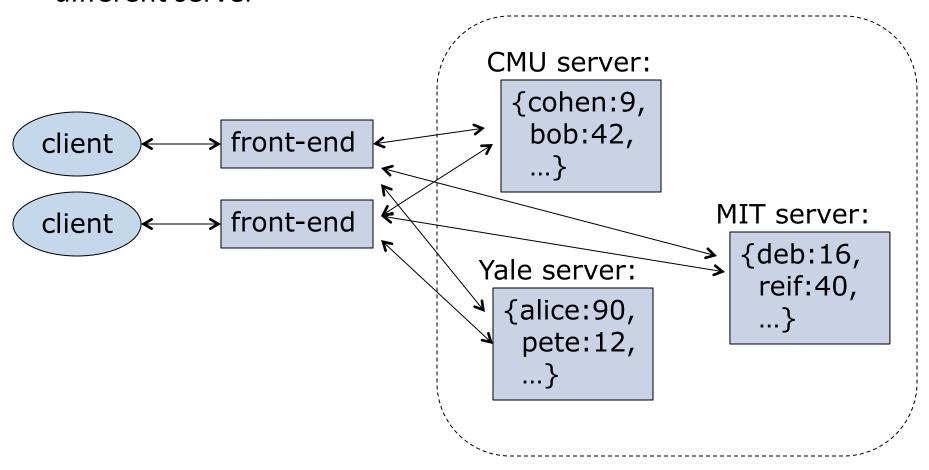
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Cache replacement policies

- Problem: caches have finite size
- Common* replacement policies
 - Optimal (Belady's) policy
 - Discard item not needed for longest time in future
 - Least Recently Used (LRU)
 - Track time of previous access, discard item accessed least recently
 - Least Frequently Used (LFU)
 - Count # times item is accessed, discard item accessed least frequently
 - Random
 - Discard a random item from the cache

Partitioning for scalability

 Partition data based on some property, put each partition on a different server



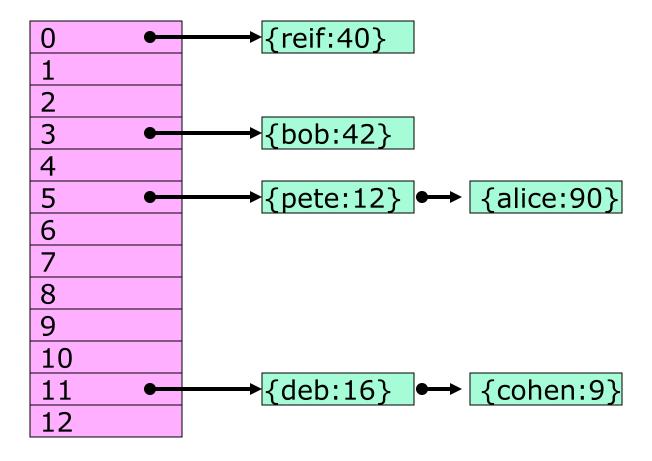
Horizontal partitioning

- a.k.a. "sharding"
- A table of data:

username	school	value
cohen	CMU	9
bob	CMU	42
alice	Yale	90
pete	Yale	12
deb	MIT	16
reif	MIT	40

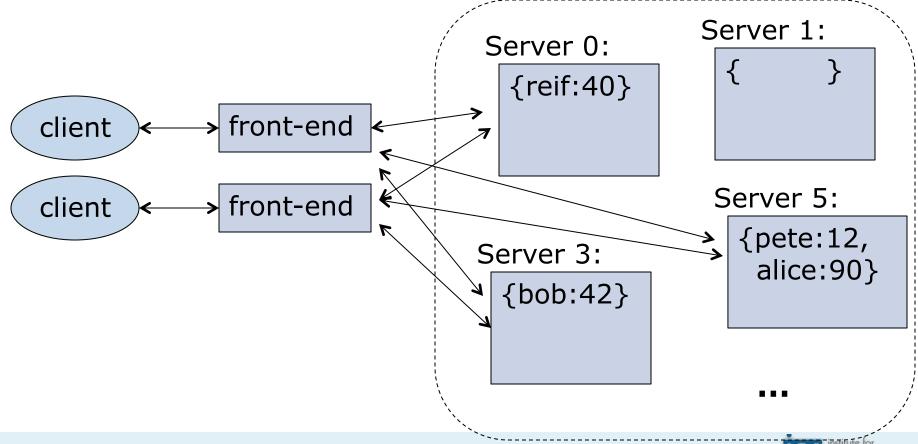
Recall: Basic hash tables

For n-size hash table, put each item X in the bucket:
 X.hashCode() % n



Partitioning with a distributed hash table

- Each server stores data for one bucket
- To store or retrieve an item, front-end server hashes the key, contacts the server storing that bucket



Consistent hashing

- Goal: Benefit from incremental changes
 - Resizing the hash table (i.e., adding or removing a server) should not require moving many objects
- E.g., Interpret the range of hash codes as a ring
 - Each bucket stores data for a range of the ring
 - Assign each bucket an ID in the range of hash codes
 - To store item X don't compute X.hashCode() % n. Instead, place X in bucket with the same ID as or next higher ID than X.hashCode()

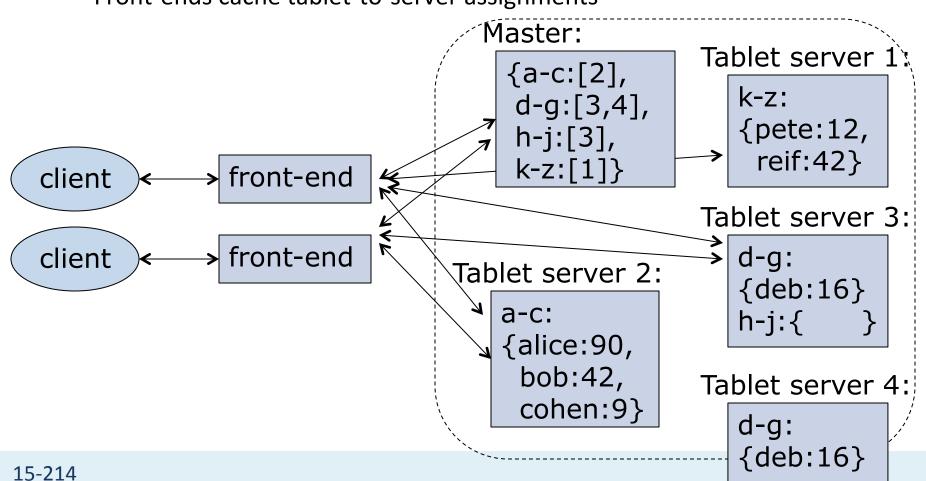
Problems with hash-based partitioning

- Front-ends need to determine server for each bucket
 - Each front-end stores look-up table?
 - Master server storing look-up table?
 - Routing-based approaches?
- Places related content on different servers
 - Consider range queries:

SELECT * FROM users WHERE lastname STARTSWITH 'G'

Master/tablet-based systems

- Dynamically allocate range-based partitions
 - Master server maintains tablet-to-server assignments
 - Tablet servers store actual data
 - Front-ends cache tablet-to-server assignments



Coming next...

- More distributed systems
 - MapReduce