

# Achieving Rapid Response Times in Large Online Services

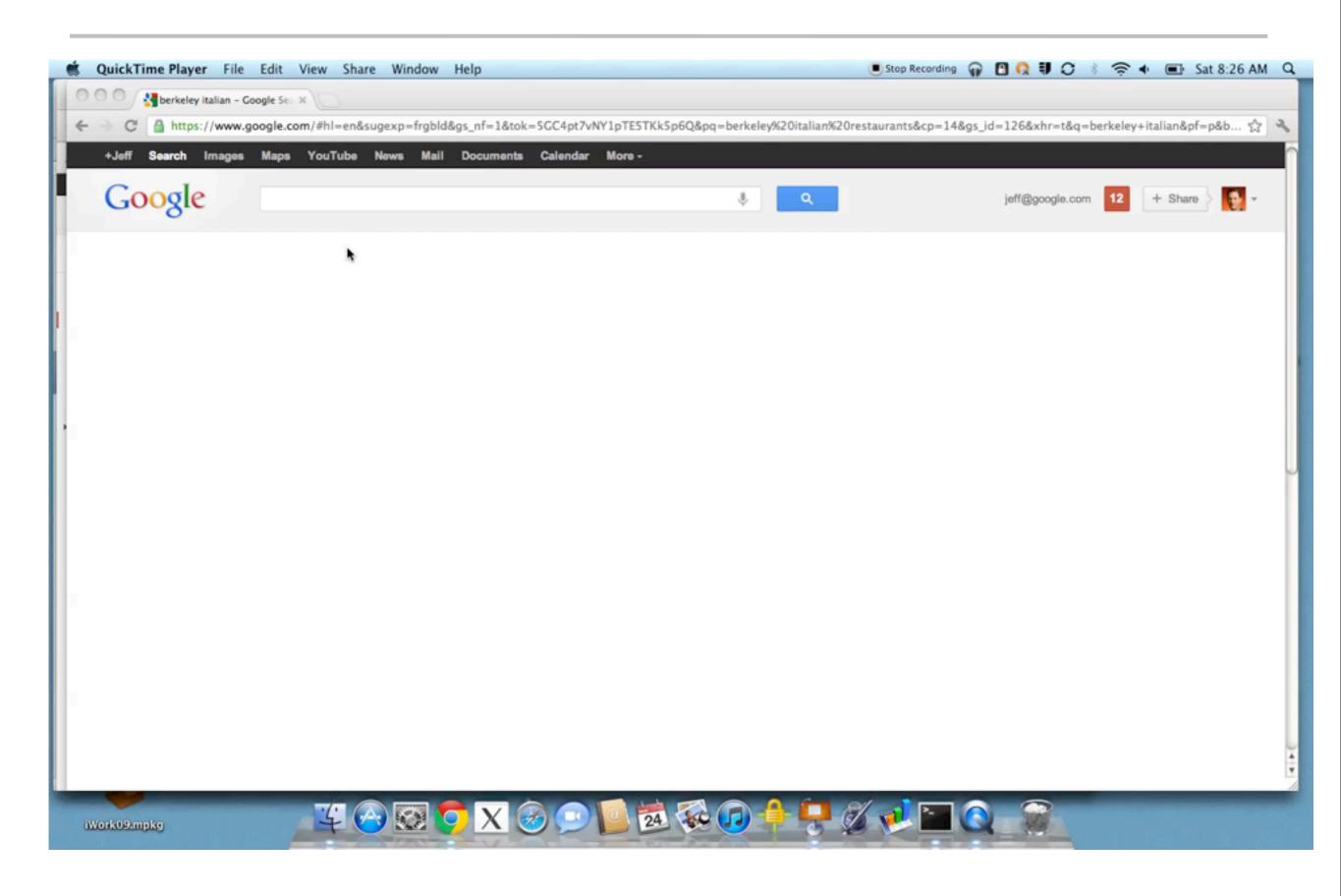
Jeff Dean Google Fellow

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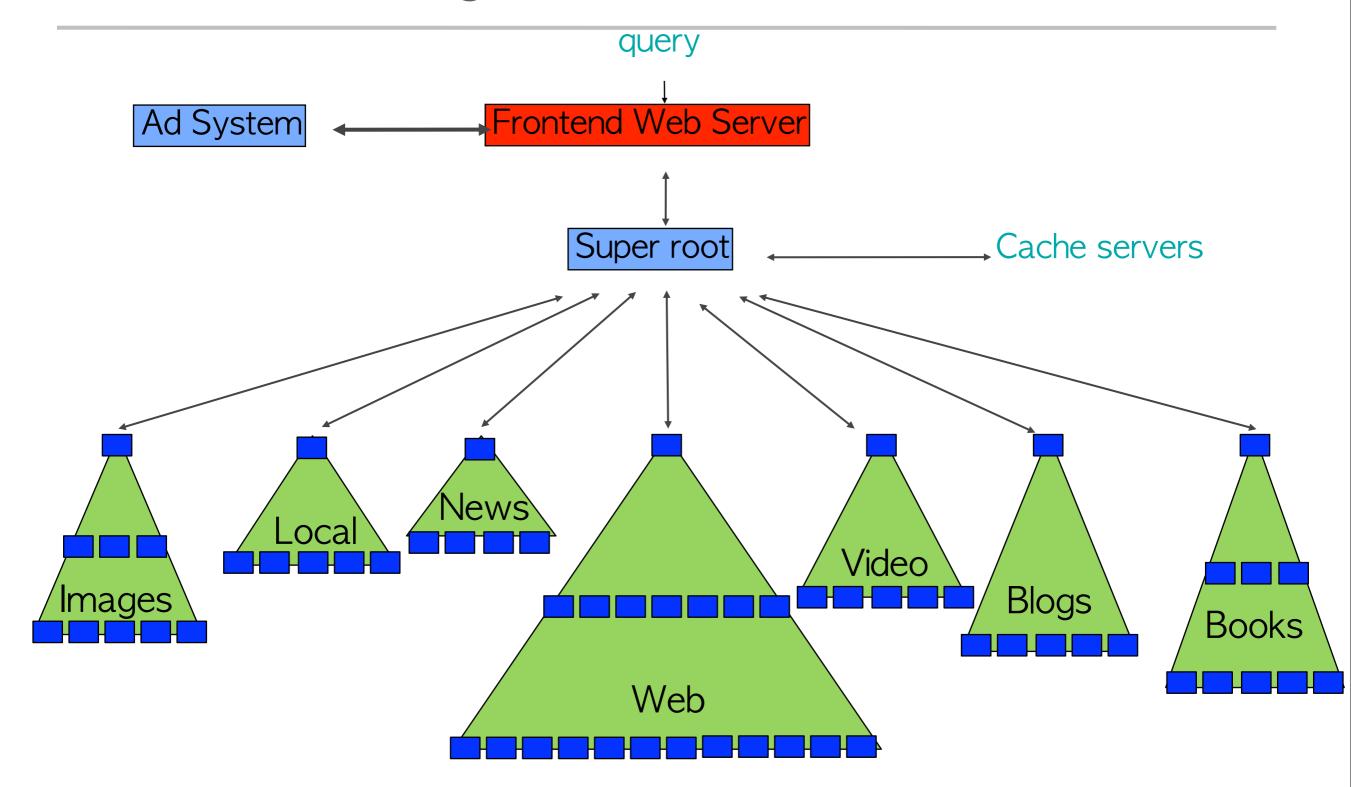
## Faster Is Better



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## Large Fanout Services





## Why Does Fanout Make Things Harder?

- Overall latency ≥ latency of slowest component
  - -small blips on individual machines cause delays
  - -touching more machines increases likelihood of delays
- Server with 1 ms avg. but 1 sec 99%ile latency
  - -touch 1 of these: 1% of requests take ≥1 sec
  - -touch 100 of these: 63% of requests take ≥1 sec



## One Approach: Squash All Variability

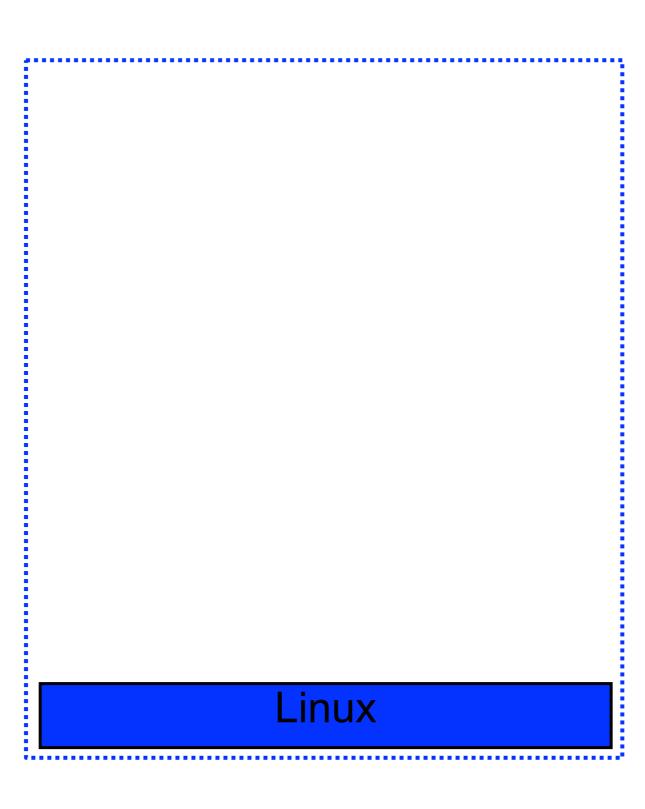
- Careful engineering all components of system
- Possible at small scale
  - –dedicated resources
  - -complete control over whole system
  - -careful understanding of all background activities
  - less likely to have hardware fail in bizarre ways
- System changes are difficult
  - -software or hardware changes affect delicate balance

Not tenable at large scale: need to share resources

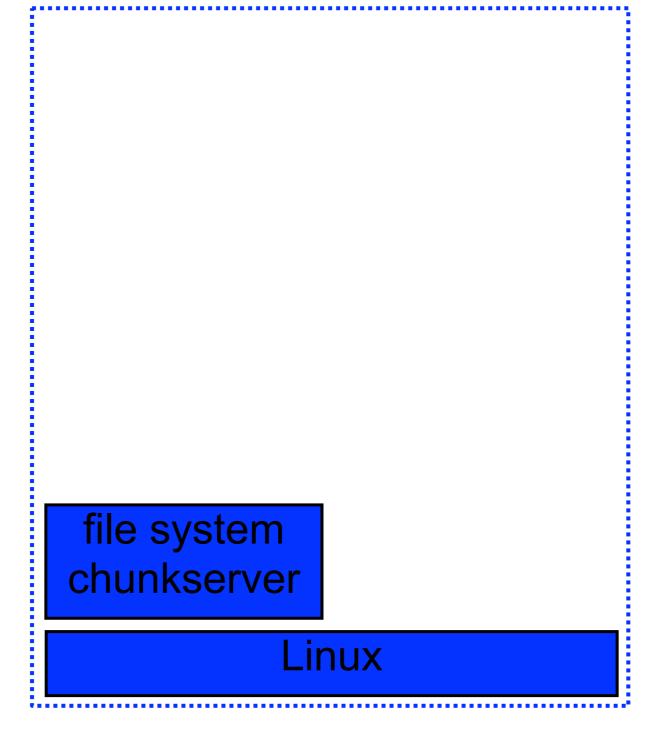


- Huge benefit: greatly increased utilization
- ... but hard to predict effects increase variability
  - –network congestion
  - background activities
  - -bursts of foreground activity
  - -not just your jobs, but everyone else's jobs, too
- Exacerbated by large fanout systems

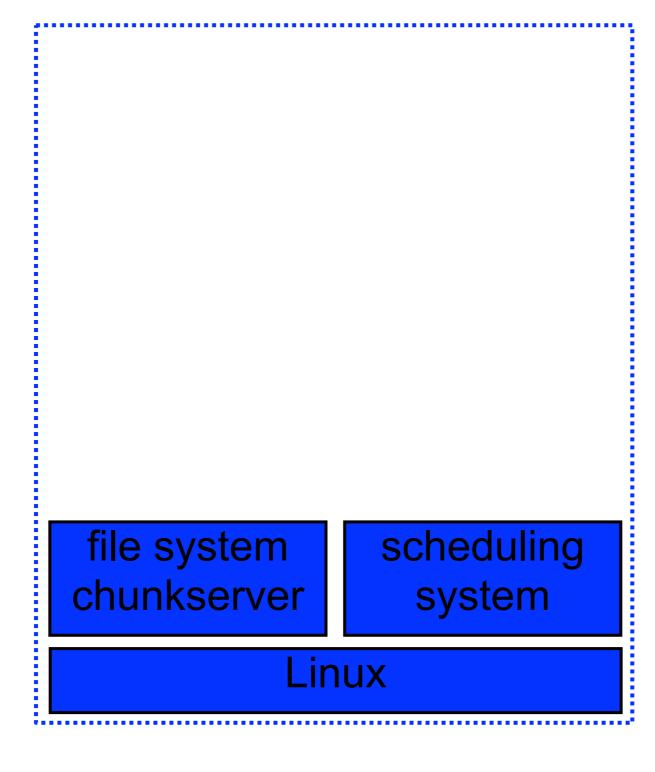




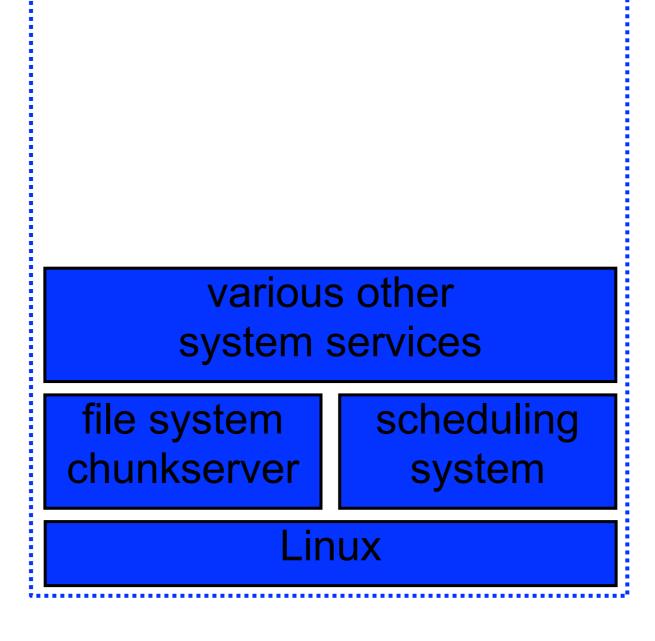




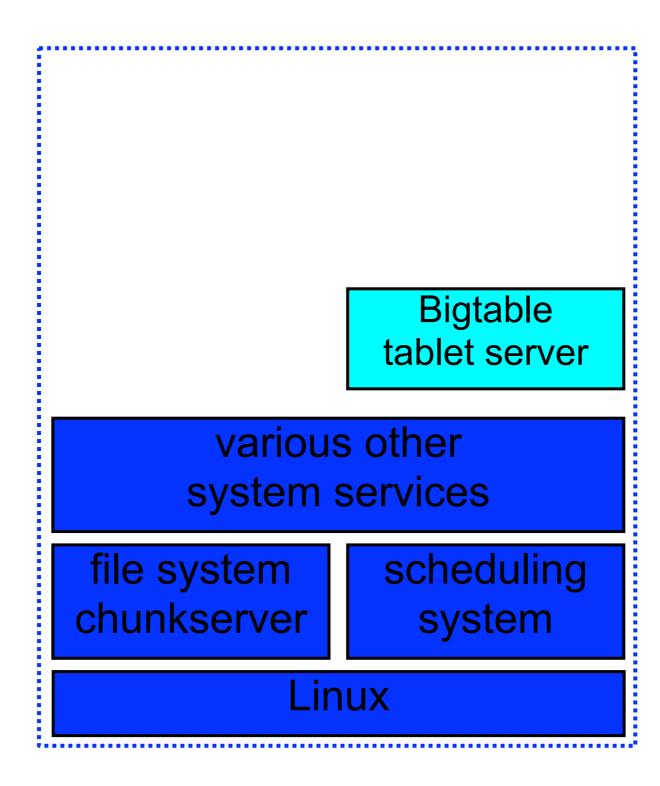




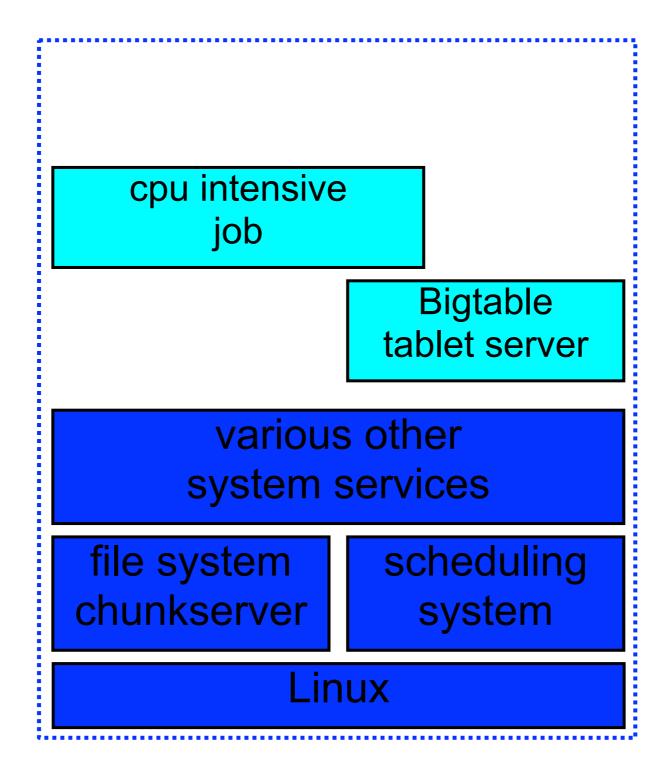




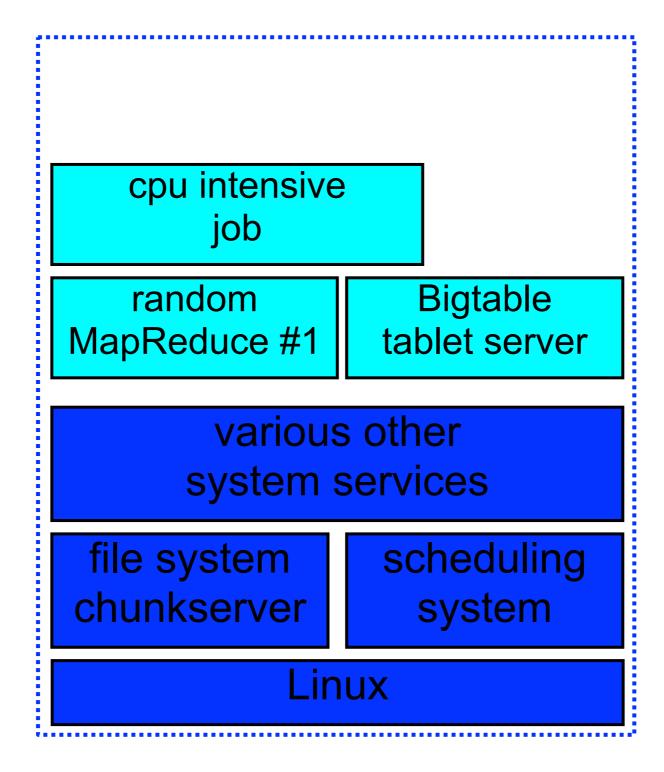




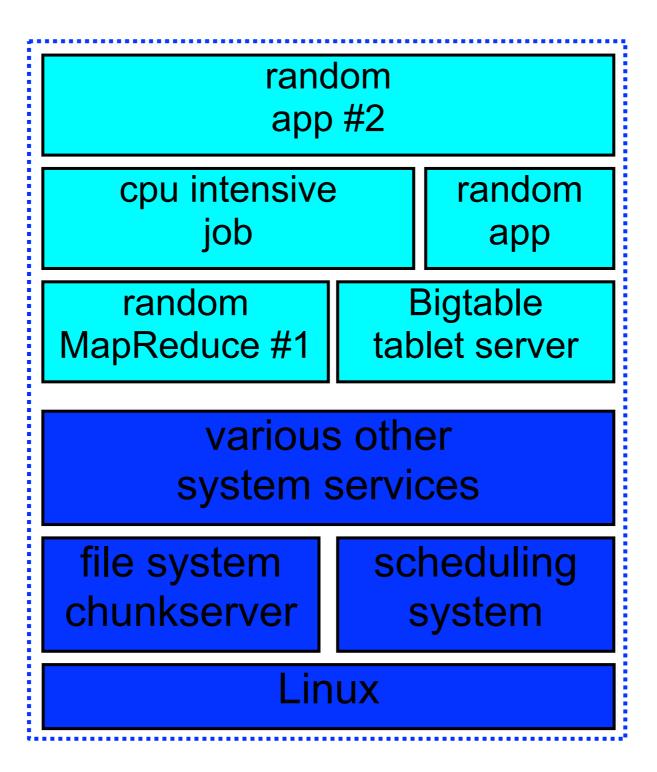














## Basic Latency Reduction Techniques

- Differentiated service classes
  - -prioritized request queues in servers
  - –prioritized network traffic
- Reduce head-of-line blocking
  - -break large requests into sequence of small requests
- Manage expensive background activities
  - –e.g. log compaction in distributed storage systems
  - -rate limit activity
  - -defer expensive activity until load is lower



## Synchronized Disruption

- Large systems often have background daemons
  - -various monitoring and system maintenance tasks
- Initial intuition: randomize when each machine performs these tasks
  - -actually a very bad idea for high fanout services
    - at any given moment, at least one or a few machines are slow
- Better to actually synchronize the disruptions
  - -run every five minutes "on the dot"
  - one synchronized blip better than unsynchronized



## Tolerating Faults vs. Tolerating Variability

#### Tolerating faults:

- rely on extra resources
  - RAIDed disks, ECC memory, dist. system components, etc.
- make a reliable whole out of unreliable parts
- Tolerating variability:
  - use these same extra resources
  - make a predictable whole out of unpredictable parts
- Times scales are very different:
  - variability: 1000s of disruptions/sec, scale of milliseconds
  - faults: 10s of failures per day, scale of tens of seconds



## Latency Tolerating Techniques

#### Cross request adaptation

- –examine recent behavior
- -take action to improve latency of future requests
- -typically relate to balancing load across set of servers
- -time scale: 10s of seconds to minutes

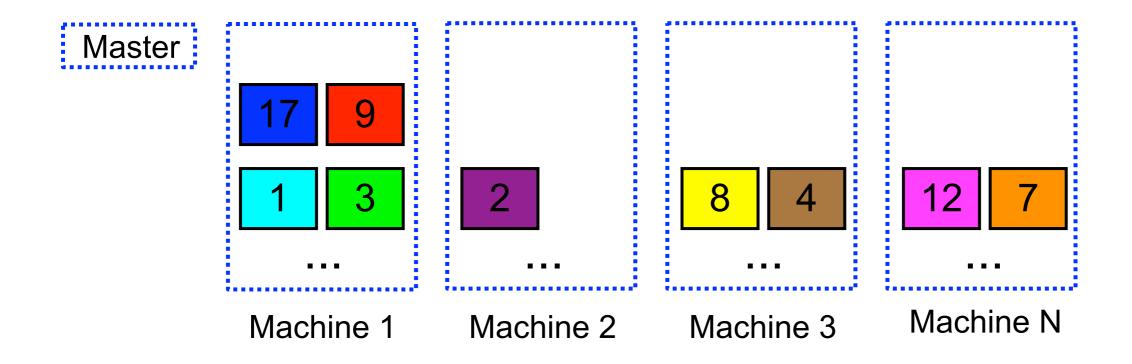
#### Within request adaptation

- –cope with slow subsystems in context of higher level request
- -time scale: right now, while user is waiting



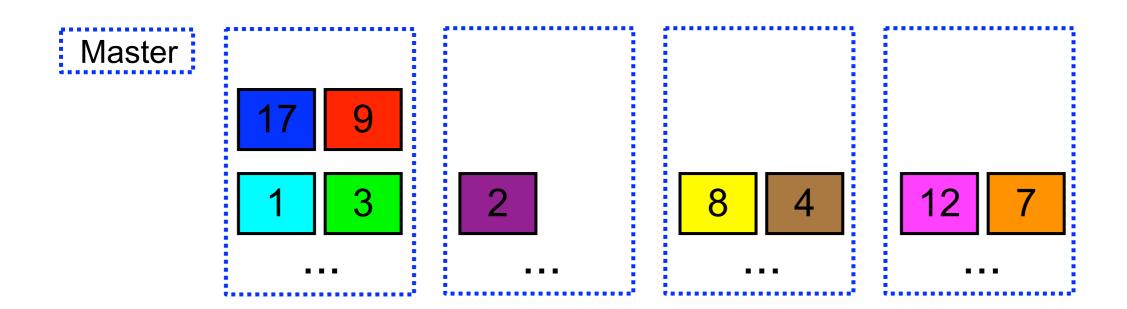
## Fine-Grained Dynamic Partitioning

- Partition large datasets/computations
  - -more than 1 partition per machine (often 10-100/machine)
  - -e.g. BigTable, query serving systems, GFS, ...



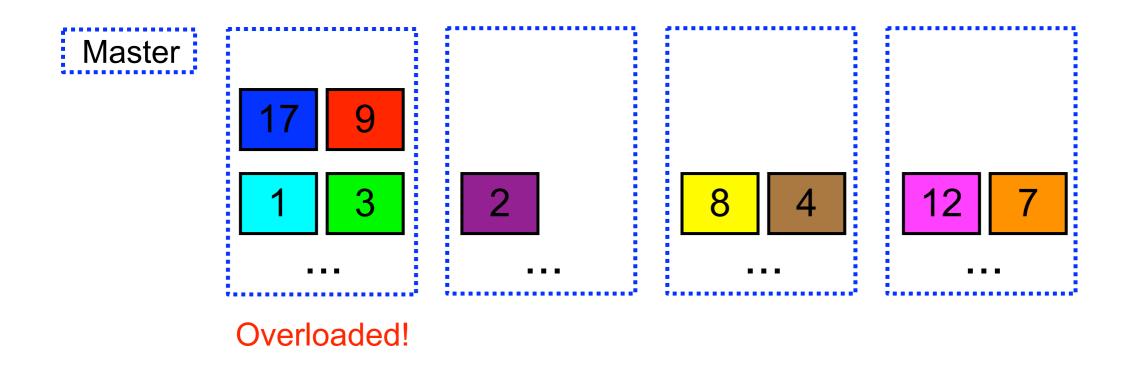


- Can shed load in few percent increments
  - -prioritize shifting load when imbalance is more severe



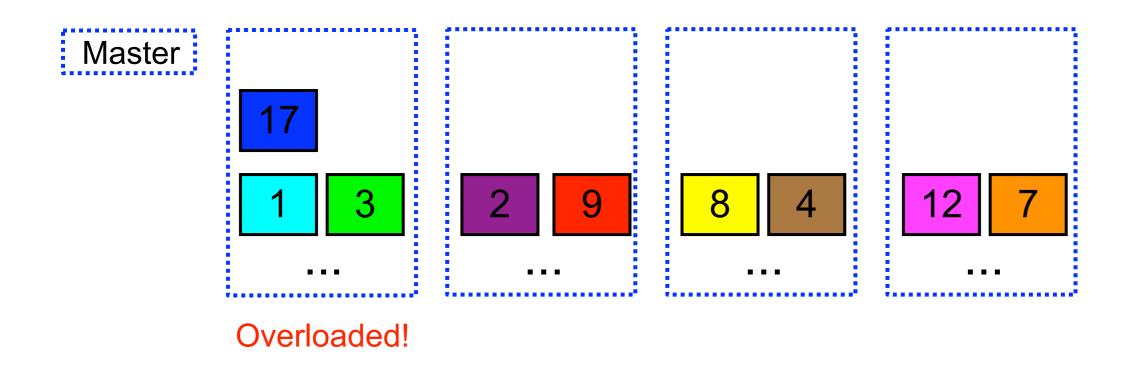


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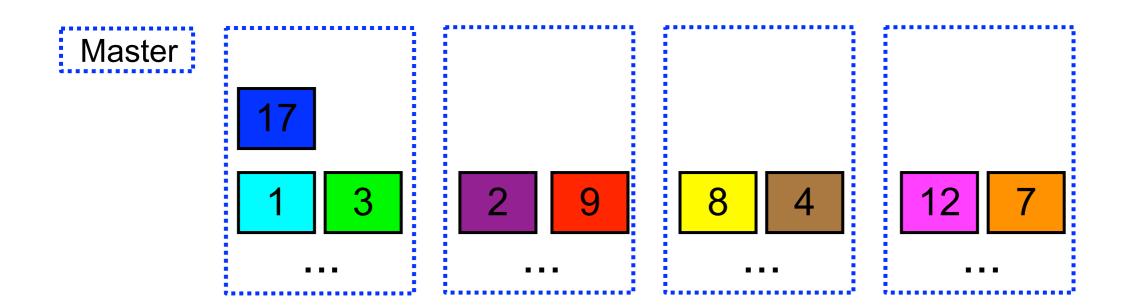


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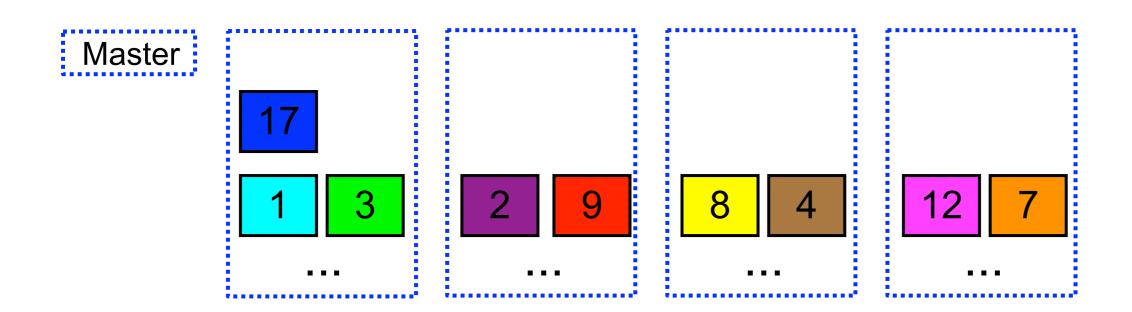
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## Speeds Failure Recovery

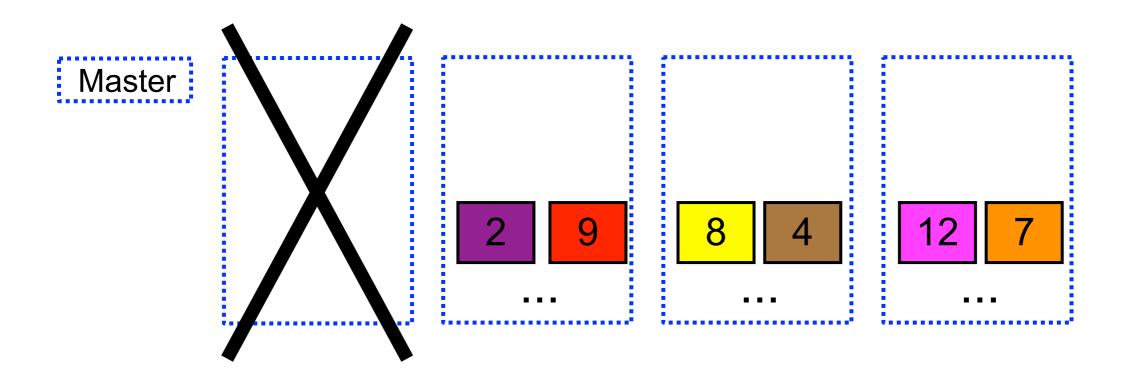
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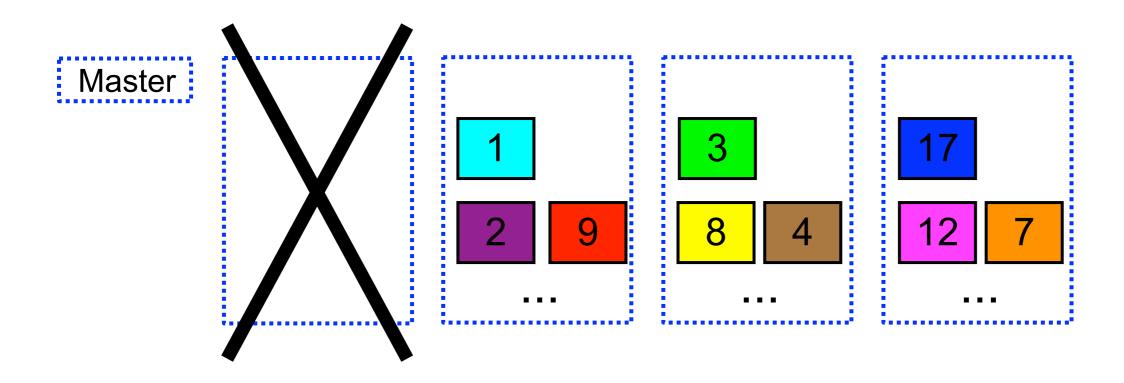
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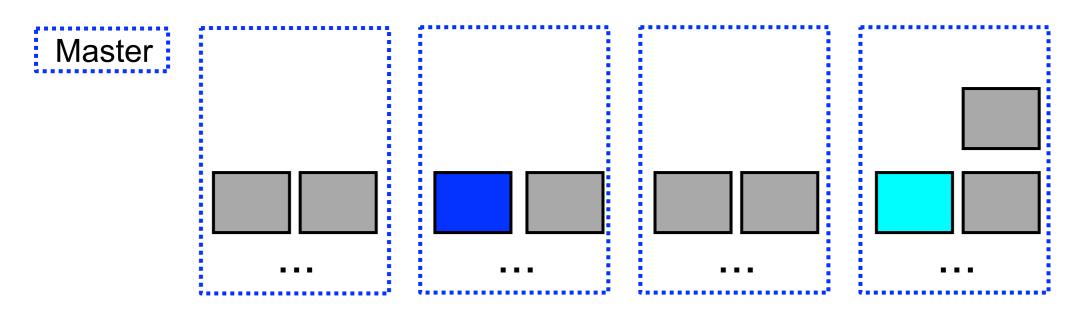
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## Selective Replication

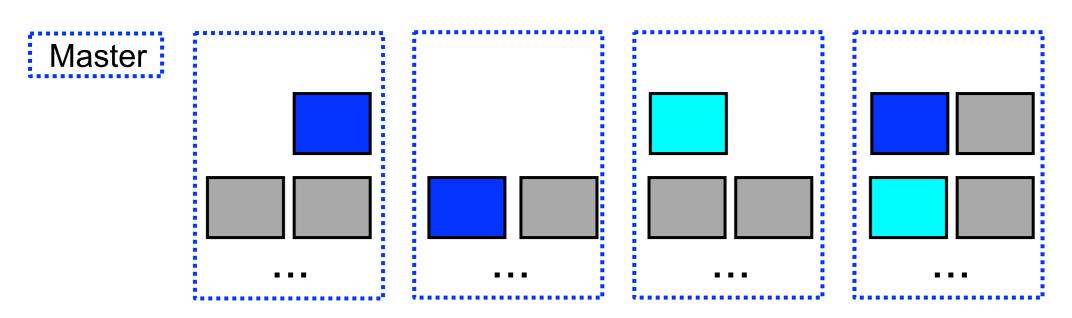
- Find heavily used items and make more replicas
  - –can be static or dynamic
- Example: Query serving system
  - -static: more replicas of important docs
  - dynamic: more replicas of Chinese documents as Chinese query load increases





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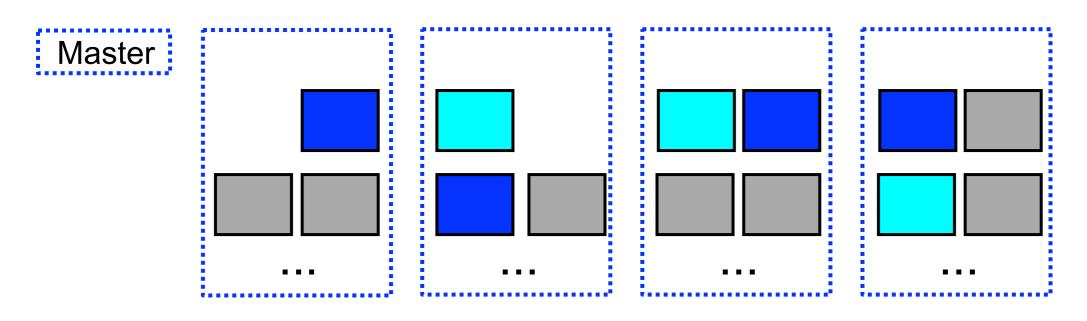
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## Latency-Induced Probation

- Servers sometimes become slow to respond
  - -could be data dependent, but...
  - -often due to interference effects
    - e.g. CPU or network spike for other jobs running on shared server
- Non-intuitive: remove capacity under load to improve latency (?!)
- Initiate corrective action
  - -e.g. make copies of partitions on other servers
  - -continue sending shadow stream of requests to server
    - keep measuring latency
    - return to service when latency back down for long enough



## Handling Within-Request Variability

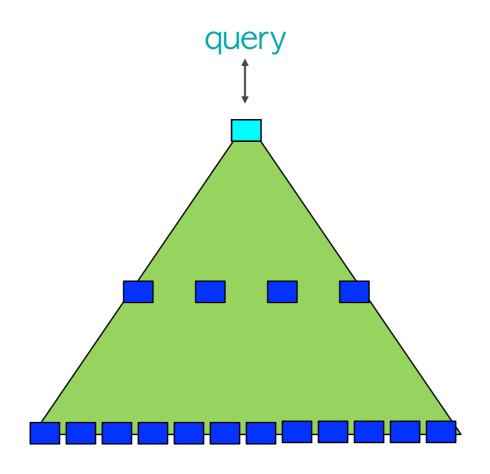
Take action within single high-level request

#### Goals:

- –reduce overall latency
- -don't increase resource use too much
- -keep serving systems safe

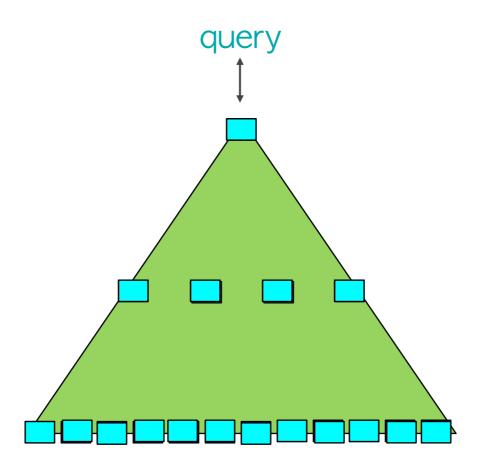


# Data Independent Failures



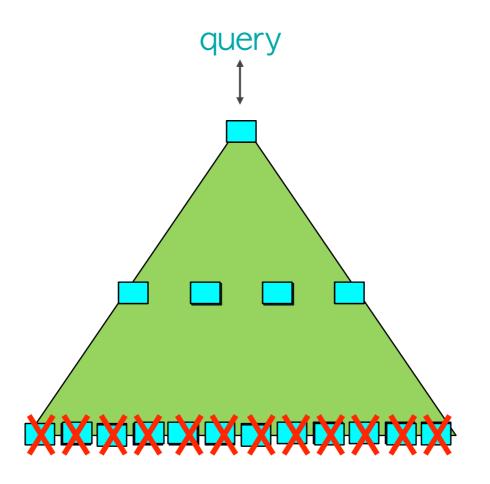


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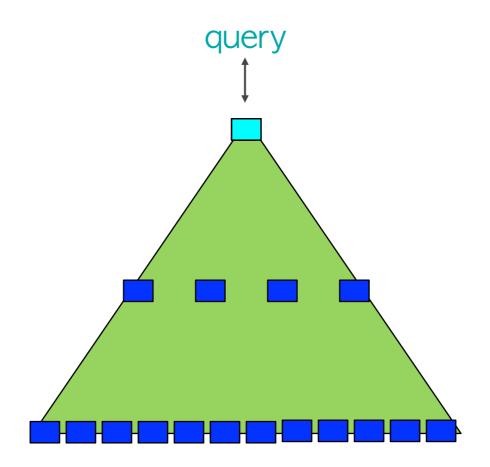


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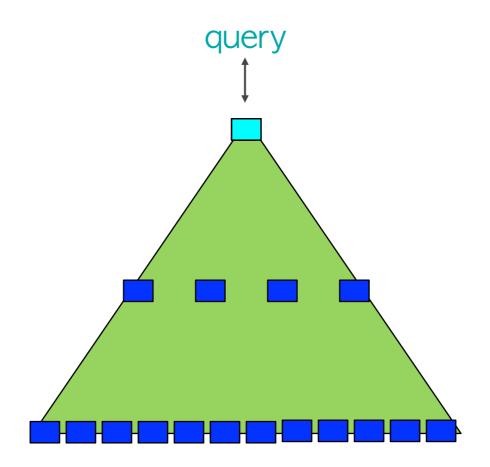


# Canary Requests (2)



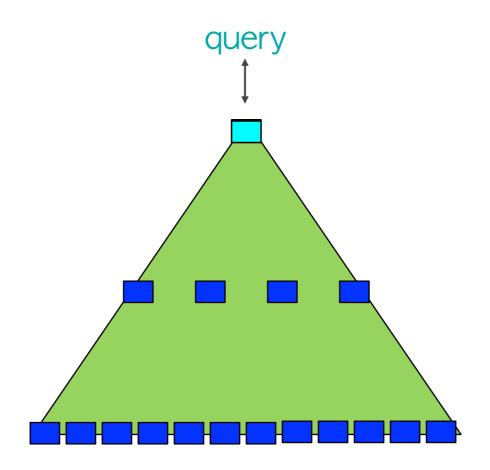


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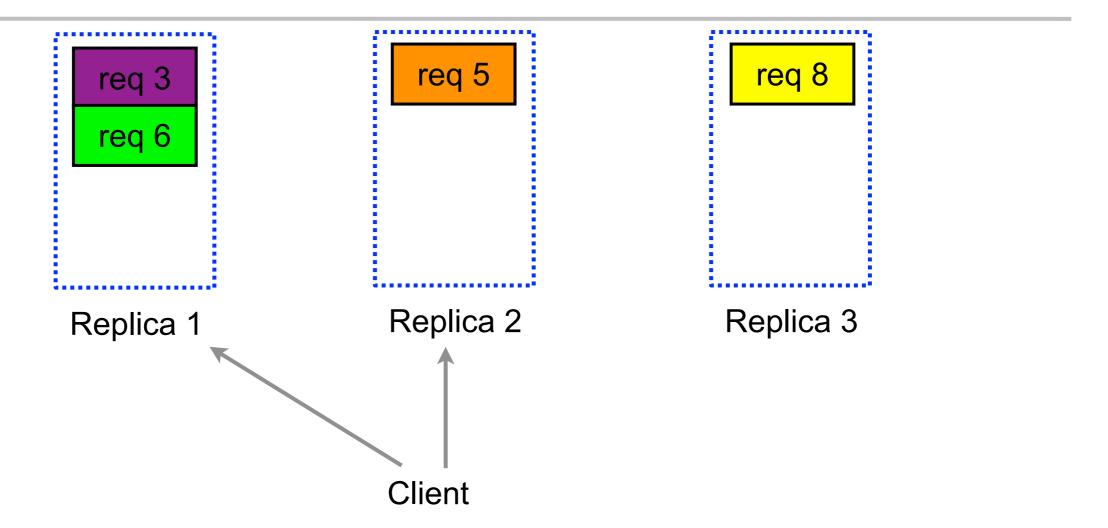




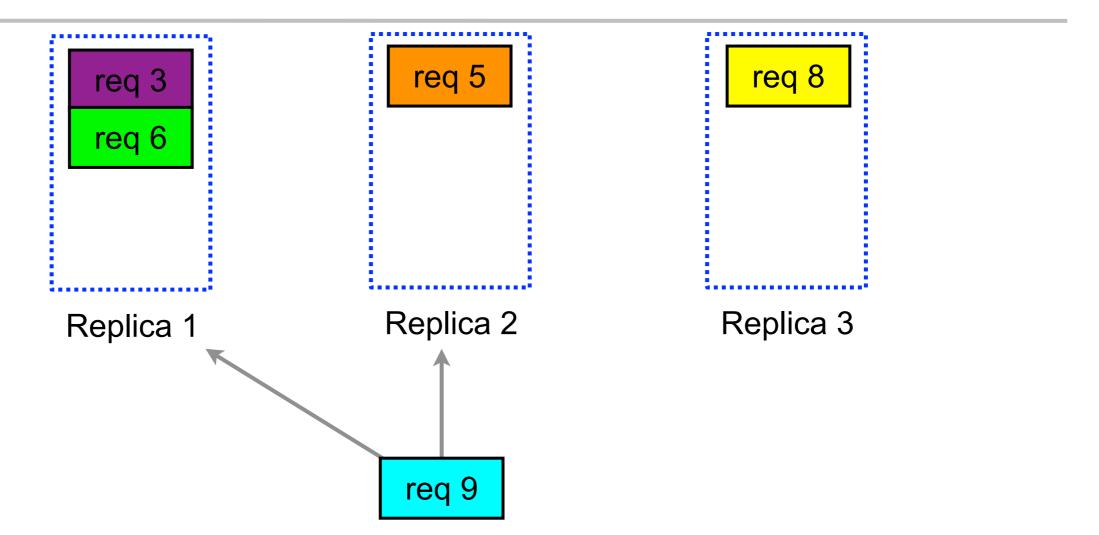
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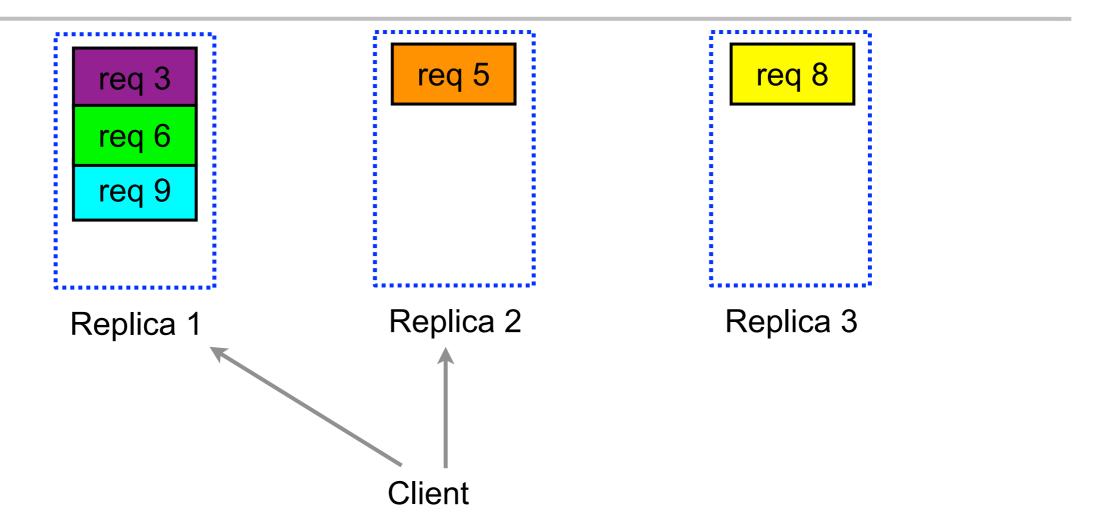




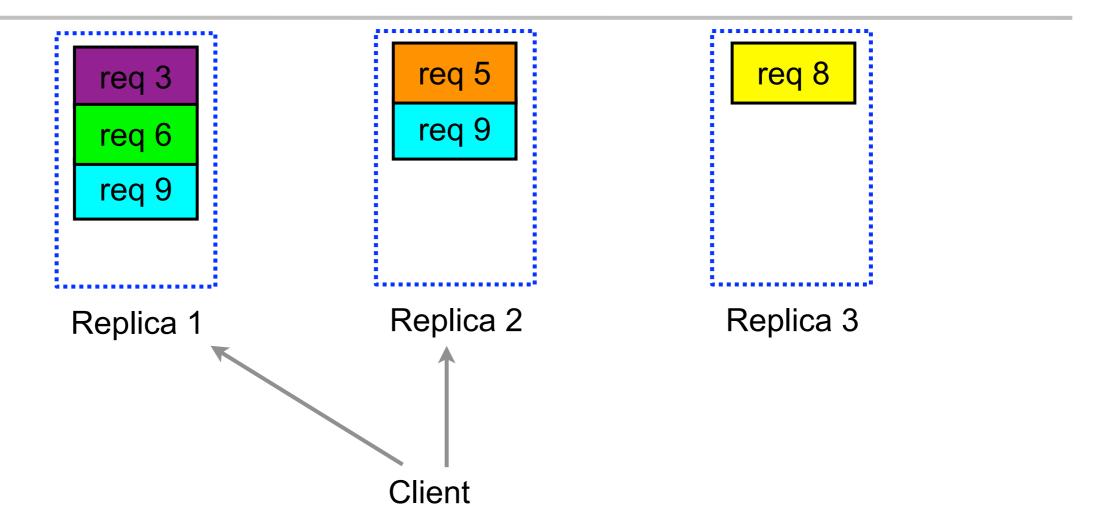




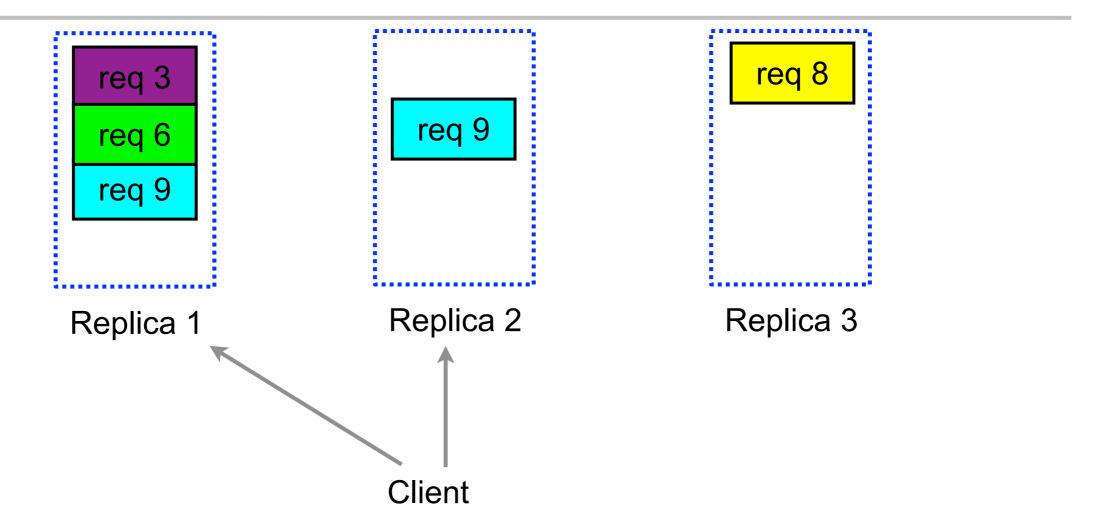




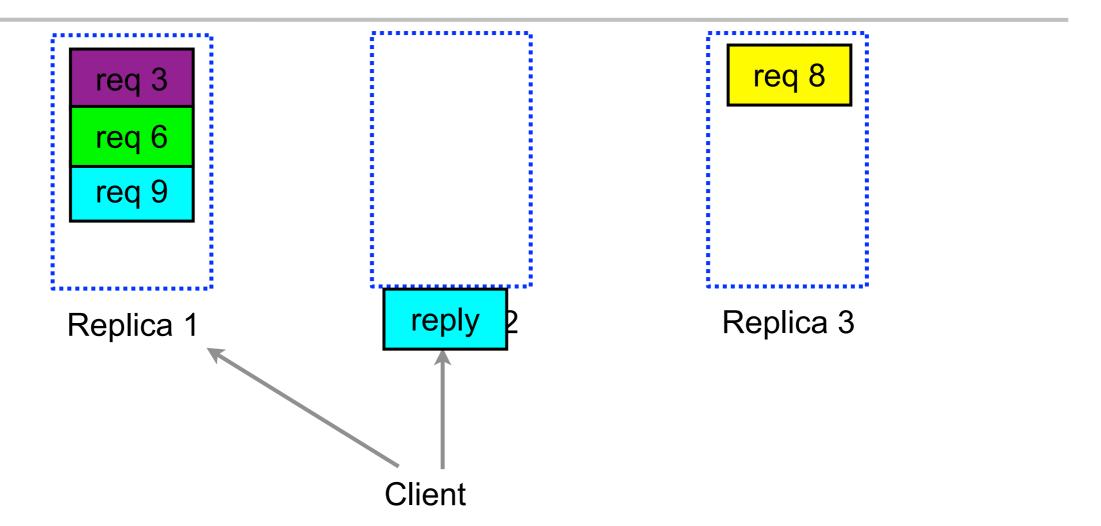




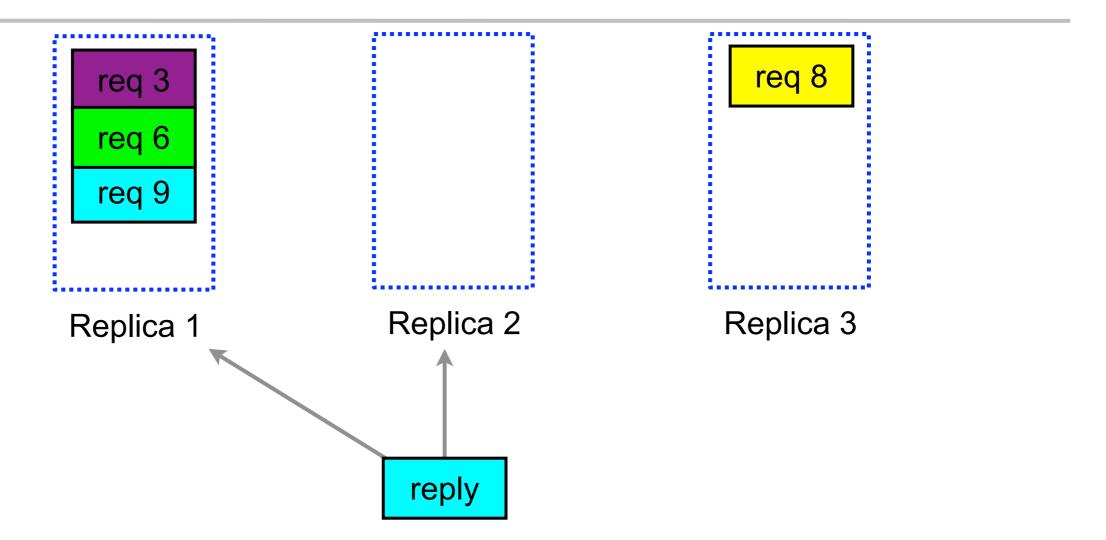




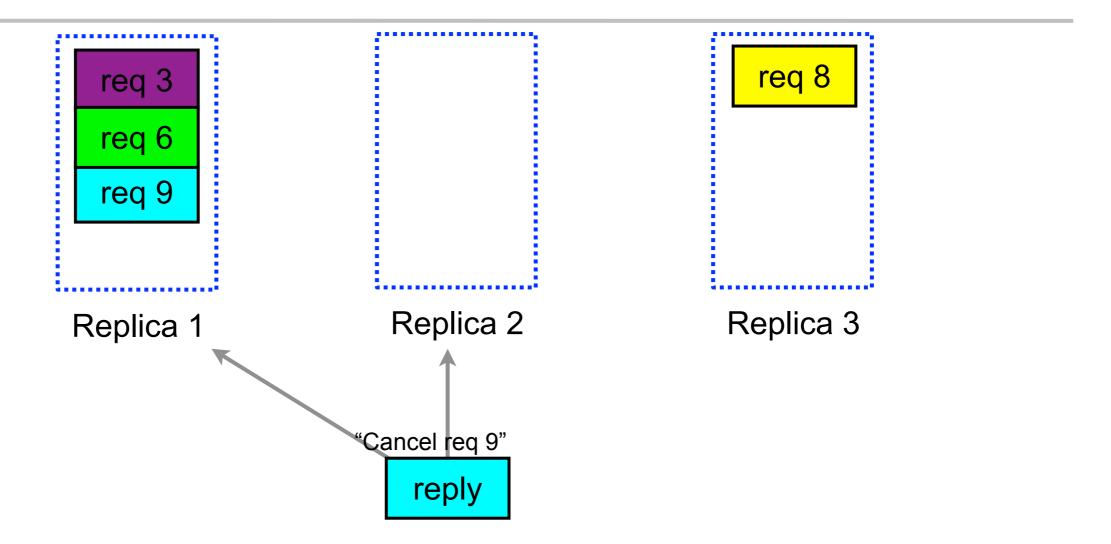




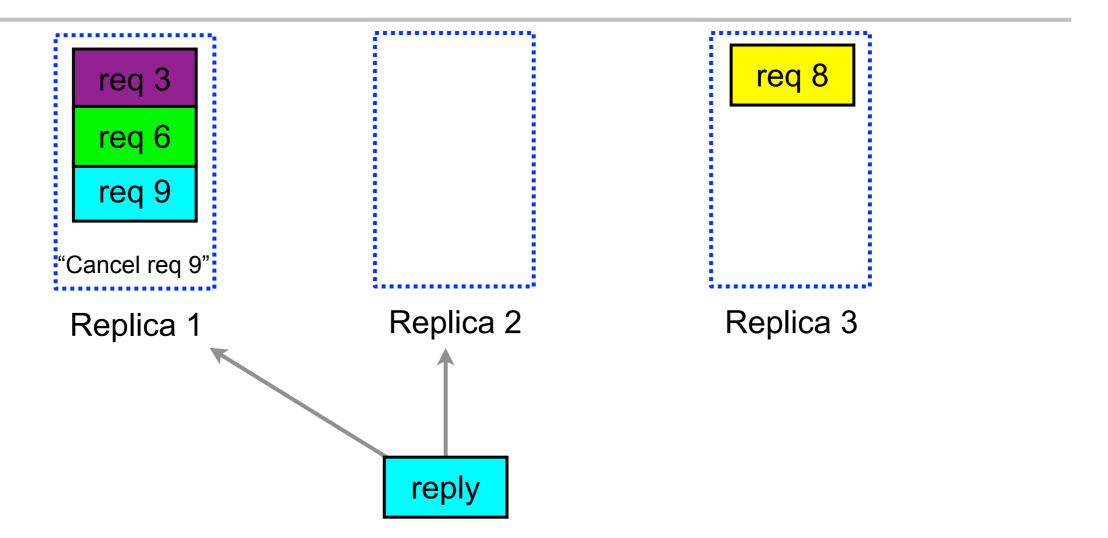




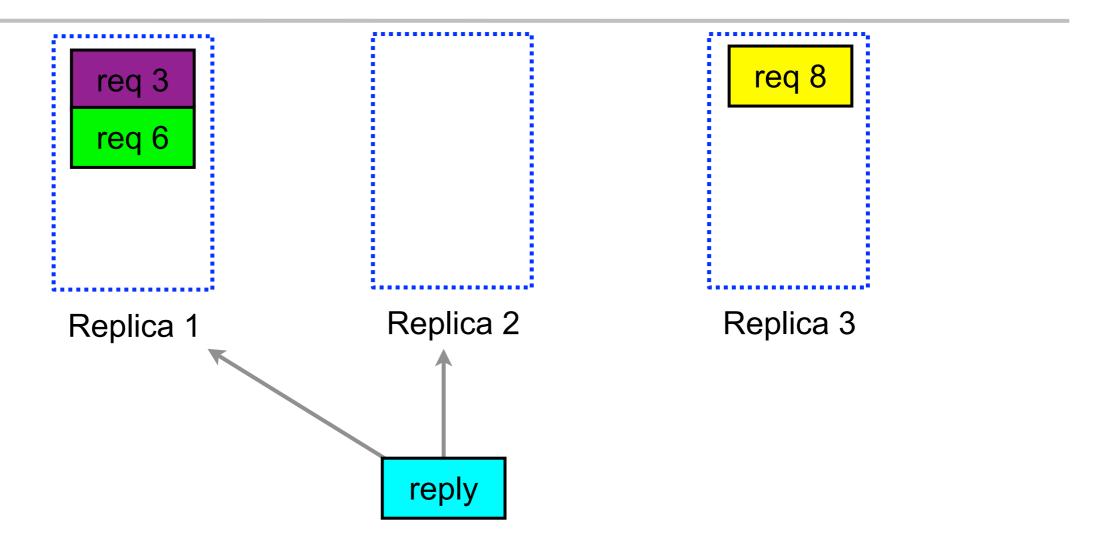














#### Backup Requests Effects

- In-memory BigTable lookups
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  - -measure elapsed time until data for last key arrives



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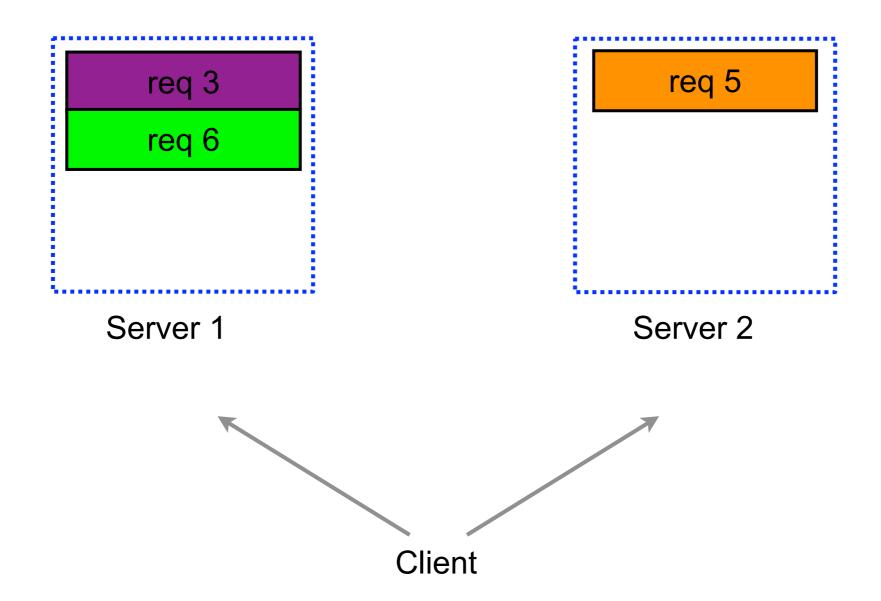
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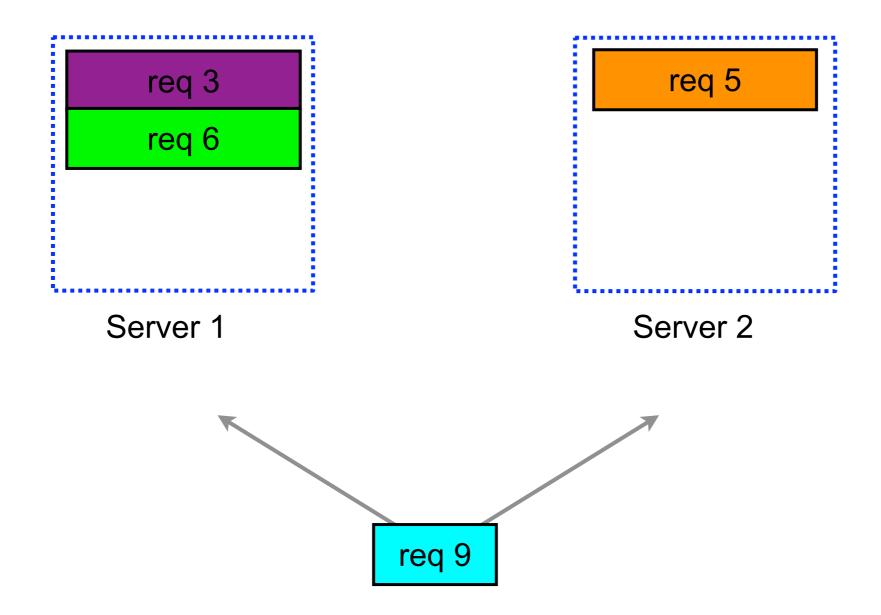
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- Modest increase in request load:
- 10 ms delay: <5% extra requests; 50 ms delay: <1%

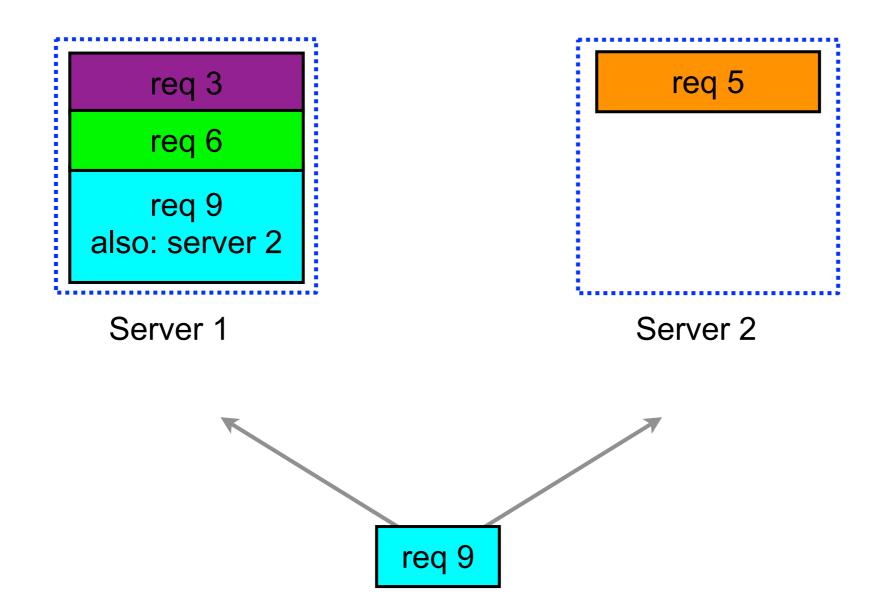




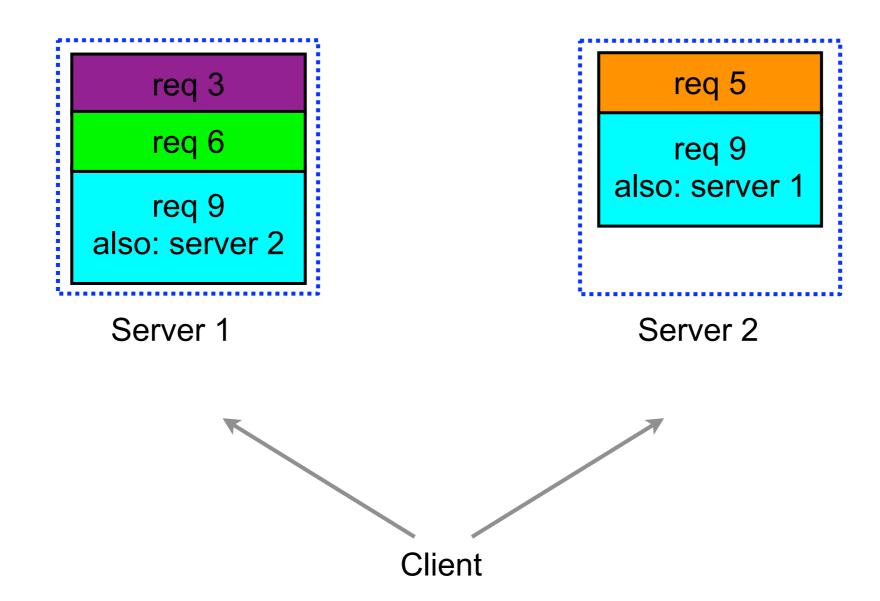




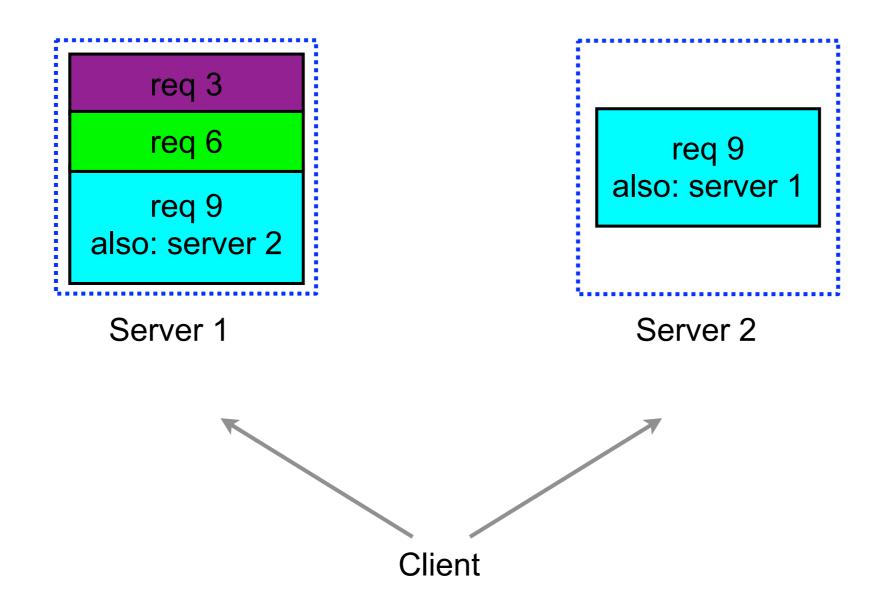




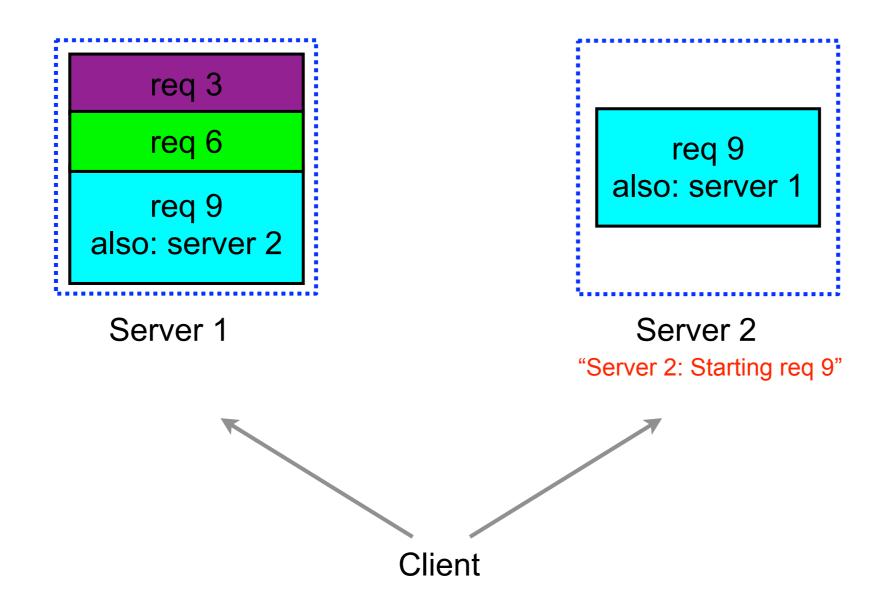




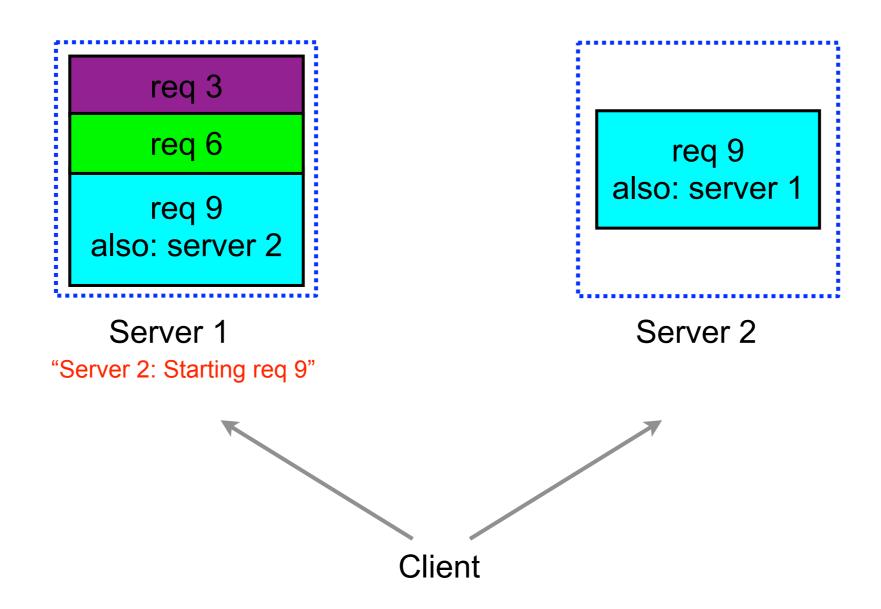




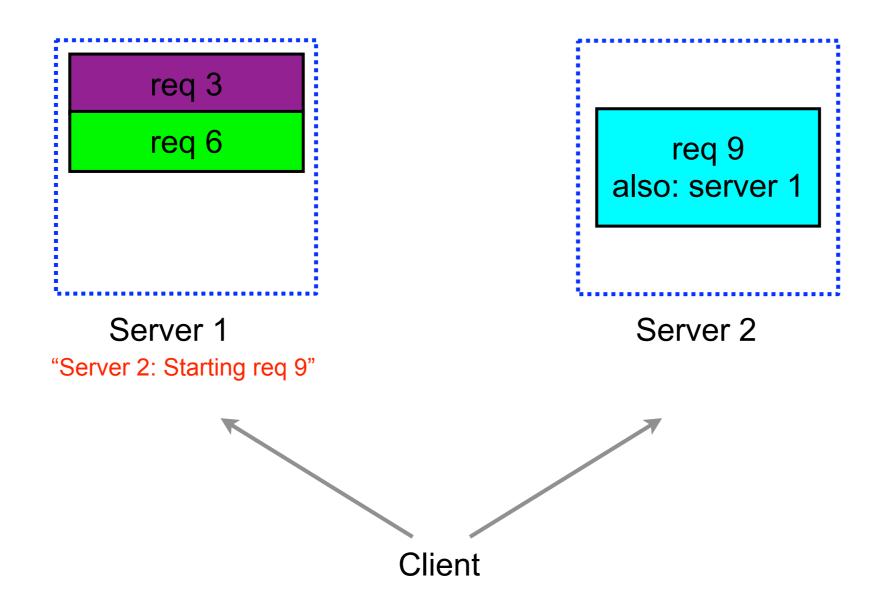




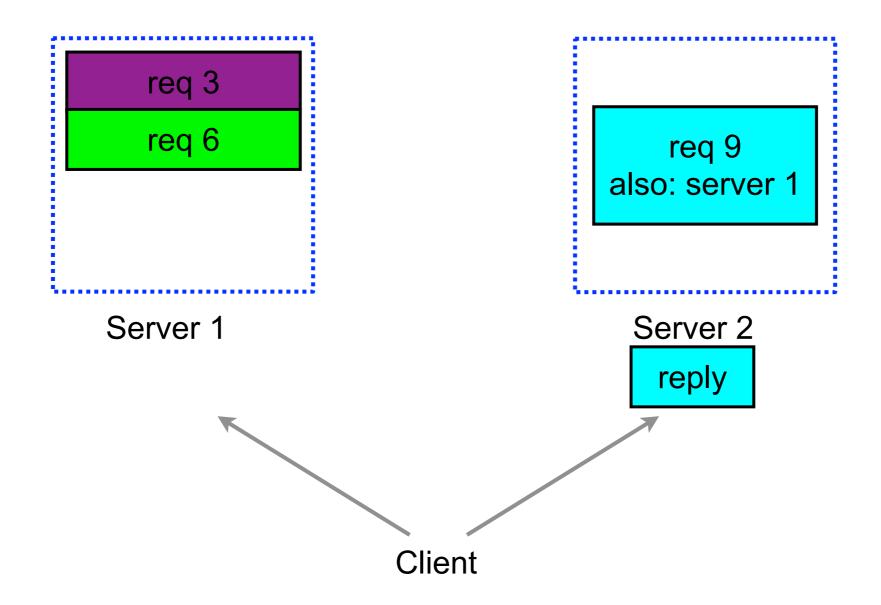




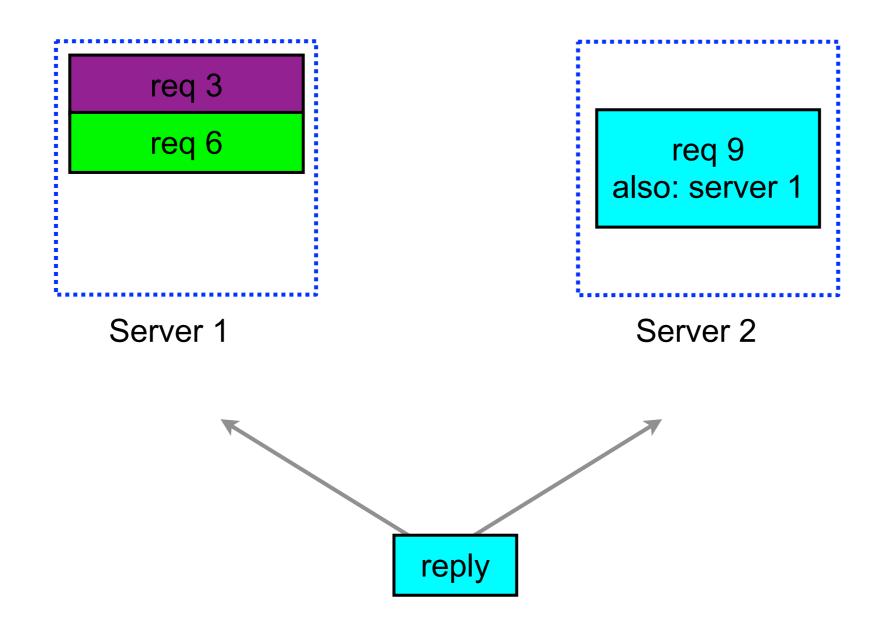




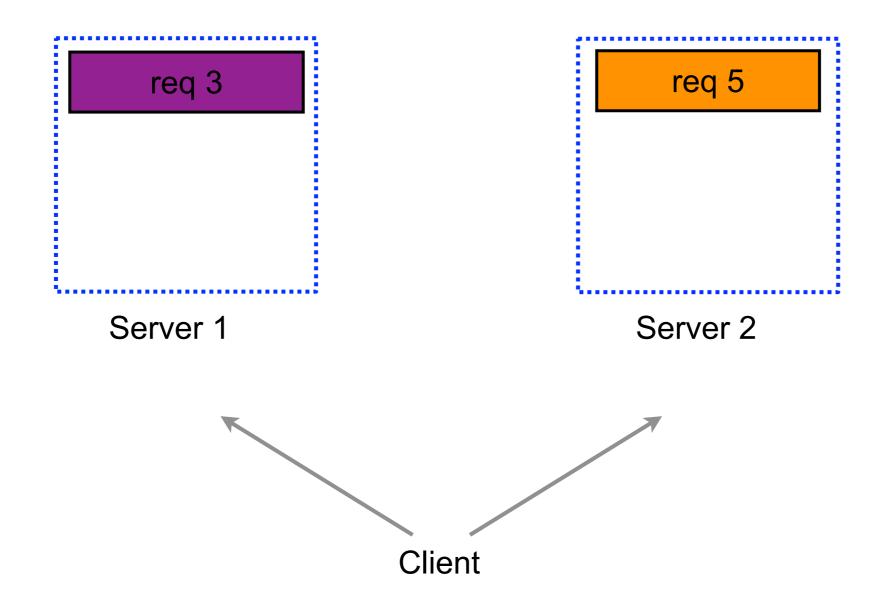




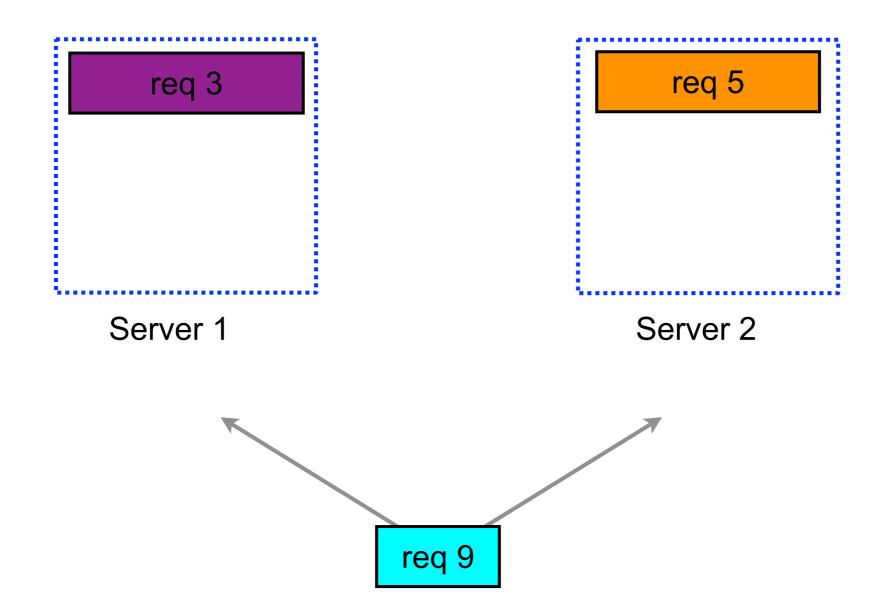




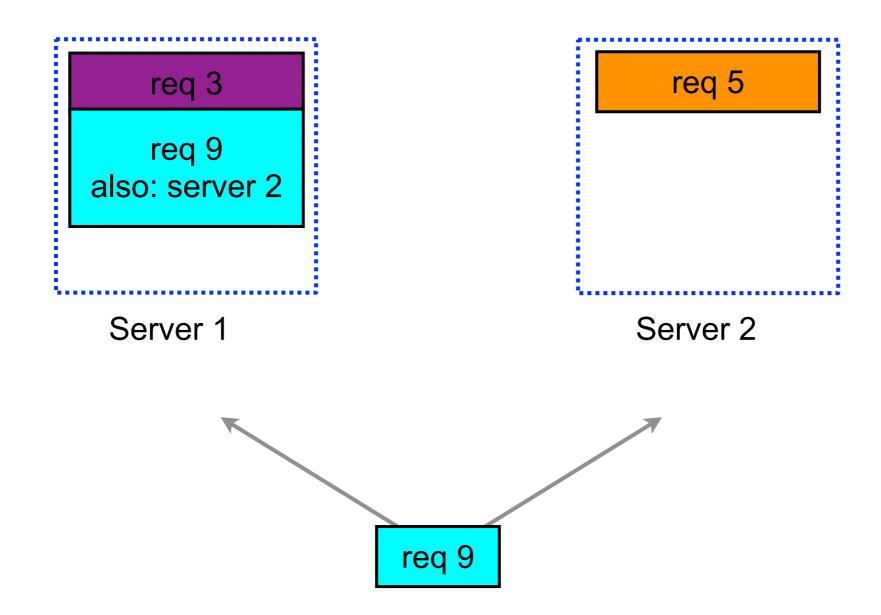




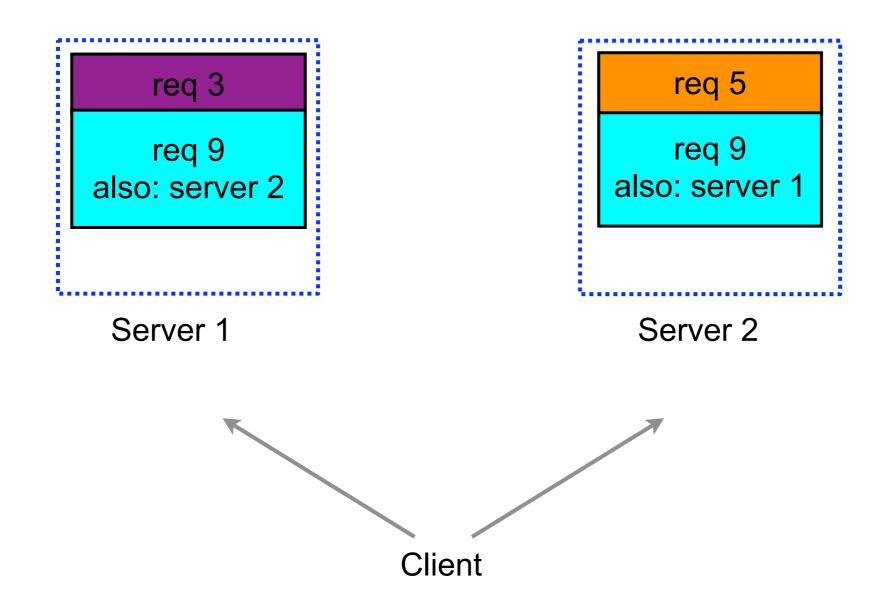




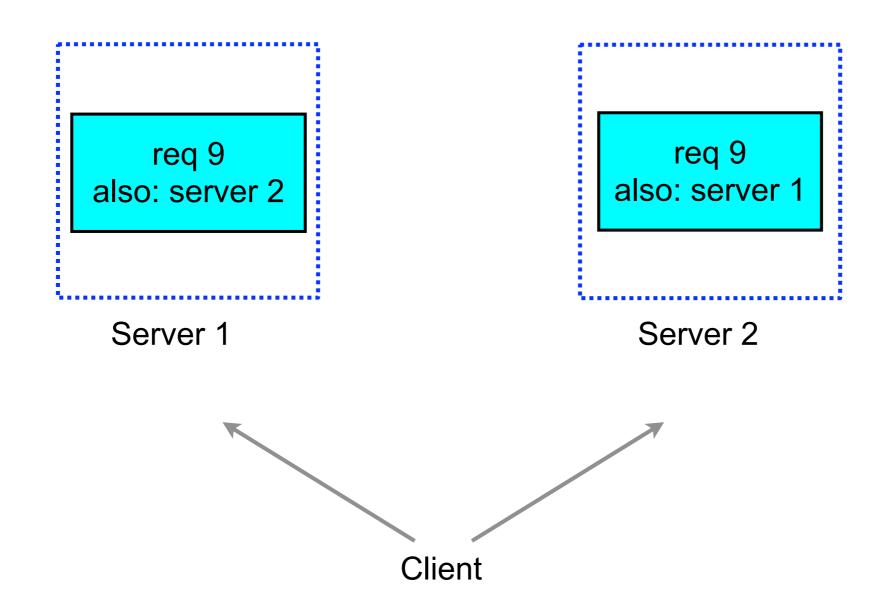




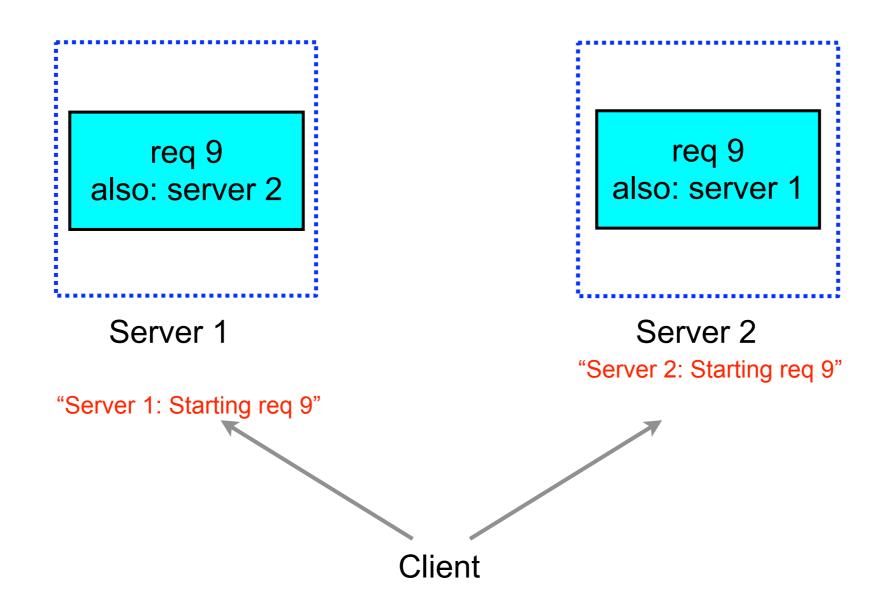




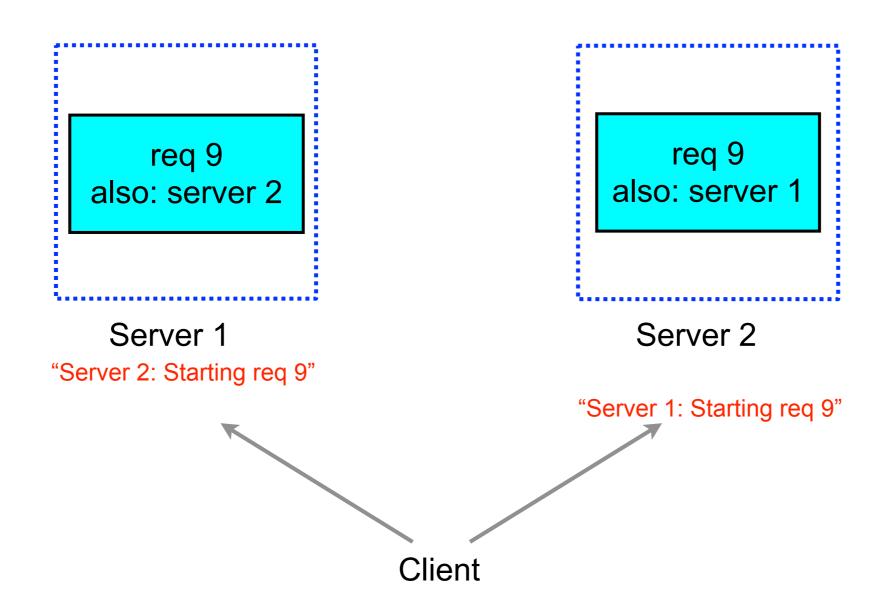




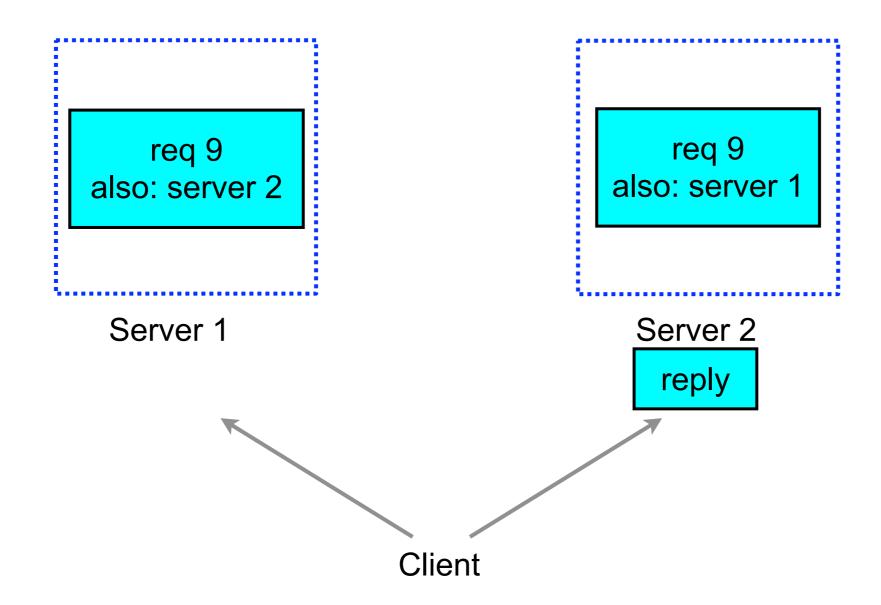




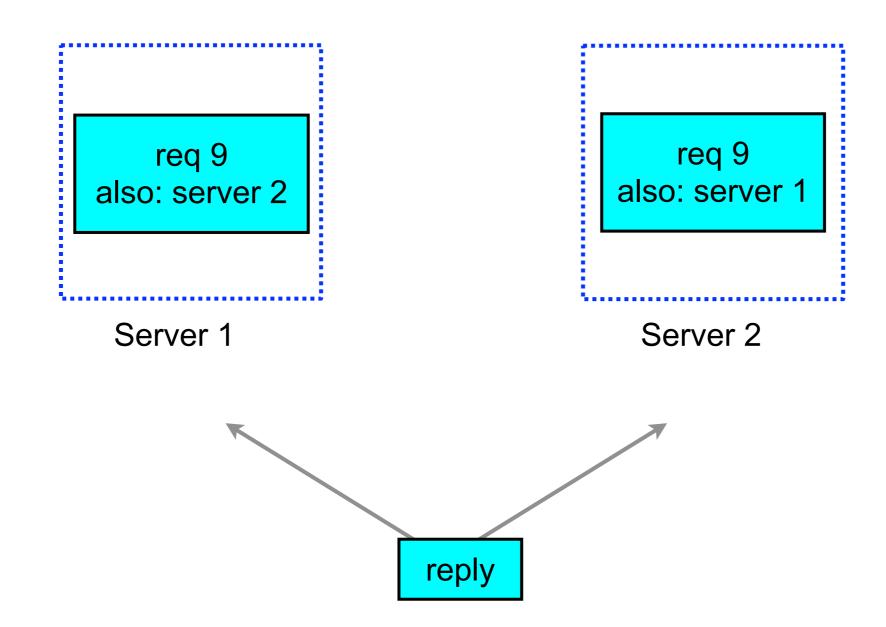














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  - -send request to first replica
  - -wait 2 ms, and send to second replica
  - -servers cancel request on other replica when starting read
- Time for bigtable monitoring ops that touch disk



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Backups cause about ~1% extra disk reads



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Backups w/big sort job gives same read latencies as no backups w/ idle cluster!



## Backup Request Variants

- Many variants possible:
- Send to third replica after longer delay
  - -sending to two gives almost all the benefit, however.
- Keep requests in other queues, but reduce priority
- Can handle Reed-Solomon reconstruction similarly



#### **Tainted Partial Results**

- Many systems can tolerate inexact results
  - -information retrieval systems
    - search 99.9% of docs in 200ms better than 100% in 1000ms
  - complex web pages with many sub-components
    - e.g. okay to skip spelling correction service if it is slow
- Design to proactively abandon slow subsystems
  - -set cutoffs dynamically based on recent measurements
    - can tradeoff completeness vs. responsiveness
  - -important to mark such results as tainted in caches



#### Hardware Trends

- Some good:
  - lower latency networks make things like backup request cancellations work better
- Some not so good:
  - plethora of CPU and device sleep modes save power, but add latency variability
  - –higher number of "wimpy" cores => higher fanout => more variability

#### Conclusions

#### Tolerating variability

- important for large-scale online services
- large fanout magnifies importance
- makes services more responsive
- saves significant computing resources

#### Collection of techniques

- -general good engineering practices
  - prioritized server queues, careful management of background activities
- -cross-request adaptation
  - load balancing, micro-partitioning
- -within-request adaptation
  - backup requests, backup requests w/ cancellation, tainted results

#### Thanks

Joint work with Luiz Barroso and many others at Google

• Questions?

