Fast and Complete: Enabling Complete Neural Network Verification with Rapid and Massively Parallel Incomplete Verifiers

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> Paper: <u>arxiv.org/abs/2011.13824</u> Code: <u>PaperCode.cc/FastAndComplete</u>

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Grand Piano

Adversarial Examples and Robustness



 $(\omega_0 + \delta) = \text{Statistical Particles}$ Bagel Adversarial Perturbation

Property to Prove: $f(x) > 0, orall x \in \mathcal{C}$

Provable Robustness Guarantee with BaB

Branch & Bound (BaB) for ReLU Neural networks



BaB requires an efficient solver for each subproblem. Typically, linear programming (LP) verifiers are used

Bound Propagation Based Verifiers



Our Main Contribution: Combining **rapid** bound propagation based incomplete verifiers such as CROWN/LiRPA with branch and bound to **scale up neural network verification** on GPUs



Results: 1-2 Magnitudes faster than LP based verifiers



Keys to success:

- 1. CROWN/LiRPA is ~100x faster than LP verifier; Massively parallelizable on GPUs;
- Optimize bounds together with intermediate layer bounds for each layer, allowing tightening the relaxation and even be tighter than regular LP verifiers



Similar or lower timeout rates CROWN cannot detect infeasible split constraints => occasional usage of LP in BaB to guarantee completeness

NEW!

See our <u>new paper</u> β -CROWN which eliminates **LP** and uses bound propagation to achieve SOTA in both **complete** and **incomplete** verification

https://arxiv.org/pdf/2103.06624

 "Efficient Neural Network Robustness Certification with General Activation Functions", Zhang, Huan et al. NeurIPS 2018
"Automatic Perturbation Analysis for Scalable Certified Robustness and Beyond", Xu, Kaidi et al., NeurIPS 2020
"A unified view of piecewise linear neural network verification. Bunel, Rudy, et al. NeurIPS. 2018.