# BinDaaS: Blockchain-Based Deep-Learning as-a-Service in Healthcare 4.0 Applications

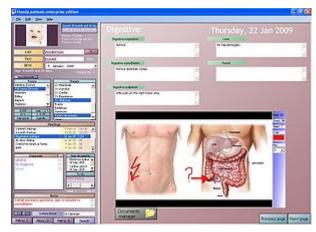
https://youtu.be/Epvu9B44ZfQ





## 개요

- EHR
  - Electronic Health Record (전자의무기록)
  - 디지털 형태로 체계적으로 수집되어 전자적으로 저장된 환자 및 인구의 건강정보
- 본 논문에서는 EHR을 블록체인 기반으로 만듦
- 또한 DaaS(Deep Learning as a Service)를 이용하여 질병을 예측



Sample view of an electronic health record



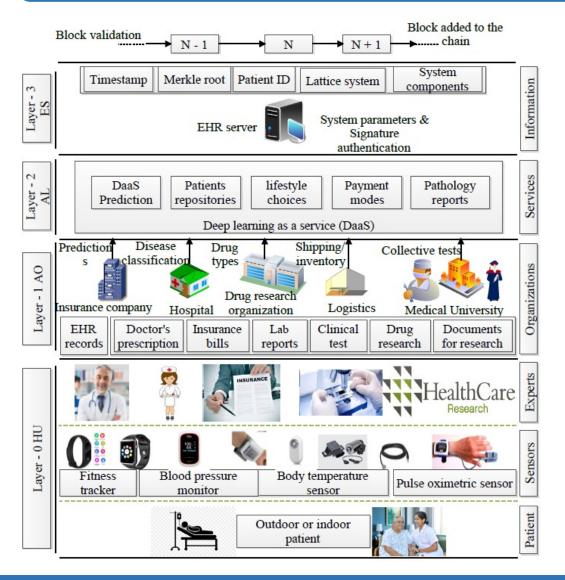
## Motivation

• healthcare 4.0 어플리케이션 개발을 위한 프레임워크 (BinDaaS) 제안

• 해당 프레임워크는 블록체인 백엔드 + 딥러닝 기술로 이루어짐



# System Architecture of BinDaaS



#### CASE-I

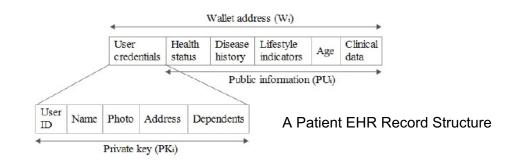
HU를 통해 수집된 데이터는 AO 레이어를 통해 인증 받아야 함 ex) 의사는 병원에 의해 인증을 받아야 함

#### **CASE-II**

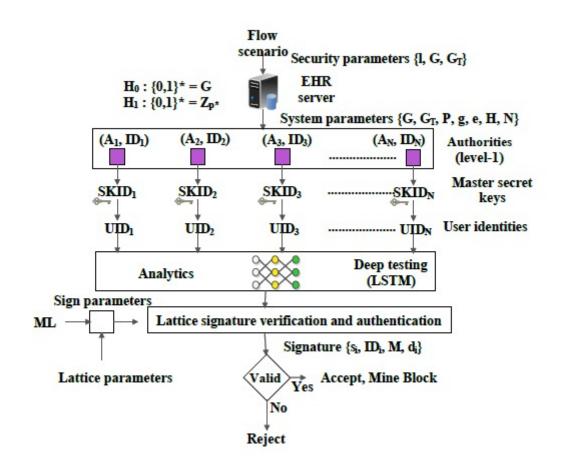
AO 레이어는 ES 서버에 의해 인증을 받아야 함

#### CASE-III

ES 서버의 파라미터들이 충족되면, ES에 의해 ES\_Notary에서 공증 작업을 수행한 뒤 새로운 블록을 생성하여 네트워크의 모든 유저에 전파



# Proposed HCAAM scheme using lattices in BinDaaS



a scheme heterogeneous collective authority authentication mechanism (HCAAM, 다종 집단 권한 인증 메커니즘)

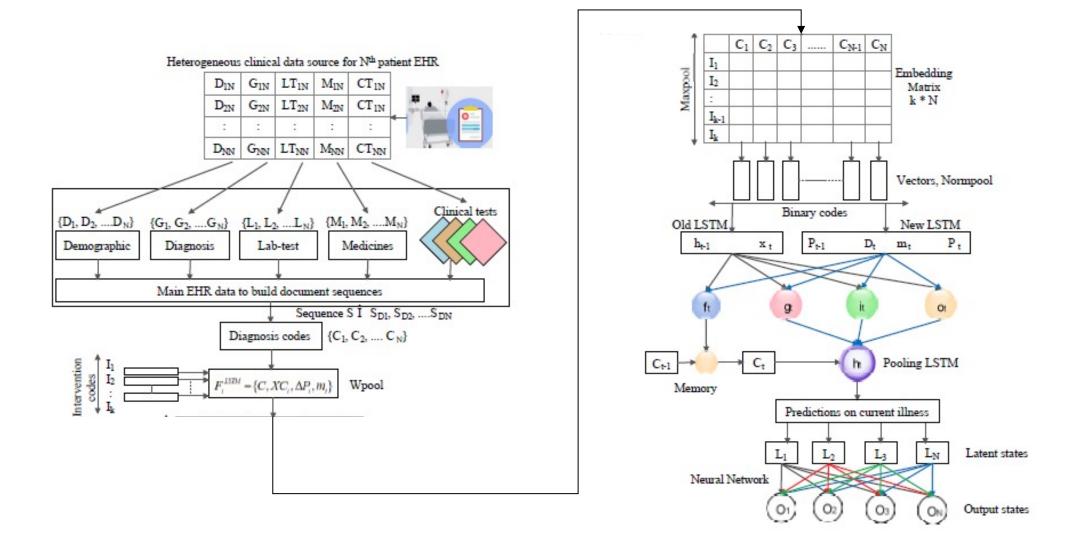
이전 페이지의 그림에서 요청은 아래에서 위로, 시큐리티 파라미터는 위에서 아래로 이동

아래 레이어(AO 등)에서는 다양한 위 레이어(AL 등)의 요청을 처리해야 하기에 다종 집단의 권한에 대한 인증이 필요함

이것을 HCAAM이 처리



### DaaS integration in BinDaaS using LSTM for future risk predictions from EHR



# LSTM DaaS for future prediction of disease

#### Algorithm 3 LSTM DaaS for future prediction of diseases

```
Input: Patient EHR records D_1, D_2, ..., D_n as sequence of admissions
S = \{S_{D_1}, S_{D_2}, ..., S_{D_n}\} for n users.
Patient diagnosis codes C = \{C_1, C_2, ..., C_n\} as feature vectors x_{C_i} \in \mathbb{R}^m
Patient interventions I = \{I_1, I_2, \dots, I_k\} as feature vectors x_{I_i} \in R^m, where m is
vector dimension length, elapsed time \Delta t for each i^{th} patient.
Admission codes from WPool with associated probabilities P(WPool)
Output: Future prediction of patient health based on outcome probability P(y|h_{1,2,\ldots,n}).
Initialization: i = 0, j = 0, memory state of LSTM c = 0;
                                                                                                                             else
                                                                                                                  24:
 1: for (i \leftarrow 1 \text{ to } n) do
                                                                                                                             end if
      F_{I,STM}^i \leftarrow \{x_{c_i}, x_{I_i}, \Delta p_i, m_i\}

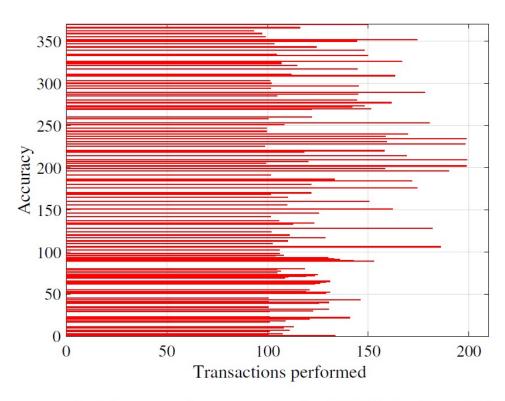
R^m \leftarrow WPool(\varrho_1, \varrho_2, \dots, \varrho_n)
                                                                                                                  26: end for
       K \leftarrow Compute\_length\ P(y|\varrho_{1,2,...,n})
         \Delta P_i^t \leftarrow D_i^t - D_i^{t-1}
 6: end for
7: for (i \leftarrow 1 \text{ to } n) do
         for (j \leftarrow 1 \text{ to } k) do
              W_{ij} \leftarrow Embed\_Matrix ((D, Z))
              B = \{b_0, b_1, \ldots, b_n\}
              x_t^i \leftarrow max\{A^{d_1}, A^{d_2}, \dots, A^{d_n}\}
                                                                                                                  35:
             p_{t}^{f} \leftarrow max\{B_{s}^{I_{1}}, B_{s}^{I_{2}}, \dots, B_{s}^{I_{k}}\}
                                                                                                                  36:
                                                                                                                             end if
          end for
                                                                                                                  37: end while
14: end for
15: for (i \leftarrow 1 \text{ to } n) do
```

```
18: for (j \leftarrow 1 \text{ to } k) do
             WPool \leftarrow \sigma(\sum w_i x_t + U_t h_{t-1})
            A_t \leftarrow \frac{1}{m_t} (\text{WPool}+b_i)
            if (m_t == 1) then
                  A_{I} > 0
                   A_{t} < 0
27: while (z>0) do
            if (P > A_t) then
                  \Delta_{t-1:t} \leftarrow |\log(e + \delta_{t-1:t})^{-1}|
           \begin{aligned} \mathbf{N}_i &\leftarrow \sigma(w_f x_t + u_f h_{t-1} + Q_f q_{\Delta t-1:t} + p_f p_{t-1} + b_f) \\ &SoftMax(z) &\leftarrow \mathbf{e}^z / \sum_{\tau} t e^{Zt} \end{aligned}
               P(d_{t+1} = c | f_t) \leftarrow SoftMax(z)
                  MeanPool \leftarrow h_{1,2,...,n}
                   MeanPool \leftarrow -\log P(y|u_{1,2,...,n})
38: e_h \leftarrow \sigma(h_t + b_h)
39: x_v \leftarrow h_t a_n + b_v
40: P(y|h_{1,2,...,n}) \leftarrow f_{prob}(x_y)
```

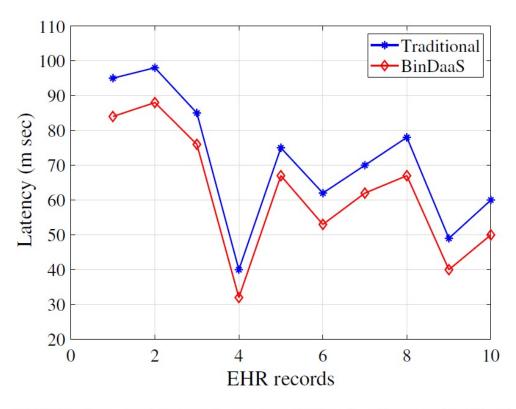
17: end for

 $NormPool \leftarrow m_t + \log(1 + \Delta t)^{-1}$ 

## **Evaluation Results**



((a)) Improved accuracy in the LSTM\_DaaS model



((b)) End-to-end latency over traditional schemes in BinDaaS



### **Evaluation Results**

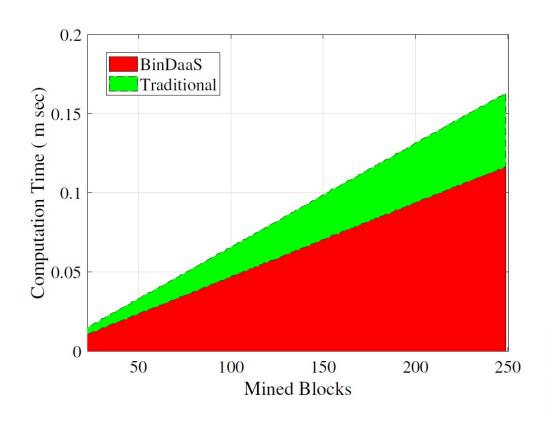


TABLE IV: Comparative Analysis with existing schemes

Parameters	<b>Bao</b> et al. [48]	Li <i>et al.</i> [49]	Hathaliya et al.[46]	<b>Aujla</b> <i>et al.</i> [47]	<b>Proposed</b> BinDaaS
A1	<b>√</b>	×	<b>√</b>	<b>√</b>	<b>√</b>
A2	×	×	$\checkmark$	$\checkmark$	✓
A3	×	×	$\checkmark$	✓	$\checkmark$
A4	×	×	$\checkmark$	$\checkmark$	$\checkmark$
A5	×	×	×	×	$\checkmark$
A6	_	✓	<b>✓</b>	×	$\checkmark$
A7	×	×	×	×	$\checkmark$
A8	×	-	×	$\checkmark$	/
A9	×	×	×	$\checkmark$	$\checkmark$
A10	-	X	✓	✓	✓

A1: Replay Attacks; A2: Side-Channel Attacks; A3:Distributed Denial-of-Service(DDoS) attacks; A4: Session-based attacks A5: Provenance and auditability attacks; A6: Tracebility of attacks; A7: Signature-forgery attacks; A8: Signature verificability; A9; Quantum attacks; A10: Known ciphertext attack; ✓ shows scheme is safe; X shows scheme is not safe; & - shows attack is not considered in the scheme.



# Q&A

