Low latency access of replicated fragments on memory constrained servers

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Evolving Digital Landscape



Rate Requirements

Global application traffic share 2021¹



Centralized Paradigm



Potential Issues

- Not scalable with traffic load
- Susceptible to hardware failures and attacks

Distributed Paradigm



Potential Issues

Susceptible to hardware failures and attacks

Large-scale distributed systems



Resilience though redundancy



Latency redundancy tradeoff

- Download speedup due to parallel access
- Increased load due to redundant access

Load balancing through file fragmentation



Shared coherent access

- Availability and better content distribution
- File segments on multiple servers

Memory constrained system



What are latency reducing storage and access schemes for replicated fragments?

- parallel access from all B servers
- α -fragment of message stored at each server

Coded Storage for single file



Single file divided into V fragments

- encoded into VR fragments
- each coded fragment stored over B = VR servers
- reconstruction by set of V coded symbols

Prior Work

MDS codes

Outperform replication codes in file access delay

▶ Huang et al(2012), Li et al(2016), Badita et al(2019)

Rateless codes

Offers near optimal performance

Mallick et al(2019)

Staircase codes

Subfragmentation improves latency performance

Bitar et al(2020)

Our model

Replication codes for a file with equal sized fragmentation over multiple servers where each can store multiple file fragments

Latency optimal storage and access



A unit size divisible message $m = (m_1, \ldots, m_V)$

• replicated $R = \alpha B/V$ times

storage: for each fragment, where to store each replica?

access: for each server, sequence of access for replicas?

File download time



- ▶ Number of useful servers after ℓ th download, N_{ℓ}
- Fragment download times are *i.i.d.* exponential with unit rate
- Rate of download at ℓ th stage is N_{ℓ}
- The mean download time is $\mathbb{E} \sum_{\ell=0}^{V-1} \frac{1}{N_{\ell}}$

Optimality criterion



Number of downloads, ℓ

Optimality condition for storage scheme Maximize the number of useful servers sequence

(VR, V) MDS code on α -B system



Optimality of MDS code

Reduction in useful servers is the least

Decoding complexity



Implementation challenges

- Requires sufficiently large alphabet or large fragment sizes
- Polynomial decoding complexity that can't be parallelized

Scaling issues of MDS coding



Encoding growing data or redundancy

- Complete re-encoding of data blocks
- Potential data loss waiting for sufficient data blocks

Replication coded storage

 α -(V, R) replication coded storage over B servers

$$S \triangleq \{(S_1, S_2, \dots, S_B) : |S_b| = \alpha V \text{ for all } b, \alpha = R/B\}.$$

 $\frac{3}{7} - (7,3)$ replicated storage



Fragment sets $S_1 = \{1, 2, 3\}, S_2 = \{2, 3, 4\}, \dots$

• Occupancy sets $\Phi_1 = \{1, 6, 7\}, \Phi_2 = \{1, 2, 7\}, \dots$

Upper bound on number of useful servers N_ℓ



Upper bound

- ▶ For $m \triangleq \lceil B/R \rceil$, we have $N_{\ell} \leqslant B\mathbb{1}_{\{\ell \leqslant V-m\}} + (V \ell)R\mathbb{1}_{\{\ell > V-m\}}$
- Normalized average of number of useful servers is upper bounded as

$$\frac{1}{BV}\sum_{\ell=0}^{V-1}N_{\ell}\leqslant 1-\frac{(m+1)}{2V}$$

Trivial case: $\alpha \ge 1$



Replication as good as MDS without memory constraint

- Each server can store all the fragments
- All servers remain useful throughout
- What if $\alpha < 1$?

Randomized (B, V, R) replication coded storage



Place the fragments on randomly chosen servers

- Each server can store all coded VR fragments
- \blacktriangleright Exponential download rate \propto the number of stored fragments

Asymptotically an α -(V, R) storage

- As V is increased with R/B fixed
- ▶ normalized storage at any server converges to $\alpha = R/B$
- service rate of servers converge to unity for almost all downloads



Asymptotic optimality

The randomized (B, V, R) storage scheme is an α -(V, R) storage scheme asymptotically in V.

Performance of Random Replication Storage

i.i.d. random storage vector Θ where $P\{\Theta_{vr} \neq b\} = (1 - 1/B)$

$$N_{\ell} = B - \sum_{b \in [B]} \prod_{v \notin I_{\ell}} \prod_{r \in [R]} \mathbb{1}_{\{\Theta_{vr} \neq b\}}.$$

$$\frac{1}{BV} \mathbb{E} N_{\ell} = \frac{1}{V} \left(1 - \left(1 - \frac{1}{B}\right)^{R(V-\ell)} \right)$$

Mean number of useful servers

For the random (B, V, R) replication storage ensemble,

$$\frac{1}{BV}\sum_{\ell=0}^{V-1}\mathbb{E}N_{\ell} = 1 - \frac{\left(1 - \frac{1}{B}\right)\left(1 - \left(1 - \frac{1}{B}\right)^{RV}\right)}{V\left(1 - \left(1 - \frac{1}{B}\right)^{R}\right)}$$

Numerical Results



Bounding the number of useful servers



Maximum overlaps

- ▶ Between fragment sets $\tau_M \triangleq \max|S_a \cap S_b|$
- ► Between occupancy sets $\lambda_M \triangleq \max |\Phi_v \cap \Phi_w|$

Bounding the number of useful servers



Universal bounds

► For
$$i \in \{0, ..., \lfloor \frac{\kappa}{\tau_M} \rfloor\}$$
 and $\ell_i \triangleq iK - i(i-1)\frac{\tau_M}{2}$

$$N_{\ell} \geqslant egin{cases} B-i, & \ell_i \leqslant \ell < \ell_{i+1}, \ (V-\ell)(R-(V-\ell-1)rac{\lambda_M}{2}), & \ell \geqslant V - \lfloor rac{R}{\lambda_M}
floor - 1 \end{cases}$$

Optimal Storage



Tightest lower bounds

- ▶ The lower bounds are maximized for $\lambda_M = \tau_M = 1$
- Less overlaps are better

How to find the good storage schemes?

Table: Correspondence between designs and storage codes

t -(V, K, λ) designs to codes	
Design parameter	Storage parameter
\mathcal{P} : Points	[V]: File fragments
B: Blocks	$(S_b : b \in [B])$: Fragment sets at servers
$ \mathcal{P} $: Number of points	V: Number of file fragments
$ \mathcal{B} $: Number of blocks	B: Number of servers
K: Size of each block	K: Storage capacity at each server
R: Replication factor for each point	R: Replication factor for each fragment

Optimal Access



- The set of useful servers evolve as a Markov chain
- Given a storage scheme, optimal access is a Markov decision process that maximizes $\mathbb{E}\left[\sum_{\ell=0}^{V-1} N_{\ell}\right]$

Greedy Scheduler



- ▶ Greedy scheduler rank $\rho_{\ell}^{g}(v) \triangleq \sum_{b \in \Phi_{v}} \mathbb{1}_{\{|S_{b}^{\ell}|=1\}}$ for every fragment
- $\blacktriangleright \mathbb{E}\left[N_{\ell+1} N_{\ell} \mid I_{\ell}\right] = \sum_{v \notin I_{\ell}} p_{I_{\ell}, I_{\ell} \cup \{v\}} \rho_{\ell}^{g}(v)$

Ranked Scheduler



- ► For each useful server schedule the fragment with highest rank $\rho_{\ell}: I_{\ell}^{c} \to \mathbb{R}$
- Harmonic rank $\rho_{\ell}^{h}(v) \triangleq \sum_{b \in \Phi_{v}} \frac{1}{|S_{\ell}^{\ell}|}$ for every fragment

Numerical Studies



Conclusion

- We studied codes for distributed storage system with storage constraints and file subfragmentation for achieving low latency
- For exponential download times, we proposed to maximize mean number of useful servers instead of minimizing latency
- We show that MDS codes are optimal
- When there are no memory constraints at the server, replication coded file can be optimally placed
- When servers have memory constraints, we show that replication coding combined with probabilistic placement are optimal asymptotically
- Placement of coded fragments depends on overlap properties of storage codes
- Optimal access sequence is a Markov decision process

Acknowledgements



References

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