Tunable Privacy for Access Controlled Data in Peer-to-Peer Systems

Rammohan Narendula, Thanasis G. Papaioannou, Zoltán Miklós and Karl Aberer School of Computer and Communication Sciences, EPFL, Switzerland Email: firstname.lastname@epfl.ch

Abstract—Peer-to-peer paradigm is increasingly employed for organizing distributed resources for various applications, e.g. content distribution, open storage grid etc. In open environments, even when proper access control mechanisms supervise the access to the resources, privacy issues may arise depending on the application. In this paper, we introduce, PANACEA, a system that offers high and tunable privacy based on an innovative resource indexing approach. In our case, privacy has two aspects: the deducibility of a resource's existence/non-existence and the discovery of the provider of the resource. We systematically study the privacy that can be provided by the proposed system and compare its effectiveness as related to conventional P2P systems. Employing both probabilistic and information-theoretic approaches, we analytically derive that PANACEA can offer high privacy, while preserving high search efficiency for authorized users. Our analysis and the effectiveness of the approach have been experimentally verified. Moreover, the privacy offered by the proposed system can be tuned according to the specific application needs.

I. INTRODUCTION

Peer-to-peer (P2P) systems are increasingly used in many distributed application domains, e.g. content distribution, file sharing, open storage grids, video streaming, etc. However, users typically expect to be able to use these systems to share access-controlled and (semi-) private data. Conventional P2P systems should be properly adapted to meet the access control requirements of such applications. Typical approaches for data access control in open environments include cryptographic methods [1], Digital Rights Management (DRM) technologies, and trust-based methods [2], which require complicated key distribution and management. We consider a simpler, yet effective, approach for data access control in P2P systems: We assume that resources reside at the publisher nodes itself, to ensure that access control is enforced safely in an untrusted P2P environment. A user directly presents his credentials to the publishing peer of a particular resource after locating the resource in the P2P overlay. The publishing peer replies the query after applying its local authorization policies.

P2P systems typically try to maximize search efficiency. To this extreme, structured P2P systems, such as Kademlia [3] etc., employ an index implemented as a Distributed Hash Table (DHT) over the P2P overlay. Such an index typically consists of index entries of the form (key, value)-pairs, where the key is the resource identifier (often produced by one-way hash functions, e.g. MD5), while the value is the peer identifier, where the resource is stored. Indeed, as shown in [3], such an index significantly improves the search costs, in terms of both query latency and communication overhead. However, as index entries are hosted on arbitrary and often untrusted nodes, access to the index entries cannot be controlled by the peers that publish their data to the index. Thus, the index reveals both the existence/non-existence and the location (i.e. publishing peer) of each queried resource, hence, data privacy is breached. We refer to the former privacy aspect concerning resource existence/non-existence as resource privacy, while the latter one concerning resource location as provider privacy. On the other extreme, unstructured P2P systems, such as Gnutella (gnutella.com), employ no index and limited-hop flooding is used for locating the queried data, which incurs high latency and communication overhead, yet, with no guarantees on the data discovery. However, if access-controlled, unstructured P2P systems can provide the highest data privacy by answering queries only to authorized users. Thus, there is a trade-off between search efficiency and data privacy in this context.

In this paper, we explore this trade-off and propose a PrivAcy preserviNg Access-ControllEd (PANACEA) P2P system that combines high data privacy (both resource and provider privacies) and high search efficiency for authorized users. We carefully quantify privacy offered by PANACEA, employing both probabilistic modeling and information-theoretic approaches. We also analytically study the search efficiency/overhead of the PANACEA, as related to structured and unstructured P2P systems. The parameters of PANACEA can be tuned so that the trade-off between privacy and search efficiency is set according to the application needs. Numerically evaluating our analytic results for practical systems and verifying them with simulation experiments, we demonstrate that, with proper values for the parameters of PANACEA, authorized users almost always find the queried resources at a very low search overhead; on the other hand, unauthorized users can deduce the existence of a resource and its provider with a very low probability. Moreover, the communication overhead is high for unauthorized users. According to our information-theoretic analysis, PANACEA can tolerate large groups of *collusive adversaries* without compromising privacy. Figure 1 illustrates the position of PANACEA as related to structured and unstructured P2P access-controlled systems in the three-dimensional space < provider privacy, resource privacy, search efficiency>, employing the terminology of [4]. A resource privacy of 0 refers to the case that the adversary cannot deduce the existence/nonexistence of a resource. To the best of our knowledge, PANACEA is the first approach

that concurrently addresses resource and provider privacies in access-controlled systems.

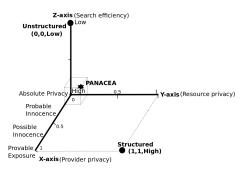


Fig. 1. Position of various systems on privacy and search efficiency axes.

Note that the specific authorization policy and the format of credentials are orthogonal to current scope of the paper. As a result, PANACEA mechanism can be employed by providers with different access control techniques, such as role-based access control, discretionary access control or attribute-based access control, all existing in the system simultaneously.

The remainder of the paper is organized as follows: In Section II, we describe the publishing and searching mechanisms in PANACEA. In Section III, we analytically derive the privacy properties and the search overhead employing a probabilistic approach. We also quantify the entropies of the system regarding resource and provider privacies. In Section IV, we verify our analysis and present our simulation experiments that demonstrate the effectiveness and the tunability of the system. In Section V, we discuss the related work. Finally, in Section VI, we conclude our work.

II. THE PANACEA SYSTEM

In this section, we present the proposed PANACEA system and explain how the resource and provider privacies are achieved. As already mentioned, resource privacy concerns hiding the existence and the non-existence of resources: an unauthorized user should not be able to determine either of them. Our system aims to combine look up efficiency of structured P2P systems with high resource and provider privacies offered by unstructured ones. PANACEA employs a DHT that hosts a resource and provider privacy-preserving (RPP) index. However, as explained later in this section, PANACEA indexes only a subset of the resources into the DHT; this is a necessary characteristic for providing resource privacy. The rest of the resources are located by flooding, similarly to the unstructured P2P systems. As a result, PANACEA acts partly as as a structured P2P system and partly as an unstructured one for the resources not indexed in the DHT.

The proposed indexing mechanism consists of tunable parameters that allow the application designer to choose between strong privacy guarantees and increased search efficiency based on the specific application needs. This tuning of the privacy parameters determines the position of the resulting system in the graph of Figure 1, as compared to structured and unstructured P2P systems. We describe the publishing

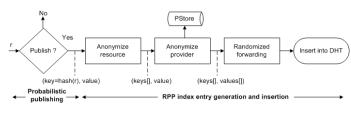


Fig. 2. Privacy preserving publishing in PANACEA

and search mechanisms of PANACEA in Section II-A and in Section II-B respectively.

A. Privacy Preserving Publishing

PANACEA achieves the resource and provider privacy goals with an innovative privacy-aware publishing mechanism, illustrated in Figure 2, which involves:

- 1) Probabilistic publishing of resources
- 2) RPP index generation
- 3) Randomized forwarding
- 4) Insertion into the DHT

This approach is described as follows:

1) Probabilistic publishing: Instead of publishing every resource into the DHT, as in structured P2P systems, PANACEA announces a resource with a system-defined probability μ (as shown in Figure 2) and creates an RPP-index entry as described later. Therefore, absence of an index entry for a specific resource key in the DHT does not necessarily mean non-existence of the corresponding resource in the system. Due to probabilistic publishing, PANACEA acts as a hybrid semi-structured P2P system. All the resources, which are not announced in the DHT, are discovered using limited-hop flooding with a system-defined time-to-live (TTL).

2) RPP index generation: We employ k-anonymization techniques [5] to achieve both resource and provider privacies for the resources selected to be announced into the DHT by the probabilistic publishing phase. A k-anonymization technique typically anonymizes a data item by hiding it inside a list of kdata items so that an adversary cannot identify it. Specifically, instead of having a (key, value) pair as an index entry for a resource, as in structured P2P systems, we propose that the index entry pair consists of a list of keys and a list of values, i.e. (key[], value[]), which is derived by applying resource and provider anonymization that are subsequently explained. We refer to such an index entry as (m, n)-index entry, where m refers to cardinality of the key list and n refers to that of the value list. In this terminology, an index entry of the conventional structured P2P systems can be seen as a (1, 1)index entry.

Regarding resource anonymization, once a new resource is chosen to be published, its corresponding (1,1)-index is converted to an (m,1)-index by adding m-1 resource keys randomly selected (that may correspond to genuine or phantom resources) from a resource namespace R to the key list of the index entry. The resource namespace R can be domainspecific or span multiple domains and even contain words and their combinations from a dictionary. Note that humanreadable plain text keys (i.e resource names) from R, which are employed by the users to refer to the resources, are mapped by a hash function h() to the system key space (i.e. resource ids) K. When a resource name r is anonymized, in fact, its key h(r) is anonymized among m keys. An adversary that is able to observe the resource keys of the (m, n)-entry may be able to derive the corresponding resource names of the m-1keys by employing a dictionary attack. However, he will not be able to deduce whether this entry corresponds to a genuine resource or not.

After resource anonymization, the resulting (m, 1)-index entry is fed to the provider anonymizer module, as depicted in Figure 2. The provider list is populated with n number of entries with the providing peer itself being one of them. The other n - 1 entries are randomly chosen from the *Provider Store* (*PStore*) - a local database of provider ids. We assume that PStores at each peer are initialized with a number of well-known peers and its neighbours in the overlay, and then incrementally expanded over time with unknown providers contained in the (m, n)-entries traversing through the peer.

3) Randomized forwarding: After an (m, n)-index entry is constructed by a publishing peer, it has to be inserted into the P2P system using the DHT put() method. However, this index entry must be published anonymously, as the next-hop node in the DHT routing could easily deduce that the initiator node is itself the publisher from the (m, n)-entry where it is contained. In order to anonymize the node that initiates the insertion request, we propose that a randomized forwarding phase (see Figure 2) preceds the DHT put() operation. Specifically, each peer that receives the insertion request decides with a system-defined probability λ to *forward* it to a node randomly selected from the *n* providers in the (m, n)-entry or *initiate* the DHT routing with the put(m, n) method with probability $1 - \lambda$. The technique of randomized forwarding to anonymize the original sender, was originally proposed in Crowds [4] for anonymizing web access. However, in [4], the next-hop node was randomly selected from the full set of cooperating nodes before contacting the web server. Clearly, our case is more complicated than the Crowds one, since the (m, n)-entry contains the publisher itself. Hence, by randomly choosing the next hop from the set of providers in the (m, n)-entry, we achieve equal probability for each of them for being the publisher. Note that randomized forwarding preceeds DHT routing, and hence it does not demand any modifications to it.

The randomized forwarding phase introduces additional communication overhead. This process can be viewed as a geometric distribution with parameter λ . Therefore, if X is the random variable that describes the number of hops of a put() request, then the probability that it travels x hops before it enters DHT routing is given by:

$$P(X = x) = \lambda^x (1 - \lambda)$$

The expected number of hops can be given by the mean of the geometric distribution, i.e. $E(X) = \lambda/1 - \lambda$. We assume that PStore caches the IP address along with each provider id and

that the IP address for each provider is stored in the provider list of the (m, n)-index entry. Thus, the relaying of put() can happen in O(1).

4) Insertion into the DHT: Finally, for the insertion of the (m, n)-entry into the DHT, put(m, n) operation is invoked, which is implemented using the conventional put() method. Note that the conventional put() method inserts only a (1, 1)-index entry. Yet, the same method can be used to insert a (1, n) entry, as the *value* field is not used in the DHT routing. Hence, we propose to convert the put(m, n) request into m number of put(1, n) requests, using each of the m keys as pivot ones.

Note that, since the keys in an index entry are chosen independently by peers, key collisions (i.e. conflicts) in the DHT are possible. Key collisions also happen when multiple copies of the same resource are inserted into the DHT. We propose that the list of providers in the new (m, n)-index entry is simply appended to the list of already existing providers for the collided key. A resource r with multiple copies is expected to have a larger provider list than that of a resource published by only one provider. However, as long as a genuine resource has smaller or equal provider list size to the maximum one n_f of a phantom resource, an adversary cannot differentiate between them. We propose an extension to our basic approach that increases the provider list sizes of phantom keys. Specifically, a peer randomly selects a small partition of the set R denoted as R_L ($R_L \subset R$) and constantly employs R_L for the resource anonymization instead of R (referred to as R_L -approach).

B. Searching

When a peer searches for a resource with key r, it executes get(r). If an (m, n)-entry was published in the DHT having r as one of its m ids, then the peer returns the provider list of this entry to the searcher. Subsequently, the searcher contacts all these providers, except for those where he knows from prior transactions that he is not authorized to for the queried resource. Once an index entry is found, a provider can be reached in O(1), since its IP address is maintained in the index entry along with the provider id. However, in case of multiple providers for same resource, an (m, n)-index entry for an existing resource may not contain all the providers of that resource in the system because of probabilistic publishing in PANACEA. In other words, the index entry is not always complete. As a result, a searcher may not be able to reach the provider where he is authorized through the RPP index. Therefore, even if an (m, n)-index entry is present in the index, the searcher may have to employ limited-hop flooding. However, the probability that a query is flooded over the overlay can be very low for resources with a few copies and with proper selection of the publishing probability μ , as shown in Section III. No provider responds to search queries from unauthorized users.

III. ANALYSIS

In this section, we analytically study the privacy offered by PANACEA by employing probability theory and information theory approaches. Moreover, we estimate the expected communication overhead of our approach.

We use the following notation in our analysis. Let N be the number of peers in the system and N_c be the expected number of copies of a genuine resource r and $N_a \leq N_c$ be the number of copies that a particular user is authorized to access. We call a user as *unauthorized* to r, when he is not authorized to access any of the N_c copies of r, i.e. $N_a = 0$.

A. Probabilistic approach

We evaluate the privacy breach that can be achieved by a user who has *complete* knowledge on the parameters of our PANACEA system and queries the system for a particular resource. We denote:

- i) $P_{K,a}$ (resp. $P_{K,u}$) as the probability for an authorized (resp. unauthorized) user to deduce the existence of a certain genuine resource.
- ii) $P_{V,a}$ (resp. $P_{V,u}$) as the probability for an authorized (resp. unauthorized) user to deduce the provider of a certain genuine resource.
- iii) P₋ as the probability for an authorized or unauthorized user to deduce the non-existence of a certain *non-existing* resource.

Definition 1: An access-controlled system is said to provide *higher* privacy if it promises:

- i) Lower probabilities for $P_{K,u}$, $P_{V,u}$, which addresses an unauthorized user deducing a resource's presence and its provider respectively.
- ii) Lower probabilities for P_{-} , which addresses a user deducing a resource's non-existence.

Under this definition of privacy, any *privacy-efficient* access control mechanism should aim to:

- Minimize $P_{K,u}$, $P_{V,u}$, P_{-} , which should ideally be 0 as in unstructured P2P systems.
- Maximize search cost $C_{s,u}$ for unauthorized users and ideally close to that of the unstructured P2P systems.

However, the *search efficiency* of the privacy-enabling mechanism should remain high, i.e.:

- $P_{K,a}$, $P_{V,a}$ should ideally be 1 (as in structured P2P systems), and
- the search communication cost $C_{s,a}$ should be kept low and ideally close to that of the structured P2P systems.

We express the privacy and search cost metrics of PANACEA in terms of the corresponding metrics of structured and unstructured P2P systems. To this end, we use superscripts U and S to denote metrics for unstructured and structured P2P systems respectively, and no superscript for PANACEA, e.g. $P_{K,u}^U$ refers to unstructured systems and the equivalent metric for PANACEA is $P_{K,u}$.

First, we quantify provider privacy for an authorized user. There are three cases that can arise:

Case (i): If any of the N_a copies, where he is authorized to, was published to the DHT, he could deduce the provider of the resource with probability 1. The probability that at least one of N_a copies was published into the DHT is $1 - (1 - \mu)^{N_a}$.

Case (ii): On the other hand, consider the case that none of the N_a copies was published in the DHT (probability of which, is $(1 - \mu)^{N_a}$), but some *i* of the remaining $N_c - N_a$ copies were published. Recall that if the user finds an (m, n)-entry for *r* in the DHT, he contacts all the providers therein. In this case, as an authorized copy of the resource would not be located by contacting $i \cdot n$ number of providers listed in the DHT entry, the user would also employ flooding. Here, we say that the user could deduce the provider with probability $\frac{1}{i \cdot n}$ in addition to what is possible to be deduced by flooding (i.e., $P_{V,a}^U$). However, the number of providers is less than $i \cdot n$, because of "collisions" (i.e. provider id conflicts) in the provider lists across the multiple (m, n) entries of the resource copies. We account for this with a *collision factor* f_v .

Case(iii): When a DHT entry is not found for the resource key (probability of which is $(1 - \mu)^{N_c})^1$, the user attempts to deduce the provider by flooding $(P_{V,a}^U)$. Hence,

$$P_{V,a} = \left[1 - (1 - \mu)^{N_a}\right] \cdot 1 + (1 - \mu)^{N_a} \cdot \sum_{i=1}^{N_c - N_a} {N_c - N_a \choose i} \mu^i (1 - \mu)^{(N_c - N_a - i)} \\ \cdot \left[\frac{1}{(i \cdot n)(1 - f_v)} + P_{V,a}^U\right] + (1 - \mu)^{N_c} \cdot P_{V,a}^U .$$

$$(1)$$

We apply similar reasoning to formulate resource privacy $P_{K,a}$ for an authorized user in addition to the following. Let the maximum provider list size n_f of a phantom resource key be known to the user. Then, if the size of provider list of the (m, n)-entry of the queried resource is larger than n_f , the user can deduce the existence of the resource in the system with probability 1. Otherwise if it is smaller than n_f , a probability $\frac{1}{m}$ can be assigned to the existence, as the queried key is mixed with m-1 other ones in the (m, n)-entry, in addition to what can also be assigned by flooding $(P_{K,a}^U)$. Therefore,

$$P_{K,a} = \left[1 - (1 - \mu)^{N_a}\right] \cdot 1 + (1 - \mu)^{N_a} \cdot (1 - (1 - \mu)^{N_c - N_a})$$
$$\mathbf{I}\left(\sum_{i=1}^{N_c - N_a} \binom{N_c - N_a}{i} \mu^i (1 - \mu)^{(N_c - N_a - i)} (in)(1 - f_v)\right)$$
$$+ (1 - \mu)^{N_c} \cdot P_{K,a}^U, \text{ where}$$
$$\mathbf{I}(l) = \left\{\begin{array}{c}1, & \text{for } l \ge n_f\\P_{K,a}^U \cdot 1 + (1 - P_{K,a}^U) \cdot \frac{1}{m}, \text{ otherwise}\end{array}\right\}.$$
(2)

Next, we quantify the search cost $C_{s,a}$ in terms of the number of messages generated because of a query from an authorized user. First, a user searches in the DHT incuring cost $C_{s,a}^S$. Thereafter, we account for two possible cases- none of N_c copies or some *i* copies of the resource are published into the DHT. The former case can happen with probability $(1 - \mu)^{N_c}$ where the user employes flooding incuring a cost of $C_{s,a}^U$. In the latter case, *i* · *n* number of providers are contacted. If

¹In fact, a DHT entry can also be present for one resource, if its key was inserted as a phantom one. However, we assume this probability as negligible, as |R| is big as compared to the number of genuine resources.

none of them has an authorized copy (probability of which, is $(1 - \frac{N_a}{N_*})^i$), the user employs flooding. Overall, $C_{s,a}$:

$$C_{s,a} = C_{s,a}^{S} + (1-\mu)^{N_c} \cdot C_{s,a}^{U} + \sum_{i=1}^{N_c} {N_c \choose i} \mu^i (1-\mu)^{(N_c-i)}$$
$$\cdot \left[i \cdot n \cdot (1-f_v) + \left(1 - \frac{N_a}{N_c}\right)^i C_{s,a}^{U} \right]$$
(3)

We can derive the resource privacy $P_{K,u}$, the provider privacy $P_{V,u}$, and the search cost $C_{s,u}$ for an unauthorized user from eq. (3) by setting Na = 0.

Finally, we derive P_- , i.e. the probability to deduce the non-existence of a non-existing resource. Given an event space $\Omega = \{\text{DHT}, \neg \text{DHT}\}$ that a non-existent resource is found or not in the DHT respectively, P_- is given by:

$$P_{-} = Pr(- \mid \Omega) = Pr(- \mid \neg \text{DHT}) \cdot Pr(\neg \text{DHT}) + Pr(- \mid \text{DHT}) \cdot Pr(\text{DHT}), \text{ where}$$

$$Pr(- \mid \neg \text{DHT}) = P_{K,u}^{U} = 0$$

$$Pr(- \mid \text{DHT}) = \frac{m-1}{m}$$

$$Pr(\text{DHT}) = \left[1 - \left(1 - \frac{1}{\mid R \mid}\right)^{\mu N_{r}(m-1)}\right]$$

$$Pr(\neg \text{DHT}) = 1 - Pr(\text{DHT})$$
(4)

 N_r is the total number of resources in the system and R is the resource namespace. $Pr(- | \neg DHT)$ expresses the probability that a resource is non-existent, given that it is not found in the DHT. This is similar to the probability of deducing the existence of an unauthorized resource for a user in unstructured P2P systems, because an existing resource is same as a non-existing resource for an unauthorized user. Pr(- | DHT) is the probability that the resource corresponding to the key does not exist. Pr(DHT) expresses the probability that a phantom resource from namespace R may have been inserted into the index, while $Pr(\neg DHT)$ is the complement of Pr(DHT). Observe that P_- is minimal (~ 0) for reasonable values of the various parameters. Also, we estimate the expected query cost to deduce the non-existence. The user first searches for an index entry and then employs flooding, hence,

$$C_{s,-} = C_{s,-}^S + C_{s,-}^U \,. \tag{5}$$

B. Information-theoretic approach

In [6], [7], an information theoretic approach was proposed to measure privacy offered by a system employing *entropy* H as an anonymity metric, which is defined as:

$$H = -\sum_{i} p_i \log_2 p_i , \qquad (6)$$

where p_i is the attacker's estimate of the probability that a participant *i* was responsible for some observed action. Entropy is maximized to $\log_2 |A|$ if equal probability is assigned to all members of the anonymity set *A*, and it is minimized at 0 when |A| = 1. According to [6], a system with entropy

H has *effective anonymity set* of size 2^{H} . As an adversary in PANACEA may have different information sets (i.e. each resulting from different observations), *conditional entropy* H_0 [7] is a more adequate metric, which is given by:

$$H_0 = \sum_{y} \Pr[Y = y] H(X|Y = y) = E_y H(X|Y = y),$$
(7)

where X is a random variable of the private aspect to be preserved and Y is a random variable modeling different observations y.

Next, we calculate the entropy of PANACEA for resource anonymity against a searcher. To this end, the random variable X models the existence of a resource, i.e. whether a resource name from the resource namespace R exists in the system or not. The random variable Y models the observations of the searcher for a requested resource: (i) an authorized copy is found in the DHT, (ii) an unauthorized copy is found in the DHT and an authorized copy is found by flooding, (iii) an unauthorized copy is found in the DHT but no authorized copy is found by flooding, (iv) no copy is found in the DHT and an authorized copy was found by flooding, and (v) no copy is found in the DHT but no authorized copy was found by flooding. In the information sets (i) and (ii), there is no anonymity, thus entropy is 0. In the case of (iii), if the expected provider list size of the queried resource is greater than the maximum provider list size of any non-existing resource that is inserted in the DHT for resource anonymization, then there is no privacy and the resource is existent. Otherwise, the probability assigned to any resource in the (m, n)-entry to be existent is 1/m, because this key was inserted with a mix of m keys. Thus, the anonymity set is m in that case. In the observation (iv), the anonymity set is 1, as the queried resource is found. Finally, in the case of (v), the anonymity set is R, as the queried resource is not found and it can be any of the resources in the resource namespace R. Therefore, the resource entropy H_k for a searcher is given by:

$$H_{K} = E_{iii}H(X|Y = iii) + E_{v}H(X|Y = v)$$

= $-(1 - \mu)^{N_{a}}(1 - (1 - \mu)^{N_{c} - N_{a}})(1 - P_{K,a}^{U})$
 $\mathbf{1}(E(n') < n_{f})\log_{2}\frac{1}{m} - (1 - \mu)_{c}^{N}(1 - P_{K,a}^{U})\log_{2}\frac{1}{|R|},$
(8)

where E(n') is the expected provider list size in the (m, n)entry for the queried resource and n_f is the maximum provider list size for a phantom resource respectively. $E(n') = \sum_{i=1}^{N_c-N_a} {N_c-N_a \choose i} \mu^i (1-\mu)^{(N_c-N_a-i)} (in) (1-f_v)$ and **1**() is an indicator function.

Also, regarding provider anonymity, the random variable X models the publisher of the requested resource and the random variable Y models the information sets for the searcher as in the case of resource anonymity. If the searcher has made the observations (i), (ii), then the provider entropy is 0. In the case (iii), where an entry for the requested resource is found in the DHT, but not for an authorized copy, and no authorized copy was found by flooding, we calculate H(X|Y = iii) according to the following logic: If the expected number of

providers for the requested resource as in the DHT is E(n'), then the publisher of the requested resource can be either one of these entries in the provider list or one peer in rest of the system. Therefore, each peer in the provider list has probability 1/(E(n') + 1) to be the publisher and any other peer has probability 1/(E(n') + 1)(N - E(n')) to be the publisher of the requested resource. If the observation (iv) is made by the searcher, then the publisher is located and thus the entropy is 0. Finally, in the case of observation (v), the anonymity set is the complete set N of peers in the system, as the publisher is not located. Therefore, the provider entropy H_V is:

$$H_{V} = E_{\text{iii}} H(X|Y = \text{iii}) + E_{v} H(X|Y = v)$$

$$= -(1-\mu)^{N_{a}}(1-(1-\mu)^{N_{c}-N_{a}})(1-P_{K,a}^{U})$$

$$(E(n')\frac{1}{E(n')+1}\log_{2}\frac{1}{E(n')+1} + (N-E(n'))\frac{1}{(N-E(n'))(E(n')+1)}$$

$$\log_{2}\frac{1}{(N-E(n'))(E(n')+1)}$$

$$-(1-\mu)^{N_{c}}(1-P_{K,a}^{U})\log_{2}\frac{1}{N}$$
(9)

For a searcher that is authorized to a copy of the queried resource, the resource and provider entropies are given by eq. (8), (9) respectively. On the other hand, the resource and provider entropies for an unauthorized searcher can be given by eq. (8), (9) respectively by having $N_a = 0$.

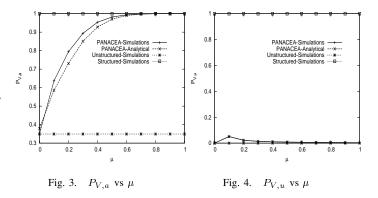
IV. EVALUATION

In this section, we verify our analysis and evaluate the privacy and search efficiencies of the PANACEA system as related to unstructured and structured access-controlled P2P systems using simulation experiments.

In our simulated system (implemented in Java), we assume N = 10000 peers that use the system both to publish and search for resources. The PStores on the peers are initialized with 25 random entries. The providers are organized in a Kademlia-like structured topology, but they are also connected over an unstructured overlay power-law network with average degree 7.5 and maximum degree 150. We conducted two types of simulation experiments, which differ in their resource distributions and the type of the generated queries. Each resource is randomly assigned a publisher peer and a list of user peers who are authorized to access the resource. Any other peer is said to be an unauthorized user for this resource and publisher pair. We compute $P_{V,a}$ and $P_{V,u}$ as follows:

- $P_{V,a} = 1$ if a user is able to contact a publisher where he is authorized to access the requested resource.
- If the resource key is found in the DHT with a value list of size n', then P_{V,a} = P_{V,u} = ¹/_{n'}.
- Otherwise, if the resource key is not found in the DHT and no authorized copy is located by flooding, then $P_{V,a} = P_{V,u} = 0.$

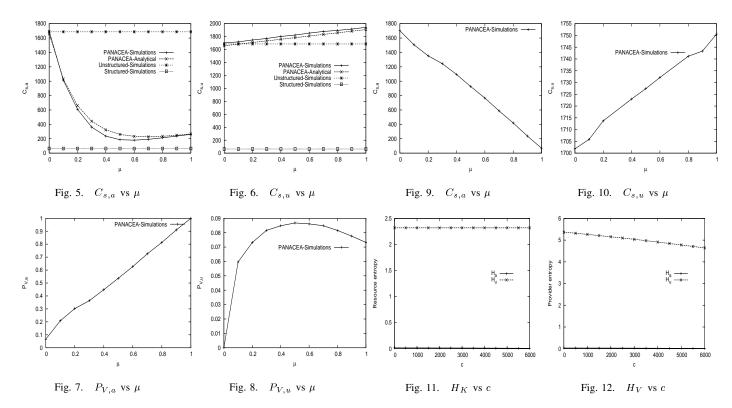
The PANACEA's parameters are taken as follows: key list size m = 5, value list size n = 5 and forwarding probability $\lambda =$



0.6. Also, ttl = 4 was employed for limited-hop flooding in the unstructured overlay.

A. Provider privacy and search cost

Initially, we aim to verify the correctness of eqs. (1), (3) using the simulation results with a rather static setting regarding resource popularity. Speciafically, we assume 100 resources with $N_c = 50$ copies for each (thus 5K resources in total). 100 peers are randomly selected, each of which is inserted into the authorization list of random $N_a = 5$ copies. Each resource is then given to randomly chosen peers that publish them using the PANACEA publishing mechanism. The collision probability for provider lists is experimentally found to be $f_v = 0.002$. We also experimentally found that by searching in the unstructured overlay $P_{V,a}^U = 0.38$, $P_{V,u} = P_- = 0$, 912 distinct nodes are visited and 1687 messages are sent per query on the average. In order to measure $P_{K,a}, C_{s,a}$, we generate authorized searches from the above 100 authorized peers for all of the 100 resources, thus 10K search queries in total. Also, in order to measure $P_{K,u}, C_{s,u}$, we randomly select 100 unauthorized users that query the system for the same 100 resources. These experiments have been run 10 times each and the mean values are plotted in Figures 3 to 6. As depicted in Figures 3, 4, the analytical equations model the privacy properties of the simulated PANACEA system very accurately. As the probability of publishing μ increases, PANACEA approaches the search efficiency of a structured system (see Figure 3). Note that for only $N_a = 5$ authorized copies in the system, a small value of $\mu = 0.6$ makes the search efficiency of PANACEA close to that of structured systems. On the other hand, for unauthorized users, provider privacy of our system is always close to that of unstructured P2P systems, as shown in Figure 4. Therefore, PANACEA design meets its privacy objectives of Section III. For $\mu = 0$, $P_{V,u} = 0$. When $\mu = 0.1$, a provider list is found in the DHT for the queried resource, reducing the privacy in comparison to $\mu = 0$. As μ increases, so is the size of provider list of the resource, and thus $P_{V,u}$ decreases. In Figure 5, the effect of μ on the search communication cost for authorized users is depicted. As μ increases, the probability to find a provider where the user is authorized also increases. After $\mu = 0.6$, there is no more search cost improvement, because the size of the provider list of the queried resource slightly increases.



As shown later, the search cost significantly decreases as μ increases. However, as observed from Figure 6, the search cost for unauthorized users significantly increases (over the cost of limited-hop flooding) with μ , which is a highly desirable property of PANACEA.

Next, we evaluate the provider privacy and the search cost of PANACEA for authorized and unauthorized users in a more realistic setting, where 10K resources whose popularity (N_c) follows Zipf distribution are published in the system. The maximum number of resource copies is 150 and their mean number is 10 thus resulting in a total of 100K resources. A random number of 5 to 50 peers are chosen to be authorized to each resource. Each resource of the 100K ones is randomly assigned to a peer that initiates PANACEA publishing. We randomly generate 20K number of authorized and unauthorized search queries separately. Again, the experiments are repeated 10 times and mean values of the results are plotted in Figure 7 to Figure 10. As depicted in Figure 7, search efficiency increases with μ for an authorized user. Also, as shown in Figure 8, provider privacy is always minimal for unauthorized users. Finally, Figure 9 depicts that the search cost for authorized users decreases with μ , as opposed to that of unauthorized users that increases with μ .

B. Resource privacy

As mentioned in Section II-A, resource privacy may be breached by observing large sizes of the provider lists. Although it is difficult to preserve resource privacy of highly popular resources, we do not focus on them as their presence in the system can be easily taken for granted. Our goal for resource privacy in PANACEA is to preserve privacy for the other resources. We observe that as long as an existing resource has a provider list size less or equal to that of a phantom resource, an adversary cannot differentiate between them. We assume again 100K resources Zipf-distributed with mean 10 and maximum 150 copies. Employing a small local static partition, of the resource namespace R for resource anonymization, i.e. the R_L -approach with $|R_L| = 25$, we observed that phantom keys have provider lists longer than those of the 87.6 percentile of the existing resources, the resource privacy for unauthorized users is $P_{K,u} = 1/m = 0.2$ and for the authorized users is $P_{K,a} = 1$ for $\mu = 1$. For more popular resources, the adversary can exploit the provider list sizes to conclude their existence.

Finally, by numerically evaluating eqs. (4) and (5) with |R| = 2M, $N_r = 100K$, $\mu = 1$, we observed the PANACEA system meets its design objectives in this case as well, as $P_- \sim P_-^U$ and $C_{s,-} \sim C_{s,-}^U$. We omit the verification of these formulas with simulations for brevity reasons.

C. System entropy

In Section III, we have defined the entropies for resource anonymity and provider anonymity. This formulation allows us to measure the privacy breach that can be achieved by collusive groups of size c based on a simple modification to the formulas: The provider list of the (m, n)-entry of the queried resource will only contain E(n')(1 - c/N) noncollusive entries. Employing the parameters of Section IV-B with $\mu = 0.6$, $n_f = 20$ and |R| = 2M, we numerically evaluate the slighly modified versions of eqs. (8) and (9). As depicted in Figures 11, 12, the anonymity set by PANACEA for unauthorized users is $4.92 \equiv m$ for resource privacy and the anonymity set for provider privacy is 36.7.

V. RELATED WORK

There is significant research work in the literature related to PANACEA, particularly in the areas of access control in P2P systems, privacy of access-controlled content, and anonymous P2P systems.

To enable access control in P2P systems, PHera [8] proposes a fine-grained access control framework based on super-peerbased P2P overlays where the access-control policies of subpeers are enforced by the super-peers. Super peers index the data of sub-peers and they could preserve data privacy by not replying to the queries from unauthorized peers. However, this approach assumes that all super-peers are unanimously trusted by their sub-peers to enforce their data privacy and access control policies, which is difficult in general [9]. In PANACEA, peers can share their resources through index hosted on untrusted nodes, and yet, can enforce access control.

Regarding the privacy of access-controlled content, a privacy-preserving approach for centralized indexing of such data is proposed in [9]. A group of data providers iteratively circulate a bloom filter representing the content hosted on the providers, bits of which are set probabilistically by the proposed algorithm. At the end of this iterative process, the index -represented by the bloom filter- emerges, which preserves data privacy regarding its location (i.e. provider privacy). However, as opposed to PANACEA, [9] does not address resource privacy. Furthermore, new resources can be easily inserted in the index of PANACEA, while index reconstruction is required in [9].

The OneSwarm system proposed in [10] employs an unstructured friend-to-friend overlay for privacy preserving content sharing. It preserves the privacy of a peer's location using cryptographic mechanisms. The system allows users to define permissions for data sharing among trusted friends. Peers search for data objects using flooding techniques, similarly to access-controlled unstructured systems.

There exists a large number of works in the area of anonymous P2P systems that achieve publisher (source) or reader (searcher) anonymity or both [1], [11], [4], [12]. Additionally, the anonymity of a node hosting an index entry (resource) is also considered [11]. In Freenet [1], resource identifiers are generated in several cryptographic ways and are inserted into the system based on these identifiers. It achieves access control and resource and provider privacies using cryptographic techniques, which however, involve complicated key distribution and management overhead. Furthermore, resource discovery is not guaranteed and involves significant search communication overhead compared to structured systems. In addition, the searchers have to be associated with the providers a priori, in order to be informed about the cryptographic keys. Instead, in our approach, search efficiency is high and new searchers can be dynamically authorized by providers to access the resources. P2P access control system based on such cryptographic indexing was discussed in [2].

A hybrid P2P system involving structured and unstructured topologies to achieve sender and receiver anonymity, was discussed in [12] and referred to as Agyaat. Agyaat, provides mutual anonymity for the sender and receiver, which is not among the goals of PANACEA. Agyaat offers three alternative resource discovery approaches: semantic groups, centralized directory service, and dynamic services. In the first case, peers that host semantically similar resources are grouped into a cloud. Then, some sort of resource and provider privacies can be provided at the expense of resource discovery, which is flooding-based, as opposed to our approach. For improving resource discovery, a centralized directory service or dynamic services can be also employed. Then, a resource is mapped to a cloud and the index is stored at a central server or at the coordinator peers of the clouds in a distributed manner. These are similar to the privacy preserving indexing employed in [9]. However, Agyaat does not describe the anonymous construction of this index and it does not analytically quantify its effectiveness, as opposed to PANACEA.

VI. CONCLUSION

In this paper, we have proposed PANACEA, a P2P infrastructure to share access-controlled data, which combines high resource and provider privacies with high search efficiency for authorized users. We have analytically derived the privacy and search efficiency properties of the system employing probabilistic and information-theoretic approaches. Our analysis was verified by simulation experiments, while we analytically and experimentally showed that PANACEA meets its design objectives. As a future work, we intend to employ the mechanism in a Kademlia client and observe the privacy offered in a real testbed.

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