

Secure Activation function using Functional Encryption

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Outline



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- Framework
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Introduction



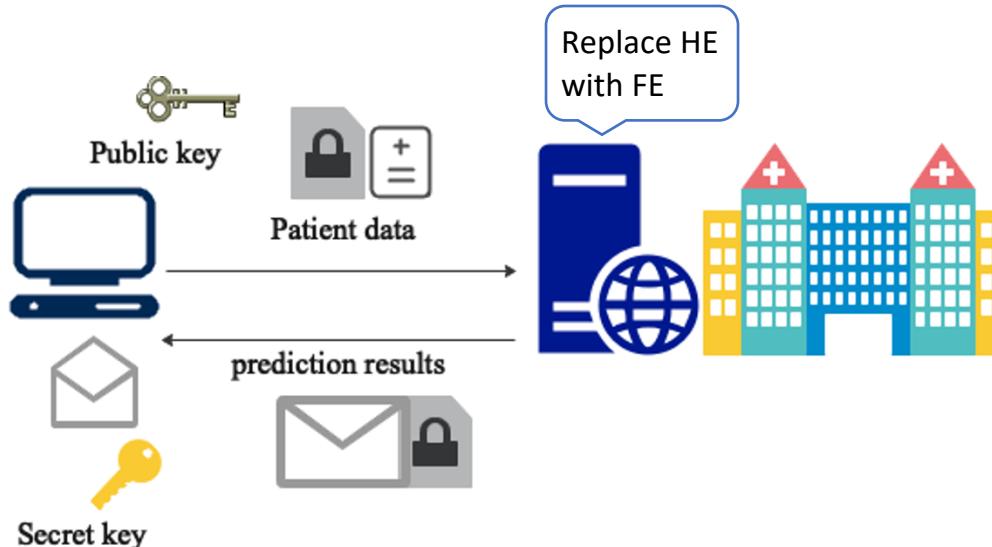
- Many AI applications in the fields like computer vision and NLP
- Google cloud AI, Microsoft azure learning (MLaaS)
- Cloud server is responsible for training a ML model for classification, regression, etc
- Trained on the data collected from the customers
- Privacy ?



Introduction



- E.g. Privacy-preserving cardiovascular disease prediction service using homomorphic encryption



Motivation



- Homomorphic Encryption and Secure Multiparty computation are already doing a great job in PPML then why use Functional Encryption?

Encrypted results, training on encrypted data is not allowed

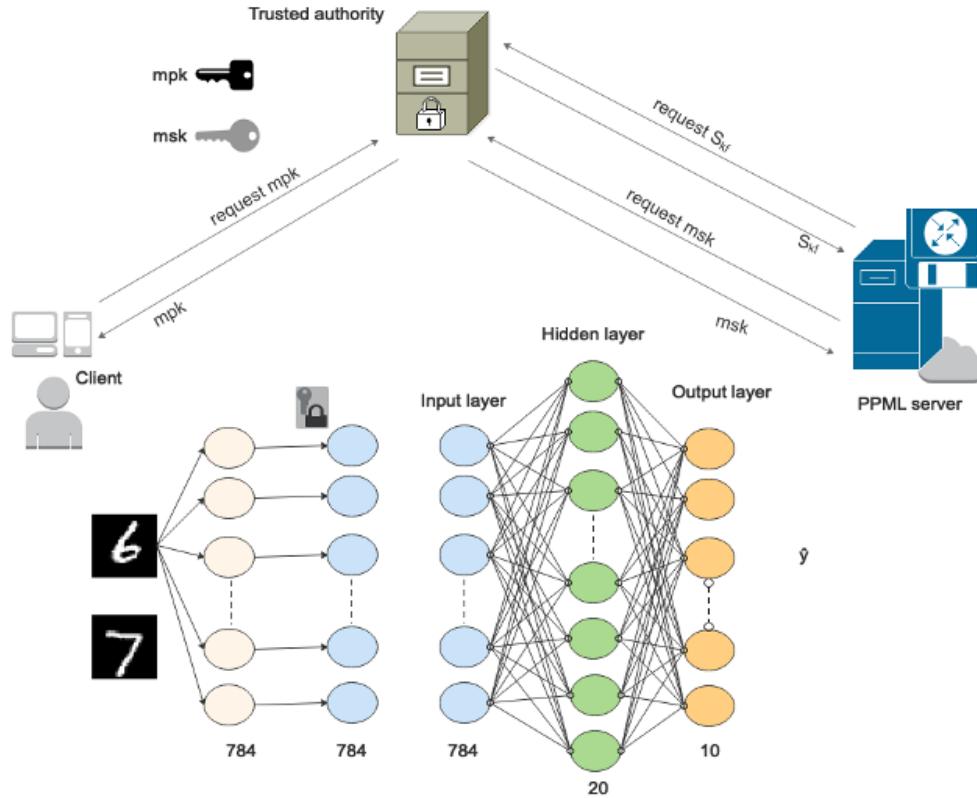
- Can the server train the model without seeing the data but reveals the result in plain?

Functional Encryption is here to help

How to do machine learning out of it ?



Framework



Live demo



SecureReLU



Algorithm 1: SecureRelu

Input: Encrypted X, W, b

Output: Relu(W.X+b)

for $i \leftarrow 0$ **to** $W.size$ **do**

$prod_i \leftarrow Decrypt(Ct_i, sk_{fe}, W_i)$

$result_i \leftarrow Relu(prod_i + b)$

return result

Implementation

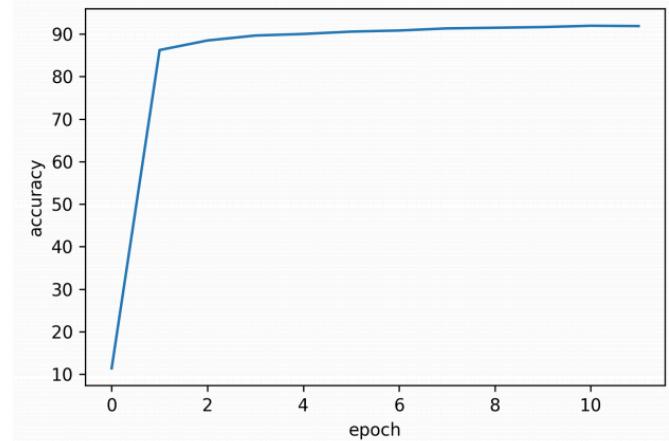


- 2-layer Neural Network, SGD, categorical cross entropy
- MNIST with 20-30% samples
- Numpy (we can't directly use sklearn and keras)
- Developed python functions to use CiFEr [7] library functions
- All computations are done on CPU

Results



MNIST	Epoch	Training time	Accuracy
20 % samples	10	> 2 hrs	85 %



Function	Time
Master keys generation	1.8032 s
FE key generation	0.00196 s
Encryption	0.0109 s
Decryption	0.021168 s

Challenges



- FE has been growing richer and richer in theoretical aspects, but Library support is still limited
- ML computations usually involve large no. of inner products; so computing and managing those many ciphertexts and keys is an overhead
- no GPU support is available for the libraries; so model takes forever to run

References



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Thank you!
Questions?