

# Abstract Interpretation

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Acknowledgement: this lecture used slides from Profs Alex Aiken, David Schmidt

# Outline

- ▶ What is abstract interpretation?
- ▶ Abstract domain
- ▶ Galois connection
- ▶ Design an abstract interpretation system
- ▶ Certify neural networks: abstract interpretation for artificial intelligence (ai<sup>2</sup>, ai4ai)

# History: Patrick Cousot, Radhia Cousot 1977

- ▶ Abstract Interpretation: A Unified Lattice Model for Static Analysis of Programs by Construction or Approximation of Fixpoints, 1977
- ▶ Methods and Logics for Proving Programs, 1990
- ▶ Completeness in Abstract Interpretation, 1995
- ▶ Directions for Research in Approximate System Analysis, 1999
- ▶ Probabilistic Abstract Interpretation, 2012
- ▶ An abstract interpretation framework for termination, 2012
- ▶ Abstract interpretation: past, present and future, 2014

# What is an abstract interpretation

An abstract interpretation consists of:

- ▶ An abstract domain  $A$  ( $+, -, 0$ ) and concrete domain  $D$  (Int)
- ▶ Concretization  $\gamma$  and abstraction functions  $\sigma$ , forming a *Galois connection*
- ▶ (sound) abstract semantic function ( $s$ )

The abstract value computed is often the property we want to prove in the concrete system

# What is abstract interpretation?

- ▶ Define an abstract domain and perform computation on abstract domain
- ▶ Soundness: the conclusions from abstract interpretation are correct comparing to the conclusions reached from concrete executions
- ▶ A theoretical framework to formalize *approximation*
- ▶ A sound approximation: the conclusion proved in the abstract domain will be held in the concrete domain
- ▶ Abstract interpretation can lose information, meaning some conclusions that can be reached by the concrete executions but cannot be reached by abstract interpretation

# An Example

See Prof. Alex Aiken's slide

# Partitioning and Abstract Domain

- ▶ Partition: abstract sets of environments/concrete inputs (D)
- ▶ Abstract domain construction: What are the properties about D that I wish to calculate?
  - ▶ Interval domain: upper and lower bound
  - ▶ Congruence domain: measure density of its values

- Intervals (nonrelational):

$$x \Rightarrow [a, b], y \Rightarrow [a', b'], \dots$$

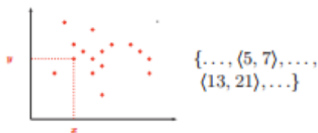
- Polyhedra (relational):

$$x + y - 2z \leq 10, \dots$$

- Difference-bound matrices (weakly relational):

$$y - x \leq 5, z - y \leq 10, \dots$$

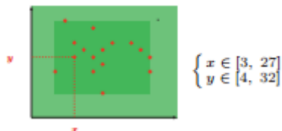
# Partitioning and Abstract Domain



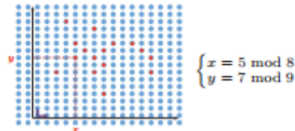
(a) [In]finite Set of Points



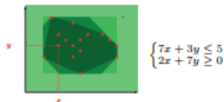
(b) Sign Abstraction



(c) Interval Abstraction



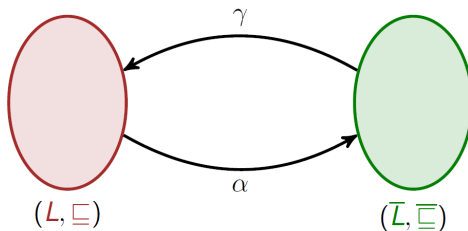
(d) Simple Congruence Abstraction



(b) Polyhedral Abstraction



# Galois Connection: intuition



## Concretization

$\gamma$  is the **concretization** function.

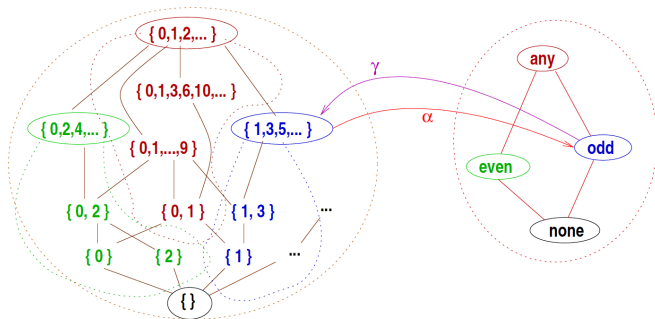
$\gamma(\bar{y})$  is the concrete value in  $L$  that is **represented** by  $\bar{y}$ .

## Abstraction

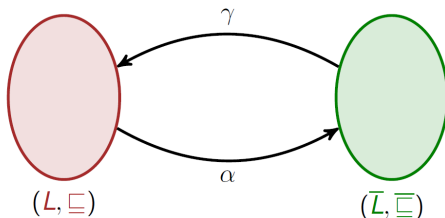
$\alpha$  is the **abstraction** function.

$\alpha(x)$  is the **most precise** abstract value in  $\bar{L}$  whose concretization **approximates**  $x$ .

# Galois connection: example



# Galois Connection: Definition



## Definition

A **Galois connection** between a lattice  $(L, \subseteq)$  and a lattice  $(\bar{L}, \bar{\subseteq})$  is a pair of functions  $(\alpha, \gamma)$ , with  $\alpha : L \rightarrow \bar{L}$  and  $\gamma : \bar{L} \rightarrow L$ , satisfying:

$$\alpha(x) \bar{\subseteq} \bar{y} \quad \text{iff} \quad x \subseteq \gamma(\bar{y}) \quad (\text{for all } x \in L, \bar{y} \in \bar{L})$$

Notation for Galois connections:  $(L, \subseteq) \xrightleftharpoons[\alpha]{\gamma} (\bar{L}, \bar{\subseteq})$

The order is preserved.

# Designing an abstract interpretation system

**Example:** We have concrete domain,  $\text{Nat}$ , and concrete operation,  $\text{succ} : \text{Nat} \rightarrow \text{Nat}$ , defined as  $\text{succ}(n) = n + 1$ .

We have abstract domain,  $\text{Parity}$ , and abstract operation,  $\text{succ}^\# : \text{Parity} \rightarrow \text{Parity}$ , defined as

$$\text{succ}^\#(\text{even}) = \text{odd}, \quad \text{succ}^\#(\text{odd}) = \text{even}$$

$$\text{succ}^\#(\text{any}) = \text{any}, \quad \text{succ}^\#(\text{none}) = \text{none}$$

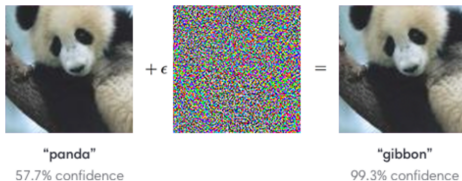
$\text{succ}^\#$  must be consistent (sound) with respect to  $\text{succ}$ :

$$\text{if } n \mathcal{R}_{\text{Nat}} a, \text{ then } \text{succ}(n) \mathcal{R}_{\text{Nat}} \text{succ}^\#(a)$$

where  $\mathcal{R} \subseteq \text{Nat} \times \text{Parity}$  relates numbers to their parities (e.g.,  $2 \mathcal{R}_{\text{Nat}} \text{even}$ ,  $5 \mathcal{R}_{\text{Nat}} \text{odd}$ , etc.).

# Abstract interpretation for robust neural networks

What does it mean to prove the robustness of a neural network?



Attack	Original	Perturbed	Diff
FGSM [12], $\epsilon = 0.3$			
Brightening, $\delta = 0.085$			

Fig. 1: Attacks applied to MNIST images [25].

# Abstract interpretation for robust neural networks

Why can we use abstract interpretation?

- ▶ Deep Neural Nets: Concrete semantics — Affine transforms + Restricted nonlinearity
- ▶ Abstract Interpretation: Scalable and precise numerical domains
- ▶ Abstract semantics should be defined on: Affine transformation (multiplication and addition), ReLU

# Abstract interpretation for robust neural networks

What is the abstract domain?

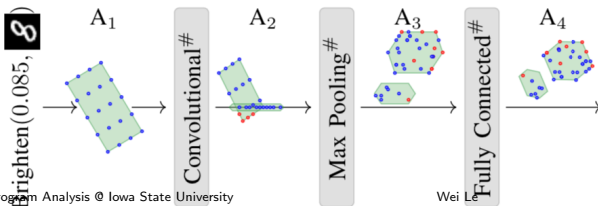
- ▶ Abstract domain: shapes expressible as a set of logical constraints
- ▶ Zonotope: a center-symmetric convex closed polyhedron [CAV09]

# Abstract interpretation for robust neural networks

High level ideas:

- ▶ abstract element:  $A_1$  is an abstract element (represent a group of inputs) that captured all perbuted inputs
- ▶ abstract layer: process abstract element
- ▶ abstract transformer: design abstract semantics for each concrete transformation available in the neural network
- ▶  $A_4$  is an overapproximation of input of interest
- ▶ verify  $A_4$  will generate the same classification

\* In particular, we can capture the entire set of brightening perturbations exactly with a single zonotope. However, in general, this step may result in an abstract element that contains additional inputs (that is, red points).





# Abstract interpretation for robust neural networks

Further reading:

AI2: Safety and Robustness Certification of Neural Networks with Abstract Interpretation