

# Hybrid Cryptographic Access Control for Cloud based Electronic Health Records Systems

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## ABSTRACT

Cloud based Electronic Health Record (EHR) systems are next generation "big data systems" for facilitating a) efficient and scalable storage, and b) to foster collaborative care, clinical research and development. Mobility and use of multiple mobile devices in collaborative healthcare intrigue robust privacy preservation. Thus, large scale EHR systems require secure access to privacy sensitive EHR data, data storage and management. We provide a comprehensive solution with i) a cryptographic role based technique to distribute session keys to establish communications and information retrieval using Kerberos protocol, ii) location and biometrics based authentication method to authorize the users and iii) a wavelet based steganographic technique to embed EHR data securely using ECG biometric as the host in a trusted cloud storage. Based on a comprehensive security analysis, our model proves to be a scalable, secure and a reliable model to access and manage EHR data.

**Keywords:** Electronic Health Record(EHR), Access Control, Location Awareness, Steganography.

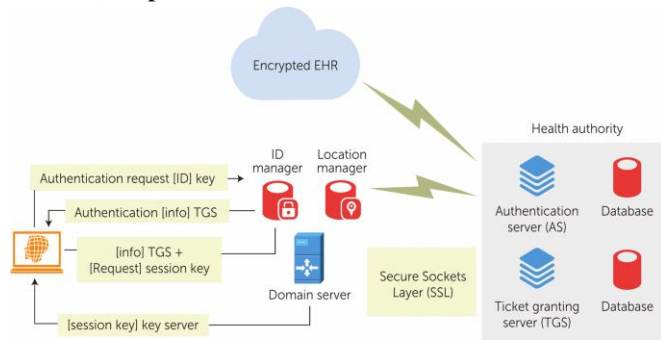
## I. INTRODUCTION

Electronic health records (EHR) Systems offer more efficient means for delivering quality ensured healthcare services and to promote collaborative clinical research. EHRs consist of data pertaining to "all aspects of care" (e.g. genomic test results, diagnosis, medication, laboratory test results and imaging data). According to Australian Bureau of statistics over 23 millions of Australian population [3], if the average EHR data file is 01 Gigabytes, then, the overall EHR data amounts to petabyte scale. According to IBM big data is defined as any situation or event that generates data with any or all of the three properties: Volume, Variety and Velocity [1]. Thus, it is evident that we have a significant "big EHR data" management problem in terms of volume, veracity and velocity. Poorly implemented EHR based systems pose significant risks for patient safety and data misuses [10]. The main security concerns in big data systems include secure-storage, secure-access and secure retrieval [2]. In addition to robust access control mechanisms, location of data access is an important aspect of secure data usage. Recently reported incidents

on illegal trade and stealing of patient data over mobile devices and technologies intrigue research on secure data usage based on location. Multi-factor user authentication using multiple features with location validations is a promising solution. Healthcare service facilities increasingly use mobile devices to improve the work flow dynamics and efficiency. However, user mobility and use of multiple mobile devices make the above mentioned security concerns more severe and demands robust privacy preserving measures. Based on the above discussion, the main research questions that we address here are: How to securely store and manage "big EHR data"? and How to ensure secure location-aware access to "big EHR data"? Cloud based utility services (e.g. storage) offer additional benefits to EHR systems such as being more cost effective with the ease of management and to collaborate with mobile technologies and devices to gather data [4][5]. In addressing the first research question we find ECG as an appropriate host signal which is a verifiable secure feature to store "big EHR data" in cloud. For the second research question we propose a cryptographic role based access control which is a scalable solution to facilitate access to large number of users. Our solutions



to users (e.g. patients, physicians, nurses, or lab workers), to perform different tasks.



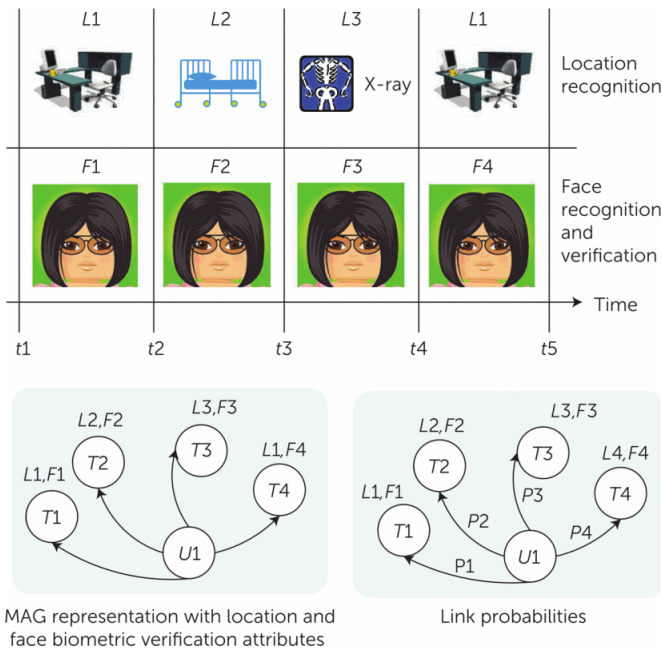
**Figure 2.** The Kerberos model infrastructure

The role hierarchy is used to identify patients, healthcare professionals (e.g. physicians, surgeons, cardiologists, gynecologists) and support staff (e.g. nurses, laboratory worker) assigned with different permissions. The AS and TGS work in a complementary manner through secure communication to automate the authentication and authorization processes for users. The contribution of key distribution process is to check the users' credentials during the authorization process and distributes appropriate session keys for a user after receiving encrypted authorization information. The IU uses as a part of authentication process to verify mobile users locations through communication with domain server. In figure. 2, the first message pair is exchanged between users and the AS to prove the identities of users. This pair of messages are fully encrypted through the use of the AS credentials. The second pair of messages involves communication between the AS and the DS to verify users locations. In Section (A) we explain the process for location verification. Validation decision from DS to AS completes the authentication process. The third pair of messages exchanged between users and the TGS to complete the authorization process. The initial user request, validated credentials and the authentication details are encrypted and forwarded to TGS. TGS accepts and issues the session key based on the role-hierarchy. (A) Location Verification-The location validation is performed during the initial authentication of the user by the HA using a validation request to the DS and before delivering paragraphs must be indented. the extracted information content to the user. In the first instance location validation is vital to ensure the service request to be from a legitimate user belong to a secure location known to the respective DS. Before transmitting the extracted content to the user, HA contacts the DS to

validate the current location of the user as a secure location. This validation is necessary as for the mobile user the location can change during the initial request was made and till the extraction process is completed. For example if the user is using an unmanaged WiFi network while travelling, then sensitive health information are susceptible to malicious eavesdropping and signal interceptions which can cause sever privacy breeches. Location-manager (see Figure 1) is able to verify the current location of the user and also can provide the trace based validity for the most recent  $0n0$  number of location changes. The affiliation of these  $n$  locations (i.e. the attributes) corresponding to two time instances (i.e. the nodes) can be computed using the multiplicative attribute graph (MAG) model [7]. The similarity between two sets of locations (known vs claimed locations) to declare the validity is interpreted as the affiliation between the attributes using the MAG model. When the Location-Manager gets the validation request from HA, it performs a proximity verification using the known location markers and the trace history information (Figure 3). Once this location markers are selected Location Manager further verifies with the Identity Manager by inferring the footage of the user by verifying the face biometric signature trace. Then the most likely location markers are selected using Multiplicative Attribute Graph (MAG) [7] (i.e. the link with the highest probability). If  $0k$  out-of- $n0$  (where  $k=n > 60\%$ ) instances prove the users location and identity, then the user is validated as a legitimate user in a secure location known to the DS. Our approach is more robust as the verification is strongly coupled with the location information and face biometric trace within that time period.

## B. EHR Embedding And Retrieval

In this model, we assume that HA is fully secure and responsible for generating the security parameters. Since we use lossy steganography, it does not increase the size as each bit of the original data is replaced by another bit of the hidden data. Simple a user  $U1$  the face biometric features



**Figure 3.** Identity verification using the location trace of for the time period ( $t_1$   $t_5$ ) is represented using MAG model. bit replacement can significantly distort the original ECG. However, minimize the distortion of the host ECG by applying wavelet. When apply wavelet signal transformation, data gets divided into lots of coefficients. Then, we randomly hide the sensitive data into the least significant coefficients to ensure minimum distortion to the data. EHR embedding include, i) organizing the EHR content as distinct sections on to a tree structure and ii) randomly allocating them to different portions of the ECG segments by specifying their Indexes I and Ends E. After splitting ECG Segment, signal transformation technique using Haar wavelet is applied. The result of signal transformation yeild two sets of coefficients: CA and CD. The coefficients are used to classify each Segment a either: 1) the most sensitive features of the original signal (i.e. coefficients approximation CA), and 2) the least significant features (i.e. coefficients detailed CD) which can be freely used in hiding EHR Sections (See Figure 4). Next, for each EHR Section, a hash value H is computed. In order to make the hiding process unique to each patient, a security key (K) will be defined for every patient which will be used to 1) encrypt each Section of EHR before hiding it, 2) reshuffle coefficients (CD), and 3) hide Section bits in a certain set of coefficients. Haar Wavelet recomposition will then be applied on both CA and CD. Consequently, a new watermarked Segment is reconstructed. Next, re-embed the watermarked Segment into the full original ECG signal of that patient using its I and E. The same process will be

repeated to hide all Sections. Finally, HA stores certain information such as I and E of each Segment, the hidden Section number and K along with unique Patient ID which is necessary for retrieving process. HA then stores the watermarked ECG along with generated number mapped to the patient ID on its cloud servers. Therefore, even these information are intercepted, they will not reveal anything. For an auzhorized user, HA will extract ECG Segment from its cloud servers and accomplish the extraction process at its local servers. Then, it uses the session key to encrypt the bits and send them to the user device.

#### IV. SECURITY ANALYSIS

In this section, we present a security analysis for the communication channel i) between users' domain and HA, ii)between HA and cloud provider. Based on the analyses we demonstrate the resilience of the proposed system against man-in-the-middle (MiM) and reply attacks.

##### A. Communications Between User Domain And HA

It provides a qualitative analysis to demonstrate robustness of the communication channel between HA and the user. Consider an instance where an intruder sends a request to CCSP via a secure domain pretending to be a legitimate user with false identities and location information. The intruder will not succeed in gaining access to the EHR system due to two security features. Firstly, CCSP validates with the particular DS to re-validate the user identity and the location. DS validates the trace of the user based on the location and the associated face biometric over n instances. So the likelihood of the intruder to forge the face biometric at all the traced locations is highly unlikely. Therefore, it is clear that a mimicry attack cannot be successful to gaining access to the EHR system via the HA. Moreover, MiM attack is prevented through utilize public key infrastructure (PKI) to perform both authentication and authorization processes. All users have to communication with the HA securely through applying both public/private keys to exchange messages combined with time stamps. The PKI support both confidentiality and integrity and the time stamps are prevent any reply attacks form intruders.

## B. Communications Between HA To Cloud Based EHR Database

Assumption that HA is fully secure for generating and securely storing the key K and PID for all patients. Consider a scenario, where an intruder has access to the watermarked ECG at the cloud servers or during the transmission between cloud DB and HA. The number of all possible combinations of ECG segment hosting a particular EHR section can be defined as follows (Equation 1),

$$P = \prod_{n=1}^n S * \sum_{r=1}^r R * \sum_{c=1}^c C * N^L \quad \dots\dots(1)$$

where P is the total number of possible combinations, n is the number of samples in the ECG, r and c are the rows and columns numbers in the reshuffled coefficients CD, L is the key length and N is its possibilities. Assume n = 1000 (i.e. an ECG of 10 second length), r = 128 and c = 32 (i.e. the size of reshuffled CD) and the key character set is 256 and its symbols length is 256 (See Eq2),

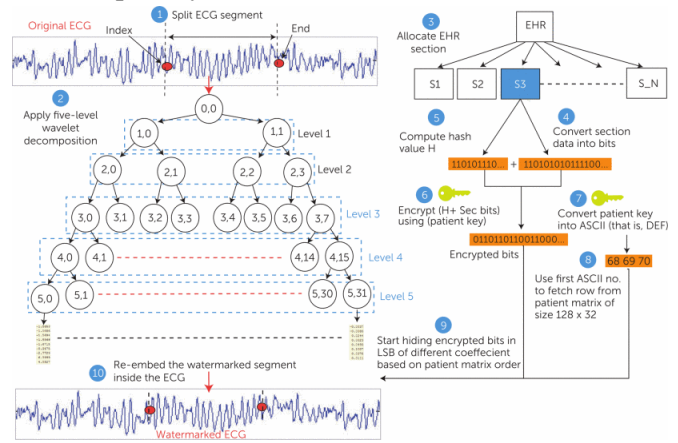
$$p = 1000! * \sum_{r=1}^{128} R! * \sum_{c=1}^{32} C! * 256^{256} \rightarrow p = \infty \quad \dots\dots(2)$$

Thus, it proves that it is highly improbable to find the intended EHR section in reasonable time as shown in Equation 2. In conclusion, the channel between the HA and the cloud based EHR host is resilient to attacks such as man-in-the-middle attack.

## V. CONCLUSION

Secure and efficient storage, access control and retrieval are crucial for big data EHR systems. In the article a hybrid cryptography based access control model for cloud based EHR systems. We use strong user authentication based on biometrics and location with Kerberos protocol to issue encrypted session tickets. The approach ensures a fail-proof two level user authentication for more robust secure interactions with the EHR system. Wavelet based steganographic technique to store the EHR data in a ECG signal which acts as a fail-proof host data to ensure the data to be accessed only to legitimate users. Furthermore, the security analysis proves the robustness of solution in terms of its resilience to some common yet notorious attacks. Thus, our solution offers a significant contribution to developing a large scale distributed EHR system with strong security features to preserve

## the privacy of the EHR data.



**Figure 4.** Block diagram shows the steps of hiding a section of EHR inside a segment of ECG host signal before re-embedding the segment in the watermarked ECG.

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