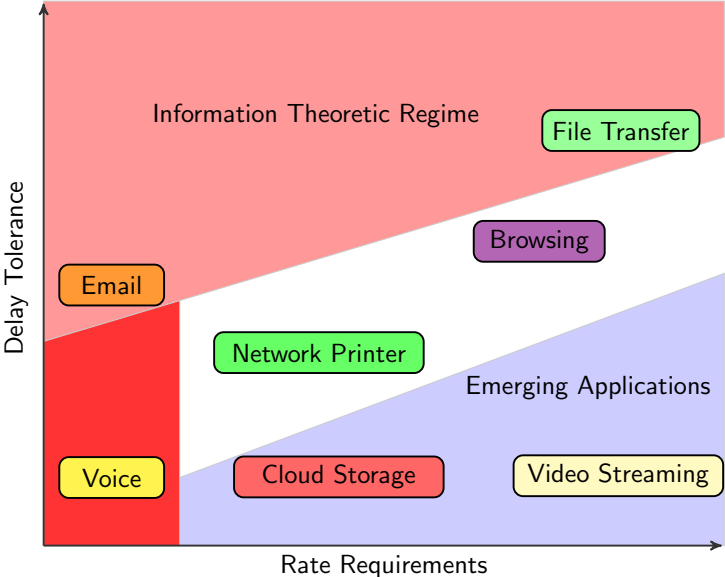# Low latency replication over memory constrained servers

Parimal Parag
Rooji Jinan
Ajay Badita
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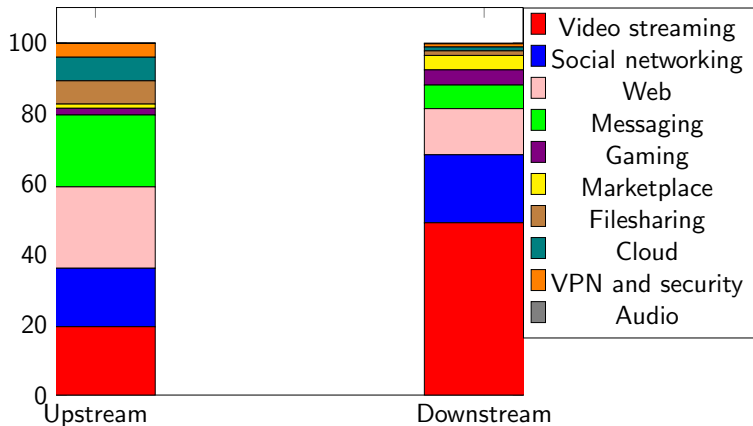
Sep 01, 2021

Evolving Digital Landscape
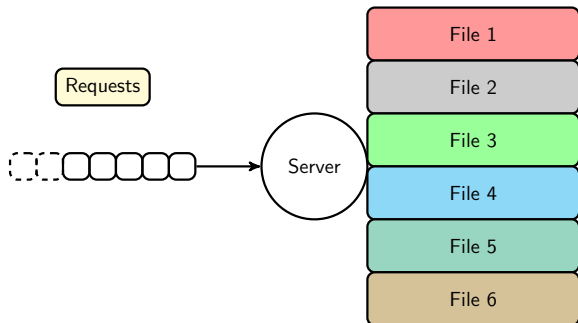
Global application traffic share 2021 [1]

[1] https://www.sandvine.com/hubfs/Sandvine_Redesign_2019/Downloads/2021/Phenomena/MIPR%20Q1%202021%20

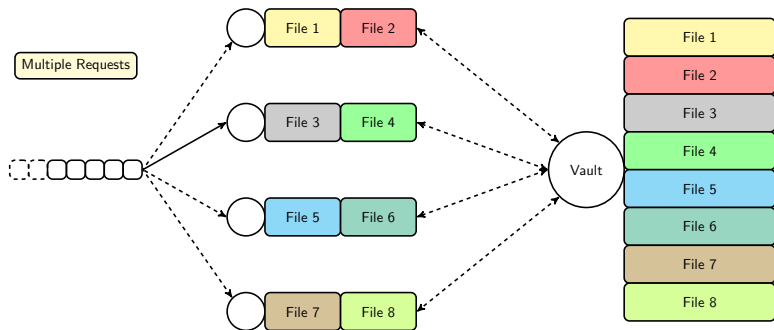# Centralized Paradigm

Potential Issues

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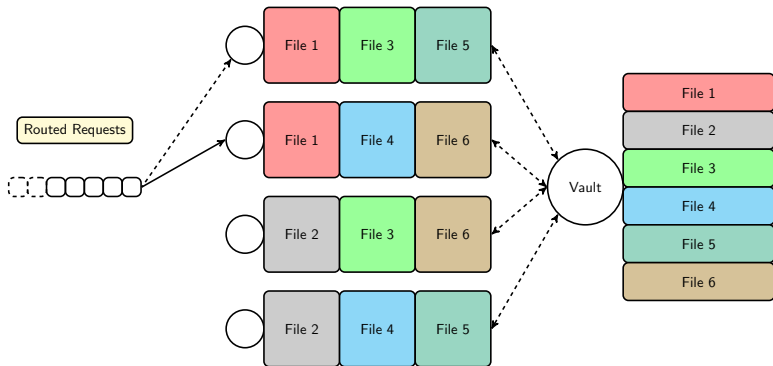- ▶ Susceptible to hardware failures and attacks

Distributed Paradigm

Potential Issues

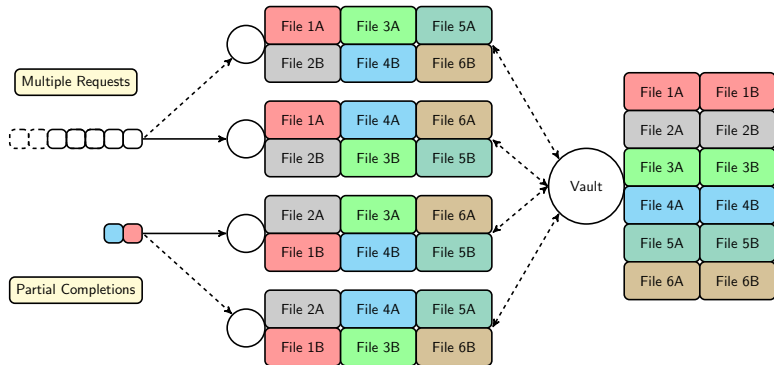▶ Susceptible to hardware failures and attacks

Resilience though redundancy

Latency redundancy tradeoff

▶ Download speedup due to parallel access

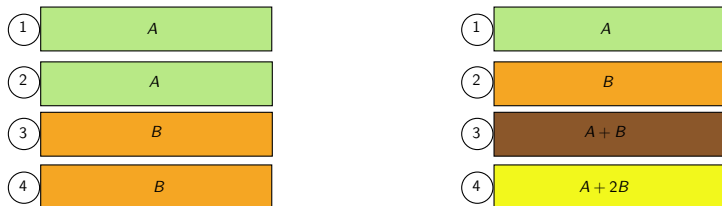▶ Increased load due to redundant access

Load balancing through file fragmentation

Shared coherent access
▶ Availability and better content distribution
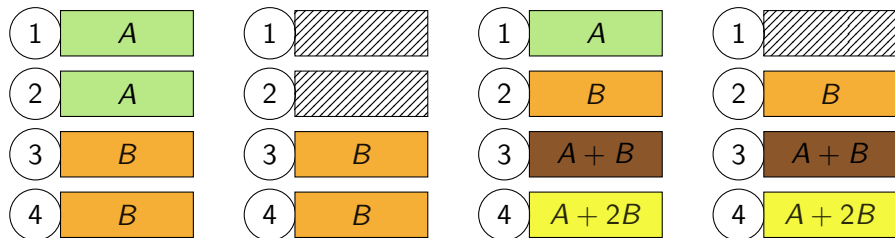▶ File segments on multiple servers

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

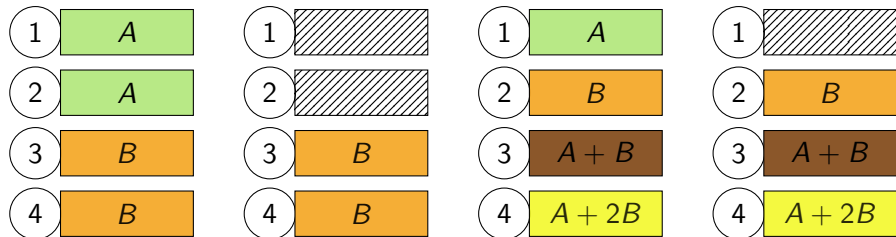# File download time

Mean file download time

- fragment downloads are *i.i.d.* and memoryless with unit rate
- parallel access from N_ℓ useful servers after ℓ downloads
- Harmonic sum of number of useful servers $\sum_{\ell=0}^{V-1} \frac{1}{N_\ell}$

File download time

Number of useful servers after ℓ downloads

- **replication:** $B - R\ell$
- **MDS coding:** $B - \ell$

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
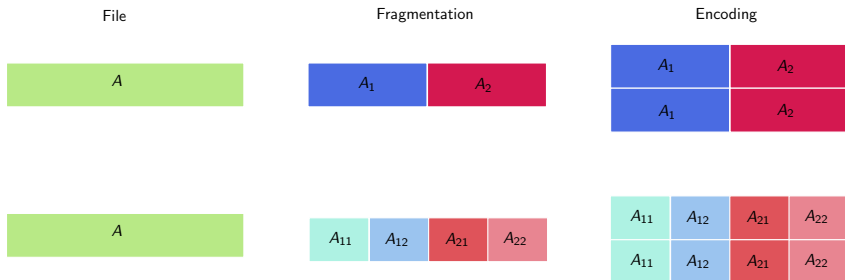Subfragmentation improves latency performance

▶ 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
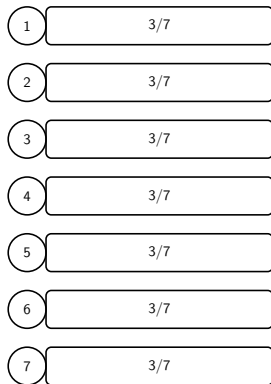
Storage model
fragmentation & encoding

File	Fragmentation	Encoding

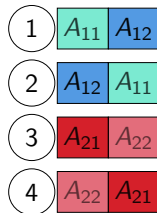▶ File divided into V fragments & encoded into VR fragments
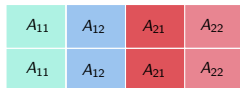
Memory constrained system

Storing αB size coded messages for a unit size message

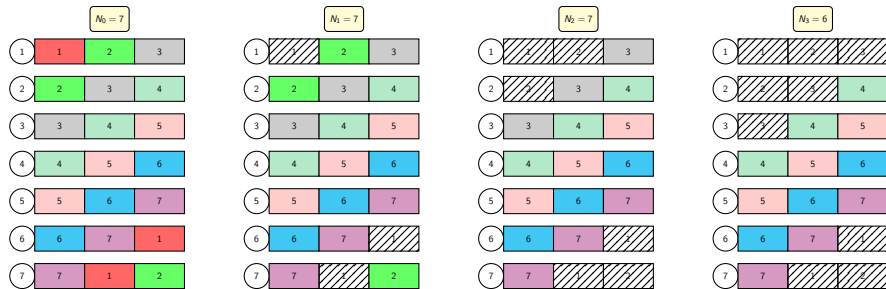- parallel access from all B servers
- α-fragment of message stored at each server

Storage model

Placement

File download time

- Number of useful servers after ℓth download, N_ℓ
- Fragment download times are *i.i.d.* exponential with unit rate
- Rate of download at ℓth 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

Normalized mean download time

$$\frac{1}{V}\mathbb{E}\sum_{\ell=0}^{V-1}\frac{1}{N_\ell} \geqslant \frac{1}{\frac{1}{V}\sum_{\ell=0}^{V-1}\mathbb{E}N_\ell}$$

Optimality condition for storage scheme

Maximize the normalized mean number of useful servers averaged

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?

MDS coded storage

Optimality of MDS coded storage

- ▶ Sequence of number of useful servers is the largest
- ▶ Latency optimal storage code

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

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

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

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

Problem statement

Find optimal storage scheme

$$S^* = \arg\max_{S \in \mathcal{S}} \frac{1}{V} \sum_{\ell=0}^{V-1} \mathbb{E} N_\ell.$$

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: $\alpha \geqslant 1$

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
▶ 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,

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

Numerical Results

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

Practical storage and access

- Placement of coded fragments depends on certain properties of storage codes
- Optimal access sequence is a Markov decision process
- We have heuristic solution to both questions
- Optimal placement remains open

Acknowledgements

References

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- R. Jinan, A. Badita, P. Sarvepalli, P. Parag. Latency optimal storage and scheduling of replicated fragments for memory-constrained servers. arXiv, Sep. 2020. Under review at TIT.

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