

# SDP-CROWN: Efficient Bound Propagation for Neural Network Verification with Tightness of Semidefinite Programming

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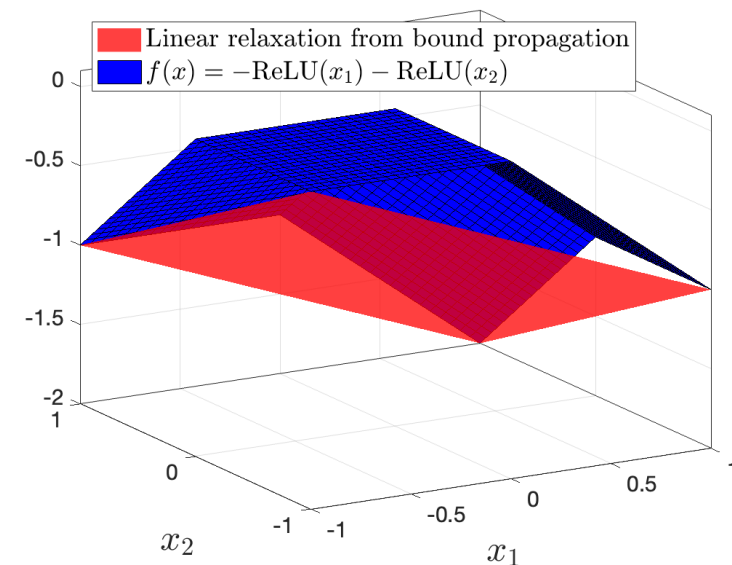
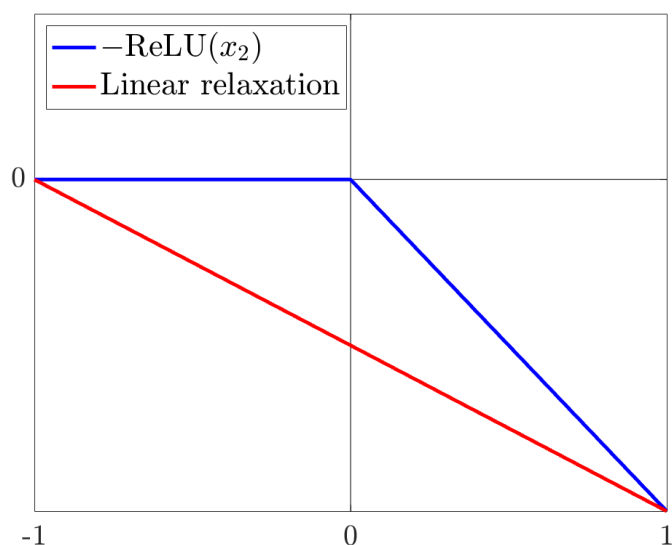
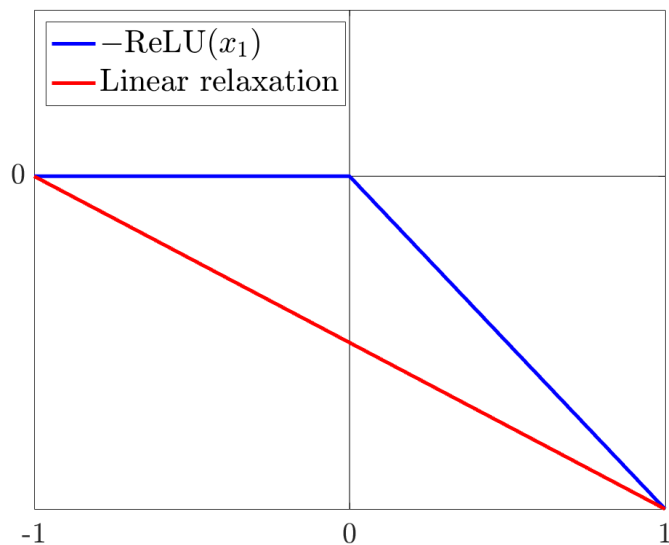
# Neural network verification

- Neural network (NN) verification aims to guarantee consistent model behavior in the presence of small perturbations to the input.
- **Bound propagation** is one of the leading methods for NN verification because it is highly scalable.
  - **For elementwise perturbations**, where each neuron is perturbed **independently**, bound propagation works remarkably well.
  - **However, for  $\ell_2$ -norm perturbations**, which introduce **inter-neuron coupling**, bound propagation becomes loose and overly conservative.

# Bound propagation

Neural network:  $f(x) = -\text{ReLU}(x_1) - \text{ReLU}(x_2)$ .

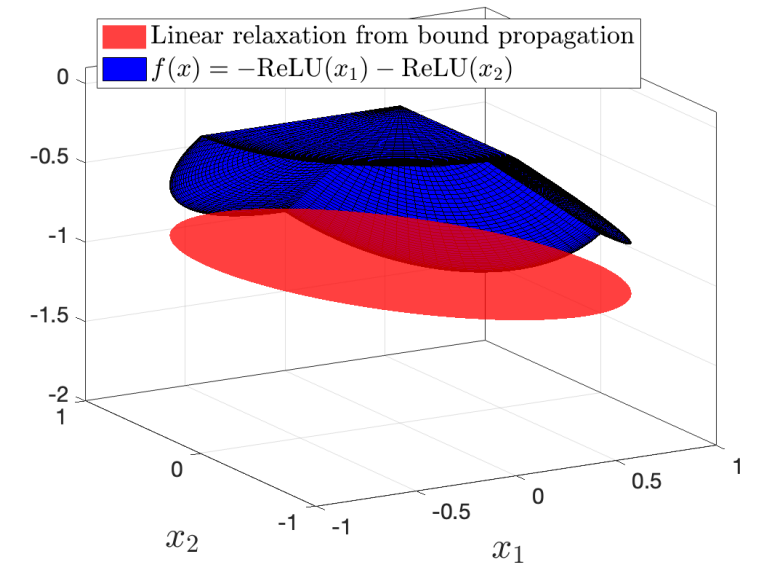
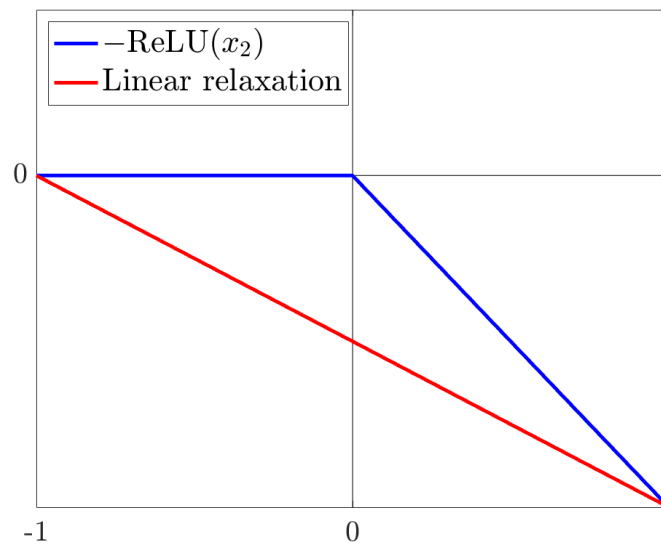
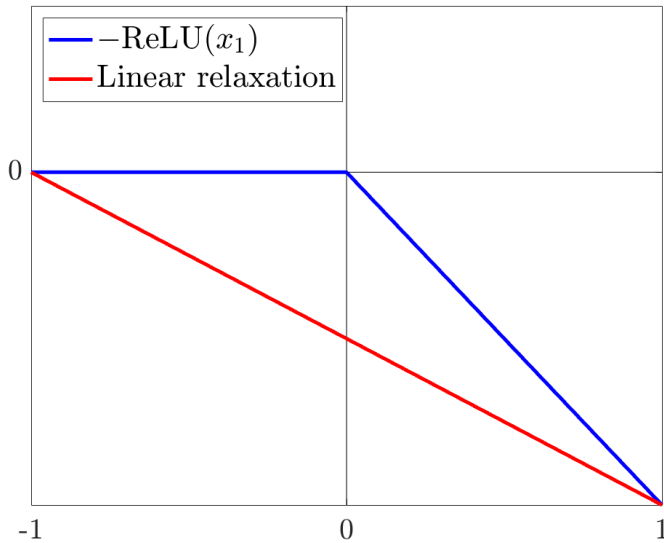
- **Elementwise perturbations:**  $-1 \leq x_1 \leq 1$  and  $-1 \leq x_2 \leq 1$ .



# Why does bound propagation fail?

Neural network:  $f(x) = -\text{ReLU}(x_1) - \text{ReLU}(x_2)$ .

- $\ell_2$ -norm perturbations:  $x_1^2 + x_2^2 \leq 1$ .

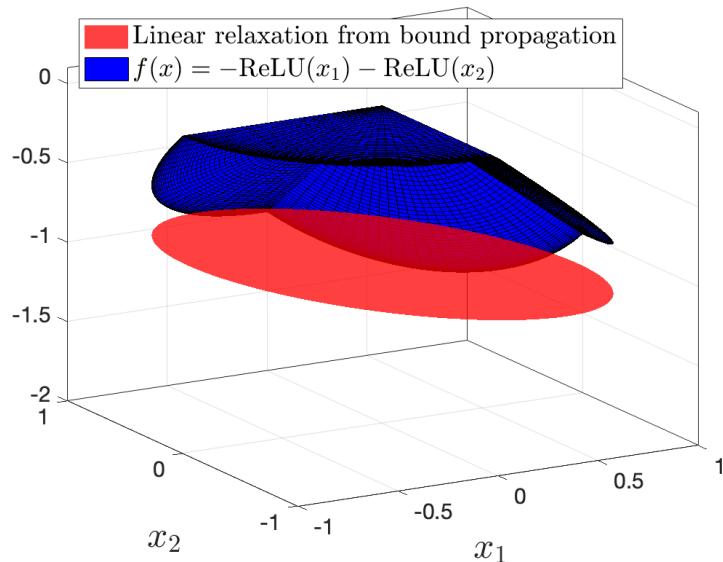


# Our method: SDP-CROWN

Neural network:  $f(x) = -\text{ReLU}(x_1) - \text{ReLU}(x_2)$

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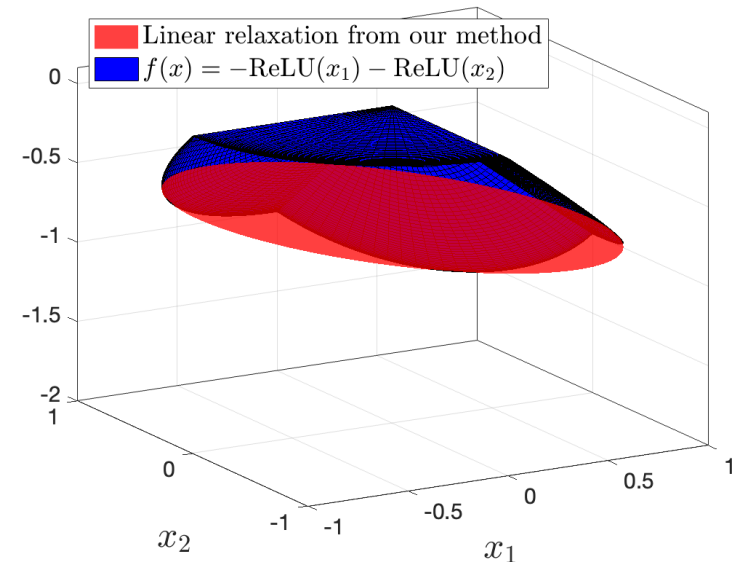
Bound propagation



Model inter-neuron  
coupling by optimizing a  
scalar variable per layer



SDP-CROWN



# Results summary

- **Significantly tighter** than existing bound propagation methods for certifying  $\ell_2$  adversaries.
- **Enjoying the same level of scalability** as bound propagation methods.

**For more details, please take a look at our paper!**

**Thank you for your attention!**