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Privacy Preserving Delegated Access Control in Public Clouds

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Abstract—Current approaches to enforce fine-grained access control on confidential data hosted in the cloud are based on fine-grained encryption of the data. Under such approaches, data owners are in charge of encrypting the data before uploading them on the cloud and re-encrypting the data whenever user credentials change. Data owners thus incur high communication and computation costs. A better approach should delegate the enforcement of fine-grained access control to the cloud, so to minimize the overhead at the data owners, while assuring data confidentiality from the cloud. We propose an approach, based on two layers of encryption, that addresses such requirement. Under our approach, the data owner performs a coarse-grained encryption, whereas the cloud performs a fine-grained encryption on top of the owner encrypted data. A challenging issue is how to decompose access control policies (ACP) such that the two layer encryption can be performed. We show that this problem is NP-complete and propose novel optimization algorithms. We utilize an efficient group key management scheme that supports expressive ACPs. Our system assures the confidentiality of the data and preserves the privacy of users from the cloud while delegating most of the access control enforcement to the cloud.

Index Terms—Privacy, identity, cloud computing, policy decomposition, encryption, access control

1 INTRODUCTION

Security and privacy represent major concerns in the adoption of cloud technologies for data storage. An approach to mitigate these concerns is the use of encryption. However, whereas encryption assures the confidentiality of the data against the cloud, the use of conventional encryption approaches is not sufficient to support the enforcement of fine-grained organizational access control policies (ACPs). Many organizations have today ACPs regulating which users can access which data; these ACPs are often expressed in terms of the properties of the users, referred to as identity attributes, using access control languages such as XACML. Such an approach, referred to as attribute-based access control (ABAC), supports fine-grained access control (FGAC) which is crucial for high-assurance data security and privacy. Supporting ABAC over encrypted data is a critical requirement in order to utilize cloud storage services for selective data sharing among different users. Notice that often user identity attributes encode private information and should thus be strongly protected from the cloud, very much as the data themselves.

Approaches based on encryption have been proposed for fine-grained access control over encrypted data [2], [3]. As shown in Fig. 1, those approaches group data items based on ACPs and encrypt each group with a different symmetric key. Users then are given only the keys for the data items they are allowed to access. Extensions to reduce the number of keys that need to be distributed to the users have been proposed exploiting hierarchical and other relationships among data items. Such approaches however have several limitations:

- As the data owner does not keep a copy of the data, whenever user dynamics changes, the data owner needs to download and decrypt the data, re-encrypt it with the new keys, and upload the encrypted data. The user dynamics refers to the operation of adding or revoking users. Notice also that this process must be applied to all the data items encrypted with the same key. This is inefficient when the data set to be re-encrypted is large.
- In order to issue the new keys to the users, the data owner needs to establish private communication channels with the users.
- The privacy of the identity attributes of the users is not taken into account. Therefore the cloud can learn sensitive information about the users and their organization.
- They are either unable or inefficient in supporting fine-grained ABAC policies.

Recently proposed approaches based on broadcast key management schemes [4], [5], [6] address some of the above limitations. We refer to these approaches as single layer encryption (SLE) approaches, since, like previous approaches, they require the data owner to enforce access control through encryption performed at the data owner. However, unlike previous approaches, SLE assures the privacy of the users and supports fine-grained ACPs.

However, while SLE addresses some limitations of previous approaches, it still requires the data owner to enforce all the ACPs by fine-grained encryption, both initially and subsequently after users are added/revoked. All these encryption activities have to be performed at the owner that thus incurs high communication and computation cost. For example, if a user is revoked, the owner...
must download from the cloud the data affected by this change, generate a new encryption key, re-encrypt the downloaded data with the new key, and then upload the re-encrypted data to the cloud. In this paper, we propose a new approach to address this shortcoming. The approach is based on two layers of encryption applied to each data item uploaded to the cloud. Under this approach, referred to as **two layer encryption** (TLE), the data owner performs a coarse grained encryption over the data in order to assure the confidentiality of the data from the cloud. Then the cloud performs fine grained encryption over the encrypted data provided by the data owner based on the ACPs provided by the data owner. It should be noted that the idea of two layer encryption is not new. However, the way we perform coarse and fine grained encryption is novel and provides a better solution than existing solutions based on two layers of encryption [7]. We elaborate in details on the differences between our approach and existing solutions in the related work section.

A challenging issue in the TLE approach is how to decompose the ACPs so that fine-grained ABAC enforcement can be delegated to the cloud while at the same time the privacy of the identity attributes of the users and confidentiality of the data are assured. In order to delegate as much access control enforcement as possible to the cloud, one needs to decompose the ACPs such that the data owner manages minimum number of attribute conditions in those ACPs that assures the confidentiality of data from the cloud. Each ACP should be decomposed to two sub ACPs such that the conjunction of the two sub ACPs result in the original ACP. The two layer encryption should be performed such that the data owner first encrypts the data based on the outer layer encryption and the cloud re-encrypts the encrypted data using the inner layer encryption. The two encryptions together enforce the ACP as users should perform two decryptions to access the data.

Example 1. We use the following running example where a hospital (Owner) supports fine-grained access control on electronic health records (EHRs) and makes these records available to hospital employees (Usrs) through a public cloud (cloud). Typical hospital employees includes Usrs playing different roles such as receptionist (rec), cashier (cas), doctor (doc), nurse (nur), pharmacist (pha), and system administrator (sys). An EHR document consists of data items including BillingInfo (BI), ContactInfo (CI), MedicationReport (MR), PhysicalExam (PE), LabReports (LR), Treatment Plan (TP) and so on. In accordance with regulations such as health insurance portability and accountability act (HIPAA), the hospital policies specify which users can access which data item(s). In our example system, there are four attributes, role (rec, cas, doc, nur, pha, sys), insurance plan, denoted as ip, (ACME, MedA, MedB, MedC), type (assistant, junior, senior) and year of service, denoted as yos, (integer).

The system has the ACP (“role = doc” ∧ “ip = 2-out-4”) ∨ (“role = doc” ∧ “yos ≥ 2”). The ACP can be decomposed as two sub ACPs “role = doc” and “ip = 2-out-4” ∧ “yos ≥ 2”. The data owner enforces the former by encrypting the data for the users satisfying the former and the cloud enforces the latter by re-encrypting the data owner encrypted data for the users satisfying the latter. Since the data owner handles the minimum number of attributes, the overhead of managing attributes at the data owner is reduced. The overhead is further reduced, as it has to enforce only subsets of complex policies and thus only needs to perform a coarse grained encryption to enforce the simplified policies. Since the cloud does not handle the condition “role = doc”, it cannot decrypt owner encrypted data and thus confidentiality is preserved. Notice that users should satisfy the original ACP to access the data by performing two decryptions. In this paper, we show that the problem of decomposing ACPs such that the data owner manages the minimum number of attribute conditions while at the same time assuring the confidentiality of the data in the cloud is NP-complete. We propose two optimization algorithms to find the near optimal set of attribute conditions and decompose each ACP into two sub ACPs.

The TLE approach has many advantages. When user dynamics changes, only the outer layer of the encryption needs to be updated. Since the outer layer encryption is performed at the cloud, no data transmission is required between the data owner and the cloud. Further, both the data owner and the cloud service utilize a broadcast key management scheme [8] whereby the actual keys do not need to be distributed to the users. Instead, users are given one or more secrets which allow them to derive the actual symmetric keys for decrypting the data.

The rest of the paper is organized as follows. Section 2 describes in detail the underlying building blocks used to construct our system. An overview of the TLE approach is given in Section 3. Section 4 provides a detailed treatment of the policy decomposition for the purpose of two layer encryption. Section 5 gives a detailed description of the TLE approach. Section 6 reports experimental results for policy decomposition algorithms and the SLE versus the TLE approaches. We briefly analyze the tradeoffs, the security and the privacy of the overall systems in Section 7. Section 8 discusses the related work and compare them with our approach. Finally, Section 9 concludes the paper providing future directions.

2 Building Blocks

In this section we first introduce broadcast encryption (BE) schemes [9], [10] and oblivious commitment based envelope (OCBE) protocols [11]. We present an abstract view of the
main algorithms of those protocols and then describe how we use them to build our privacy-preserving attribute based group key management (PP AB-GKM) scheme [8]. We then present an overview of the SLE approach [4], [5], [6] which is used as the base model for comparison with the TLE approach proposed in this paper.

2.1 Broadcast Encryption

Broadcast encryption [9] was introduced to solve the problem of how to efficiently encrypt a message and broadcast it to a subset of the users in a system. The subset of users can change dynamically. In the broadcast encryption literature, these users are called privileged and the non-authorized users revoked. We denote the set of users by \( U \), the set of revoked users \( R \). The set of privileged users is thus \( U \setminus R \). We set \( N = |U| \) and \( r = |R| \). While all users can get the encrypted message, only the privileged users can decrypt it. The most simplest broadcast encryption scheme simply consists of encrypting a message for each privileged user separately and then broadcasting all the encrypted messages. Obviously, this scheme is very inefficient as the message length is prohibitively large \( O(N - r) \).

We use an algorithm based on subset-cover algorithm that supports broadcast encryption with stateless users. The algorithm builds a binary tree and assigns users to the leaf nodes and thus results in a predefined user grouping. Each such group is called a subset. A user can be a member of several subsets. The cover, denoted by \( C \), is defined as the set of subsets that contains all the privileged users, that is, users in \( U \setminus R \). The subsets in the cover are disjoint and hence each privileged user belongs to only one subset.

**Definition 1 (Broadcast encryption).** A subset-cover based broadcast encryption scheme consists of the algorithms: Setup, GetSecKeys, GetCover, Broadcast, KeyDer and Decrypt. Each of these algorithms are described below:

- **Setup(\( \ell, N \))**: The server constructs a binary tree \( \Lambda \) where there are at least \( N \) leaf nodes for the security parameter \( \ell \) which denotes the bit length. Each node in \( \Lambda \) is either assigned a unique key whose length is decided by \( \ell \), or can computationally derive a unique key. The user \( u_i \), \( i = 1, 2, \ldots, N \), is assigned the \( i \)th leaf node.
- **GetSecKeys(\( u \))**: The server gives all the keys assigned to \( u \) in \( \Lambda \).
- **GetCover(\( U \setminus R \))**: Given the privileged user set \( U \setminus R \), the server outputs the cover \( C \), that is, the set of disjoint subsets that cover all the privileged users.
- **Broadcast(\( M, C \))**: The server generates a session key \( K \) and encrypts the message \( M \) with \( K \) and encrypts \( K \) with each key in the cover \( C \).
- **KeyDer(\( u, C \))**: The user \( u \) identifies its subset in the cover \( C \), outputs the key that decrypts the session key.
- **Decrypt(\( C, K \))**: It decrypts the encrypted message \( C \) with the key \( K \), to output the message \( M \).

Having defined the algorithms, we give a high-level description of the basic subset-cover technique. In the basic scheme, \( N \) users are organized as the leaves of a balanced binary tree of height \( \log N \). A unique secret is assigned to each vertex in the tree. Each user is given \( \log N \) secrets that correspond to the vertices along the path from its leaf node to the root node. In order to provide forward secrecy when a single user is revoked, the updated tree is described by \( \log N \) subtrees formed after removing all the vertices along the path from the user leaf node to the root node. To rekey, the server uses the \( \log N \) secrets corresponding to the roots of these subtrees. Many improved subset-cover based broadcast encryption algorithms have been proposed. In this work, we consider the complete subtree algorithm [10]. The complete subtree algorithm improves the basic technique for simultaneously revoking \( r \) users and describing the privileged users using \( r \log(N/r) \) subsets. Each user stores \( \log N \) keys.

2.2 Oblivious Commitment Based Envelope Protocols

The oblivious commitment based envelope protocols, proposed by Li and Li [11], provide a mechanism to obliviously deliver a message to the users who satisfy certain conditions. There are three entities in these protocols, a server \( Svr \), a user \( Usr \), and a trusted third party, called the identity provider (IdP). IdP issues to \( Usr \) identity tokens, expressed as Pedersen commitments [12], corresponding to the identity attributes of \( Usr \).

**Definition 2 (OCBE).** An OCBE protocol consists of the algorithms: Setup, GenCom and GetData. Each of these algorithms are described below:

- **Setup(\( \ell \))**: The \( IdP \) runs a Pedersen commitment setup protocol to generate the system parameters, a finite cyclic group \( G \) of large prime order \( p \), two generators \( g \) and \( h \) of \( G \). The size of \( p \) is dependent on the security parameter \( \ell \).
- **GenCom(\( x \))**: A \( Usr \) wants to commit to the value \( x \). It submits \( x \) to the \( IdP \). The \( IdP \) computes the Pedersen commitment \( c = g^x h^{\ell} \) where \( r \) is randomly chosen from \( \mathbb{F}_p \). The \( IdP \) digitally signs \( c \) and sends \( r, c \) and the signature of \( c \) to the \( Usr \).
- **GetData(\( c, cond, d \))**: The \( Usr \) sends the signed commitment \( c \) and indicates the \( Svr \)’s condition \( cond \) that it wants to satisfy. \( cond \) has the format “name predicate value” where the predicate can be \( \geq, \leq, <, >, \) or \( = \). For example, \( cond \) “\( age \geq 18 \)” and \( c \) are for the attribute \( age \). After an interactive session, the \( Svr \) encrypts the data \( d \) and sends the encrypted data, called envelope, to the \( Usr \). The \( Usr \) can decrypt and access the data only if it satisfies the condition.

As mentioned above, the OCBE protocols always guarantee the following properties:

- The \( Svr \) does not learn the identity attributes of the users.
- A \( Usr \) can open the envelope only if its committed attribute value satisfies the condition.
- A \( Usr \) cannot submit fake commitments in order to satisfy a condition as the commitments are signed by the \( IdP \).

2.3 Privacy Preserving Attribute Based Group Key Management

The privacy preserving attribute based group key management scheme uses the BE scheme introduced in Section 2.1 and the OCBE protocols introduced in Section 2.2. Such scheme combines the previous work on AB-GKM scheme [8], [13] and privacy preservation in Broadcast Group Key Management (BGKM) [4], [6].
The BGKM schemes are a special type of GKM scheme where the rekey operation is performed with a single broadcast without requiring the use of private communication channels. Unlike conventional GKM schemes, the BGKM schemes do not give users the private keys. Instead users are given a secret which is combined with public information to obtain the actual private keys. Such schemes have the advantage of requiring a private communication only once for the initial secret sharing. The subsequent rekeying operations are performed using one broadcast message. Further, in such schemes achieving forward and backward security requires only to change the public information and does not affect the secret shares given to existing users. In general, a BGKM scheme consists of the following five algorithms: Setup, SecGen, KeyGen, KeyDer, and ReKey. Our overall construction is based on the AB-GKM scheme \cite{8, 13} which is an expressive construct of the access control vector BGKM (ACV-BGKM) scheme \cite{4, 6}.

Before we present details of the PP AB-GKM protocol, we first define certain terms that are useful to describe the protocol. Attribute conditions and access control policies are formally defined as follows.

**Definition 3 (Attribute condition).** An attribute condition \( C \) is an expression of the form: “\( \text{name}_{\text{attr}}\text{op}t \)” where \( \text{name}_{\text{attr}} \) is the name of an identity attribute \( \text{attr} \), \( \text{op} \) is a comparison operator such as \( =, \lt, \gt, \leq, \geq, \neq, \text{ and } l \) is a value that can be assumed by the attribute \( \text{attr} \).

**Definition 4 (Access control policy).** An access control policy \( ACP \) is a tuple \((s, o)\), where \( o \) denotes a set of data items \\{\( D_1, \ldots, D_i \)\}; and \( s \) is a monotonic expression\(^1\) over a set of attribute conditions that must be satisfied by a \( \text{Usr} \) to have access to \( o \).

We denote the set of all attribute conditions as \( \text{ACB} \) and the set of all \( ACPs \) as \( \text{ACP} \).

Before providing a detailed description of the PP AB-GKM scheme, we present the intuition and abstract algorithms. A separate BGKM instance for each attribute condition is constructed. The \( ACP \) is embedded in an access structure \( T \). \( T \) is a tree in which the internal nodes represent threshold gates and the leaves represent BGKM instances for the attributes. \( T \) can represent any monotonic policy. The goal of the access tree is to allow the derivation of the group key for only the users whose attributes satisfy the access structure \( T \).

A high-level description of the access tree is as follows. Each threshold gate in the tree is described by its child nodes and a threshold value. The threshold value \( t_x \) of a node \( x \) specifies the number of child nodes that should be satisfied in order to satisfy the node. Each threshold gate is modeled as a Shamir secret sharing polynomial \cite{14} whose degree equals to one less than the threshold value. The root of the tree contains the group key and all the intermediate values are derived in a top-down fashion. A user who satisfies the access tree derives the group key in a bottom-up fashion.

Due to space constraints, we only provide the abstract algorithms of the PP AB-GKM scheme. As a convention, we use the format “\( <\text{protocol name}> :: <\text{algorithm}> \)” to refer to the constructs introduced earlier or elsewhere. Specifically, \( \text{BE:: <algorithm>>}, \text{OCB:: <algorithm>}, \text{ and ACV-BGKM:: <algorithm>} \) \cite{4} to refer to algorithms of \( \text{BE}, \text{OCB} \) and \( \text{ACV-BGKM} \) protocols respectively.

**Definition 5 (PP AB-GKM).** The PP AB-GKM scheme consists of five algorithms: Setup, SecGen, KeyGen, KeyDer and ReKey. An abstract description of these algorithms are given below:

- Setup(\( \ell, N, N_a \)): It takes the security parameter \( \ell \), the maximum group size \( N \), and the number of attribute conditions \( N_a \) as input and initializes the system. It invokes BE::Setup(\( \ell, N, (\ell) \)) and ACV-BGKM::Setup(\( \ell, N \)) algorithms.
- SecGen(\( \gamma \)): The secret generation algorithm gives a \( \text{Usr} \), \( 1 \leq j \leq N \) a set of secrets for each commitment \( \text{com}_i \), \( 1 \leq i \leq m \). It invokes BE::GetSecGen and OCBE::GetData algorithms.
- KeyGen(\( \text{ACP} \)): The key generation algorithm takes the access control policy \( \text{ACP} \), the input and outputs a symmetric key \( K \), a set of public information tuples \( \text{PI} \) and an access tree \( T \). It invokes BE::GetCover() and ACV-BGKM::KeyGen algorithms.
- KeyDer(\( \beta, \text{PI}, T \)): Given the set of identity attributes \( \beta \), the set of public information tuples \( \text{PI} \) and the access tree \( T \), the key derivation algorithm outputs the symmetric \( K \) only if the identity attributes in \( \beta \) satisfy the access structure \( T \). It invokes BE::KeyDer and ACV-BGKM::KeyDer algorithms.
- ReKey(\( \text{ACP} \)): The rekey algorithm is similar to the KeyGen algorithm. It is executed whenever the dynamics in the system change.

### 2.4 Single Layer Encryption Approach

The SLE scheme \cite{6} consists of the four entities, \( \text{Owner}, \text{Usr}, \text{IdP} \) and \( \text{cloud} \). They play the following roles:

- **Owner**, the data owner defines \( \text{ACPs} \), and uploads encrypted data to the \( \text{Cloud} \), the cloud storage service.
- **The Cloud** hosts the encrypted data of the \( \text{Owner} \).
- **IdP**, the identity provider, a trusted third party, issues identity tokens to users based on the attribute attributes that users have. An identity token is a signed Pedersen commitment that binds the identity attribute value to a \( \text{Usr} \) while hiding it from others. There can be one or more certified \( \text{IdPs} \). We assume that all \( \text{IdPs} \) issue identity tokens in the same format.
- **Usr**, the user, uses one or more identity tokens to gain access to the encrypted data hosted in the cloud.

As shown in Fig. 2, the SLE approach follows the conventional data outsourcing scenario where the \( \text{Owner} \) enforces all \( \text{ACPs} \) through selective encryption and uploads encrypted data to the untrusted \( \text{cloud} \). The system goes through five different phases. We give an overview of the five phases below:

- **Identity token issuance.** \( \text{IdPs} \) issue identity tokens to \( \text{Usrs} \) based on their identity attributes.
- **Identity token registration.** \( \text{Usrs} \) register all their identity tokens to obtain secrets in order to later decrypt the data that they are allowed to access.

**Data encryption and uploading.** Based on the secrets issued and the \( \text{ACPs} \), the \( \text{Owner} \) encrypts the data using the keys
generated using the AB-GKM::KeyGen algorithm and uploads to the cloud.

Data downloading and decryption. Users download encrypted data from the cloud and decrypt using the key derived from the AB-GKM::KeyDer algorithm.

Encryption evolution management. Over time, either access control policies or user credentials may change. Further, already encrypted data may go through frequent updates. In such situations, it may be required to re-encrypt already encrypted data. The Owner alone is responsible to perform such re-encryptions. The Owner downloads all affected data from the cloud, decrypts them and then follows the data encryption and upload step.

3 OVERVIEW

We now give an overview of our solution to the problem of delegated access control to outsourced data in the cloud. A detailed description is provided in Section 5. Like the SLE system described in Section 2.4, the TLE system consists of the four entities, Owner, User, IdP and cloud. However, unlike the SLE approach, the Owner and the cloud collectively enforce ACPs by performing two encryptions on each data item. This two layer encryption allows one to reduce the load on the Owner and delegates as much access control enforcement duties as possible to the cloud. Specifically, it provides a better way to handle data updates, and user dynamics changes. Fig. 3 shows the system diagram of the TLE approach. The system goes through one additional phase compared to the SLE approach. We give an overview of the six phases below:

Identity token issuance. IdPs issue identity tokens to Users based on their identity attributes.

Policy decomposition. The Owner decomposes each ACP into at most two sub ACPs such that the Owner enforces the minimum number of attributes to assure confidentiality of data from the cloud. It is important to make sure that the decomposed ACPs are consistent so that the sub ACPs together enforce the original ACPs. The Owner enforces the confidentiality related sub ACPs and the cloud enforces the remaining sub ACPs.

Identity token registration. Users register their identity tokens in order to obtain secrets to decrypt the data that they are allowed to access. Users register only those identity tokens related to the Owner’s sub ACPs and register the remaining identity tokens with the cloud in a privacy preserving manner. It should be noted that the cloud does not learn the identity attributes of Users during this phase.

Data encryption and uploading. The Owner first encrypts the data based on the Owner’s sub ACPs in order to hide the content from the cloud and then uploads them along with the public information generated by the AB-GKM::KeyGen algorithm and the remaining sub ACPs to the cloud. The cloud in turn encrypts the data based on the keys generated using its own AB-GKM::KeyGen algorithm. Note that the AB-GKM::KeyGen at the cloud takes the secrets issued to Users and the sub ACPs given by the Owner into consideration to generate keys.

Data downloading and decryption. Users download encrypted data from the cloud and decrypt the data using the derived keys. Users decrypt twice to first remove the encryption layer added by the cloud and then by the Owner. As access control is enforced through encryption, Users can decrypt only those data for which they have valid secrets.

Encryption evolution management. Over time, user credentials may change. Further, already encrypted data may go through frequent updates. In such situations, data already encrypted must be re-encrypted with a new key. As the cloud performs the access control enforcing encryption, it simply re-encrypts the affected data without the intervention of the Owner.

4 POLICY DECOMPOSITION

Recall that in the SLE approach, the Owner incurs a high communication and computation overhead since it has to manage all the authorizations when user dynamics change. If the access control related encryption is somehow delegated to the cloud, the Owner can be freed from the responsibility of managing authorizations through re-encryption and the overall performance would thus improve. Since the cloud is not trusted for the confidentiality of the outsourced data, the Owner has to initially encrypt the data and upload the encrypted data to the cloud. Therefore, in order for the cloud to allow to enforce authorization policies through encryption and avoid re-encryption by the Owner, the data may have to be encrypted again to have two encryption layers. We call the two encryption layers as inner encryption
layer (IEL) and outer encryption later (OEL). IEL assures the confidentiality of the data with respect to the cloud and is generated by the Owner. The OEL is for fine-grained authorization for controlling accesses to the data by the users and is generated by the cloud.

An important issue in the TLE approach is how to distribute the encryptions between the Owner and the cloud. There are two possible extremes. The first approach is for the Owner to encrypt all data items using a single symmetric key and let the cloud perform the complete access control related encryption. The second approach is for the Owner and the cloud to perform the complete access control related encryption twice. The first approach has the least overhead for the Owner as the Owner does not manage any attributes and perform fine-grained access control related encryption. However it has the highest information exposure risk due to collisions between Usrs and the cloud as one malicious Usr revealing the Owner’s encryption key exposes all sensitive data to the cloud. Further, IEL updates require re-encrypting all data items. The second approach has the least information exposure risk due to collisions as the fine grained access control is enforced in the first encryption. However, it has the highest overhead on the Owner as the Owner has to perform the same task initially as in the SLE approach and, further, needs to manage all identity attributes. An alternative solution is based on decomposing ACPs so that the information exposure risk and key management overhead are balanced. The problem is then how to decompose the ACPs such that the Owner has to manage the minimum number of attributes while delegating as much access control enforcement as possible to the cloud without allowing it to decrypt the data. In what follow we propose such an approach to decompose and we also show that the policy decomposition problem is hard.

### 4.1 Policy Cover

We define the policy cover problem as the optimization problem of finding the minimum number of attribute conditions that “covers” all the ACPs in the ACPB. We say that a set of attribute conditions covers the ACPB if in order to satisfy any ACP in the ACPB, it is necessary that at least one of the attribute conditions in the set is satisfied. We call such a set of attribute conditions as the attribute condition cover. For example, if ACPB consists of the three simple ACPs \{“role = doc”\land “ip = 2-out-4”, “role = doc”\land “yos \geq 2”, “role = nur”\}, the minimum set of attributes that covers ACPB is \{“role = doc”, “role = nur”\}. “role = doc” should be satisfied in order to satisfy the first two ACPs. Notice that satisfying “role = doc” is not sufficient to satisfy the ACPs. The set is minimum since the set obtained by removing either “role = doc” or “role = nur” does not satisfy the cover relationship. While one can compute the minimum cover for simple examples such as the above one, below we show that it is a hard problem in general and there is no polynomial time algorithm to do. We define the related decision problem as follows.

**Definition 6 (POLICY-COVER).** Determine whether ACPB has a cover of k attribute conditions.

The following theorem states that this problem is NP-complete.

**Theorem 1.** The POLICY-COVER problem is NP-complete.

**Proof.** We first show that POLICY-COVER \(\in\) NP. Suppose that we are given a set of ACPs \(ACP\) which contains the attribute condition set \(AC\), and integer \(k\). For simplicity, we assume that each ACP is a conjunction of attribute conditions. However, the proof can be trivially extended to ACPs having any monotonic Boolean expression over attribute conditions. The certificate we choose has a cover of attribute conditions \(AC' \subset AC\). The verification algorithm affirms that \(|AC'| = k\), and then it checks, for each policy in the ACPB, that at least one attribute condition in \(AC'\) is in the policy. This verification can be performed trivially in polynomial time. Hence, POLICY-DECOM is NP.

Now we prove that the POLICY-COVER problem is NP-hard by showing that the vertex cover problem, which is NP-Complete, is polynomial time reducible to the POLICY-COVER problem. Given an undirected graph \(G = (V, E)\) and an integer \(k\), we construct a set of ACPs \(ACP\) that has a cover set of size \(k\) if and only if \(G\) has a vertex cover of size \(k\).

Suppose \(G\) has a vertex cover \(V' \subset V\) with \(|V'| = k\). We construct a set of ACPs \(ACP\) that has a cover of \(k\) attribute conditions as follows. For each vertex \(v_i \in V\), we assign an attribute condition \(C_i\). For each vertex \(v_j \in V'\), we construct an access control policy by obtaining the conjunction of attribute conditions as follows.

- Start with the attribute condition \(C_j\) as the ACP \(P_j\).
- For each edge \((v_j, v_k)\), add \(C_k\) to the ACP as a conjunctive literal (for example, if the edges are \((v_j, v_{a}), (v_j, v_{b})\) and \((v_j, v_{c})\), we get \(P_j = C_j \land C_a \land C_b \land C_c\)).

At the end of the construction we have a set of distinct access control policies \(ACP\) with size \(k\). We construct the attribute condition set \(AC = \{C_1, C_2, \ldots, C_k\}\) such that \(C_i\) corresponds to each vertex in \(V'\). In order to satisfy all access control policies, the attribute conditions in \(AC\) must be satisfied. Hence, \(AC\) is an attribute condition cover of size \(k\) for the ACPs \(ACP\).

Conversely, suppose that \(ACP\) has an attribute condition cover of size \(k\). We construct \(G\) such that each attribute condition corresponds to a vertex in \(G\) and an edge between \(v_i\) and \(v_j\) if they appear in the same access control policy. Let this vertex set be \(V_i\). Then we add the remaining vertices to \(G\) corresponding to other attribute conditions in the access control policies and add the edges similarly. Since the access control policies are distinct there will be at least one edge \((v_i, u)\) for each vertex \(v_i\) in attribute condition cover such that \(u \notin V_i\). Hence \(G\) has a vertex cover of size \(V_1 = k\).

Since the POLICY-COVER problem is NP-complete, one cannot find a polynomial time Algorithm for finding the minimum attribute condition cover. In the following section we present two approximation algorithms for the problem.

The random cover Algorithm 2 takes as input the set of ACPs \(ACP\) and returns a set of attribute conditions whose size is guaranteed to be no more than twice the size of an optimal attribute condition cover. Random cover utilizes the GEN-GRAH Algorithm 1 to first represent \(ACP\) as a graph.
We give a high-level overview of the GEN-GRAPH Algorithm 1. It takes the ACPB as the input and converts each ACP into DNF (disjunctive normal form). The unique conjunctive terms are added to the set $C$. For each attribute condition in each conjunctive term in $C$, it creates a new vertex in $G$ and adds edges between the vertices corresponding to the same conjunctive term. Depending on the ACPs, the algorithm may create a graph $G$ with multiple disconnected subgraphs.

Algorithm 1 GEN-GRAPH

1. $C = \emptyset$
2. for Each $ACP_i \in ACPB$, $i = 1$ to $N_p$ do
3.   $ACP'_i \leftarrow$ Convert $ACP_i$ to DNF
4.   for Each conjunctive term $c$ of $ACP'_i$ do
5.     Add $c$ to $C$
6. end for
7. end for
8. //Represent the conditions as a graph
9. $G = (E, V)$, $E = \phi$, $V = \phi$
10. for Each conjunctive term $c_i \in C$, $i = 1$ to $N_c$ do
11.   Create vertex $v$, if $v \notin V$, for each $AC$ in $c_i$
12.   Add an edge $e_i$ between $v$, and each vertex already added for $c_i$
13. end for
14. Return $G$

As shown in the random cover Algorithm 2, it takes the ACPB as the input and outputs a near-optimal attribute condition cover $ACC$. First the algorithm converts the ACPB to a graph $G$ as shown in the GEN-GRAPH Algorithm 1. Then for each disconnected subgraph $G_i$ of $G$, it finds the near optimal attribute condition cover and add to the $ACC$. The attribute condition to be added is related at random by selecting a random edge in $G_i$. Once an edge is considered, all its incident edges are removed from $G_i$. The algorithm continues until all edges are removed from each $G_i$. The running time of the algorithm is $O(V + E)$ using adjacency lists to represent $G$. It can be shown that the random cover Algorithm is a polynomial-time 2-approximation algorithm.

We now present the idea behind our second approximation algorithm, greedy cover, which uses a heuristic to select the attribute conditions. This algorithm is similar to the random cover Algorithm 2 except that instead of randomly selecting the edges to be included in the cover, it selects the vertex of highest degree and removes all of its incident edges.

We now illustrate the idea behind random cover algorithm using the same running example. The following is the complete set of ACPs of the hospital.

As shown in the random cover Algorithm 2, it takes the ACPB as the input and outputs a near-optimal attribute condition cover $ACC$. First the algorithm converts the ACPB to a graph $G$ as shown in the GEN-GRAPH Algorithm 1. Then for each disconnected subgraph $G_i$ of $G$, it finds the near optimal attribute condition cover and add to the $ACC$. The attribute condition to be added is related at random by selecting a random edge in $G_i$. Once an edge is considered, all its incident edges are removed from $G_i$. The algorithm continues until all edges are removed from each $G_i$. The running time of the algorithm is $O(V + E)$ using adjacency lists to represent $G$. It can be shown that the random cover Algorithm is a polynomial-time 2-approximation algorithm.

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As mentioned in the greedy cover Algorithm, single vertex graphs are trivially included in the $ACC$. It then picks "role = nur" as the node has the highest number of incident edges and removes all the incident edges from $G_i$.
the graph. From the remaining graph, it similarly picks “role = dat” and “role = doc”. At this point, all attribute conditions are covered and the greedy algorithm outputs the final answer.

4.2 Policy Decomposition
The Owner manages only those attribute conditions in ACC. The cloud handles the remaining set of attribute conditions, ACP/ACC. The Owner re-writes its ACPs such that they cover ACC. In other words, the Owner enforces the parts of the ACPs related to the ACs in ACC and cloud enforces the remaining parts of the policies along with some ACs in ACC. The POLICY-DECOMPOSITION Algorithm 3 shows how the ACPs are decomposed into two sub ACPs based on the attribute conditions in ACC.

Algorithm 3 POLICY-DECOMPOSITION

1. $ACP_{Owner} = \phi$
2. $ACP_{Cloud} = \phi$
3. for Each $ACP_i$ in $ACP_{Owner}$ do
4. Convert $ACP_i$ to DNF
5. $ACP_i(\text{owner}) = \phi$
6. $ACP_i(\text{cloud}) = \phi$
7. if Only one conjunctive term then
8. Decompose the conjunctive term $c_i$ into $c_1$ and $c_2$ such that ACs in $c_1 \in ACC$, ACs in $c_2 \notin ACC$ and $c = c_1 \land c_2$
9. $ACP_i(\text{owner}) = c_1$
10. $ACP_i(\text{cloud}) = c_2$
else if At most one term has more than one AC then
12. for Each single AC term $c$ of $ACP_i$ do
13. $ACP_i(\text{owner}) \lor = c$
14. $ACP_i(\text{cloud}) \lor = c$
15. end for
16. Decompose the multi AC term $c_i$ into $c_1$ and $c_2$ such that ACs in $c_1 \in ACC$, ACs in $c_2 \notin ACC$ and $c = c_1 \land c_2$
17. $ACP_i(\text{owner}) \lor = c_1$
18. $ACP_i(\text{cloud}) \lor = c_2$
else
20. for Each conjunctive term $c$ of $ACP_i$ do
21. Decompose $c$ into $c_1$ and $c_2$ such that ACs in $c_1 \in ACC$, ACs in $c_2 \notin ACC$ and $c = c_1 \land c_2$
22. $ACP_i(\text{owner}) \lor = c_1$
23. end for
24. $ACP_i(\text{cloud}) = ACP_i$
25. end if
26. Add $ACP_i(\text{owner})$ to $ACP_{Owner}$
27. Add $ACP_i(\text{cloud})$ to $ACP_{Cloud}$
28. end for
29. Return $ACP_{Owner}$ and $ACP_{Cloud}$

Algorithm 3 takes the $ACP_{Owner}$ and $ACP_{Cloud}$ as input and produces the two sets of ACPs $ACP_{Owner}$ and $ACP_{Cloud}$ that are to be enforced at the Owner and the cloud respectively. It first converts each policy into DNF and decomposes each conjunctive term into two conjunctive terms such that one conjunctive term has only those ACs in ACC and the other term may or may not have the ACs in ACC. It can be easily shown that the policy decomposition is consistent. That is, the conjunction of corresponding sub ACPs in $ACP_{Owner}$ and $ACP_{Cloud}$ respectively produces an original ACP in $ACP_{Owner}$.

Example 2. For our example ACPs, the Owner handles the following sub ACPs.

- ("role = rec" $\lor$ "role = nur", CI)
- ("role = cas" $\lor$ "role = pha", BI)
- ("role = doc", CR)
- ("role = doc" $\lor$ "role = pha", TR)
- ("role = doc" $\lor$ "role = nur" $\lor$ "role = pha", MR)
- ("role = nur" $\lor$ "role = dat" $\lor$ "role = doc", LR)
- ("role = nur" $\lor$ "role = dat", PE)

As shown in Algorithm 3, the Owner re-writes the ACPs that the cloud should enforce such that the conjunction of the two decomposed sub ACPs yields an original ACP. In our example, the sub ACPs that the cloud enforces look like follows.

- ("role = rec" $\lor$ "type $\geq$ junior", CI)
- ("role = cas" $\lor$ "role = pha", BI)
- ("ip = 2-out-4", CR)
- ("ip = 2-out-4" $\lor$ "role = pha", TR)
- ("role = doc" $\land$ "ip = 2-out-4") $\lor$ ("role = nur" $\land$ "yos $\geq$ 5") $\lor$ ("role = pha", MR)
- ("role = nur" $\land$ "type $\geq$ junior") $\lor$ ("role = dat" $\land$ "type $\geq$ junior") $\lor$ ("role = doc" $\land$ "yos $\geq$ 2") $\lor$ ("role = nur" $\land$ "type = senior") $\lor$ ("role = dat" $\land$ "yos $\geq$ 4") $\lor$ ("role = dat", PE)

5 TWO LAYER ENCRYPTION APPROACH

In this section, we provide a detailed description of the six phases of the TLE approach introduced in Section 3. The system consists of the four entities, Owner, Usr, IdP and cloud. Let the maximum number of users in the system be $N$, the current number of users be $n (n < N)$, and the number of attribute conditions $N_a$.

5.1 Identity Token Issuance

IdPs are trusted third parties that issue identity tokens to Usrs based on their identity attributes. It should be noted that IdPs need not be online after they issue identity tokens. An identity token, denoted by $IT$ has the format $\{nym, id-tag, c, \sigma\}$, where $nym$ is a pseudonym uniquely identifying a Usr in the system, $id-tag$ is the name of the identity attribute, $c$ is the Pedersen commitment for the identity attribute value $x$ and $\sigma$ is the IdP’s digital signature on $nym$, id-tag and $c$.

5.2 Policy Decomposition

Using the policy decomposition Algorithm 3, the Owner decomposes each ACP into two sub ACPs such that the Owner enforces the minimum number of attributes to assure confidentiality of data from the cloud. The algorithm produces two sets of sub ACPs, $ACP_{Owner}$ and $ACP_{Cloud}$. The Owner enforces the confidentiality related sub ACPs in $ACP_{Owner}$ and the cloud enforces the remaining sub ACPs in $ACP_{Cloud}$.
5.3 Identity Token Registration

Users register their TTs to obtain secrets in order to later decrypt the data they are allowed to access. Users register their TTs related to the attribute conditions in \( AC^C \) with the owner, and the rest of the identity tokens related to the attribute conditions in \( AC^B/AC^C \) with the cloud using the AB-GKM::SecGen algorithm.

When users register with the owner, the owner issues them two sets of secrets for the attribute conditions in \( AC^C \) that are also present in the sub-ACP in \( AC^P_{cloud} \). The owner keeps one set and gives the other set to the cloud. Two different sets are used in order to prevent the cloud from decrypting the owner encrypted data.

5.4 Data Encryption and Upload

The owner encrypts the data based on the sub-ACPs in \( AC^P_{owner} \) and uploads them along with the corresponding public information tuples to the cloud. The cloud in turn encrypts the data again based on the sub-ACPs in \( AC^P_{cloud} \). Both parties execute AB-GKM::KeyGen algorithm individually to first generate the symmetric key, the public information tuple \( PI \) and access tree \( T \) for each sub-ACP. We now give a detailed description of the encryption process.

The owner arranges the sub-ACPs such that each data item has a unique ACP. Note that the same policy may be applicable to multiple data items. Assume that the set of data items \( D = \{d_1, d_2, \ldots, d_n\} \) and the set of sub-ACPs \( AC^P_{owner} = \{ACP_1, ACP_2, \ldots, ACP_n\} \). The owner assigns a unique symmetric key, called an ILE key, \( K_i^{ILE} \) for each sub-ACP \( i \in AC^P_{owner} \) encrypts all related data with that key and executes the AB-GKM::KeyGen to generate the public \( PI \) and \( T_i \), the owner uploads those encrypted data \((id, E_{K^{ILE}}(d_i), i)\) along with the indexed public information tuples \((i, PI, T_i)\), where \( i = 1, 2, \ldots, n \) to the cloud. The cloud handles the key management and encryption based access control for the ACPs in \( AC^P_{cloud} \). For each sub-ACP \( i \in AC^P_{cloud} \), the cloud assigns a unique symmetric key \( K_i^{ILE} \) and produces the tuple \((id, E_{K^{ILE}}(d_i)), i, j)\), where \( i \) and \( j \) gives the index of the public information generated by the owner and the cloud respectively.

5.5 Data Downloading and Decryption

Users download encrypted data from the cloud and decrypt twice to access the data. First, the cloud generates public information tuple is used to derive the OLE key and then the owner generates public information tuple is used to derive the ILE key using the AB-GKM::KeyGen algorithm. These two keys allow a user to decrypt a data item only if the user satisfies the original ACP applied to the data item.

For example, in order to access a data item \( d_i \), users download the encrypted data item \( E_{K^{OLE}}(E_{K^{ILE}}(d_i)) \) and the corresponding two public information tuples \( PI_i \) and \( PL_i \), \( PL_j \) is used to derive the key of the outer layer encryption \( K^{OLE} \) and \( PL_i \) used to derive the key of the inner layer encryption \( K^{ILE} \). Once those two keys are derived, two decryption operations are performed to access the data item.

5.6 Encryption Evolution Management

After the initial encryption is performed, affected data items need to be re-encrypted with a new symmetric key if credentials are added/removed. Unlike the SLE approach, when credentials are added or revoked, the owner does not have to involve the cloud. The cloud generates a new symmetric key and re-encrypts the affected data items. The cloud follows the following conditions in order to decide if re-encryption is required:

1. For any ACP, the new group of users is a strict subset of the old group of users, and backward secrecy is enforced.
2. For any ACP, the new group of users is a strict subset of the old group of users, and forward secrecy is enforced for the already encrypted data items.

When ACPs are modified, the owner has to perform the policy decomposition again. If the set \( AC^P_{owner} \) remains unchanged after the policy decomposition, the owner does not have to re-encrypt, and only needs to provide the updated \( AC^P_{cloud} \) to the cloud which enforces the new set of sub-ACPs by re-encrypting. Otherwise, both the owner and the cloud are required to re-encrypt all the affected data items.

6 Experimental Results

In this section we first present experimental results concerning the policy decomposition algorithms. We then present an experimental comparison between the SLE and TLE approaches.

The experiments were performed on a machine running GNU/Linux kernel version 2.6.32 with an Intel Core 2 Duo CPU T9300 2.50 GHz and 4 Gbytes memory. Only one processor was used for computation. Our prototype system is implemented in C/C++. We use V. Shoup’s NTL library [15] version 5.4.2 for finite field arithmetic, and SHA-1 and AES-256 implementations of OpenSSL [16] version 1.0.0d for cryptographic hashing and incremental encryption. We use boost library [17] version 0.1.13 to convert policies into DNF. Adjacency list representation is used to construct policy graphs used in the two approximation algorithms for finding a near optimal attribute condition cover.

We utilized the AB-GKM scheme with the subset cover optimization. We used the complete subset algorithm introduced by Naor et al. [10] as the subset cover. We assumed that 5 percent of attribute credentials are revoked for the AB-GKM related experiments. All finite field arithmetic operations in our scheme are performed in an 512-bit prime field. We set the total attribute condition count between 100-1,500 and the attribute conditions per policy count between 2-20. We generate random Boolean expressions consisting of conjunctions and disjunctions as policies. Each term in the Boolean expression represents a attribute condition.

Fig. 5 shows the size of the attribute condition cover, that is, the number of attribute conditions the data owner enforces, for systems having 500 and 1,500 attribute conditions as the number of attribute conditions per policy is increased. In all experiments, the greedy policy cover algorithm
performs better than the random cover algorithm. As the number of attribute conditions per policy increases, the size of the attribute condition cover also increases. This is due to the fact that as the number of attribute conditions per policy increases, the number of distinct disjunctive terms in the DNF increases.

Fig. 6 shows the break down of the running time for the complete policy decomposition process. In this experiment, the number of attribute condition is set to \(100, 500, 1,000\) and the maximum number of attribute conditions per policy is set to 5. The total execution time is divided into the execution times of three different components of our scheme. The “DNF + Graph” time refers to the time required to convert the policies to DNF and construct an in-memory graph of policies using an adjacency list. The “Cover” time refers to the time required to find the optimal cover and the “Decompose” time refers to time required to create the updated policies for the data owner and the cloud based on the cover. As can be seen from the graphs, most of the time is spent on finding a near optimal attribute condition cover. It should be noted that the random approximation algorithm runs faster than the greedy algorithm. One reason for this behavior is that each time the latter algorithm selects a vertex it iterates through all the unvisited vertices in the policy graph, whereas the former algorithm simply picks a pair of unvisited vertices at random. Consistent with the worst-case running times, the “DNF + Graph” and “Decompose” components demonstrate near linear running time, and the “Cover” component shows a non-linear running time.

Fig. 7 reports the average time spent to execute the ABGKM::KeyGen with SLE and TLE approaches for different group sizes. We set the number of attribute conditions to 1,000 and the maximum number of attribute conditions per policy to 5. We utilize the greedy algorithm to find the attribute condition cover. As seen in the diagram, the running time at the Owner in the SLE approach is higher since the Owner has to enforce all the attribute conditions. Since the TLE approach divides the enforcement cost between the Owner and the cloud, the running time at the Owner is lower compared to the SLE approach. The running time at the cloud in the TLE approach is higher than that at the Owner since the cloud performs fine grained encryption whereas the Owner only performs coarse grained encryption. As shown in Fig. 8, a similar pattern is observed in the ABGKM::KeyDer as well.

7 Analysis

In this section, we first compare the SLE and the TLE approaches, and then give a high level analysis of the security and the privacy of both approaches.

7.1 SLE versus TLE

Recall that in the SLE approach, the Owner enforces all ACPs by fine-grained encryption. If the system dynamics change, the Owner updates the keys and encryptions.
The cloud merely acts as a storage repository. Such an approach has the advantage of hiding the ACPs from the cloud. Further, since the Owner performs all access control related encryptions, a Usr colluding with the cloud is unable to access any data item that is not allowed to access. However, the SLE approach incurs high overhead. Since the Owner has to perform all re-encryptions when user dynamics or policies change, the Owner incurs a high overhead in communication and computation. Further, it is unable to perform optimizations such as delayed AB-GKM::ReKey or re-encryption as the Owner has to download, decrypt, re-encrypt and re-upload the data, which could considerably increase the response time if such optimizations are to be performed.

The TLE approach reduces the overhead incurred by the Owner during the initial encryption as well as subsequent re-encryptions. In this approach, the Owner handles only the minimal set of attribute conditions and most of the key management tasks are performed by the cloud. Further, when identity attributes are added or removed, or the Owner updates the cloud’s ACPs, the Owner does not have to re-encrypt the data as the cloud performs the necessary re-encryptions to enforce the ACPs. Therefore, the TLE approach reduces the communication and computation overhead at the Owner. Additionally, the cloud has the opportunity to perform delayed encryption during certain dynamic scenarios as the cloud itself manages the OEL keys and encryptions. However, the improvements in the performance comes at the cost of security and privacy. In this approach, the cloud learns some information about the ACPs.

7.2 Security and Privacy
The SLE approach correctly enforces the ACPs through encryption. In the SLE approach, the Owner itself performs the attribute based encryption (ABE) based on ACPs. The AB-GKM scheme makes sure that only those Usrs who satisfy the ACPs can derive the encryption keys. Therefore, only the authorized Usrs are able to access the data.

The TLE approach correctly enforces the ACPs through two encryptions. Each ACP is decomposed into two ACPs such that the conjunction of them is equivalent to the original ACP. The Owner enforces one part of the decomposed ACPs through attribute based encryption. The cloud enforces the counterparts of the decomposed ACPs through another attribute based encryption. Usr can access a data item only if it can decrypt both encryptions. As the AB-GKM scheme makes sure that only those Usrs who satisfy these decomposed policies can derive the corresponding keys, a Usr can access a data item by decrypting twice only if it satisfies the two parts of the decomposed ACPs, that is, the original ACPs.

In both approaches, the privacy of the identity attributes of Usrs is assured. Recall that the AB-GKM::SecGen algorithm issues secrets to users based on the identity tokens which hide the identity attributes. Further, at the end of the algorithm neither the Owner nor the cloud knows if a Usr satisfies a given attribute condition. Therefore, neither the Owner nor the cloud learns the identity attributes of Usrs. Note that the privacy does not weaken the security as the AB-GKM::SecGen algorithm makes sure that Usrs can access the issued secrets only if their identity attributes satisfy the attribute conditions.

8 RELATED WORK
Fine-grained access control. Fine-grained access control allows one to enforce selective access to the content based on expressive policy specifications. Research in FGAC can be categorized into two dissemination models: push-based and pull-based models. Our work focuses on the pull-based model.

In the push-based approaches [2], [3] subdocuments are encrypted with different keys, which are provided to users at the registration phase. The encrypted subdocuments are then broadcasted to all users. However, such approaches require that all [2] or some [3] keys be distributed in advance during user registration phase. This requirement makes it difficult to assure forward and backward key secrecy when user groups are dynamic. Further, the rekey process is not transparent, thus shifting the burden of acquiring new keys on users. Shang et al. [4] proposes approach to solve such problem. It lays the foundation to make rekey transparent to users and protect the privacy of the users who access the content. However, it does not support expressive access control policies as in our approach and also it is not directly applicable to pull based approaches.

Under the pull-based model, the content publisher is required to be online in order to provide access to the content. Recent research efforts [5], [18], [19], [20] have proposed approaches to construct privacy preserving access control systems using a third-party storage service. In such approaches, the data owner has to enforce the ACPs and the privacy of the users from the content publisher is not protected. Further, in some approaches, multiple encryptions of the same document are required which is inefficient.

A major drawback of all the above approaches is that they do not consider the management of encrypted data hosted in a third party when users are added or removed from the system or when the ACPs/subdocuments are updated. All the approaches require the data owner to handle encryption. Di Vimercati et al. [7] first identifies this problem and proposes an initial solution. While their solution improves over existing solutions, such solution does not support expressive attribute based policies and does not protect the privacy of the users.

Attribute based encryption. The concept of attribute-based encryption has been introduced by Sahai and Waters [21]. The initial ABE system is limited only to threshold policies in which there are at least $k$ out of $n$ attributes common between the attributes used to encrypt the plaintext and the attributes users possess. Pirretti et al. [22] gave an implementation of such a threshold ABE system using a variant of the Sahai-Waters Large Universe construction [21]. Since this initial threshold scheme, a few variants have been introduced to provide more expressive ABE systems. Goyal et al. [23] introduced the idea of key-policy ABE (KP-ABE) systems and Bethencourt et al. [24] introduced the idea of ciphertext-policy ABE (CP-ABE) systems. Even though these constructs are
expressive and provably secure, they are not suitable for group management and especially in supporting forward security when a user leaves the group (i.e. attribute revocation) and in providing backward security when a new user joins the group. Some of the above schemes suggest using an expiration attribute along with other attributes. However, such a solution is not suitable for a dynamic group where joins and departures are frequent.

Proxy re-encryption (PRE). In a proxy re-encryption scheme [25] one party A delegates its decryption rights to another party B via a third party called a "proxy." More specifically, the proxy transforms a ciphertext computed under party A's public key into a different ciphertext which can be decrypted by party B with B's private key. In such a scheme neither the proxy nor party B alone can obtain the plaintext. Recently Liang et al. [26] has extended the traditional PRE to attribute based systems and independently Chu et al. [27] have extended the traditional PRE to support conditions where a proxy can re-encrypt only if the condition specified by A is satisfied. These improved PRE techniques alone or combined with ABE schemes [28] could be utilized to implement delegated access control in the cloud. However, they do not protect the identity attributes of the users who access the system and are difficult to manage.

9 Conclusions

Current approaches to enforce ACPS on outsourced data using selective encryption require organizations to manage all keys and encryptions and upload the encrypted data to the remote storage. Such approaches incur high communication and computation cost to manage keys and encryptions whenever user credentials change. In this paper, we proposed a two layer encryption based approach to solve this problem by delegating as much of the access control enforcement responsibilities as possible to the cloud while minimizing the information exposure risks due to colluding Users and cloud. A key problem in this regard is how to decompose ACPS so that the Owner has to handle a minimum number of attribute conditions while hiding the content from the cloud. We showed that the policy decomposition problem is NP-Complete and provided approximation algorithms. Based on the decomposed ACPS, we proposed a novel approach to privacy preserving fine-grained delegated access control to data in public clouds. Our approach is based on a privacy preserving attribute based key management scheme that protects the privacy of users while enforcing attribute based ACPS. As the experimental results show, decomposing the ACPS and utilizing the two layer of encryption reduce the overhead at the Owner. As future work, we plan to investigate the alternative choices for the TLE approach further. We also plan to further reduce the computational cost by exploiting partial relationships among ACPS.

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References

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