

# Formal Verification of Neural Networks

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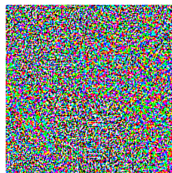
# Motivation



“panda”

57.7% confidence

+ .007 ×



noise

=



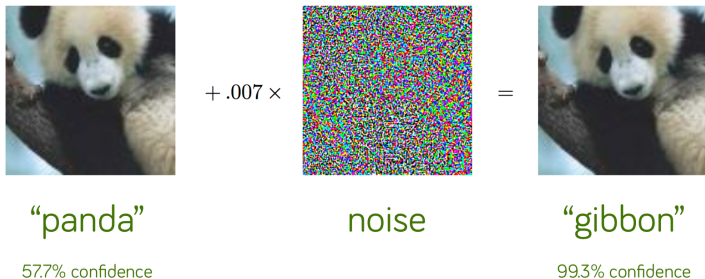
“gibbon”

99.3% confidence

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<sup>1</sup>Ian Goodfellow, Jonathon Shlens, and Christian Szegedy. “Explaining and harnessing adversarial examples”. In: *International Conference on Learning Representations*. 2015.

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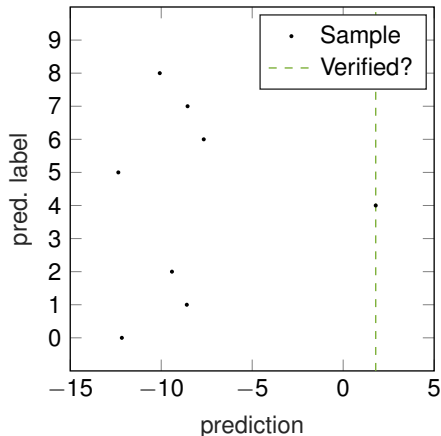
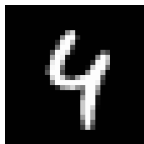


Adversarial examples<sup>1</sup> limit the applicability of neural networks in cyber-physical systems!

<sup>1</sup>Goodfellow, Shlens, and Szegedy, “Explaining and harnessing adversarial examples”.

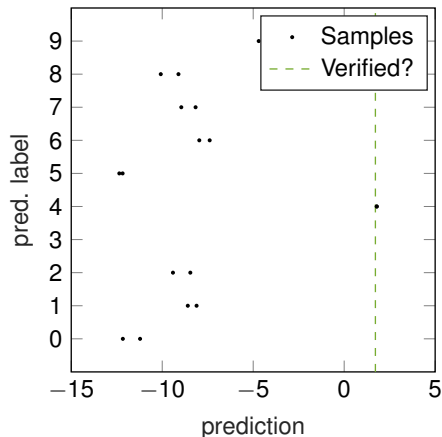
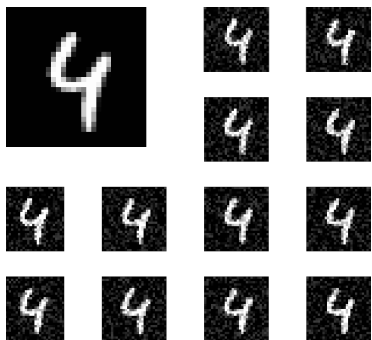
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Let us demonstrate the formal verification of neural networks by an example:



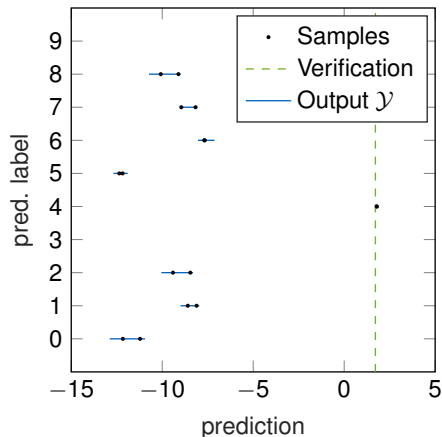
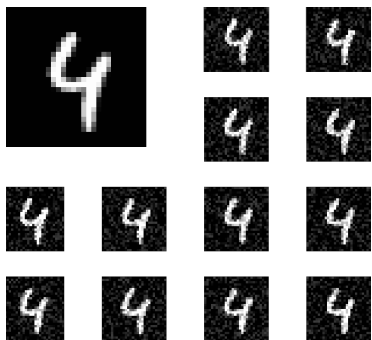
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## Your tasks:

- Learn about formal verification of neural networks
- Topics:
  - ① Design network architectures for verification
  - ② Implement and evaluate adversarial attacks
  - ③ Robust reinforcement learning

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As an example, please check NNV in CORA<sup>2</sup>

## Interested? Contact us!

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<sup>2</sup><https://cora.in.tum.de/pages/neural-networks/>