

# USENIX'23 Artifact Appendix: Precise and Generalized Robustness Certification for Neural Networks

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# A Artifact Appendix

## A.1 Abstract

We provide code and data of our paper in this artifact. Our artifact is publicly available at https://github.com/ Yuanyuan-Yuan/GCert with detailed documents. Using our tool, users can certify neural network robustness towards various semantic-level mutations.

# A.2 Description & Requirements

### A.2.1 Security, privacy, and ethical concerns

None

#### A.2.2 How to access

An archived copy of the initial version is available at: https: //zenodo.org/record/8062051.

Our artifact is actively maintained at: https://github.com/Yuanyuan-Yuan/GCert.

#### A.2.3 Hardware dependencies

We do not have any particular requirements for the hardware. Our artifact may need GPUs to speed up the certification; we suggest evaluators having at least one GPU.

#### A.2.4 Software dependencies

Our tool is built based on Pytorch; evaluators need to first install Pytorch. See detailed instructions in our documents.

#### A.2.5 Benchmarks

None.

## A.3 Set-up

## A.3.1 Installation

Users only need to install Pytorch first. See details in our documents.

## A.3.2 Basic Test

To test the basic functionality, evaluators can first run cd experiments to change the current directory. Then run python augment\_geometrical.py. This script will start training a generative model with regulation proposed in our paper.

Detailed instructions are provided in our documents.

# A.4 Version

Based on the LaTeX template for Artifact Evaluation V20220926. Submission, reviewing and badging methodology followed for the evaluation of this artifact can be found at https://secartifacts.github.io/usenixsec2023/.