Homomorphic Evaluation of Convolutional Neural Networks

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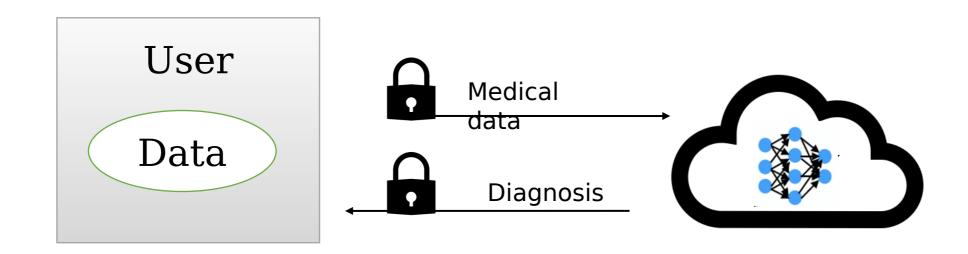
October 25th, Università di Padova

Machine Learning As a Service

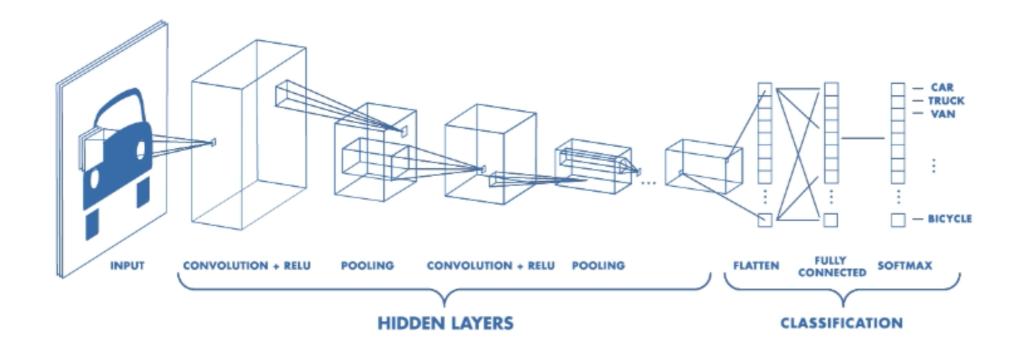


\$1.0 billion in 2019 [] \$8.48 billion by 2025*

Privacy-Preserving Machine Learning



CNN



Activation function: ReLU, Max, sign, Sigmoid, etc.

Fully Homomorphic Encryption (FHE)

Given encrypted_data and a computation F, there exists an alternative F's.t.

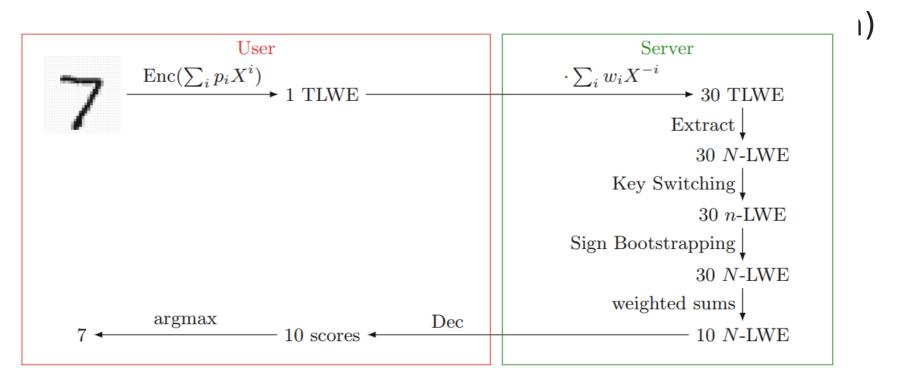


Fig. 1: homomorphic evaluation of a 784:30:10 neural network with activation function sign [Crypto 2018]

Prior Frameworks Are Not the Best

- From the activation function side
 - Only evaluate the sign function
 - Others are converted to polynomials
- From the message space side
 - Limited to 4-5 bits (eg. [0,128]□-1, [128,255]□1)
- Batch
 - Not SIMD, only SISD

We Propose a New Framework

	Activation Func.	Large MSG Space (# of Boot, if yes)	Batch
[BMMP, Crypt`18]	Sign	×	×
[LMP, Asiacrypt`22]	Sign	√ (2)	×
[LW, EUROCRYPT`2 3]	Sign	X	√ (msg: 0/1)
Our work	Sign/Max/ReLU	✓ (1)	✓

Learning With Errors (LWE)

RSA: (discrete logarithm problem)

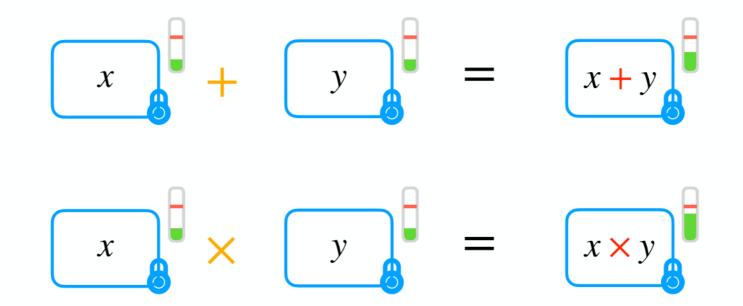
LWE hardness:

LWE encryption: mod, msg space

decryption: mod

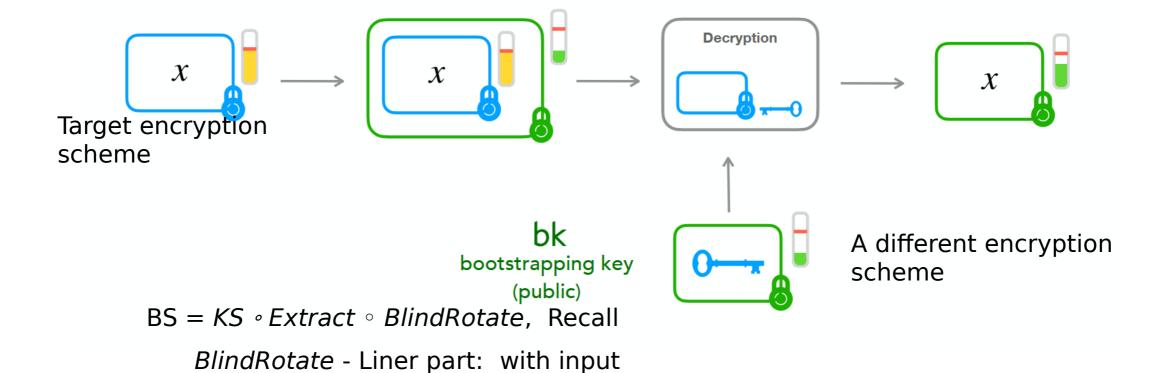
addition homomorphism:

Errors Increased After Homomorphic Operations



Noise grows too much % ⇒ decryption incorrect 8

Bootstrapping [Gen09]



[Chillotti. Introduction to FHE and the TFHE scheme]

(msb)Extract – Nonlinear part: