

# Secure Collaborative Design of Experiments with Homomorphic Encryption

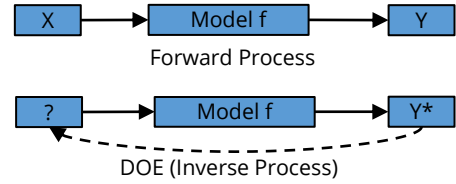
Jin Chao, Khin Mi Mi Aung, Zhang Xin

Institute for Infocomm Research, A\*STAR

Enables the client to outsource the DOE process to the model owner while keeping the target output confidential

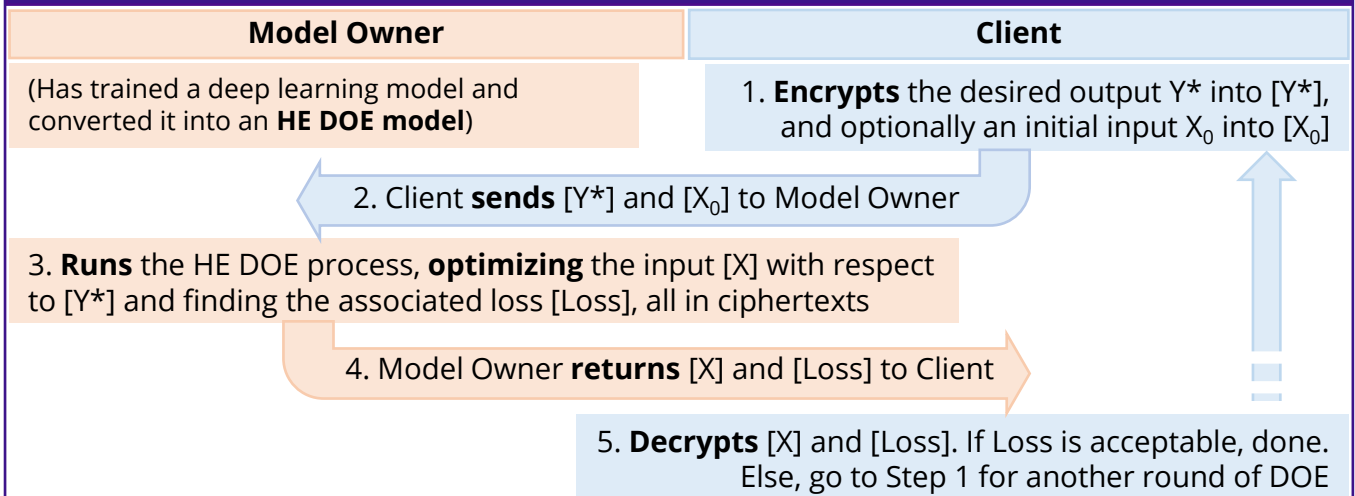
## Design of Experiments (DOE)

Given a **model  $f$**  and a **target output  $Y^*$** , DOE is the **inverse** process of finding an input  $X$  such that  $Y = f(X)$  is close to  $Y^*$ , and the **loss** measures the distance between  $Y$  and  $Y^*$ .



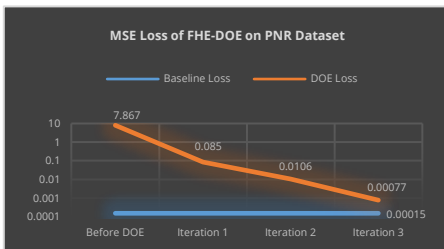
We focus on **deep learning** models, whose DOE process can be implemented by the **backpropagation** algorithm optimizing the inputs with respect to the output loss.

## Homomorphic Encryption (HE) DOE Protocol



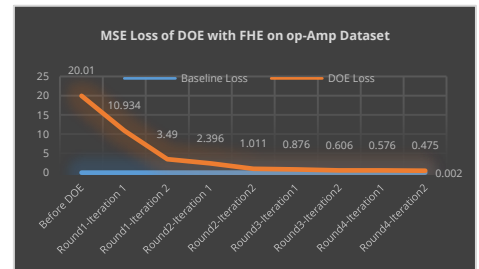
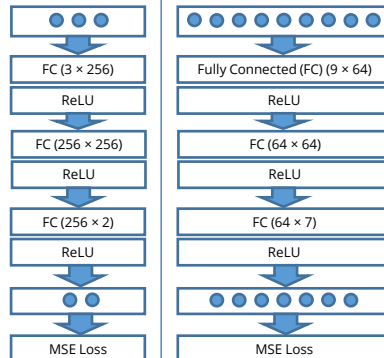
## Test Outcomes

- We use the CKKS scheme for HE, which works with **polynomial** computations. Hence, we **approximate** activation functions using polynomials in the HE DOE process.
- The approximations and the noise in fixed-point arithmetic introduce errors, but the convergence trends in the following graphs show **the errors are small**. The loss in the HE DOE process is close to the loss in the DOE process on plaintext data after the same number of iterations.



PNR: a simulation dataset generated by pre-defined functions  
MSE Loss of the plaintext model (after 500 epochs): 0.00015

For one-core CPU, one iteration on encrypted data under 128-bit security parameters takes 245 seconds, while one iteration on plaintext takes 0.523 seconds.



Op-Amp: a simulation dataset for manufacturing conditions (e.g., temperature, pressure) and the corresponding outputs  
MSE loss of the plaintext model (after 6000 epochs): 0.002

This research is supported by Institute for Infocomm Research, A\*STAR Research Entities under its RIE2020 Advanced Manufacturing and Engineering (AME) Programmatic Programme (Award A19E3b0099).