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# RAPTOR CODE BASED SECURE STORAGE IN CLOUD COMPUTING

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**Abstract:**-The end of this decade is marked by a paradigm shift of the industrial information technology towards a pay-per-use service business model known as cloud computing. Cloud data storage redefines the security issues targeted on customer's outsourced data. To ensure the correctness of users' data in the cloud, we propose an effective and flexible distributed scheme with two salient features, opposing to its predecessors. By utilizing the homomorphic token with distributed verification of raptor coded data, our scheme achieves the integration of storage correctness insurance and data error localization, i.e., the identification of misbehaving server(s). Using this new scheme it further support security and dynamic operations on data block. Our result shows that, our proposed model provides a secure storage for data in cloud.

**Keywords:** *raptor code, homomorphic token, dependable distributed storage, error localization, data dynamics.*

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## I. INTRODUCTION

Several trends are opening up the era of Cloud Computing, which is an Internet-based development and use of computer technology. The ever cheaper and more powerful processors together with the software as a service (SaaS) computing architecture, are transforming data centers into pools of computing service on a huge scale. The increasing network bandwidth and reliable yet flexible network connections make it even possible that users can now subscribe high quality services from data and software that reside solely on remote data centers.

Moving data into the cloud offers great convenience to users since they don't have to care about the complexities of direct hardware management. The pioneer of Cloud Computing vendors, Amazon Simple Storage Service (S3) and Amazon Elastic Compute Cloud (EC2) [1] are both well known examples. While these internet-based online services do provide huge amounts of storage space and customizable computing resources, this computing platform shift, however, is eliminating the responsibility of local machines for data maintenance at the same time. As a result, users are at the mercy of their cloud service providers for the availability and integrity of their data.

From the perspective of data security, which has always been an important aspect of quality of service, Cloud Computing inevitably poses new challenging security threats for number of reasons. Firstly, traditional cryptographic primitives for the purpose of data security protection cannot be directly adopted due to the users' loss control of data under Cloud Computing. Therefore, verification of correct data storage in the cloud must be conducted without explicit knowledge of the whole data. Considering various kinds of data for each user stored in the cloud

and the demand of long term continuous assurance of their data safety, the problem of verifying correctness of data storage in the cloud becomes even more challenging. Secondly, Cloud Computing is not just a third party data warehouse.

The data stored in the cloud may be frequently updated by the users, including insertion, deletion, modification, appending, reordering, etc. To ensure storage correctness under dynamic data update is hence of paramount importance. However, this dynamic feature also makes traditional integrity insurance techniques futile and entails new solutions. Last but not the least, the deployment of Cloud Computing is powered by data centers running in a simultaneous, cooperated and distributed manner. Individual user's data is redundantly stored in multiple physical locations to further reduce the data integrity threats. Therefore, distributed protocols for storage correctness assurance will be of most importance in achieving a robust and secure cloud data storage system in the real world. However, such important area remains to be fully explored in the literature.

Recently, the importance of ensuring the remote data integrity has been highlighted by the following research works [3]–[7]. These techniques, while can be useful to ensure the storage correctness without having users possessing data, cannot address all the security threats in cloud data storage, since they are all focusing on single server scenario and most of them do not consider dynamic data operations. As an complementary approach, researchers have also proposed distributed protocols [8]–[10] for ensuring storage correctness across multiple servers or peers. Again, none of these distributed schemes is aware of dynamic data operations. As a result, their applicability in cloud data storage can be drastically limited.

In this paper, we propose an effective and flexible distributed scheme with explicit dynamic data support to ensure the correctness of users' data in the cloud. We rely on rateless erasure correcting code in the file distribution preparation to provide redundancies and guarantee the data dependability. This construction drastically reduces the communication and storage overhead as compared to the traditional replication-based file distribution techniques. By utilizing the homomorphic token with distributed verification of rateless erasure-coded data, our scheme achieves the storage correctness insurance as well as data error localization: whenever data corruption has been detected during the storage correctness verification, our scheme can almost guarantee the simultaneous localization of data errors, i.e., the identification of the misbehaving server(s).

The rest of the paper is organized as follows. Section II introduces the system model, adversary model, our design goal and notations. Then we provide the detailed description of our scheme in Section III. Finally, Section IV gives the concluding remark of the whole paper.

## II. PROBLEM STATEMENT

### A. System Model

A representative network architecture for cloud data storage is illustrated in Figure 1. Three different network entities can be identified as follows:

- User: users, who have data to be stored in the cloud and rely on the cloud for data computation, consist of both individual consumers and organizations.
- Cloud Service Provider (CSP): a CSP, who has significant resources and expertise in building and managing distributed cloud storage servers, owns and operates live Cloud Computing systems.
- Third Party Auditor (TPA): an optional TPA, who has expertise and capabilities that users may not have, is trusted to assess and expose risk of cloud storage services on behalf of the users upon request.

In cloud data storage, a user stores his data through a CSP into a set of cloud servers, which are running in a simultaneous, cooperated and distributed manner. Data redundancy can be employed with technique of erasure-correcting code to further tolerate faults or server crash as user's data grows in size and importance. Thereafter, for application purposes, the user interacts with the cloud servers via CSP to access or retrieve his data. In some cases, the user may need to perform block level operations on his data. The most general forms of these operations we are considering are block update, delete, insert and append.

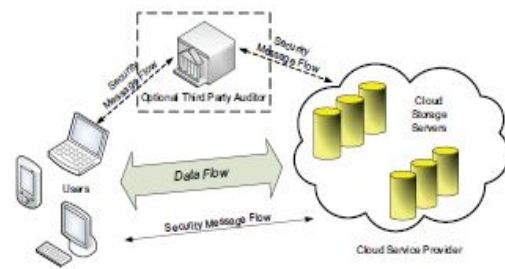


Fig. 1: Cloud data storage architecture

As users no longer possess their data locally, it is of critical importance to assure users that their data are being correctly stored and maintained. That is, users should be equipped with security means so that they can make continuous correctness assurance of their stored data even without the existence of local copies. In case that users do not necessarily have the time, feasibility or resources to monitor their data, they can delegate the tasks to an optional trusted TPA of their respective choices. In our model, we assume that the point-to-point communication channels between each cloud server and the user is authenticated and reliable, which can be achieved in practice with little overhead. Note that we don't address the issue of data privacy in this paper, as in Cloud Computing, data privacy is orthogonal to the problem we study here.

### B. Adversary Model

From user's perspective, the adversary model has to capture all kinds of threats toward his cloud data integrity. Because cloud data do not reside at user's local site but at CSP's address domain, these threats can come from two different sources: internal and external attacks. For internal attacks, a CSP can be self-interested, untrusted, and possibly malicious. Not only does it desire to move data that has not been or is rarely accessed to a lower tier of storage than agreed for monetary reasons, but it may also attempt to hide a data loss incident due to management errors, Byzantine failures, and so on. For external attacks, data integrity threats may come from outsiders who are beyond the control domain of CSP, for example, the economically motivated attackers. They may compromise a number of cloud data storage servers in different time intervals and subsequently be able to modify or delete users' data while remaining undetected by CSP.

Therefore, we consider the adversary in our model has the following capabilities, which captures both external and internal threats toward the cloud data integrity. Specifically, the adversary is interested in continuously corrupting the user's data files stored on individual servers. Once a server is comprised, an adversary can pollute the original data files by modifying or introducing its own fraudulent data to prevent the original data from being retrieved by the

user. This corresponds to the threats from external attacks. In the worst case scenario, the adversary can compromise all the storage servers so that he can intentionally modify the data files as long as they are internally consistent. In fact, this is equivalent to internal attack case where all servers are assumed colluding together from the early stages of application or service deployment to hide a data loss or corruption incident.

### C. Design Goals

To ensure the security and dependability for cloud data storage under the aforementioned adversary model, we aim to design efficient mechanisms for dynamic data verification and operation and achieve the following goals: (1) Storage correctness: to ensure users that their data are indeed stored appropriately and kept intact all the time in the cloud. (2) Fast localization of data error: to effectively locate the malfunctioning server when data corruption has been detected. (3) Dynamic data support: to maintain the same level of storage correctness assurance even if users modify, delete or append their data files in the cloud. (4) Dependability: to enhance data availability against Byzantine failures, malicious data modification and server colluding attacks, i.e. minimizing the effect brought by data errors or server failures. (5) Lightweight: to enable users to perform storage correctness checks with minimum overhead.

## III. ENSURING CLOUD DATA STORAGE

In cloud data storage system, users store their data in the cloud and no longer possess the data locally. Thus, the correctness and availability of the data files being stored on the distributed cloud servers must be guaranteed. One of the key issues is to effectively detect any unauthorized data modification and corruption, possibly due to server compromise and/or random Byzantine failures. Besides, in the distributed case when such inconsistencies are successfully detected, to find which server the data error lies in is also of great significance, since it can be the first step to fast recover the storage errors.

To address these problems, our main scheme for ensuring cloud data storage is presented in this section. The first part of the section is devoted to a review of basic tools from coding theory that is needed in our scheme for file distribution across cloud servers. Then, the homomorphic token is introduced. The token computation function we are considering belongs to a family of universal hash function [11], chosen to preserve the homomorphic properties, which can be perfectly integrated with the verification of erasure-coded data [8] [12]. Subsequently, it is also shown how to derive a challenge response protocol for verifying the storage correctness as well as identifying misbehaving servers. Finally, the procedure for file retrieval and

error recovery based on erasure-correcting code is outlined.

### A. File Distribution Preparation

It is well known that rateless erasure-correcting code may be used to tolerate multiple failures in distributed storage systems. In cloud data storage, we rely on this technique mainly raptor code to disperse the data file  $F$  redundantly across a set of  $n = m+k$  distributed servers. Raptor codes pre-encode the source symbols using a fixed length block code, and then encode these new symbols with an LT code. The main advantage is that, for correct decoding, it is no longer necessary that the LT decoding succeeds for all the symbols. Thus, it is possible to use a simpler degree distribution that does not recover all the symbols but makes the decoding process faster.

The decoding algorithm is composed of two steps. The inner LT decoder returns a hard bit-reliability vector. This latter is processed by the outer LDPC decoder, based on belief propagation algorithm.

### B. Challenge Token Precomputation

In order to achieve assurance of data storage correctness and data error localization simultaneously, our scheme entirely relies on the pre-computed verification tokens. The main idea is as follows: before file distribution the user pre-computes a certain number of short verification tokens on individual vector  $G(j)$  ( $j \in \{1, \dots, n\}$ ), each token covering a random subset of data blocks. Later, when the user wants to make sure the storage correctness for the data in the cloud, he challenges the cloud servers with a set of randomly generated block indices. Upon receiving challenge, each cloud server computes a short "signature" over the specified blocks and returns them to the user.

The values of these signatures should match the corresponding tokens pre-computed by the user. Meanwhile, as all servers operate over the same subset of the indices, the requested response values for integrity check must also be a valid codeword determined by secret matrix  $P$ .

Suppose the user wants to challenge the cloud servers  $t$  times to ensure the correctness of data storage. Then, he must pre-compute  $t$  verification tokens for each  $G(j)$  ( $j \in \{1, \dots, n\}$ ), using a PRF  $f(\cdot)$ , a PRP  $\Phi(\cdot)$ , a challenge key  $K_{\text{chal}}$  and a master permutation key  $K_{\text{PRP}}$ . After token generation, the user has the choice of either keeping the pre-computed tokens locally or storing them in encrypted form on the cloud servers. In our case here, the user stores them locally to obviate the need for encryption and lower the bandwidth overhead during dynamic data operation which will be discussed shortly. The details of token generation are shown in Algorithm 1.

**Algorithm 1** Token Pre-computation

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1: procedure
2:   Choose parameters  $l, n$  and function  $f, \phi$ ;
3:   Choose the number  $t$  of tokens;
4:   Choose the number  $r$  of indices per verification;
5:   Generate master key  $K_{PRP}$  and challenge  $k_{chat}$ ;
6:   for vector  $G^{(j)}, j \leftarrow 1, n$  do
7:     for round  $i \leftarrow 1, t$  do
8:       Derive  $\alpha_i = f_{k_{chat}}(i)$  and  $k_{PRP}^{(i)}$  from  $K_{PRP}$ ;
9:       Compute  $v_i^{(j)} = \sum_{q=1}^r \alpha_i^q * G^{(j)}[\phi_{k_{PRP}^{(i)}}(q)]$ 
10:    end for
11:  end for
12:  Store all the  $v_i$ s locally.
13: end procedure

```

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**C. Correctness Verification and Error Localization**

Error localization is a key prerequisite for eliminating errors in storage systems. However, many previous schemes do not explicitly consider the problem of data error localization, thus only provide binary results for the storage verification. Our scheme outperforms those by integrating the correctness verification and error localization in our challenge-response protocol: the response values from servers for each challenge not only determine the correctness of the distributed storage, but also contain information to locate potential data error(s). Once the inconsistency among the storage has been successfully detected, we can rely on the pre-computed verification tokens to further determine where the potential data error(s) lies in. Algorithm 2 gives the details of correctness verification and error localization.

**Algorithm 2** Correctness Verification and Error Localization

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```

1: procedure CHALLENGE( $i$ )
2:   Recompute  $\alpha_i = f_{k_{chat}}(i)$  and  $k_{PRP}^{(i)}$  from  $K_{PRP}$ ;
3:   Send  $\{\alpha_i, k_{PRP}^{(i)}\}$  to all the cloud servers;
4:   Receive from servers:
    $\{R_i^{(j)} = \sum_{q=1}^r \alpha_i^q * G^{(j)}[\phi_{k_{PRP}^{(i)}}(q)] | 1 \leq j \leq n\}$ 
5:   for  $(j \leftarrow m+1, n)$  do
6:      $R^{(j)} \leftarrow R^{(j)} - \sum_{q=1}^r f_{k_j}(s_{I_q, j}) \cdot \alpha_i^q, I_q = \phi_{k_{PRP}^{(i)}}(q)$ 
7:   end for
8:   if  $((R_i^{(1)}, \dots, R_i^{(m)}) \cdot P = (R_i^{(m+1)}, \dots, R_i^{(n)}))$  then
9:     Accept and ready for the next challenge.
10:  else
11:    for  $(j \leftarrow 1, n)$  do
12:      if  $(R_i^{(j)} \neq v_i^{(j)})$  then
13:        return server  $j$  is misbehaving.
14:      end if
15:    end for
16:  end if
17: end procedure

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**D. File Retrieval and Error Recovery**

Since our layout of file matrix is systematic, the user can reconstruct the original file by downloading the data vectors from the first  $m$  servers, assuming that

they return the correct response values. Notice that our verification scheme is based on random spot-checking, so the storage correctness assurance is a probabilistic one. However, by choosing system parameters (e.g.,  $r, l, t$ ) appropriately and conducting enough times of verification, we can guarantee the successful file retrieval with high probability. On the other hand, whenever the data corruption is detected, the comparison of pre-computed tokens and received response values can guarantee the identification of misbehaving server(s), again with high probability, which will be discussed shortly. Therefore, the user can always ask servers to send back blocks of the  $r$  rows specified in the challenge and regenerate the correct blocks by rateless erasure correction, shown in Algorithm 3, as long as there are at most  $k$  misbehaving servers are identified. The newly recovered blocks can then be redistributed to the misbehaving servers to maintain the correctness of storage.

**Algorithm 3** Error Recovery

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```

1: procedure
   % Assume the block corruptions have been detected
   among
   % the specified  $r$  rows;
   % Assume  $s \leq k$  servers have been identified misbehaving
2:   Download  $r$  rows of blocks from servers;
3:   Treat  $s$  servers as erasures and recover the blocks.
4:   Resend the recovered blocks to corresponding servers.
5: end procedure

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**IV. CONCLUSION**

In this paper, we investigated the problem of data security in cloud data storage, which is essentially a distributed storage system. To ensure the correctness of users' data in cloud data storage, we proposed an effective and flexible distributed scheme with explicit dynamic data support, including block update, delete, and append. We rely on raptor code in the file distribution preparation to provide redundancy parity vectors and guarantee the data dependability. By utilizing the homomorphic token with distributed verification of raptor coded data, our scheme achieves the integration of storage correctness insurance and data error localization, i.e., whenever data corruption has been detected during the storage correctness verification across the distributed servers, we can almost guarantee the simultaneous identification of the misbehaving server(s). Through detailed security and performance analysis, we show that our scheme is highly efficient and resilient to Byzantine failure, malicious data modification attack, and even server colluding attacks. We believe that data storage

security in Cloud Computing, an area full of challenges and of paramount importance, is still in its infancy now, and many research problems are yet to be identified. We envision several possible directions for future research on this area. The most promising one we believe is a model in which public verifiability is enforced. Public verifiability, supported in [6] [4] [11], allows TPA to audit the cloud data storage without demanding users' time, feasibility or resources. An interesting question in this model is if we can construct a scheme to achieve both public verifiability and storage correctness assurance of dynamic data. Besides, along with our research on dynamic cloud data storage, we also plan to investigate the problem of fine-grained data error localization.

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