Specified keywords search scheme for EHR sharing

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Abstract
Searchable encryption allows data users to search for encrypted files by keywords without restriction. However, electronic health record (EHR) contains sensitive information, and data users should search for and share EHR with restriction. If data users are not restricted when EHR is searched and shared, there is a high risk that EHR will be misused and reveal large amounts of private patient information. This paper proposes a specified keywords search scheme for EHR sharing based on searchable encryption and proxy re-encryption to address this problem. In the scheme, the data user searches with the keywords specified by the doctor and obtains EHR from the medical cloud. Proxy re-encryption is used to implement the sharing of EHR and privacy preservation securely. The security proof demonstrates that our scheme is secure against chosen keyword attack. Furthermore, the experimental results show that the scheme achieves computational efficiency

Keywords Electronic health record · Searchable encryption · Specified keywords · Conjunctive keywords

1 Introduction

With the development of information technology, more and more hospitals use EHR (Heart et al. 2017) to record treatment information. The use of EHR reduces the burden on doctors and brings convenience to clinical work. Moreover, compared with the traditional medical record, EHR reduces the cost of storage and can be stored for a long time. In the process of patient diagnosis, the transmission of EHR between doctors or hospitals is inevitable. Since the EHR contains patients' private information, the EHR must be encrypted, and it is necessary to focus on data security during EHR sharing (Riad et al. 2019; Chi et al. 2019; Gautam et al. 2019). Therefore, how to retrieve and share encrypted EHR while ensuring data security and privacy preservation is an essential research direction.

Searchable encryption and proxy re-encryption can effectively ensure the security of encrypted data search and sharing (Wu et al. 2016; Yu et al. 2021; Chen et al. 2021). Searchable symmetric encryption (SSE) scheme was proposed by Song et al. (2000). The scheme realizes the search for encrypted data, but the key management is complicated. In 2004, Boneh et al. (2004) introduced the public key cryptosystem into the searchable encryption and proposed the first public key encryption with keyword search scheme (PEKS). Still, the scheme can only be used under the secure channel. Baek et al. (2008) proposed a public key searchable encryption scheme under public channel. Currently, public key searchable encryption is widely used in a variety of scenarios (Jiang et al. 2018; Hassan et al. 2019; Xu et al. 2020). Blaze et al. (1998) put forward the proxy re-encryption (PRE). Shao et al. (2010) proposed a proxy re-encryption scheme with a keyword search by combining searchable encryption and proxy re-encryption, which achieves search and sharing of ciphertext.

Many schemes for searching and sharing EHR based on searchable encryption and proxy re-encryption have been
proposed in recent years. Wang et al. (2019) proposed a scheme for sharing EHR among multi-institution based on cloud. The scheme supports conjunctive keyword search and guarantees access legitimacy through an identity-based access control mechanism. To reduce the overhead, Ying et al. (2018) provided a cloud-based lightweight EHR sharing scheme. Ciphertext policy attribute-based encryption (Bethencourt et al. 2007) is used to achieve fine-grained access control of EHR in the scheme. With the development of blockchain (Nakamoto 2008), some scholars apply it to the EHR searching and sharing scheme. Wang et al. (2019) proposed a cloud-assisted EHR data sharing scheme via blockchain. The scheme stores keyword index and EHR ciphertext in blockchain and cloud, respectively. Searchable encryption and proxy re-encryption are used to realize data sharing security and privacy preservation. Niu et al. (2020) introduced a blockchain-based EHR sharing scheme that implements multi-keyword search and uses attribute-based encryption to ensure the confidentiality and fine-grained access control of EHR. For the integrity and access control of EHR, a security protocol for a cloud-assisted EHR system via blockchain is proposed by Kim et al. (2020). In (Qin et al. 2021), a secure sharing scheme for EHR based on consortium blockchain is presented to achieve controllable sharing and precise search of EHR. Dagher et al. (2018) introduced a blockchain-based EHR sharing framework. Under this framework, patients, data providers, and third parties can access EHR safely and efficiently. Chen et al. (2019) proposed a blockchain-based searchable encryption scheme for EHR, and the scheme ensures the traceability and integrity of the index using blockchain. In the above schemes, when searching EHR, either the data owner generates a trapdoor for the data user, or the data user constructs a trapdoor by himself. However, the former will result in low search autonomy, while the latter can search unrestricted, which is not conducive to privacy preservation.

Considering that EHR contains patient privacy and sensitive information, EHR should be restricted from searching and sharing to achieve data security and privacy protection. We propose a specified keywords search scheme for EHR sharing. In our scheme, the doctor specifies the keywords for the data user, and the data user can only search with the keywords specified. Therefore, the data user can only obtain EHR containing keywords specified by the doctor, preventing malicious data users from abusing EHR data. Multi-keyword search and keyword access control, privacy preservation, and EHR data security are implemented in our scheme. The contributions of our scheme are as follows.

- The keyword index is generated using searchable encryption and stored in the public cloud. The data user can make conjunctive keyword search in the cloud, which supports multi-keyword search and reduces the communication overhead of search.
- Searchable encryption and proxy re-encryption are used to realize the search of specified keywords and the secure sharing of ciphertext, respectively, protecting patient privacy and ensuring EHR security.
- Before the data user obtains data from the medical cloud, the medical cloud verifies the identity of the data user to determine its legitimacy and ensure the security of EHR data.

The paper is organized as follows: Sect. 2 reviews preliminaries. Section 3 presents the system model. Section 4 provides the algorithm framework and security model. Section 5 describes the scheme construction and security proof in detail. The experiments and efficiency analyses are performed in Sect. 6. Finally, Sect. 7 concludes our work.

2 Preliminaries

In this section, we review the preliminaries required for this paper.

2.1 Bilinear pairing

Let $G_1$ and $G_2$ be cyclic groups of prime order $p$, a bilinear pairing $e : G_1 \times G_1 \rightarrow G_2$ must satisfy the following properties.

1. Bilinear: For any $x, y \in G_1$ and $a, b \in \mathbb{Z}_p^*$, we have $e(x^a y^b) = e(x)^a e(y)^b$.
2. Non-degenerate: $x, y \in G_1$ exists such that $e(x, y) \neq 1$.
3. Computable: For any $x, y \in G_1$, an efficient algorithm computes $e(x, y)$.

2.2 Complexity assumptions

**Definition 1** DDH Assumption (Bao et al. 2003) For any $a, b \in \mathbb{Z}_p^*$, given a tuple $(g, g^a, g^b, T \in G_1)$ as input. We define an algorithm $A$ has advantage $\varepsilon$ as $\Pr[A(g, g^a, g^b, T) = 1] - \Pr[A(g, g^a, g^b, T) = 1]$. We say that the decisional Diffie–Hellman assumption holds if advantage $\varepsilon$ is negligible for all probabilistic polynomial time $A$.

**Definition 2** $q$-ABDHE Assumption (Fang et al. 2009) For any $x, z \in \mathbb{Z}_p^*$, given a tuple $(g, g^x, g^{x^2}, \ldots, g^{x^q}, g^z, g^{x z^2}, \ldots, g^{x z^{q-1}}, T \in G_2)$ as input. We define an algorithm $A$ has advantage $\varepsilon$ as $\Pr[A(g, g^x, g^{x^2}, \ldots, g^{x^q}, g^z, g^{x z^2}, \ldots, g^{x z^{q-1}}, T \in G_2] = 1] - \Pr[A(g, g^x, g^{x^2}, \ldots, g^{x^q}, g^z, g^{x z^2}, \ldots, g^{x z^{q-1}}, T = 1)]$. We say that the augmented bilinear pairing assumption holds if advantage $\varepsilon$ is negligible for all probabilistic polynomial time $A$. 
decisional Diffie–Hellman exponent assumption holds if advantage \( \varepsilon \) is negligible for all probabilistic polynomial time \( A \).

### 3 System model

This section presents the system model of the specified keywords search scheme for EHR sharing.

There are five entities in the system: patient, doctor, data user, public cloud, and medical cloud, as shown in Fig. 1. The five entities are described as follows.

**Patient** The patient needs to register and obtain a unique identity \((ID_i)\) from the registration system. Before interacting with the doctor, the patient shows \( ID_i \) to the doctor as an authorization to manage EHR. When the patient needs an EHR, they generate a trapdoor to search for the EHR ciphertext storage address \((address)\) on the public cloud and obtain the EHR ciphertext from the medical cloud through the \( address \) and \( ID_i \).

**Doctor** The doctor generates an EHR for the patient and extracts keywords. The EHR is encrypted and stored in the medical cloud. Meanwhile, keywords are encrypted by doctors with searchable encryption and stored in the public cloud. When a data user requests an EHR, the doctor specifies keywords for the data user. In addition, the doctor acts as a trusted third party to generate proxy re-encryption keys for the data user.

**Data user** The data user generates a trapdoor to search within the keywords specified by the doctor. Then, the data user gets \( address \) from the public cloud. Then, the data user searches \( address \) and search token to the medical cloud to obtain the EHR ciphertext.

**Public cloud** The public cloud is used to store keyword index. The public cloud is highly open and can be accessed by any public cloud user, but only patients and authorized data users can search in the public cloud. Before data users search, the public cloud acts as a proxy to re-encrypt the keyword index.

**Medical cloud** The medical cloud is used to store EHR ciphertext and return \( address \) to the doctor. The medical cloud is exclusive and can be accessed by authenticated users. When the EHR is shared, the medical cloud acts as a proxy to re-encrypt the EHR ciphertext.

### 4 Scheme framework and security model

In this section, we provide the scheme framework and security model.

#### 4.1 Scheme framework

In the scheme, we consider two different cases of EHR search and sharing (Case I and Case II). Case I describes searching and obtaining EHR by the patient, including eight polynomial algorithms. Case II describes searching and sharing EHR by the data user, including nine polynomial algorithms. The definition of each polynomial algorithm is as follows.

- **Setup** \((\lambda) \rightarrow PP\) : The algorithm takes a security parameter \( \lambda \) as input, and outputs the public parameters \( PP \).
- **KeyGen** \((PP) \rightarrow (sk, pk)\) : The algorithm inputs \( PP \) and outputs the public/private key pairs \((sk, pk)\) for the patient, doctor and data user.
- **Enc** \((PP, m, sk_d, pk_p) \rightarrow C_m\) : Given \( PP \), an EHR \( m \), the doctor’s private key \( sk_d \), and the patient’s public key \( pk_p \), the algorithm encrypts the EHR and outputs the EHR ciphertext \( C_m \).
- **IndGen** \((PP, W, sk_d, pk_p) \rightarrow I\) : Given \( PP \), the keyword set \( W = \{w_1, w_2, \ldots, w_n\} \), \( sk_d \), and \( pk_p \), the algorithm outputs the keyword index \( I \).

**Case I: the patient searches and obtains EHR**

- **TrapdoorGen** \((PP, W', sk_p, pk_d) \rightarrow Tw\) : Given \( PP \), the search keyword set \( W' = \{w'_1, w'_2, \ldots, w'_l\} \), the patient’s private key \( sk_p \), and the doctor’s public key \( pk_d \), this algorithm outputs the trapdoor \( Tw \).
- **DataSearch** \((I, Tw) \rightarrow address \text{ or } \bot\) : The algorithm takes \( I \) and \( Tw \) as input, and outputs \( address \) or \( \bot \).
- **DataAccess** \((address, ID_i) \rightarrow C_m \text{ or } \bot\) : The algorithm takes \( address \) and \( ID_i \) as input, and outputs \( C_m \) or \( \bot \).
- **Dec** \((C_m, sk_p) \rightarrow m\) : The algorithm takes \( C_m \) and \( sk_p \) as input, and outputs \( m \).

**Case II: the data user searches and obtains EHR**

- **IndReEnc** \((I, rk_1) \rightarrow I'\) : The algorithm inputs \( I \) and the re-encryption key \( rk_1 \), and outputs the re-encrypted keyword index \( I' \).
- **TrapdoorGen** \((PP, W'', sk_p, pk_d) \rightarrow Tw'\) : The algorithm inputs \( PP \), the specified keyword set \( W'' = \{w''_1, w''_2, \ldots, w''_l\} \), the data user’s private key \( sk_d \) and \( pk_p \), and outputs the trapdoor \( Tw' \).
- **DataSearch** \((I', Tw') \rightarrow address \text{ or } \bot\) : The algorithm takes \( I' \) and \( Tw' \), and outputs \( address \) or \( \bot \).
- **DataAccess** \((address, C_m, rk_2, token) \rightarrow C'_m \text{ or } \bot\) : The algorithm inputs \( address \), \( C_m \), the re-encryption key \( rk_2 \), and the token \( token \), and outputs the re-encrypted EHR ciphertext \( C'_m \) or \( \bot \).
4.2 Security model

The security is based on the DDH assumption and \( q \)-ABDHE assumption. Indistinguishability of chosen keyword attack (IND-CKA) game is designed to prove the security. There are two types of attackers in the IND-CKA game.

**IND-CKA game**: Suppose \( A \) is a polynomial time adversary, \( B \) is the challenger, \( A \) and \( B \) play the following two games:

**Game 1**: \( A \) is assumed to be an outside attacker.
- **Setup**: The algorithms \( \text{Setup}(\lambda) \) and \( \text{KeyGen}(PP) \) are executed. Then, \( B \) sends \((PP, pk_p, sk_p, pk_d)\) to \( A \).
- **Query phase 1**: \( A \) makes keyword index query.
  - **Keyword index query**: \( A \) can adaptively ask \( B \) for the keyword index \( I = (R_0, R_1, R_2, I_{w_i}) \) as a response.
  - **Challenge**: \( A \) outputs a keyword pair \((w_0, w_1)\). (Neither \( w_0 \) nor \( w_1 \) has been queried in phase 1). \( B \) randomly chooses \( \delta \in \{0, 1\} \), let \( w^* = w_\delta \), and responds the keyword index \( I^* = (R^*_0, R^*_1, R^*_2, I^*_{w_i}) \) to \( A \).
- **Query phase 2**: \( A \) continues to adaptively query as in query phase 1, \( w_i \neq w_0, w_1 \).
  - **Guess**: \( A \) outputs \( \delta' \in \{0, 1\} \), if \( \delta' = \delta \), the adversary \( A \) wins the game.

**Game 2**: \( A \) is assumed to be an inside attacker.
- **Setup**: The algorithms \( \text{Setup}(\lambda) \) and \( \text{KeyGen}(PP) \) are executed. Then, \( B \) sends \((PP, pk_p, sk_d)\) to \( A \).
- **Private key query**: \( A \) makes private key query, \( B \) sends \( sk_d \) and maintains \( List = (sk_d, c) \).
- **Trapdoor query**: \( A \) can adaptively ask \( B \) for the trapdoor for \( w_i \), \( B \) constructs \( Tw = (Tw_1^*, Tw_2^*) \) as a response.
  - **Challenge**: \( A \) outputs a keyword pair \((w_0, w_1)\). (Neither \( w_0 \) nor \( w_1 \) has been queried in phase 1). \( B \) randomly chooses \( \delta \in \{0, 1\} \), let \( w^* = w_\delta \), and responds the keyword index \( Tw^* = (Tw_1^*, Tw_2^*) \) to \( A \).
  - **Query phase 2**: \( A \) continues to adaptively query as in query phase 1, \( w_i \neq w_0, w_1 \).
  - **Guess**: \( A \) outputs \( \delta' \in \{0, 1\} \), if \( \delta' = \delta \), the adversary \( A \) wins the game.

5 Scheme construction and security proof

This section presents our scheme and the security proof in detail.

5.1 Scheme construction

The scheme comprises three phases: system setup, data generation and storage, data search and access.

- **Setup(\( \lambda \))**: Let \( \lambda \) be the security parameter and \((g, p, G_1, G_2, e)\) be the bilinear pairing. Select a one-way collision-resistant hash function \( H : \{0, 1\}^* \rightarrow \mathbb{Z}_p^* \). Set \( Y = \)
The public parameters are \( PP = (g, p, e, G_1, G_2, H, Y) \).

- \( KeyGen(PP) \): The patient randomly chooses \( sk_p \in \mathbb{Z}_p^* \) as the private key, and computes the public key \( pk_p = g^{sk_p} \). What is more, the patient needs to register in hospital’s system and obtain a unique identity \( ID_i \). The doctor randomly chooses \( sk_d \in \mathbb{Z}_p^* \) as the private key, and computes the public key \( pk_d = g^{sk_d} \). Similarly, the data user randomly chooses \( sk_u \in \mathbb{Z}_p^* \) as the private key, and computes the public key \( pk_u = g^{sk_u} \).

**Phase 2: Data generation and storage**

- \( Enc(PP, m, sk_d, pk_p) \): The patient shows \( ID_i \) to the doctor as authorization for the doctor to manage his/her EHR. The doctor generates an EHR \( m \) and extracts keywords \( W = \{w_1, w_2, \ldots, w_n\} \) after diagnosis. Then, the doctor encrypts the EHR as follows.

\[
\begin{align*}
\text{Randomly chooses } & \alpha \in \mathbb{Z}_p^* \text{ and calculates } C_1 = m \cdot \gamma^{\alpha sk_d}, \\
C_2 & = pk_p^{\alpha sk_d}, \\
C_3 & = g^{\alpha h_m}.
\end{align*}
\]

- Computes \( h_{1D_i} = H(ID_i), 1 \leq i \leq n \) and constructs a polynomial \( h(x) = (x - h_{1D_1})(x - h_{1D_2}) \cdots (x - h_{1D_n}) \).

Let the keyword index \( I = \{R_0, R_1, R_2, I_l\} \) by the following operations.

- Randomly chooses \( r \in \mathbb{Z}_p^* \) and computes \( R_0 = Y^r, R_1 = pk_p^r, R_2 = \gamma^{sk_d r} \).
- Computes \( I_i = g^{-rh_{w_i}}, \text{ where } h_{w_i} = H(w_i), 1 \leq i \leq n \).

The EHR ciphertext can be denoted as \( C_m = (C_1, C_2, C_3) \). Then, the doctor uploads \( C_m \) and \( h(x) \) to the cloud and obtains \( address \).

- \( IndGen(PP, W, sk_d, pk_p) \): Moreover, the doctor generates a keyword index of the keyword set \( W = \{w_1, w_2, \ldots, w_n\} \) by the following operations.

- Randomly chooses \( r \in \mathbb{Z}_p^* \) and computes \( R_0 = Y^r, R_1 = pk_p^r, R_2 = \gamma^{sk_d r} \).
- Computes \( I_i = g^{-rh_{w_i}}, \text{ where } h_{w_i} = H(w_i), 1 \leq i \leq n \).

Let the keyword index \( I = \{R_0, R_1, R_2, I_l\} \). Then, the doctor sends \( I \) and \( address \) to the public cloud.

**Phase 3: Data search and access**

**Case I: the patient searches for and obtains EHR**

When a patient is transferred to another hospital or requests a claim from an insurance company, the patient needs access to the electronic medical record. Case I describes the patient searching and obtaining EHR. The patient generates the trapdoor and sends it to the public cloud. After receiving the trapdoor, the public cloud runs the search algorithm and returns the EHR ciphertext storage address to the patient. The patient sends the address and identity to the medical cloud. If the medical cloud verifies the identity, the patient will obtain the EHR ciphertext. Finally, the patient decrypts the EHR ciphertext to get EHR. The system model of case I is shown in Fig. 2. In addition, we give a flow diagram, as shown in Fig. 3 to make our scheme clearer.

- \( TrapdoorGen(PP, W', sk_p, pk_d) \): The patient generates the trapdoor of the keyword set \( W' = \{w'_1, w'_2, \ldots, w'_l\} \) that he/she wants to search, where \( l < n \).

\[
\begin{align*}
\text{Randomly chooses } & \varphi \in \mathbb{Z}_p^* \text{ and sets } T w_1 = \varphi, \\
\text{Computation: } & T w_2 = (pk_d g^{-\psi})^{-1} \cdot H(w'_1), \text{ where } h'_{w_i} = H(w'_i).
\end{align*}
\]

Then, the doctor sends \( T w = (T w_1, T w_2) \) to the public cloud.

- \( DataSearch(I, Tw) \): After receiving the trapdoor from the patient, the public cloud verifies the equation \( e(R_1 \prod I_i, Tw_2)R_0^{Tw_1} = R_2 \). If the equation holds, the public cloud returns \( address \) to the patient. Otherwise, the public cloud returns \( \perp \).

**Correctness:**

\[
e(R_1 \prod I_i, Tw_2)R_0^{Tw_1} = e(pk_p^r \prod g^{-rh_{w_i}}, (pk_d g^{-\psi})^{-1} \cdot H(w'_1)) \psi
\]
Fig. 3 Flow diagram of algorithms (Case I)

DataAccess(address, IDi) : The patient sends IDi and address to the medical cloud. The medical cloud computes \( h_{IDi} = H(ID_i) \) and verifies the equation \( h(h_{IDi}) = 0 \). If the equation holds, the medical cloud returns \( C_m \) at address to the patient. Otherwise, the medical cloud returns ⊥.

Correctness: \( h(h_{IDi}) = (h_{ID1} - h_{ID2}) (h_{ID1} - h_{ID2}) \cdots (h_{ID1} - h_{IDn}) = 0 \)

Dec(\( C_m, sk_p \)) : The patient calculates \( C_1/e(C_2, g^{1/skp}) \) to get \( m \) and checks whether the equation \( e(g^{h_{m,sk_p}}, C_2) = e(C_3, pk_d) \) holds or not. If the equation holds, the decryption is correct.

Correctness:

\[
C_1/e(C_2, g^{1/skp}) = mY^{a_{sk_d}/e(pk_p, g^{1/skp})}
= mY^{a_{sk_d}/Y^{a_{sk_d}}}
= m
\]

Case II: the data user searches for and shares EHR

When a research institute or pharmaceutical company investigates the symptoms of a certain disease (e.g., COVID-19) to formulate treatment plans better, the research institute or pharmaceutical company needs to search and share the EHR for the specified keywords as a data user. Case II describes the data user searching and sharing EHR. When the data user requests to share EHR, the doctor specifies a set of search keywords for the data user. For example, suppose the data user requests to share EHR about COVID-19. In that case, the doctor specifies the search keywords such as “nasal congestion”, “headache”, “cough”, “fever”, and other COVID-19 related keywords. Still, the data user cannot search for “stomachache”, “heart disease”, and other non-COVID-19 related keywords. The doctor, the patient and the data user interact to generate proxy re-encryption keys \( rk_1 = sk_d/skp \) and \( rk_2 = g^{ek_1} \). The doctor uploads \( rk_1 \) to the public cloud. Then, the public cloud re-encrypts the keyword index. The data user generates the trapdoor and sends it to the public cloud. After receiving the trapdoor, the public cloud runs the search algorithm and returns the EHR ciphertext storage address to the data user. The doctor sends \( rk_2 \) to the medical cloud, and the medical cloud re-encrypts the
EHR ciphertext. At the same time, the data user computes the token and sends it and the address to the medical cloud. If the medical cloud verifies the token, the data user will obtain the re-encrypted EHR ciphertext. Finally, the data user decrypts the ciphertext to get EHR. The system model of case II is shown in Fig. 4. In addition, we give a flow diagram, as shown in Fig. 5 to make our scheme clearer.

- **IndRe/enc(I, rk1)**: The doctor uploads $rk_1$ to the public cloud, then the public cloud computes $R'_1 = R_1^{k_1}$. Let the new keyword index $I' = (R_0, R_1', R_2, I_1)$.

- **TrapdoorGen(PP, W′′, skp, pkd)**: For the specified keyword set $W'' = (w''_1, w''_2, \ldots, w''_l)$, the data user constructs the trapdoor.

  - Randomly chooses $\varphi' \in Z_p^*$ and sets $T w'_1 = \varphi'$.
  - Computes $T w'_2 = (p_{k_d} g^{-\varphi'})^{i_{ku} - \sum h''_{w_i}}$, where $h''_{w_i} = H(w''_i)$.

  The data user sends $T w' = (T w'_1, T w'_2)$ to the public cloud.

- **DataSearch(I', T w')**: Upon receiving the trapdoor, the public cloud checks whether the equation $e(R'_1 \prod I_1, T w'_2) R_0^{T w'_1} = R_2$ holds or not. If the equation holds, the public cloud returns $address$ to the patient. Otherwise, the public cloud returns ⊥.

**Correctness:**

$$e(R'_1 \prod I_1, T w'_2) R_0^{T w'_1}$$

$$= e(p_{k_d} g^{-\varphi'})^{i_{ku} - \sum h''_{w_i}} Y^r \varphi'$$

$$= e(g^{i_{ku} - \sum h_{w_i}}, (g^{i_{kd} - \varphi'})^{i_{ku} - \sum h''_{w_i}}) Y^r \varphi'$$

$$= e(g^r, g^{\varphi'}) e(g^{r'}, g^{\varphi'})$$

$$= e(g^r, g^{\varphi'})$$

$$= R_2$$

- **DataAccess(address, Cm, rk2, token)** : The doctor sends $rk_2$ to the medical cloud. At the same time, the data user computes $token = p_{k_p}^{1/sk_u}$ and sends $token$ and $address$ to the medical cloud. Upon receiving $token$ and $address$, the medical cloud verifies the equation $e(rk_2, token) = Y$. If the equation holds, the medical cloud computes $C'_m = e(C_2, rk_2)$ to re-encrypt the EHR ciphertext stored at $address$. Then, the medical cloud returns the re-encrypted EHR ciphertext $C'_m = (C_1, C'_2, C_3)$ to the data user. Otherwise, the medical cloud returns ⊥.

**Correctness:**

$$e(rk_2, token)$$

$$= e(p_{k_p^{1/sk_u}})$$

$$= e(g^{sk_u/sk_p}, g^{sk_p/sk_u})$$
\[ e(g, g) = Y \]

- \( \text{Dec}(C'_m, sk_a) \): The data user computes \( C'_1/C'_2^{1/sk_a} \) to get \( m \) and verifies the equation \( C'_2^{sk_m}/e(C'_3^{sk_a}, pk_d) = 1 \). If the equation holds, the decryption is correct.

**Correctness:**

\[
\begin{align*}
C_1/C_2^{1/sk_a} & = mY^{\alpha sk_d}/e(g^{\alpha sk_d}, g^{sk_a})^{1/sk_u} \\
& = mY^{\alpha sk_d}/Y^{\alpha sk_kd} \\
& = m \\
C_2^{sk_m}/e(C_3^{sk_a}, pk_d) & = e(g^{\alpha sk_kd}, g^{sk_a})^{h_m}/e(g^{\alpha h_m sk_a}, g^{sk_d}) \\
& = e(g, g)^{\alpha sk_kd sk_m h_m}/e(g, g)^{\alpha sk_a sk_kd h_m} \\
& = 1
\end{align*}
\]

**5.2 Security proof**

In this subsection, we analyze security. The analysis of games is as follows.

**Theorem 1** If the DDH assumption and q-ABDHE assumption hold, the scheme is secure against a chosen keyword attacks.

**Lemma 1** The proposed scheme is secure against a chosen keyword attack in Game 1, assuming the DDH problem is intractable.

**Proof** Suppose there exists a polynomial time adversary \( A \) in Game 1, which can attack our scheme. We build a challenger \( B \) that can solve the DDH problem.

\( B \) inputs a DDH instance \((g, A = g^a, B = g^b, T)\), and has to distinguish \( T = g^{ab} \).

**Setup:** Let \( \lambda \) be the security parameter and \((g, p, G_1, G_2, e)\) be the bilinear pairing parameters. Select a one-way collision-resistant hash function \( H : \{0, 1\}^* \rightarrow \mathbb{Z}_p^* \). The public parameters are \( PP = (g, p, e, G_1, G_2, H, Y) \). \( B \) randomly chooses \( sk_d, x \in \mathbb{Z}_p^* \) and sets the doctor’s public key as \( pk_d = g^{sk_d} \). Let \( pk_p = A^x = g^{ax} \) and \( sk_p = x \) as the patient’s public key and private key. Then, \( B \) sends \((PP, pk_p, sk_p, pk_d)\) to \( A \).

**Query phase 1:** \( A \) makes keyword index query.

**Keyword index query:** \( A \) makes keyword index query on the keyword on \( w_i \). \( B \) randomly chooses \( r \in \mathbb{Z}_p^* \) and computes...
\textbf{Trapdoor query:} \(R_0 = e(g, g)^{\gamma}, R_1 = pk_p, R_2 = e(g, g)^{sk_d}\) and \(I_{w_i} = g^{-h_{w_i}}, H_{w_i} = H(w_i), \) B sends \(I = (R_0, R_1, R_2, I_{w_i})\) to \(A.\)

**Challenge:** \(A\) outputs a keyword pair \((w_0, w_1).\) \(B\) randomly chooses \(\delta \in \{0, 1\},\) let \(w^* = w_\delta\) and computes

\[
R_0^* = e(g, B)^{\gamma/2}, \\
R_1^* = T, \\
R_2^* = e(g, B)^{sk_d/2}, \\
I_{w^*} = B^{-h_{w^*}/2}, h_{w^*} = H(w^*).
\]

Then, \(B\) responds \(I^* = (R_0^*, R_1^*, R_2^*, I_{w^*})\) to \(A.\)

**Query phase 2:** \(A\) continues to adaptively query as in query phase 1, \(w_i \neq w_0, w_1.\)

**Guess:** \(A\) outputs \(\delta' \in \{0, 1\},\) if \(\delta' = \delta,\) then output 1 meaning \(T = g^{ab} ;\) else output 0 meaning \(T \neq g^{ab}.\)

**Lemma 2** The proposed scheme is secure against a chosen keyword attack in Game 2, assuming q-\text{ABDHE} problem is intractable.

**Proof** Suppose there exists a polynomial time adversary \(A\) in Game 2, which can attack our scheme. We build a challenger \(B\) that can solve the q-\text{ABDHE} problem.

Let \(q_k\) be the number of trapdoor queries, \(q > q_k + 1.\) \(B\) inputs a q-\text{ABDHE} instance \((g, g^x, g^{x^2}, \cdots, g^{x^q}, g^z, g^{x+y+1}, T)\) and has to distinguish \(T = e(g, g)^{x+y+1}.\)

**Setup:** Let \(\lambda\) be the security parameter and \((g, p, G_1, G_2, e)\) be the bilinear pairing parameters. Select a one-way collision-resistant hash function \(H : \{0, 1\}^* \rightarrow Z_p^*.\) The public parameters are \(PP = (g, p, e, G_1, G_2, H, Y).\) Choose a random degree \(q)\) polynomial \(f(x).\) \(B\) sets \(pk_p = g^x\) and \(pk_d = g^f(x)\) as the patient’s public key and the doctor’s public key, respectively. Then, \(B\) sends \((PP, pk_p, pk_d)\) to \(A.\)

**Query phase 1:** \(A\) makes the following queries.

**Private key query:** \(B\) maintains \(List = (sk_d, c).\) \(A\) makes private key query. \(B\) checks if \(sk_d\) is in the \(List.\) If \(sk_d\) exists, \(B\) returns \(sk_d\) to \(A,\) or randomly chooses \(c = \{0, 1\}.\) If \(c = 1,\) \(B\) returns \(sk_d = f(x)\) to \(A.\) Otherwise, \(B\) outputs a random number and aborts.

**Trapdoor query:** When \(A\) makes trapdoor query on the keyword \(w_i, B\) queries \(List.\) If \(c = 0, B\) sets \(T w_1 = f(h_{w_1})\) and computes \(T w_2 = g^{f(h_{w_1})/2}(h_{w_1})\), where \(h_{w_1} = H(w_1).\) Then \(B\) sends \(T w = (T w_1, T w_2)\) to \(A.\) When \(q > q_k + 1, T w_1 = f(h_{w_1})\) is a random value, since \(f(x)\) is a random degree \(q)\) polynomial. Otherwise, \(B\) returns an error message and aborts.

**Challenge:** \(A\) outputs a keyword pair \((w_0, w_1).\) \(B\) randomly chooses \(\delta \in \{0, 1\},\) let \(w^* = w_\delta.\) Then, \(B\) runs the above algorithms to get \(List = (sk^*, c^*).\) If \(c^* = 1, B\) outputs an error message and aborts. Otherwise, \(B\) sets \(T w_1^* = f(h_{w^*})\) and computes \(T w_2^* = g^{f(h_{w^*})/2}(h_{w^*}),\) where \(h_{w^*} = H(w^*).\) Defines the degree \(q + 1)\) polynomial \(F(x) = (x^{q+2} + h_{w^*})/(x - h_{w^*}) = \sum_{i=0}^{q+1} (F_i x^i),\) Computes

\[
R_0^* = T F_{q+1}^{e}(g^z, \prod_{i=0}^{q}(g^{x^i})^{F_i}), \\
R_1^* = g^x \prod_{i=0}^{q}(F_i x^i), \\
R_2^* = e(R_1, w^*, T w_2) R_0^T w_1^*, \\
I_{w^*} = g^{-F(x)h_{w^*}}.
\]

\(B\) sends \(I^* = (R_0^*, R_1^*, R_2^*, I_{w^*})\) to \(A.\) Let \(r^* = z F(x),\) if \(T = e(g, g)^{x+y+1}\) then

\[
R_0^* = T F_{q+1}^{e}(g^z, \prod_{i=0}^{q}(g^{x^i})^{F_i}), \\
R_1^* = g^x \prod_{i=0}^{q}(F_i x^i), \\
R_2^* = e(R_1, w^*, T w_2) R_0^T w_1^*, \\
I_{w^*} = g^{-F(x)h_{w^*}}.
\]

**Query phase 2:** \(A\) continues to adaptively query as in query phase 1, \(w_i \neq w_0, w_1.\)

**Guess:** \(A\) outputs \(\delta' \in \{0, 1\},\) if \(\delta' = \delta,\) then output 1 meaning \(T = e(g, g)^{x+y+1};\) else output 0 meaning \(T \neq e(g, g)^{x+y+1}.\)

**6 Scheme construction and security proof**

In this section, we analyze the performance and efficiency of the proposed scheme, and compare the scheme with Wu et al. (2016), Wang et al. (2019), Xue (2022) and Liu et al. (2021).

**6.1 Theory analysis**

(1) Functionality comparison

The functions of the four schemes are compared, as shown in Table 1. It is evident that all the schemes support secure search. Whereas our scheme and the scheme of Wang et al. (2019) support multi-keyword search, only ours supports specified keywords search.

(2) Comparison of communication overhead

We compare the communication overhead of the four schemes in terms of EHR storage, data search, and data access, as shown in Table 2. We define the element length of \(G_1, G_2,\) and \(Z_p^*\) as \(|G_1|, |G_2|,\) and \(|Q|\). Since the data communicated in each scheme belongs to the elements in \(G_1, G_2,\) and \(Z_p^*,\) the communication overhead of the scheme can be effectively represented by counting the above parameters. Since the support of conjunctive keyword search, our

\[\sum_{i=0}^{q+1} (F_i x^i),\]
### Table 1 Comparison of Functionality

|                  | Secure search | Multi-keyword search | Specified keywords search |
|------------------|---------------|----------------------|---------------------------|
| Wu et al. (2016) | ✓             | ×                    | ×                         |
| Wang et al. (2019)| ✓             | ✓                    | ×                         |
| Xue (2022)       | ✓             | ×                    | ×                         |
| Ours             | ✓             | ✓                    | ✓                         |

### Table 2 Comparison of communication overhead

|                  | EHR storage | Data search | Data access |
|------------------|-------------|-------------|-------------|
| Wu et al. (2016) | $(n + 2)|G_1| + |G_2| | 2|G_1| | 2|G_1| + |G_2| |
| Wang et al. (2019)| $(2n + 7)|G_1| + |G_2| | (3 + l)|G_1| | 4|G_1| + |2|G_2| |
| Xue (2022)       | $5|G_1| + (n + 1)|G_2| | |G_1| | 7|G_1| + |2|G_2| |
| Ours             | $(n + 4)|G_1| + |2|G_2| | |G_1| | |Q| | |G_1| | 2|G_1| + |G_2| + 2|G_2| |

### Table 3 System configuration and mathematical parameters

|                  |                   |
|------------------|-------------------|
| CPU              | Intel(R) Core(TM) i7-5500U CPU @ 2.40GHz |
| OS               | Linux and Ubuntu10.10 |
| Program language | The C Programming Language |
| Program library  | Pairing-Based Cryptography (PBC) |
| Pairing          | Type A |
| Elliptic curve   | $y^2 = x^3 + x$ |
| Base field       | 512 bit |
| Group order      | $2^{159} + 2^{107} + 1$ |

scheme has a low data search communication overhead. Our communication overhead is close to that of Wu et al. (2016) in the EHR storage and data access phase.

### 6.2 Numerical analysis

We emulate the proposed protocol, Wu et al. (2016), Wang et al. (2019), Xue (2022), and Liu et al. (2021). Liu et al. (2021) is currently a popular attribute-based searchable encryption (ABSE) scheme. Since the encryption and decryption of ABSE are related to the number of attributes, we only compare the search algorithm of Liu et al. (2021). The experiments are implemented using C language and Pairing-Based Cryptography (PBC) on a PC with Linux operating system. The system configuration and mathematical parameters are shown in Table 3. For Pairing, Type A is a common type, and 512 bit is a safe length for base field. The experimental results are shown in Figs. 6, 7, and 8.

Figure 6 shows that the encryption running time of the proposed scheme and Wu et al. (2016) is much less than that of Wang et al. (2019) and Xue (2022). Since the encryption of Wang et al. (2019) and Xue (2022) is related to the number of keywords, the time cost is more as the number of keywords increases. In addition, our encryption time is very close to that of Wu et al. (2016), but a bit longer than Wu et al. (2016).

From Fig. 7, our keyword search algorithm is the most efficient, while Wu et al. (2016) is the lowest. Compared with the more current ABSE (Liu et al. 2021), the search efficiency of
Fig. 8 Time cost of decryption

our scheme is slightly better. With the increase in keywords, our advantage will become more and more apparent, which means that our scheme is suitable for big data search.

In Fig. 8, the decryption time of the Case II is much less than the scheme of Wu et al. (2016) and the scheme of Xue (2022), and the decryption running time of Case I is most. Since the decryption algorithm of the Case II and Wang et al. (2019) contain similar operations, the decryption time cost of Case II is very close to that of Wang et al. (2019).

7 Conclusion

This paper proposes an EHR search and sharing scheme based on searchable encryption with specified keyword search and proxy re-encryption, achieving multi-keyword search and protecting the privacy and EHR security. The scheme is secure against chosen keyword attack. In addition, the experimental results show that the search of the scheme is more effective than other existing schemes. What is not perfect is that the proposed scheme’s encryption and decryption running time is slightly more than that of the comparison scheme. Therefore, how to improve the encryption and decryption efficiency while ensuring security is the focus of the subsequent research. For example, in a mutliuser scenario, we can consider applying an attribute-based encryption algorithm or broadcast encryption algorithm to the scheme to improve the efficiency of encryption and decryption.

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Data availability Data available on request from the authors.

Declarations

Conflict of interest The authors have not disclosed any competing interests.

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