

A new idea for graph code decoding

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Introduction

Let (n, k, d) be the parameters of a component code. Giving a bipartite graph code we will consider an equivalent representation of their codewords as matrix \mathbf{M} with the following properties:

- $m \geq n$.
- Each row (column) of \mathbf{M} contains n open cells and n closed cells;
- Total number of open cells is equal to the length $N = mn$ of a graph code.
- Open cells are numbered in any order from 1 to N and i -th open cell is filled by i -th symbol of a codeword.
- Symbols written in open cells of any row (column) have to belong to a codeword of a component code.
- A codeword in \mathbf{M} consist of $2m$ component codewords in m rows and m columns.
- Each code symbol belongs to two component codewords (row and column).

Reference decoding algorithm

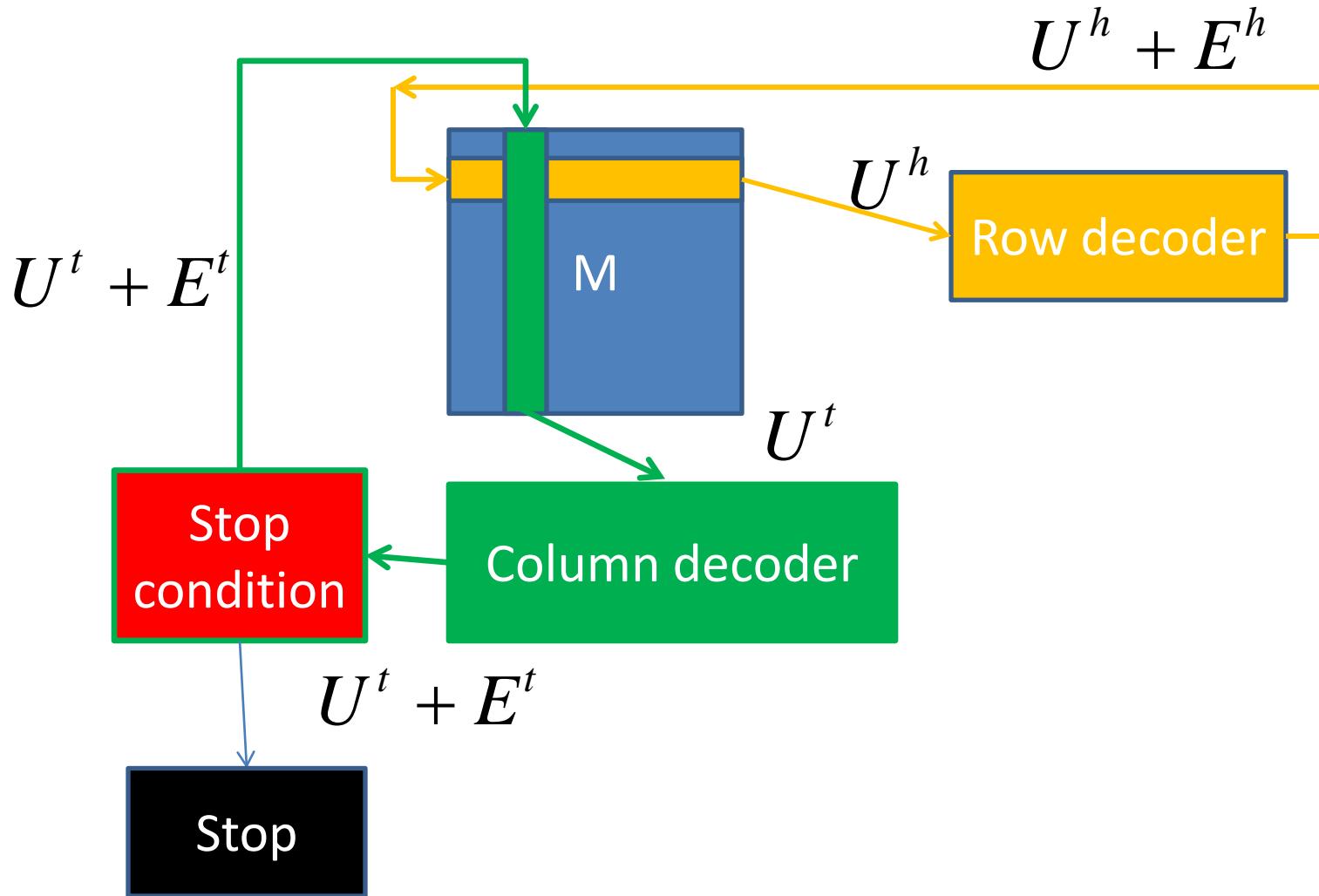
Giving a received word with errors and erasures:

1. *Decode all rows by decoding algorithm of a component code and apply the corrections to the received word.*
2. *Decode all columns by decoding algorithm of a component code and apply the corrections to the received word.*
3. *Check the stop condition and continue steps 1 – 3 if not.*

In [1, 3] was shown that the reference algorithm works well when component code corrects 5 or more errors and erasures.

For component Reed-Solomon code the conditional probability of decoding errors when more than T errors occur is approximately $1/T!$. This value is insignificant when the minimal distance of component code is $>=10$.

Reference decoding algorithm



The goal

- a new iterative decoding algorithm of a graph code when the component codes has a small distance (or low error-correcting capability)

$T \leq 2$.

The main idea

M is filled by a received word with errors and erasures.

Decode all rows and columns of **M** by a component code decoder and save all corrections in array **E**.

Notations:

- $U^h = (U_1^h, U_2^h, \dots, U_n^h), h = 1, 2, \dots, 2m,$ is a row (col.) of **M**.
- $U^{j(h,i)} = (U_1^{j(h,i)}, U_2^{j(h,i)}, \dots, U_n^{j(h,i)}),$ column (or row) of **M**
 $j(h,i) = 1, 2, \dots, 2m, i = 1, 2, \dots, n, j(h,i) \neq h,$

is a component *block* that intersects U^h in *i-th* position.

- $U_i^h = U_{l(h,i)}^{j(h,i)} = u \in \{GF(q) \cup \emptyset\}$ is one of **N** received symbols.

The main idea

$$E^h = (e_1^h, e_2^h, \dots, e_n^h),$$

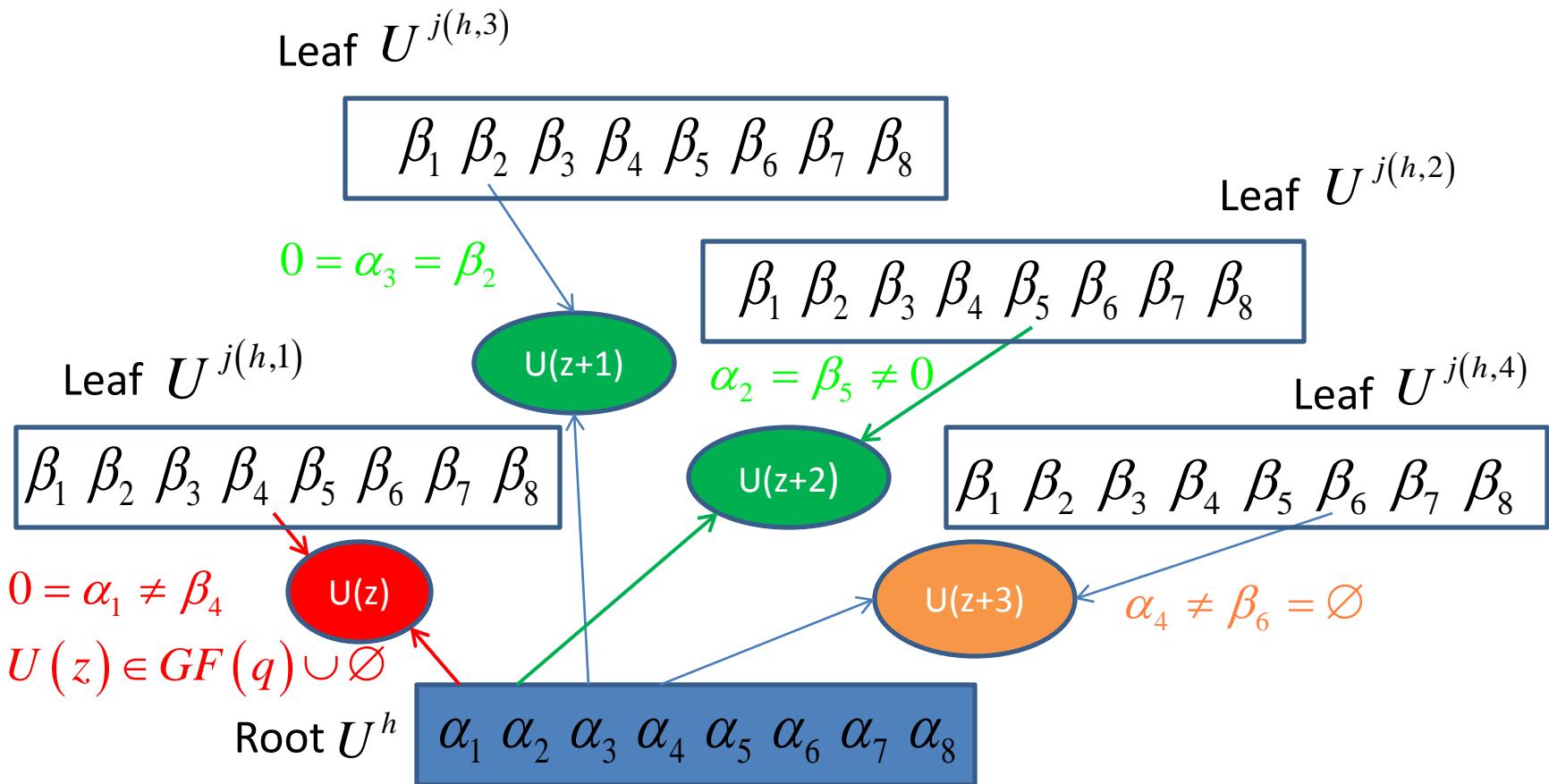
$$\alpha_i^h = e_i^h \rightarrow u \leftarrow e_{l(h,i)}^{j(h,i)} = \beta_i^h$$

$\alpha = \beta$ - *coincidence* of corrections from both sides, or
 $\alpha \neq \beta$ - *conflict* (or different corrections from both sides).

Facts:

- *if a conflict takes place then one of two decoding results of component blocks is wrong,*
- *the more is number of conflicts the less probable is correctness of the decoding result E^i .*

A part of code tree. RS(8,6,3)



The main idea (algorithm)

- Calculate for every root E^h the ratio

$$\Lambda^h = \frac{\Pr(G | E_{root}, E_{leaf}, U_{root})}{\Pr(B | E_{root}, E_{leaf}, U_{root})} = \frac{\Pr(E_{leaf} | G, E_{root}, U_{root})}{\Pr(E_{leaf} | B, E_{root}, U_{root})} \frac{\Pr(G | E_{root}, U_{root})}{\Pr(B | E_{root}, U_{root})}, h = 1, \dots, 2m,$$

G - *correct decoding* of a root (was transmitted) and
B, on the contrary, be a *wrong decoding* (was not transmitted).

- Find all the roots such that $\Lambda^h > \Lambda^{j(h,i)}$ over the root conflict positions $l(h,i), i = 1, \dots, n, h = 1, \dots, 2m$, and apply the corrections to the received (long) word in **M**.
- Do nothing with all other roots.

Proposition 1

$$\frac{\Pr(G | E_{root}, U_{root})}{\Pr(B | E_{root}, U_{root})} = \frac{\Pr(G | t_{root})}{\Pr(B | t_{root})}$$

Proposition 2

$$\frac{\Pr(E_{leaf} | G, E_{root}, U_{root})}{\Pr(E_{leaf} | B, E_{root}, U_{root})} = \frac{\Pr(\beta_{leaf} | G, \alpha_{root}, t_{leaf}, U_{root}^\emptyset)}{\Pr(\beta_{leaf} | B, \alpha_{root}, t_{leaf}, U_{root}^\emptyset)}$$

Proposition 3

$$\begin{aligned} \frac{\Pr(\beta_{leaf} | G, \alpha_{root}, t_{leaf}, U_{root}^\emptyset)}{\Pr(\beta_{leaf} | B, \alpha_{root}, t_{leaf}, U_{root}^\emptyset)} &= \frac{\Pr((\beta_1^h, \beta_2^h, \dots, \beta_n^h) | G, (\alpha_1^h, \alpha_2^h, \dots, \alpha_n^h), (t_1^h, t_2^h, \dots, t_n^h), U_{root}^\emptyset)}{\Pr((\beta_1^h, \beta_2^h, \dots, \beta_n^h) | B, (\alpha_1^h, \alpha_2^h, \dots, \alpha_n^h), (t_1^h, t_2^h, \dots, t_n^h), U_{root}^\emptyset)} = \\ &= \prod_{i=1}^n \frac{\Pr(\beta_i^h | G, \alpha_i^h, t_i^h, U_{root}^\emptyset)}{\Pr(\beta_i^h | B, \alpha_i^h, t_i^h, U_{root}^\emptyset)} \end{aligned}$$

Proposition 4

$$\Lambda = \frac{\Pr(\beta | G, \alpha, t', U_{root}^\emptyset)}{\Pr(\beta | B, \alpha, t', U_{root}^\emptyset)} = \begin{cases} \frac{1}{[\Pr(G' | t') / \Pr(B' | t')] + 1}, & \alpha \neq \beta, \\ \frac{\Pr(G' | t') / \Pr(B' | t') + (1 - t' / n)}{[\Pr(G' | t') / \Pr(B' | t') + 1] (1 - t' / n)}, & \alpha = \beta = 0 \text{ \& } u \neq \emptyset, \\ \frac{\Pr(G' | t') / \Pr(B' | t') + (t' / nq)}{[\Pr(G' | t') / \Pr(B' | t') + 1] (t' / nq)}, & \alpha = \beta \neq 0 \text{ \& } u \neq \emptyset, \\ \frac{\Pr(G' | t') / \Pr(B' | t') + (1 / q)}{[\Pr(G' | t') / \Pr(B' | t') + 1] (1 / q)}, & \alpha = \beta \text{ \& } u = \emptyset \end{cases}$$

Proposition 5

$$\Lambda = \frac{\Pr(\beta | G, \alpha, \emptyset)}{\Pr(\beta | B, \alpha, \emptyset)} = \begin{cases} 1 - p^*, & \alpha = \beta = 0, \\ p^*, & \alpha \neq \beta = 0. \end{cases}$$

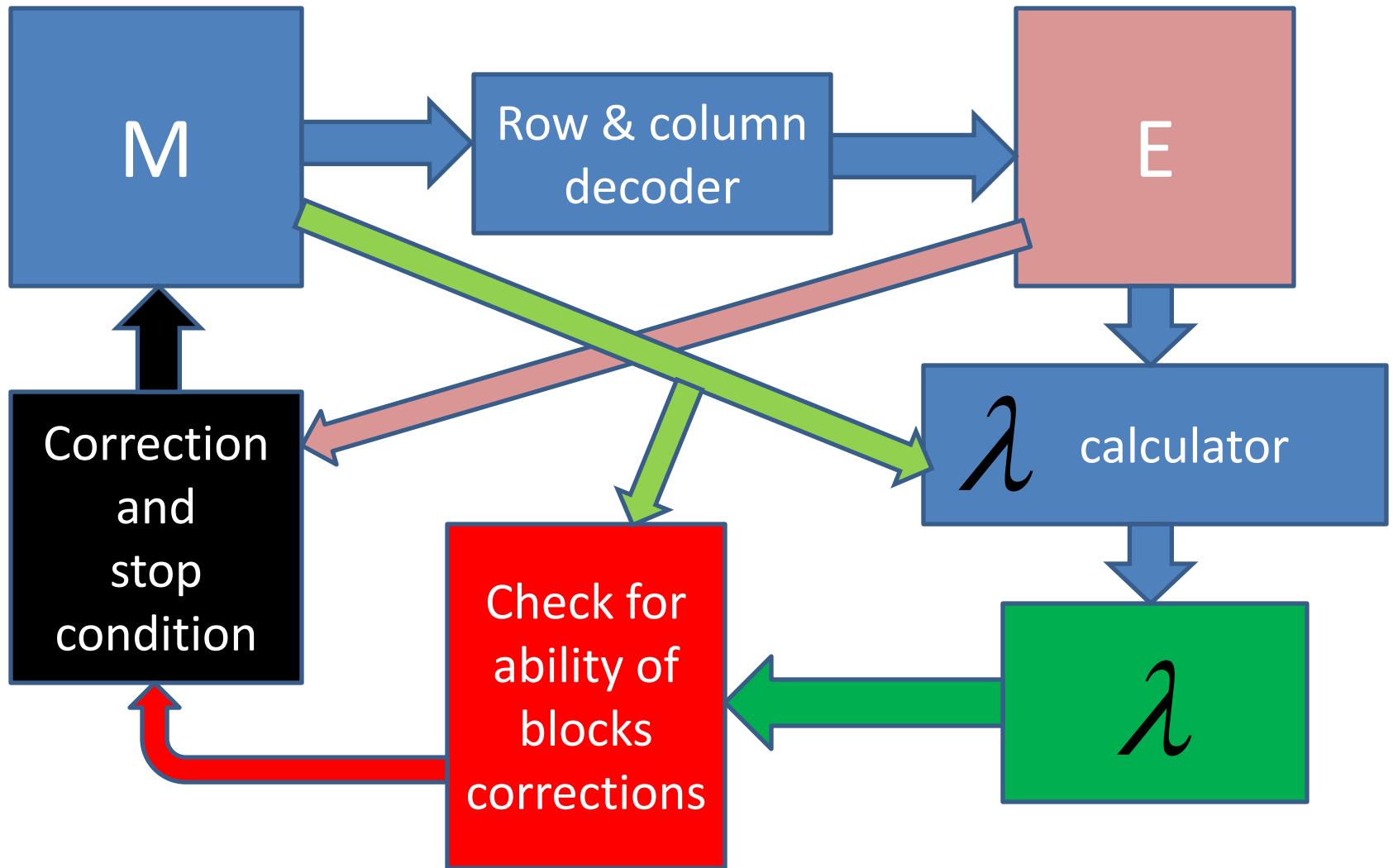
Definition

$$\lambda^h = \log \Lambda^h = \sum_{i=1}^n \lambda_i^h + \lambda_0^h, \lambda_i^h = \log \frac{\Pr(\beta_i^h | G, \alpha_i^h, t_i^h, U_{root}^\emptyset)}{\Pr(\beta_i^h | B, \alpha_i^h, t_i^h, U_{root}^\emptyset)}, \lambda_0^h = \log \frac{\Pr(G | t_{root})}{\Pr(B | t_{root})}.$$

Estimation

$$\lambda_i^h \approx \begin{cases} -\log \left[\Pr(G' | t_i^h) / \Pr(B' | t_i^h) + 1 \right] \approx \delta(d - \gamma t_i^h) > 0, & \alpha \neq \beta \\ \log \left(1 - \frac{t_i^h / n}{\Pr(G' | t_i^h) / \Pr(B' | t_i^h) + 1} \right) - \log(1 - t_i^h / n) \approx c \frac{t_i^h}{n}, & \alpha = \beta = 0 \& u \neq \emptyset, c < 1, \\ \log \left(\frac{nq}{t_i^h} \right) \approx 2 \log q - \log t_i^h \approx 2 \log q - (t_i^h - 1) / 2, & \alpha = \beta \neq 0 \& u \neq \emptyset, \\ \log q + \log \left(1 - \frac{1 - 1/q}{\Pr(G' | t_i^h) / \Pr(B' | t_i^h) + 1} \right) \approx \log q - \varepsilon, 0 \leq \varepsilon < 1, & \alpha = \beta \& u = \emptyset, \\ \log(1 - p^*), & t_i^h = \emptyset \& \alpha = \beta = 0 \\ \log p^*, & t_i^h = \emptyset \& \alpha \neq \beta = 0 \end{cases}$$

New Algorithm



Simulation results

- Code parameters: size of matrix \mathbf{M} – 256x256,
- Component code: Generalized Reed-Solomon code over $GF(16)$, $n=16$, $k=14$ or 12 , code distance – 3 or 5.
- General graph code: total length $N=4096$, total dimension $K=3072$ or 2048 , code distance (estimate) – 28 or 461.
- Simulation results for conditional error probability per a code symbol are given in the table below:
 $\delta=1.5, \gamma=.5$ for RS $d=3$ and
 $\delta=2.0, \gamma=.5$ for RS $d=5$.

S Eras.	T error	New decod er	Refer. decod er	S Eras.	T error	New decod er	Refer. decod er	S Eras.	T error	New decod er	Refer. decod er
0	250	1.e-2	3.e-2	40	130	1.e-6	4.e-4	200	90	1.2e-5	1.2e-3
0	200	8.3e-4	7.4e-3	40	100	1.1e-7	5.3e-5	200	60	2.1e-6	1.3e-4
0	190	3.5e-4	4.8e-3	50	100	3.5e-7	6.7e-5	200	40	8.3e-7	2.5e-5
0	170	5.3e-5	1.7e-3	60	145	2.7e-5	1.4e-3	200	20	7.8e-8	2.3e-6
0	150	5.2e-6	5.6e-4	80	190	3.9e-3	1.5e-2	200	15	0	1.e-6
0	140	1.2e-6	3.e-4	80	170	2.e-4	7.3e-3	300	100	1.4e-3	1.4e-2
0	130	4.2e-7	1.5e-4	80	150	9.9e-5	2.8e-3	300	20	1.8e-7	2.7e-5
0	120	8.5e-8	8.2e-5	80	135	1.5e-5	1.2e-3	400	70	1.5e-3	1.6e-2
0	110	4.9e-8	4.2e-5	80	110	1.7e-6	2.6e-4	400	10	2.9e-8	4.6e-5
10	200	1.1e-3	8.5e-3	80	80	1.9e-7	3.4e-5	500	40	1.5e-3	1.4e-2
10	160	2.6e-5	1.2e-3	100	170	1.6e-3	9.9e-3	500	20	6.1e-6	2.2e-3
10	130	4.1e-7	1.9e-4	100	140	7.4e-5	2.5e-3	500	10	5.3e-7	3.7e-4
10	100	4.6e-8	2.7e-5	100	100	1.4e-6	2.1e-4	600	20	2.3e-3	1.2e-2
20	190	6.2e-4	6.5e-3	120	140	1.7e-4	3.7e-3	600	5	1.5e-6	7.2e-4
20	140	2.6e-6	4.5e-4	120	60	4.2e-7	1.8e-5	620	2	8.1e-7	2.4e-4
20	100	2.7e-8	3.4e-5	160	100	8.e-6	8.7e-4	750	0	0	0

GRS code over $GF(16)$, $n=16$, $k=12$, code distance – 5.

Graph code: total length $N=4096$, total dimension $K=2048$, code distance (estimate) – 461.

S Eras.	T error	New decod er	Refer. decod er	S Eras.	T error	New decod er	Refer. decod er	S Eras.	T error	New decod er	Refer. decod er
0	720	8.3e-5	1.e-1	200	550	2.e-6	1.4e-2	650	350	1.7e-3	1.2e-1
0	700	1.e-6	7.6e-2	200	500	0	5.6e-5	650	300	6.e-6	4.6e-2
0	600	0	2.7e-5	250	450	0	2.3e-6	650	250	0	3.2e-4
0	580	0	0	250	440	0	0	650	220	0	1.8e-6
20	680	0	5.7e-2	400	460	1.5e-5	7.4e-2	650	200	0	0
40	670	1.5e-6	5.6e-2	400	400	0	1.2e-3	800	220	0	3.9e-2
70	650	0	4.7e-2	400	370	0	1.3e-5	800	100	0	0
70	600	0	1e-3	450	410	0	3.7e-2	1400	30	3.7e-3	7.e-2
150	600	0	4.1e-2	550	400	1.e-3	1.1e-1	1400	10	0	6.3e-4
150	520	0	7.1e-6	550	370	0	7.4e-2	1600	10	1.3e-1	1.4e-1
160	550	0	1.3e-3	550	350	0	3.4e-2	1600	0	0	0
160	500	0	0	550	300	0	1.7e-4	1650	0	2.5e-4	2.7e-4
				550	270	0	1.6e-6	1700	0	9.8e-2	9.8e-2

REFERENCES

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