LT Codes For Efficient and Reliable Distributed Storage Systems Revisited

Yongge Wang
Department of SIS, UNC Charlotte
9201 University City Blvd., Charlotte, NC 28223, USA
Email: yongge.wang@uncc.edu

Abstract—LT codes and digital fountain techniques have received significant attention from both academics and industry in the past few years. There have also been extensive interests in applying LT code techniques to distributed storage systems such as cloud data storage in recent years. However, Plank and Thomason’s experimental results show that LDPC code performs well only asymptotically when the number of data fragments increases and it has the worst performance for small number of data fragments (e.g., less than 100). In their INFOCOM 2012 paper, Cao, Yu, Yang, Lou, and Hou proposed to use exhaustive search approach to find a deterministic LT code that could be used to decode the original data content correctly in distributed storage systems. However, by Plank and Thomason’s experimental results, it is not clear whether the exhaustive search approach will work efficiently or even correctly. This paper carries out the theoretical analysis on the feasibility and performance issues for applying LT codes to distributed storage systems. By employing the underlying ideas of efficient Belief Propagation (BP) decoding process in LT codes, this paper introduces two classes of codes called flat BP-XOR codes and array BP-XOR codes (which can be considered as a deterministic version of LT codes). We will show the equivalence between the edge-colored graph model and degree-one-and-two encoding symbols based array BP-XOR codes. Using this equivalence result, we are able to design general array BP-XOR codes using graph based encoding. Similarly, based on this equivalence result, we are able to get new results for edge-colored graph models using results from array BP-XOR codes.

I. INTRODUCTION

Due to the widespread adoption of cloud computing technologies, there has been an increased interest for individuals and business entities to move their data from traditional private data center to cloud servers. Indeed, even popular storage service providers such as Dropbox use third party cloud storage providers such as Amazon’s Simple Storage Service (S3) for data storage.

With the wide adoption of cloud computing and storage technologies, it is important to consider data security and reliability issues that are strongly related to the underlying storage services. Though it is interesting to consider general data security issues in cloud computing environments, this paper will concentrate on the basic question of reliable data storage in the cloud and specifically the coding techniques for data storage in the cloud. There has been extensive research in reliable data storage on disk drives. For example, redundant array of independent disks (RAID) techniques have been proposed and widely adopted to combine multiple disk drive components into a logical unit for better resilience, performance, and capacity. The well known solutions to address the data storage reliability are to add data redundancy to multiple drivers. There are basically two ways to add the redundancy: data mirror (e.g., RAID 1) and data stripping with erasure codes (e.g., RAID 2 to RAID 6). Though data mirror (or data replication) provides the straightforward way for simple data management and repair of data on corrupted drives, it is very expensive to implement and deploy due to its high demand for redundancy. In addition to data replication techniques, erasure codes can be used to achieve the required data reliability level with much less data redundancy. Note that though error correcting codes (e.g., Reed-Solomon codes) could also be used for reliable data storage and correcting errors from failed disk drives, it is normally not used for data storage since it needs expensive computation for both encoding and decoding processes.

Erasure codes that have been used for reliable data storage systems are mainly binary linear codes which are essentially XOR-operation based codes. For example, flat XOR codes are erasure codes in which parity disks are calculated as the XOR of some subset of data disks. Though it is desirable to have MDS (maximal distance separable) flat XOR codes, it may not be available for all scenarios. Non-MDS codes have also been used in storage systems (e.g., the replicated RAID configurations such as RAID 10, RAID 50, and RAID 60). However, we have not seen any systematic research in designing non-MDS codes with flat XOR operations for storage systems.

In order to achieve better fault tolerance with minimal redundancy in data storage systems, there has also been active research in XOR based codes which are not necessarily flat XOR codes. For example, Blaum, Brady, Bruck, and Menon proposed the array code EVENODD for tolerating two disk faults and correcting one disk errors. Blaum, Bruck, and Vardy and Huang have extended the construction of EVENODD code to general codes for tolerating three disk faults. Other non-flat XOR based codes include (but are not limited to) [2k, k, d] chain code, Simple Product Code (SPC), Row-Diagonal Parity (RDP), and X-code.

The techniques that we have discussed above have been originally designed for data storage on disk drives and in storage area networks. It may not be directly applicable to distributed storage services such as the storage cloud. There have been many researches addressing the data storage
reliability issues in distributed environments. For example, Weatherspoon and Kubiatowicz [28] have compared erasure coding based solutions and replication based solutions for reliable distributed data storage systems.

Based on the seminal work of low-density parity-check (LDPC) codes by Gallager [15], several important techniques (see, e.g., Luby, Mitzenmacher, Shokrollahi, Spielman, and Stemmann [21]) have been developed for networked and communication systems such as Digital Fountain for content streaming. There have also been extensive interests in applying digital fountain techniques (such as LT codes) to distributed storage systems (see, e.g., Plank and Thomason [23] and Cao, Yu, Yang, Lou, and Hou [7]). However, it is not clear whether these applications of LT codes to distributed storage systems have advantages over other techniques that have been extensively adopted in disk drives and storage area networks such as flat XOR codes or array XOR codes.

Plank and Thomason’s experimental results [23] show that LDPC code performs well only asymptotically when the number of data packages increases and it has the worst performance for small number of data fragments (e.g., less than 100). Cao, Yu, Yang, Lou, and Hou [7] proposed to use exhaustive search approach to find a deterministic LT code that could be used to decode the original data content correctly in distributed storage systems. However, by Plank and Thomason’s experimental results, it is not clear whether the exhaustive search approach will work efficiently or even correctly. In this paper, we carry out a theoretical analysis on the feasibility and performance issues for applying LT codes to distributed storage systems. By employing the underlying ideas of efficient Belief Propagation (BP) decoding process in LT codes [20], we introduce two classes of codes called flat BP-XOR codes and array BP-XOR codes. The BP-XOR codes and array BP-XOR codes can be considered as a deterministic version of LT codes though flat BP-XOR codes are different from LT codes. Edge-colored graph models were introduced by Wang and Desmedt in [27] to model homogeneous faults in networks. We will show the equivalence between the edge-colored graph model and degree-one-and-two encoding symbol based array BP-XOR codes. Using this equivalence result, we are able to design general array BP-XOR codes using graph based results. Similarly, based on this equivalence result, we are able to get new results for edge-colored graph models using results from array BP-XOR codes. We have implemented an online software package for users to generate array BP-XOR code with their own specification and to verify the validity of their array BP-XOR codes (see [26]).

The structure of this paper is as follows. In Section II, we briefly review several coding techniques proposed for distributed storage systems. Sections III presents the challenges in applying LT codes to distributed storage systems. Section IV introduces flat BP-XOR codes for distributed storage systems and investigates the necessary and sufficient bounds for the existence of the codes. Section V introduces array BP-XOR codes for distributed storage systems and establish the equivalence between edge-colored graph models and array BP-XOR codes with degree one and two encoding symbols. Section VI presents constructions of array BP-XOR codes from graph based results (e.g., perfect one factorization of complete graphs), and Section VII proves a theorem on the limitation of array BP-XOR codes using only degree one and two encoding symbols.

II. CODING TECHNIQUES FOR DISTRIBUTED STORAGE SYSTEMS

In the seminal paper [24], Rabin proposed the Information Dispersal Algorithm (IDA) to code a data file into n pieces that will be stored among n servers such that the recovery of the information is possible when there are at most t = n − k failed servers (inactive but not malicious Byzantine style servers). Rabin’s scheme is essentially a kind of Reed-Solomon codes and needs relatively expensive finite field operations for encoding and decoding.

Krawczyk [19] extended Rabin’s IDA scheme to address the Byzantine style malicious servers which may intentionally modify their pieces of the information. Krawczyk called his scheme as Secure Information Dispersal Algorithm (SIDA).

There have been extensive interests in applying random linear coding techniques to distributed storage systems. For example, Dimakis, Ramchandran, Wu, and Suh [11] and Dimakis, Godfrey, Wu, Wainwright, and Ramchandran [10] used information flow graphs and random linear coding to achieve information theoretic minimum functional repair bandwidth γmin = k ln n + kd. In another word, if we divide the file F into k pieces, store the encoded fragments in n storage servers, and if one of these server fails, then a new comer storage server could functionally repair the system only if it could communicate γmin/d bits from each of the d surviving servers.

Inspired by network coding, Acedanski, Medard, and Koetter [1] proposed to use random linear coding for distributed networked storage with one centralized server and multiple storage servers. Dimakis, Prabhakaran, and Ramchandran [12] considered the problem from a different approach: There are n storage servers and many distributed data sources (that is, data are not from a central location and there is no centralized server). Each data source node picks one out of the n storage servers randomly, pre-routes its packet and repeat d(k) = c ln k times. Each storage server multiplies what it receives with coefficients selected uniformly and independently from F_q and stores the results together with the coefficients.

III. CHALLENGES IN APPLYING LT CODES TO DISTRIBUTED STORAGE SYSTEMS

Luby [20] pointed out that one of the potential applications of LT codes is distributed data storage systems. Several authors have continued these ideas with fruitful outcomes. For example, Plank and Thomason [23] have considered the practical implementations of LDPC codes for peer-to-peer and distributed storage systems and several experimental results have been reported. In particular, the experimental results in [23] show that LDPC “codes display their worst performance for 10 < n < 100” where n is the number of data fragments. Furthermore, their experiments show that “generating good instances of the codes is a black art...there is an opportunity...”
for theoretical research on codes for small \( n \) to have a very wide-reaching impact”. Recently, Cao, Yu, Yang, Lou, and Hou [7] proposed a LT code based secure cloud storage service (LTCS). In the LTCS scheme, a data file \( F \) is split into \( \lambda \) packets, each of which is \( |F|/\lambda \) bits. LT coding process is then used to generate \( n\alpha \) encoded packets, where \( \alpha = \lambda/k \cdot (1 + \varepsilon) \). These encoded packets are divided into \( n \) groups and each of the \( n \) storage servers receives \( \alpha \) packets. One of the basic requirements from the paper [7] is that the original \( \lambda \) data packets should always be recoverable from any \( k \) healthy servers. For LT codes, there is a small probability that one may not be able to reconstruct the original data packets from the \( k \) servers. In order to address this challenge, the authors in [7] recommended an exhaustive search method to divide the encoded symbols into \( n \) groups and check the decodability for each \( k \) combinations of groups. The process continues until one finds out one valid LT coding approach. This approach is not efficient when \( k \) and \( n \) are relatively larger. Furthermore, there is no guarantee that the exhaustive search method will end with a valid LT code. The authors did not give any analysis on how efficient this approach could be or any proof whether that is feasible. For the case of \( \alpha = 1 \), the coding scheme in [7] is essentially a flat XOR code that requires a belief propagation (BP) decoder ([20]). The following example shows that the exhaustive search may not succeed for some cases with \( \alpha = 1 \).

**Example 3.1:** For \( n = 5 \) and \( k = 3 \), the original data is divided into three fragments \( v_1, v_2, v_3 \) and coding symbols are stored in 5 servers \( (S_1, \ldots, S_5) \) such that the original data could be recovered from any 3 servers. In order for the belief propagation (BP) decoder to work, we must start from an original copy of \( v_1 \) or \( v_2 \) or \( v_3 \). Since there could be two erasure faulty servers, three servers have to store original copies of the data fragments. Without loss of generality, assume that \( S_1, S_2, \) and \( S_3 \) store \( v_1, v_2, \) and \( v_3 \) respectively. Again, since there could be two erasure servers, each data fragment need to be stored in at least three servers. Thus both \( S_1 \) and \( S_2 \) need to store \( v_1 \oplus v_2 \oplus v_3 \). Now if both \( S_1 \) and \( S_2 \) are faulty, neither \( v_1 \) nor \( v_2 \) could be recovered. \( \square \)

For the case of \( \alpha > 1 \), the coding scheme in [7] is a kind of array XOR codes (not flat XOR codes). In this case, the robust soliton distribution will be used to generate the \( n\alpha \) encoding symbols. In order for the analysis and bounds of the LT code to work, the numbers \( n, \alpha, \) and \( \lambda \) in [7] have to be sufficiently large (this has been confirmed by the experiments in [23]). For smaller values, these bounds may not work and the exhaustive search methods in [7] may never end with a successful code. However, for large enough \( n, \alpha, \) and \( \lambda \), the exhaustive search method will be inefficient and may be infeasible.

The experiment results from [23] and the potential challenges in the scheme [7] show that it is necessary and important to systematically study the encoding symbol generation problems for applying LT code to distributed storage systems. In the following sections, we will show what we could achieve and what we could not achieve with LT codes when applied to distributed storage systems.

**IV. Flat BP-XOR codes**

In this section, we introduce a class of codes called flat BP-XOR codes. In short, flat BP-XOR codes are flat XOR codes that could be decoded with the Belief Propagation (BP) algorithm for erasure codes. The BP algorithm for binary symmetric channels is present in Gallager [15] and is also used in artificial intelligence community [22]. In our paper, we use the BP algorithm for binary erasure channels (see [21], [20]).

Let \( M = \{0, 1\}^l \) be the message symbol set. The length \( l \) could be any number and it does not have impact on the coding. An \([n, k, d]\) flat BP-XOR code is a binary linear code determined by a \( k \times n \) zero-one valued generator matrix \( G \) such that for a given message vector \( x \in M^k \), the corresponding code \( y \in M^n \) is computed as \( y = xG \) where the addition of two strings in \( M \) is defined as the XOR on bits. Furthermore, a flat \([n, k, d]\) BP-XOR code requires that if at most \( d - 1 \) components in \( y \) are missing, \( x \) could be recovered from the remaining components of \( y \) with the Belief Propagation algorithm.

It is easy to see that each flat BP-XOR code is a flat XOR code, but the other direction may not hold. We can consider flat BP-XOR codes as one kind of applications of LT codes to reliable storage system design with deterministic decoding.

Example 3.1 shows that flat BP-XOR version of the LT code may not be applicable to threshold based distributed storage systems as proposed in [7]. In the following, we first mention the folklore fact to support our arguments.

**Fact 4.1:** Let \( n \geq k + 2 \), \( k \geq 2 \), and \( d = n - k + 1 \). Then there is no \([n, k, d]\) BP-XOR code.

The fact could be easily proved by the following observation: Let \( H = [\beta_1^T, \ldots, \beta_k^T | I_{n-k}] \) be an \((n-k) \times n\) parity check matrix. If every \( n-k \) columns in the matrix \([\beta_i^T | I_{n-k}]\) are linearly independent, then \( wt(\beta_i) = n - k \), where \( wt(\cdot) \) is the Hamming weight. Thus for \( n \geq k + 2 \), there is neither binary linear \([n, k, d]\) code nor \([n, k, d]\) BP-XOR code.

**Fact 4.1** shows the impossibility of designing flat \([n, k, d]\) BP-XOR codes for \( n \geq k + 2 \) and \( d = n - k + 1 \). Since flat BP-XOR codes are extremely efficient for encoding and decoding in practice, we are also interested in flat BP-XOR codes that are not MDS (maximal distance separable). In the following we show theoretical bounds designing flat BP-XOR codes for distributed data storage systems.

For an MDS \([n, k, d]\) code with \( d = n - k + 1 \), we can tolerate \( d-1 \) erasure faults. The question that we are interested in is: for given \( n \geq k + 2 \), what is best distance \( d \) we could achieve for a flat \([n, k, d]\) BP-XOR code? Fact 4.1 shows that \( d \) must be less than \( n - k + 1 \).

**Tolerating one erasure fault:** Let \( \alpha \in \{1\}^k \). The generator matrix \([I_k | \alpha^T]\) corresponds to the MDS flat \([k+1, k, 2]\) BP-XOR code that could tolerate one erasure fault.

**Tolerating two erasure faults:** Fact 4.1 shows that two parity check servers are not sufficient to tolerate two erasure faults for flat BP-XOR codes. In order to tolerate two erasures, we have to consider codes with \( n \geq k + 3 \). For \( n = k + 3 \), the following generator matrices show the existence of flat [5, 2, 3], [6, 3, 3], and [7, 4, 3] BP-XOR codes for tolerating two erasure
Indeed, the above three codes are the only flat $[k+3,k,3]$ BP-XOR codes tolerating two erasure faults with three redundancy columns.

**Theorem 4.2:** For $n \geq k+3$ and $k \geq 3$, there exists a flat $[n,k,3]$ BP-XOR code if and only if $k \leq 2^{n-k} - (n-k) - 1$.

**Proof.** Let $H = [\beta_1^T, \cdots, \beta_k^T]_I_{n-k}$ be an $(n-k) \times n$ parity check matrix. The code determined by $H$ has minimum distance 3 if and only if every 2 columns in $H$ are linearly independent. This implies that $H$ is the parity check matrix of a flat $[n,k,3]$ BP-XOR code if and only if for every $\beta_i$, we have $wt(\beta_i) \geq 2$ where $wt(\cdot)$ is the Hamming weight. By the fact that $|\{\beta \in \{0,1\}^{n-k} : wt(\beta) \geq 2\}| = 2^{n-k} - (n-k) - 1$, it follows that there exists a flat $[n,k,3]$BP-XOR code if and only if $k \leq 2^{n-k} - (n-k) - 1$. \qed

**Note:** It should be noted that the codes we have constructed in Theorem 4.2 is the well known Hamming code when $k = 2^{n-k} - (n-k) - 1$. For $k < 2^{n-k} - (n-k) - 1$, it is a truncated version of the Hamming code.

By Theorem 4.2, there is no flat $[k+4,k,3]$ BP-XOR code for $k \geq 12$. Table I lists the required redundancy for tolerating two erasure faults when the value of $k$ changes. All the codes that we have constructed in Theorem 4.2 are systematic. Based on the proof of Theorem 4.2, we have the following corollary.

**Corollary 4.3:** For $n > k$ and $k \geq 3$, there exists an $[n,k,3]$ binary linear code if and only if there exists a systematic $[n,k,3]$ binary linear code and if and only if there exists a systematic flat $[n,k,3]$ BP-XOR code.

**Tolerating three erasure faults:** We first prove the following theorem for the convenience of proving the existence of systematic flat XOR codes.

**Theorem 4.4:** For $n > k$ and $d \leq n-k+1$, there exists an $[n,k,d]$ binary linear code if and only if there exists an $(n-k) \times k$ matrix $A = (\beta_1^T, \cdots, \beta_k^T)$ with the following properties:
1) $\beta_i \in \{0,1\}^{n-k}$ for $1 \leq i \leq k$
2) Let $d_1 + d_2 = d-1$. If we remove $d_2$ rows from $A$, then every $d_1$ columns of the remaining matrix are linearly independent.

**Proof.** First, it is straightforward to show that the condition 2 in the Theorem implies that $wt(\beta_i) \geq d-1$ for $1 \leq i \leq k$, where $wt(\cdot)$ is the Hamming weight. It is also straightforward to show that the condition 2 in the Theorem implies that every $d-1$ columns in the matrix $[A|I_{n-k}]$ are linearly independent. Thus, the linear code corresponding to the parity check matrix $[A|I_{n-k}]$ is a binary linear $[n,k,d]$ code. Note that the generator matrix corresponding to the parity check matrix $[A|I_{n-k}]$ is $G = [I_k|A^T]$.

For the other direction, assume that there exists a $k \times n$ generator matrix $G$ for an $[n,k,d]$ binary linear code. Let $\alpha_1, \cdots, \alpha_n$ be the $n$ columns of $G$. Without loss of generality, we may assume that $\alpha_1, \cdots, \alpha_k$ are linearly independent. We may also assume that
$$\left( \begin{array}{c} \alpha_1^T \\ \vdots \\ \alpha_k^T \end{array} \right) = \left( \begin{array}{c} I_k \\ A \end{array} \right) \left( \begin{array}{c} \alpha_1^T \\ \vdots \\ \alpha_k^T \end{array} \right)$$
where $A = (\beta_1^T, \cdots, \beta_k^T)$ and $\beta_i \in \{0,1\}^{n-k}$.

Since the code has the minimum distance $d$, the remaining generator matrix $G$ should have a rank of $k$ after removing any $d-1$ columns from $G$. Let $d_1 + d_2 = d-1$ and assume that we remove $d_1$ columns $\alpha_{i_1}, \cdots, \alpha_{i_d}$ for $1 \leq d_2$ columns $\alpha_{i_1}, \cdots, \alpha_{i_d}$ for $1 \leq i_1 < i_2 < \cdots < i_d \leq n-k$ from the generator matrix $G$. Then $\alpha_{i_1}, \cdots, \alpha_{i_d}$ should be able to be linearly generated from the columns $\alpha_i$ for $i \geq k+1$ and $i \neq j_1, \cdots, j_d$. This is equivalent to the requirements that the rows $i_1, \cdots, i_d$ of $I_k$ could be linearly generated from the remaining rows of $A$ after removing the rows $j_1, \cdots, j_d$ from $A$. It follows that the remaining columns $i_1, \cdots, i_d$ of $A$ are linearly independent after removing the rows $j_1, \cdots, j_d$ from $A$. This completes the proof of the Theorem.

By Theorem 4.4, we have the following results.

**Theorem 4.5:** For $n \geq k+4$, there exists a systematic flat XOR $[n,k,4]$ code if and only if
$$k \leq \begin{cases} 2^{n-k-1} - n + k & \text{if } n-k \text{ is even} \\ 2^{n-k-1} - n + k - 1 & \text{if } n-k \text{ is odd} \end{cases}$$

**Proof.** Let $X = \{\beta : \beta \in \{0,1\}^{n-k}, wt(\beta) = 3,5,7,\cdots\}$. Then
$$|X| = \sum_{i \geq 3, i \text{ is odd}} \binom{n-k}{i} = \sum_{i \geq 3, i \text{ is odd}} \left( \binom{n-k-1}{i-1} + \binom{n-k-1}{i} \right) = \begin{cases} 2^{n-k-1} - n + k & \text{if } n-k \text{ is even} \\ 2^{n-k-1} - n + k - 1 & \text{if } n-k \text{ is odd} \end{cases}$$

Define an $(n-k) \times k$ matrix $A = (\beta_1^T, \cdots, \beta_k^T)$ where $\beta_i \in X$. It is straightforward to show that this matrix $A$ satisfies the condition 2 of Theorem 4.4 for $d = 4$ (alternatively, every three columns in the parity check matrix $[A|I_{n-k}]$ are linearly independent). Thus the binary linear code corresponding to the parity check matrix $[A|I_{n-k}]$ (or the generator matrix $[I_k|A^T]$) is a flat XOR $[n,k,4]$ code.

For the other direction, it suffices to show that $X$ is a
maximal set that satisfies the condition 2 of Theorem 4.4 for $d = 4$. This is proved by observing the fact that every even Hamming weight vector $\beta \in \{0,1\}^{n-k}$ is equal to $\beta_1 + \beta_2$ for some $\beta_1, \beta_2 \in X$. This completes the proof of the theorem.

In Theorem 4.5, we established a necessary and sufficient condition for designing systematic flat XOR codes tolerating three erasure faults. However, the codes we constructed in Theorem 4.5 are not necessarily flat BP-XOR codes. For example, let $n = 7, k = 3, d = 4$, and $\beta_1 = (1, 1, 1, 0), \beta_2 = (0, 1, 1, 1)$, and $\beta_3 = (1, 0, 1, 1)$. Then the corresponding generator code has the following generator matrix:

$$
\begin{bmatrix}
I_3 & 1 & 1 & 1 & 0 \\
0 & 1 & 1 & 1 \\
1 & 0 & 1 & 1
\end{bmatrix}
$$

It is straightforward that this is not a flat BP-XOR code since if we remove the first three columns from the above generator matrix, no column in the remaining generator matrix has Hamming weight 1. Indeed, it is easy to show that for $n = 7, k = 3$, and $d = 4$, there is no flat [7, 3, 4] BP-XOR code. The reason is that in order for a [7, 3, 4] linear code to be a flat BP-XOR code, we have to have four columns with Hamming weight 1 in the generator matrix. Furthermore, we need to have Hamming weight 4 for each row. Without loss of generality, we may assume that the column $(1, 0, 0)^T$ occurs twice in the generator matrix. Then we have to have three columns in the generator matrix with the format $(b, 1, 1)^T$ where $b = 0, 1$. In another word, two columns of the generator matrix are identical, which will reduce the code distance to 3.

The above discussion shows that the condition in Theorem 4.5 is not valid for the existence of flat BP-XOR code tolerating three erasure faults. Though it is interesting to identify necessary and sufficient conditions for the existence of BP-XOR codes tolerating three or more erasure faults, it is sufficient for us to use the flat XOR codes in distributed storage systems since a simple XOR based Gauss elimination methods could be used to recover the original data content in front of erasure faults. This observation tells us that LT code (i.e., the flat BP-XOR code) may not be the best choices for distributed storage systems in some cases.

As an example, Table II lists the required redundancy for tolerating three erasure faults when the value of $k$ changes.

| $k$ | required redundancy | flat XOR code |
|-----|---------------------|---------------|
| $2 \leq k \leq 4$ | 4 | $[k + 4, k, 4]$ |
| $5 \leq k \leq 10$ | 5 | $[k + 5, k, 4]$ |
| $11 \leq k \leq 26$ | 6 | $[k + 6, k, 4]$ |
| $27 \leq k \leq 56$ | 7 | $[k + 7, k, 4]$ |

**Tolerating four or more erasure faults:** In general, we are also interested in designing flat BP-XOR codes for tolerating more than three erasure faults. For distributed storage systems we could generally use nested techniques (e.g., the similar techniques as nested RAID array). In the following, we present several sufficient conditions for tolerating four or more erasure faults. Normally these conditions are not necessary. It should be noted that for general binary linear codes, there are well known bounds (see, e.g., Verhoef [25]). However, the codes corresponding to these bounds are not necessarily flat BP-XOR codes.

**Theorem 4.6:** For $n \geq k + 5$, there exists a systematic flat XOR $[n, k, 5]$ code if $k$ is less than

$$
\frac{n-k-2}{2} + 2 \left( \left\lfloor \frac{n-k}{2} \right\rfloor - 2 \right) / 2 + 2 \left( \left\lfloor \frac{n-k}{4} \right\rfloor - 2 \right) / 2.
$$

**Proof.** Let $U = \{a_1, \ldots, a_{n-k}\}$ be an $n-k$ element set. In the following, we construct four-element subsets of $U$ so that the characteristic sequences of these subsets could be used as the columns of the parity check matrix. It will be convenient for the reader to understand the following subset definitions if the elements of $U$ are interpreted as leaf nodes on a binary tree of depth $\lfloor \log_2(n-k) \rfloor$.

Let $\beta_1, \ldots, \beta_n$ be the characteristic sequences of the above sets. Then it is straightforward that the parity check matrix $H = [\beta_1^T, \ldots, \beta_n^T]_{n-k}$ corresponds to a systematic flat XOR code of minimum distance 5. The code has distance 5 since every 4 columns in $H$ are linearly independent by the facts that (1) for any $\beta_1, \beta_2$, we have $\text{wt}((\beta_1 + \beta_2)) > 2$; and (2) any three or four $\beta$ are linearly independent. The two facts follow from the construction. This completes the Proof of the Theorem.

As an example, Table III lists the required redundancy of flat XOR codes for tolerating 4 erasure faults based on Theorem 4.6. Though the conditions in Theorem 4.6 are not necessary in general. The bounds in Table III matches the bounds for general binary linear codes (see [25]). Thus the conditions in Theorem 4.6 are also necessary for $k \leq 9$.

**TABLE III**

| $k$ | required redundancy | flat XOR code |
|-----|---------------------|---------------|
| $k \leq 2$ | 6 | $[k + 6, k, 5]$ |
| $3 \leq k \leq 4$ | 7 | $[k + 7, k, 5]$ |
| $5 \leq k \leq 9$ | 8 | $[k + 8, k, 5]$ |

**V. ARRAY BP-XOR CODES FOR DISTRIBUTED STORAGE SYSTEMS**

Array codes have been studied extensively for burst error correction in communication systems and storage systems (see, e.g., [3], [4], [5], [6], [8], [29], [30]). Array codes are linear
codes where information and parity data are placed in a two dimensional matrix array. Appropriately designed array codes such as EVENODD [3], RDP [9], STAR [17], X-code [30] are very useful for high speed storage application systems since they enjoy low-complexity decoding and low update complexity. Most of these array codes are designed for RAID array based storage systems with the specific requirements such as systematic code, efficient decoding algorithm, and minimum update complexity, where update complexity refers to the number of encoding data symbols that need to be updated if one information data symbol is changed.

For distributed storage systems such as cloud storage, we may not need the code to be systematic. As studied in [20], [23], [7], LT code or digital fountain techniques could be a better choice for distributed storage systems. However, as we have mentioned in previous sections and as supported by the experimental results in [23], the probabilistic bounds in LT code performs well only asymptotically when the number \( n \) of encoding symbols increase. For small numbers of \( n < 100 \), the codes display their worst performance. Thus it is important to study the applicable coding schemes with better performance for distributed storage systems. As we have noticed, one of the major advantages that contribute to the efficiency of LT decoding process is the Belief Propagation (BP) process. In the following, we design a kind of array codes that could be efficiently decoded using the BP-process. We will call such kind of codes array BP-XOR codes. Appropriately designed array BP-XOR codes could achieve the MDS property from both communication and storage aspects: for \( k \) blocks of the original data, surviving storage servers only need to store \( k \) blocks of encoding blocks. Note that in LT codes, in order to decode \( k \) blocks of data with probability \( 1 - \delta \), \( k + O(\sqrt{n} \ln^2 (k/\delta)) \) blocks of encoding blocks are needed.

Array BP-XOR code is defined as follows. Let \( \alpha_1, \ldots, \alpha_k \in \{0,1\}^l \) be the data fragments that we want to encode, where \( l \) is any fixed number. A \( t \)-erasure tolerating array BP-XOR code is an \( m \times n \) matrix \( C = [\sigma_{i,j}]_{1 \leq i \leq m, 1 \leq j \leq n} \) such that:

1) Each \( \sigma_{i,j} \) is the XOR of one or more elements from the data fragments \( \alpha_1, \ldots, \alpha_k \).
2) \( \alpha_1, \ldots, \alpha_k \) could be recovered from any \( n - t \) columns of the matrix using the binary erasure channel based BP algorithm.

If we add the restriction that each element in \( C = [\sigma_{i,j}]_{1 \leq i \leq m, 1 \leq j \leq n} \) be the XOR of at most two elements from the data fragments \( \alpha_1, \ldots, \alpha_k \), then the restricted array BP-XOR codes are equivalent to the edge-colored graph models introduced by Wang and Desmedt in [27] for tolerating network homogeneous faults.

A. Edge-colored graphs

In this section, we first describe the edge-colored graph model by Wang and Desmedt [27]. The reader should be reminded that the edge-colored graph model in [27] is slightly different from the edge-colored graph definition in most literatures. In most literatures, the coloring of the edges is required to meet the condition that no two adjacent edges have the same color. This condition is not required in the definition of [27].

**Definition 5.1:** (Wang and Desmedt [27]) An edge-colored graph is a tuple \( G(V, E, C, f) \), with \( V \) the node set, \( E \) the edge set, \( C \) the color set, and \( f \) a map from \( E \) onto \( C \). The structure

\[
Z_{C,t} = \{ Z : Z \subseteq E \text{ and } |f(Z)| \leq t \}.
\]

is called a \( t \)-color adversary structure. Let \( A, B \in V \) be distinct nodes of \( G \). \( A, B \) are called \((t + 1)\)-color connected for \( t \geq 1 \) if for any color set \( C_t \subseteq C \) of size \( t \), there is a path \( p \) from \( A \) to \( B \) in \( G \) such that the edges on \( p \) do not contain any color in \( C_t \). An edge-colored graph \( G \) is \((t + 1)\)-color connected if and only if for any two nodes \( A \) and \( B \) in \( G \), they are \((t + 1)\)-color connected.

As an example, Figure 1 contains two 3-color connected edge-colored graphs \( G_{4,1} \) and \( G_{4,2} \). \( G_{4,1} \) contains 5 nodes, 8 edges, and 4 colors. \( G_{4,2} \) contains 7 nodes, 12 edges, and 4 colors.

![Fig. 1. 3-color connected edge-colored graphs](image-url)

A general 3-color connected edge-colored graph with \( k \) nodes can be constructed as follows.

1) For \( k = 4r+1 \), the \( v_1 \) node of \( r \) copies of \( G_{4,1} \) are glued together to form a 3-color connected edge-colored graph \( G \) with \( 4r + 1 \) nodes and \( 8r \) edges.
2) For \( k = 4r+3 \), the \( v_1 \) node of \( r - 1 \) copies of \( G_{4,1} \) and one copy of \( G_{4,2} \) are glued together to form a 3-color connected edge-colored graph \( G \) with \( 4r+3 \) nodes, \( 8r+4 \) edges.
3) For \( k = 4r+2 \) (respectively \( k = 4r+4 \)) with \( r \geq 1 \), one node is added to the 3-color connected edge-colored graph \( G \) with \( 4r + 1 \) nodes (respectively \( k = 4r+3 \) nodes) by connecting this node to any 3 nodes within the graph with distinct colors. The resulting graph is a 4-color connected edge-colored graph with \( 4r+2 \) (respectively \( 4r+4 \)) nodes and \( 8r+3 \) (respectively \( 8r+7 \)) edges.

For convenience, an edge-colored graph could also be represented by a table, where the edges with same colors are put in the same column. For example, \( G_{4,1} \) and \( G_{4,2} \) in Figure 1 are represented in Table IV.

Wang and Desmedt [27] showed several constructions of edge-colored graphs with certain color connectivity. In the following, we present a general construction of \((t + 1)\)-color connected edge-colored graphs using perfect one-factorizations of...
complete graphs. We use \( K_n = (V, E) \) to denote the complete graph with \( n \) nodes. For an even \( n \), a one-factor of \( K_n \) is a set of pairwise disjoint edges that partition the set of nodes in \( V \). A one-factorization of \( K_n \) (\( n \) is even) is a set of one-factors that partition the set of edges \( E \). A one-factorization is called perfect if the union of every two distinct one-factors is a Hamiltonian circuit. It is shown (see, e.g., Anderson [2]) that perfect one-factorizations for \( K_2p \), and certain \( K_{2n} \) do exist, where \( p \) is a prime number.

Theorem 5.2: Let \( n \) be an even number such that there is a perfect one-factorization \( F_1, \ldots, F_{n-1} \) for \( K_n \). For each \( t \leq n-3 \), there exists a \((t+1)\)-color connected edge-colored graph \( G \) with \( n-1 \) nodes, \((t+2)(n/2-1)\) edges, and \( t+2 \) colors.

Proof. Let \( V = \{v_1, \ldots, v_{n-1}\} \), \( F'_i = F_i \setminus \{v_n, v\} \), and \( E = F'_1 \cup \cdots \cup F'_{t+2} \) and color all edges in \( F'_i \) with color \( c_i \) for \( i \leq t+2 \). Then it is straightforward to check that the edge-colored graph \((V, E)\) is \((t+1)\)-color connected, \(|V| = n-1\), and \(|E| = (t+2)(n/2-1)\).

Remarks on Proof of Theorem 5.2 Since only node connectivity instead of Hamiltonian circuit is required for \((t+1)\)-color connected graphs, we could use \( F'_i \) instead of \( F_i \) to construct the edge-colored graphs. By using \( F'_i \), we reduce \( t+2 \) edges and one node in the resulting edge-colored graph. This will help us to keep the minimum cost for connectivity.

B. Constructing array BP-XOR codes from edge-colored graphs

We now use edge-colored graphs to construct array BP-XOR codes. As an example, we first give the BP-XOR code corresponding to the graph \( G_{4,2} \) in Table IV. As Step 1, the \( G_{4,2} \) part in Table IV is converted to the code in Table V. In the step 2, choose any fixed node and remove all of its occurrence from the code in Table IV. For convenience, we choose to remove the occurrence of \( v_7 \) and get the BP-XOR code in Table VI.

It is easy to check that the data fragments \( v_1, \ldots, v_6 \) can be recovered from any two columns of coding symbols. It is also straightforward to observe that the code in Table VI achieves optimal space and communication bandwidth in the event of two column erasures.

In the following, we give the general construction of BP-XOR code from edge-colored graphs. Let \( v_1, v_2, \ldots, v_k \in \{0, 1\}^l \) be data blocks that we want to encode, where \( l \) is any fixed length. Let \( G(V, E, C, f) \) be a \((t+1)\)-color connected edge-colored graph with \( V = \{v_1, \ldots, v_k, v_{k+1}\} \), \(|E| = \lambda \), and \( C = \{c_1, c_2, \ldots, c_n\} \). If we consider the nodes in the edge-colored graph \( G(V, E, C, f) \) as data blocks, edges as their parity check blocks of the adjacent nodes, and colors on the edges as labels for placing the parity checks into different columns of the array codes, then following steps construct an \( m \times n \) array BP-XOR codes, where \( m = \max_{c \in C}\{|Z| : Z \subseteq E, f(Z) = c\} \).

1) For \( 1 \leq i \leq n \), let \( \beta'_i = \{v_i \oplus v_j : \langle v_i, v_j \rangle \in E, f(\langle v_i, v_j \rangle) = c_i\} \).
2) For each \( \beta'_i \), replace the entry \( v_{k+1} \oplus v \) with \( v \) if such entry exists. Furthermore, if \( |\beta'_i| \) is smaller than \( m \), add empty element to \( \beta'_i \) to make it an \( m \)-length vector \( \beta_i \).
3) The array BP-XOR code is then specified by the \( m \times n \) matrix \( C_G = (\beta'_1, \ldots, \beta'_n) \).

Next we show that the above defined array BP-XOR code \( C_G \) can tolerate \( t \) column erasure faults. Let \( C_t \subset C \) be any set of \( t \) colors of the graph \( G \) and assume that \( t \) columns corresponding to the color set \( C_t \) are missing in \( C_G \). Since the graph \( G \) is \((t+1)\)-color connected, for any node \( v_{i0} \in V \), we have a path \( p = (v_{k+1}, v_{i1}, v_{i2}, \ldots, v_{ij}, v_{i0}) \) without using any colors in \( C_t \). Thus \( v_{i0} \) could be recovered by the following equation

\[
v_{i0} = v_{i1} \oplus (v_{i1} \oplus v_{i2}) \oplus \cdots \oplus (v_{ij} \oplus v_{i0})
\]

where \( v_{i1}, v_{i2}, \ldots, v_{ij} \) are all contained in the non-missing columns. Thus the Belief Propagation process could be used to recover the entire data blocks \( v_1, \ldots, v_k \) from the non-missing columns with only \( k \) XOR operations on the encoding symbols.

C. Constructing edge-colored graphs from array BP-XOR codes

In this section, we show that for each array BP-XOR code, we could construct a corresponding edge-colored graph.

Theorem 5.3: Let \( C \) be an \( m \times n \) array BP-XOR code with the following properties:

1) \( C \) is \( t \)-erasure tolerating;
2) \( C \) contains \( k \) information symbols; and
3) \( C \) contains only degree one and two encoding symbols.

Then there exists a \((t+1)\)-color connected edge-colored graph \( G(V, E, C, f) \) with \(|V| = k+1\), \(|E| = mn\), and \(|C| = n\).

Proof. Let \( v_1, \ldots, v_k \) be the information symbols of \( C = [v_{i,j}]_{(i,j) \in [1,m] \times [1,n]} \) and \( v_{i1}, \ldots, v_{in} \) be a list of degree one
encoding symbols in $C$. Then the $(t+1)$-color connected edge-colored graph $G(V,E,C,f)$ is defined by the following steps:

1. $V = \{v_1, \cdots, v_k, v_{k+1}\}$;
2. $E = \bigcup_{i \in [1,u]} \{ (v_{k+1}, v_i) \} \cup \{ (v_{i'}, v_{j'}) : a_{i,j} = v_{i'} \oplus v_{j'} \in C \}$;
3. $C = \{c_1, \cdots, c_n\}$;
4. for each $a_{i,j} = v_{i'} \oplus v_{j'} \in C$, let $f((v_{i'}, v_{j'})) = c_j$ and for each $a_{i,j} = v_{i'} \in C$ let $f((v_{k+1}, v_{i'})) = c_j$

Let $C_t$ be a color set of size $t$ and $v_1$ and $v_2$ be two nodes. Since the code $C$ is $t$-erasure tolerating, both $v_1$ and $v_2$ could be recovered from encoding symbols not contained in the columns corresponding to the colors in $C_t$. Thus there exists a path $p$ ($q$ respectively) connecting $v_{k+1}$ to $v_1$ (to $v_2$) respectively without using $C_t$-colored edges. It follows that $G(V,E,C,f)$ is $(t+1)$-color connected.

VI. EXAMPLES OF BANDWIDTH OPTIMAL ARRAY BP-XOR CODES

In this section, we use edge-colored graphs in Theorem 5.2 to construct $m \times n$ BP-XOR codes that could tolerate $n - 2$ erasure columns. The general process is as follows: For a given number $n$ of code columns and a number $t$ of erasure columns, we first design $(t+1)$-color connected edge-colored graphs with $n$ colors and the smallest number of graph edges. The resulting edge-colored graph is then converted to the BP-XOR code with the process described in the previous section.

In order to design an $m \times n$ BP-XOR code tolerating $n - 2$ erasure columns, find the smallest $p$ (or $2p$) such that $n \leq p$ (or $n \leq 2p - 1$), where $p$ is an odd prime. Suppose $p$ is such a prime with $n \leq p$. Then Table VII defines a $(p-1)$-color connected edge-colored graphs with $p$ nodes and $p$ colors (based on the perfect one-factorization of $K_{p+1}$ in [13]). In Table VII, if we consider the first column as a sequence of numbers: $1, p - 1; 2, p - 2; \cdots; (p - 1)/2, (p + 1)/2$, then the $i$th column of the table is defined by the following sequence of numbers (operations are mod $p$ and 0 is replaced with $p$):

$$1 + i, p - 1 + i; 2 + i, p - 2 + i; \cdots; (p - 1)/2 + i, (p + 1)/2 + i.$$  

The above edge-colored graph is then converted to the $(p - 1)/2 \times p$ BP-XOR code in Table VIII where $m = (p - 1)/2$. Then an $m \times n$ BP-XOR code is obtained by taking any of the $n$ columns in Table VIII. Since the edge-colored graph in Table VII is $(p-1)$-color connected, it follows that the above constructed $m \times n$ BP-XOR code could tolerate $n - 2$ erasure columns. In another word, the original data content $F$ is divided into $p - 1$ fragments $v_1, \cdots, v_{p-1} \in \{0,1\}$ of equal length ($l$ bits) and are stored in $n$ servers according to the BP-XOR codes, then the original data content $F$ could always be recovered from any two surviving servers. It should also be noted that each storage server stores $(p - 1)/2$ bits of data and the total data stored at two storage servers are $(p - 1)l = |F|$ bits. Thus the BP-XOR code is optimal in bandwidth and space.

We should also note that the $(p - 1)/2 \times p$ BP-XOR code in Table VIII is equivalent to the code designed by Zaitsev, Zinov’ev, and Semakov [14] which was reformulated later as the dual code of B-code in [29] using perfect one-factorization of complete graphs.

VII. THE LIMITATION OF DEGREE TWO ENCODING SYMBOLS

In this section we analyze the limitation of array BP-XOR codes when only degree one and two encoding symbols are allowed. Using the results for array BP-XOR codes, we will get new results for edge-colored graph models.

For a $t$-erasure tolerating array BP-XOR code of size $m \times n$, we could achieve space and bandwidth optimal property if there are $k = (n-t)m$ information symbols of same length. The following theorem provides a necessary condition for the existence of array BP-XOR codes when only degree one and two encoding symbols are used.

Theorem 7.1: Let $C = \{a_{i,j} : (i,j) \in [1,m] \times [1,n]\}$ be a $t$-erasure tolerating array BP-XOR code with $k = (n-t)m$ information symbols and $C$ only use degree one and two encoding symbols. Assume that $n_0 = n - t > 2$, then we have

$$n \leq n_0 - 2 \left( n_0 - \frac{2}{(n_0 - 2)m + 1} \right).$$

Proof. By the fact that $C$ is $t$-erasure tolerating, each information symbol must occur in at least $t + 1$ columns. Since there are $n_0 m$ information symbols (data fragments) to encode, the total number of information symbol occurrences in $C$ is at least $n_0 m (t + 1)$.

In order for the BP decoding process to work, we must start from a degree one encoding symbol. Thus we need to have at least $t + 1$ degree one encoding symbols in distinct columns of $C$. This implies that we could use at most $mn - (t + 1)$ cells to hold encoding symbols for degree two. In another word, $C$ contains at most $2mn - (t + 1) + t + 1$ occurrences of information symbols. By the above fact, we must have

$$n_0 m (t + 1) \leq 2 (mn - (t + 1)) + t + 1.$$ 

By rearranging the terms, we get

$$n_0 mn - n_0 m (n_0 - 1) \leq 2 mn - (n_0 - 1).$$ 

If we move all terms to the right hand side, we get

$$n_0 (n_0 - 1)m - ((n_0 - 2)m + 1)n + (n_0 - 1) \geq 0.$$ 

\begin{table}[h]
\begin{tabular}{|c|c|}
\hline
Table VII & (p - 1)-COLOR CONNECTED EDGE-COLORED GRAPHS \\
\hline
$\langle v_1, v_{p-1} \rangle$ & $\cdots$  \\
$\langle v_2, v_{p-2} \rangle$ & $\cdots$  \\
$\cdots$ & $\cdots$  \\
$\langle v_{(p-1)/2}, v_{(p+1)/2} \rangle$ & $\cdots$  \\
\hline
\end{tabular}
\end{table}

\begin{table}[h]
\begin{tabular}{|c|c|}
\hline
Table VIII & (p - 1)/2 \times p BP-XOR CODE \\
\hline
$v_1 \oplus v_{p-1}$ & $\cdots$  \\
$v_2 \oplus v_{p-2}$ & $\cdots$  \\
$\cdots$ & $\cdots$  \\
$v_m \oplus v_{m+1}$ & $\cdots$  \\
\hline
$v_{p-1} \oplus v_{p-3}$ & $\cdots$  \\
$v_{p-4}$ & $\cdots$  \\
$\cdots$ & $\cdots$  \\
$v_{m-2} \oplus v_{m-1}$ & $\cdots$  \\
$\cdots$ & $\cdots$  \\
$v_{m-1} \oplus v_m$ & $\cdots$  \\
\hline
\end{tabular}
\end{table}
Finally, the above inequality could be rewritten as
\[(n_0 - 2)m + 1) \left( \frac{n_0(n_0 - 1) - n}{n_0 - 2} \right) \geq 2n_0 - 2\]
That is,
\[n \leq \frac{n_0 - 1}{n_0 - 2} \left( n_0 - \frac{2}{(n_0 - 2)m + 1} \right) \quad (1)\]
Based on equation (1), we get the necessary conditions for \(n\) for different \(n_0\) in Table IX. The values in Table IX show that for degree one and two encoding symbols based array BP-XOR codes, if we want to recover the information symbols from more than three columns (i.e., \(n_0 \geq 3\)) of encoding symbols, then we could only have one column redundancy for \(n \geq 4\).

Combining Theorem 7.2 and values in Table IX we get the following results for edge-colored graphs.

**Theorem 7.2:** For a color set \(C\) with \(|C| \geq 5\), if we want to design an edge-colored graph \(G(V, E, C, f)\) (or a network with more than \(|C|\) kinds of homogeneous devices) with minimum cost, then the edge-colored graph is robust against at most one color failures (or one brand of homogeneous devices failures).

**Proof.** Based on the results in Theorems 5.3, 7.1 and values in Table IX we can have the following conclusion: Given integers \(n_0, m, n\), an edge-colored graph \(G(V, E, C, f)\) with \(|C| = n, |V| = n_0 m + 1, |E| = nm - (n - n_0)\), \(G(V, E, C, f)\) is \((n - n_0)\)-color connected only if \(n = n_0 + 1\). Thus the theorem follows. □

\section*{VIII. Conclusion}

Based on the BP (Belief Propagation) decoding process and the edge-colored graph model \([27]\), we introduced flat BP-XOR codes and array BP-XOR codes. We have established the equivalence between edge-colored graphs and degree one and two based array BP-XOR codes. In particular, we used results in array BP-XOR codes to get new results in edge-colored graphs. For array BP-XOR codes with higher degree encoding symbols, we do not have general results yet. It would be interesting to have a complete characterization of the existence and bounds for array BP-XOR codes with higher degree encoding symbols. These characterizations may be used to design more efficient LT codes or digital fountain techniques. We have implemented an online software package for users to generate array BP-XOR codes with their own specification and to verify the validity of their array BP-XOR codes (see \([26]\)).

\section*{ACKNOWLEDGEMENTS}

I would like to thank Duan Qi for some discussion on Hamming code and Theorem \([4,2]\) and thank Prof. Doug Stinson and Yvo Desmedt, for some discussions on edge-colored graphs, Hamiltonian circuit, and factorization of complete graphs.

\section*{REFERENCES}

[1] S. Acedanski, S. Deb, M. Mdard, and R. Koeter. How good is random linear coding based distributed networked storage. In NetCod, 2005.
[2] B.A. Anderson. Symmetry groups of some perfect 1-factorizations of complete graphs. *Discrete Mathematics*, 18(3):227–234, 1977.
[3] M. Blaum, J. Brady, J. Bruck, and J. Menon. EVENODD: An efficient scheme for tolerating double disk failures in raid architectures. *IEEE Trans. Computers*, 44(2):192–202, 1995.
[4] M. Blaum, J. Bruck, and E. Vardy. MDS array codes with independent parity symbols. *IEEE Trans. on Information Theory*, 42:529–542, 1996.
[5] M. Blaum and R. M. Roth. New array codes for multiple phased burst correction. *IEEE Trans. on Information Theory*, 39(1):66–77, 1993.
[6] M. Blaum and R. M. Roth. On lowest-density MDS codes. *IEEE Trans. on Information Theory*, 45:46–59, 1999.
[7] N. Cao, S. Yu, Z. Yang, W. Lou, and T. Hou. Lt codes-based secure and reliable cloud storage service. In *Proceedings of INFOCOM*, 2012.
[8] Yuval Cassuto and Jehoshua Bruck. Cyclic lowest density mds array codes. *IEEE Trans. Inf. Theor.*, 55(4):1721–1729, April 2009.
[9] P. Corbett, R. English, A. Goel, T. Grcanca, S. Kleiman, J. Leong, and S. Sankar. Row-diagonal parity for double disk failure correction. In *FAST*, pages 1–14, 2004.
[10] A. Dimakis, P. Godfrey, Y. Wu, M. Wainwright, and K. Ramchandran. Network coding for distributed storage systems. *IEEE Trans. Inf. Theor.*, 56(9):4539–4551, 2010.
[11] A. Dimakis, K. Ramchandran, Y. Wu, and C. Suh. A survey on network codes for distributed storage. *CoRR*, abs/1004.4438, 2010.
[12] R. Dimakis, V. Prabhakaran, and K. Ramch. Decentralized erasure codes for distributed networked storage. *IEEE Trans. Inf. Theor.*, 55(8):3809–3816, 2006.
[13] P. Elias. Error-free coding. *Technical Report* 285, Massachusetts Institute of Technology (Boston), 1954.
[14] N. V. Semakov G. V. Zaitsev, V. A. Zinov’ev. Minimum-check-density codes for correcting bytes of errors, erasures, or defects. *Problems Inform. Transmission*, 19(3):197–204, 1983.
[15] R. G. Gallager. Low density Parity Check Codes. MIT Press, 1963.
[16] K. Greenan, X. Li, and J. Wylie. Flat xor-based erasure codes in storage systems: Constructions, efficient recovery, and tradeoffs. In *Proc. MSST*, pages 1–14. IEEE Computer Society, 2010.
[17] C. Huang and L. Xu. STAR: an efficient coding scheme for correcting triple storage node failures. In *FAST*, pages 197–210, 2005.
[18] M. Kobayashi. On perfect one-factorization of the complete graph \(K_n\). *Graphs and Combinatorics*, 5(1):351–353, 1989.
[19] Hugo Krawczyk. Distributed fingerprints and secure information dispersal. In *Proc. PODC*, pages 207–218. ACM Press, 1993.
[20] M. Luby. LT codes. In *Proc. FOCS*, pages 271–280, 2002.
[21] M. Luby, M. Mitzenmacher, M. Shokrollahi, D. Spelman, and V. Stein. Practical loss-resilient codes. In *Proc. 29th ACM STOC*, pages 150–159. ACM, 1997.
[22] J. Pearl. Probabilistic Reasoning in Intelligent Systems: Networks of Plausible Inference. Morgan Kaufmann, 1988.
[23] J. Plank and M. Thomason. A practical analysis of low-density parity-check erasure codes for wide-area storage applications. In *Factors*, pages 2015–2016. ACM, 1997.
[24] Michael O. Rabin. Efficient dispersal of information for security, load balancing, and fault tolerance. *J. ACM*, 36(2):335–348, April 1989.
[25] Tom Verhoeff. An updated table of minimum-distance bounds for binary linear codes. *IEEE Trans. Inf. Theor.*, 33(5):665–680, September 1987.
[26] Yongge Wang. Array BP-XOR Code Generation and Verification Webpage. [http://coitweb.uncc.edu/~ywong/wang/bpxor/](http://coitweb.uncc.edu/~ywong/wang/bpxor/), 2012.
[27] Yongge Wang and Yvo Desmedt. Edge-colored graphs with applications to homogeneous faults. *Inf. Process. Lett.*, 111(13):634–641, July 2011.
[28] H. Weatherspoon and J. Kubiatowicz. Erasure coding vs. replication: A quantitative comparison. In *Revised Papers from the First IPTPS’01*, pages 328–338. Springer-Verlag, 2002.
[29] L. Xu, V. Bohossian, J. Bruck, and D. Wagner. Low density mds codes and factors of complete graphs. *IEEE Trans. Inf. Theor.*, 45:1817–1826, 1998.
[30] L. Xu and J. Bruck. X-code: Mds array codes with optimal encoding. 
IEEE Trans. on Information Theory, 45:272–276, 1999.