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# Node Repair on Connected Graphs

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## Node repair in distributed storage



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#### Node repair in distributed storage



Each codeword symbol stored in a node.

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#### Node repair in distributed storage



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- Total required transmission bounded by the Cut-set bound<sup>1</sup>

$$B \leqslant \sum_{i=0}^{k-1} \min\{l, (d-i)\beta\}$$

<sup>1</sup>Dimakis et. al, 2010

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Different pairs of (*l*, *β*) satisfying the above with equality give rise to different points on the storage-bandwidth trade-off.

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# Moving away from traditional setting

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#### Question 1 : Is it possible to process the data to reduce communication?

Question 2 : If so, then to what extent?

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## The bounds: How much can we process?

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$$H(S_i^f|X_i) = 0, H(X_f|S_1^f, \cdots, S_d^f) = 0.$$

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#### Lemma

Let  $v_f, f \in [n]$  be the failed node. For a subset of the helper nodes  $E \subset D$  let  $R_E^f$  be a function of  $S_E^f$  such that

$$H(X_f|R_E^f, S_{D\setminus E}^f) = 0.$$

1) If  $|E| \ge d - k + 1$ , then

 $H(R_E^f) \ge l.$ 

2) If  $|E| \leq d - k$ , then

$$H(R_E^f) \geqslant \frac{|E|l}{d-k+1}.$$

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## Proof of Lemma

• Given  $X_{D\setminus E}$  the information contained in  $R_E^f$  is sufficient to repair  $v_f$ , i.e.,

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• Take a set  $A \subset E$  with  $|A| = k - 1 - |D \setminus E|$ . Now,

$$H(R_{E}^{f}, W_{D\setminus E}, W_{A}) = H(R_{E}^{f}, W_{D\setminus E}, W_{f}, W_{A}) \ge kl$$

by the MDS property.

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$$H(R_E^f, W_{D\setminus E}, W_A) \leqslant H(R_E^f) + H(W_{D\setminus E}, W_A)$$
$$= H(R_E^f) + (k-1)l$$

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The proof of Part (2) is similar.

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## Lower bound on communication

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# Lower bound on communication

#### Definition

Repair graph: The subgraph spanned by the failed node and *d* helper nodes closest to it in terms of graph distance.

 $\Gamma_i(v_f)$ : set of helper nodes at distance *i* from the failed node.

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## Proposition

Let  $R_j^f$  be the random variable denoting the information flow from the *j*-th layer to the (j-1)-th layer. Then

$$H(R_j^f) \ge \min\left\{l, \frac{|\cup_{i=j}^t \Gamma_i(v_f)| \cdot l}{d-k+1}\right\}$$

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Proof.

Take  $E = \bigcup_{i=j}^{t} \Gamma_i(v_f)$ .

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Let  $J_f = \{v \in V(T_f) \setminus \{v_f\} : |D^*(v)| \ge d - k + 2\}$ . The total communication complexity  $\beta_{total}$  for the repair of node  $v_f$  on the repair tree  $T_f$  is bounded as

$$\beta_{total} \ge \sum_{v \in J_f} l + \sum_{v \in V(T_f) \setminus (\{v_f\} \cup J_f)} \frac{|D^*(v)|l}{d - k + 1}.$$

where  $D^*(v)$  : set of descendants of v including itself.

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- ▶ Needs to transmit at least  $|D^*(v)| \cdot l/(d-k+1)$  symbols to its immediate parent in  $T_f$ .

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- Every node  $v \in J_f$  needs to transmit at least *l* symbols to its immediate parent.

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# Achieving the bounds: The Product Matrix Framework<sup>2</sup>

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▶ Paremeters:  $[n, k, d = 2(k - 1), l = k - 1, \beta = 1, M = k(k - 1)]$ 

<sup>2</sup>Rashmi et al., 2011

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Recall: What we want to prove

A set of size at least d - k + 1 needs to transmit exactly *l* symbols for repair.

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- For  $h \in D$  define  $l^{(h)}(z) = \sum_{j=0}^{d-1} l_j^h z^j := \prod_{\substack{i \in D \\ i \neq h}} \frac{z-a_i}{a_h-a_i}$

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- The set A transmits the *l*-vector

$$\xi(f,A) = \sum_{h \in A} g^{(h)}(a_f) \begin{bmatrix} l_0^h + a_f^{k-1} l_{k-1}^h \\ l_1^h + a_f^{k-1} l_k^h \\ \vdots \\ l_{k-2}^h + a_f^{k-1} l_{2k-3}^h \end{bmatrix}$$

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# Achieving the bounds: Proving correctness

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### Claim

The content of the failed node f coincides with the vector  $\xi(f,D),$  i.e.,  $g^{(f)}(z)=\sum_{i=0}^{l-1}(\xi(f,D))_i\,z^i.$ 

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• Consider the polynomial  $H(z) = s_1(a_f, z) + z^{k-1}s_2(a_f, z) = \sum_{j=0}^{d-1} h_j z^j$ ,  $\deg(H) \leq 2k - 3 = d - 1$ .

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> The polynomial corresponding to the failed node defined before can be written as

$$g^{(f)}(z) = \sum_{j=0}^{k-2} (h_j + a_f^{k-1} h_{k-1+j}) z^j.$$

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Rephrasing, the contents of the node f is

$$(h_0 + a_f^{k-1}h_{k-1}, h_1 + a_f^{k-1}h_k, \dots, h_{k-2} + a_f^{k-1}h_{2k-3})^{\mathsf{T}}.$$

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At the same time, we can write H(z) in the Lagrange form  $H(z) = \sum_{h \in D} H(a_h) l^{(h)}(z)$ . where  $H(a_h) = g^{(h)}(a_f)$  due to the symmetry of  $s_1, s_2$ .

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Rephrasing, the contents of the node f is

$$(h_0 + a_f^{k-1}h_{k-1}, h_1 + a_f^{k-1}h_k, \dots, h_{k-2} + a_f^{k-1}h_{2k-3})^{\mathsf{T}}.$$

At the same time, we can write H(z) in the Lagrange form  $H(z) = \sum_{h \in D} H(a_h) l^{(h)}(z)$ . where  $H(a_h) = g^{(h)}(a_f)$  due to the symmetry of  $s_1, s_2$ .

It follows that the coefficients of the polynomial  $g^{(f)}(z)$  is nothing but  $\xi(f, D)$ .

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► This framework can be extended to get an MSR code with  $n, k, d = \frac{(k-1)t}{t-1}, l = \binom{k-1}{t-1}, M = t\binom{k}{t}$  for any  $t \ge 2$ .

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- The same ideas of intermediate processing applies.

Random Graphs

Extension to Interior Point Codes: Moving away from MSR
Random Graphs

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# Application to random graphs

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Random Graphs

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  - Node repair possible with high probability.
- > Threshold behavior: We say that *t*-layer repair of the failed node *v* is possible if

 $P(|N_t(v)| \ge d) \to 1 \text{ as } n \to \infty.$ 

Random Graphs

# Application to random graphs: Repair threshold

### Proposition

Let  $d = \delta n, 0 < \delta < 1$  be a constant and let t be a fixed integer. Then t is the threshold depth for repair if

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  - If  $(np)^{t-1} = o(n)$  and  $d = \Theta(n)$ , then with high probability, d nodes are not reached in t 1 layers [Chung et. al. 2001].

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#### Theorem

Let *t* be the threshold for repair. For  $d = \Theta(n)$  let  $d - k = \chi(n)$  be a function of *n* such that  $\chi(n)n^{s-1}p^s \to 0$  where  $s \leq t-1$  is the largest integer for which this condition holds. Then  $\mathbb{P}(\beta_{\mathrm{IP}} \leq (t-s)d + o(n)) \to 1$ .

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Let *t* be the threshold for repair. For  $d = \Theta(n)$  let  $d - k = \chi(n)$  be a function of *n* such that  $\chi(n)n^{s-1}p^s \to 0$  where  $s \leq t-1$  is the largest integer for which this condition holds. Then  $\mathbb{P}(\beta_{\mathrm{IP}} \leq (t-s)d + o(n)) \to 1$ .

• Using Intermediate Processing, the scaling of the bandwidth can be brought down from t to t - s.

Converse Results

Achievability

Random Graphs

## Concluding remarks

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- ▶ The intermediate processing technique is applicable to all *𝔽*-linear MSR codes as well as interior point codes and achieves the minimum possible communication in some cases.

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- In sparsely connected graphs, it is possible to do better than simple relaying.
- ▶ The intermediate processing technique is applicable to all *𝔽*-linear MSR codes as well as interior point codes and achieves the minimum possible communication in some cases.
- For random graphs  $\mathcal{G}_{n,p}$ , in certain regimes, intermediate processing can give significant reductions in communication overhead.