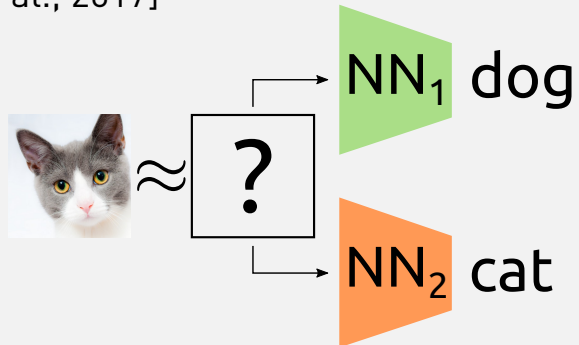


DL2: Training and Querying Neural Networks with Logic

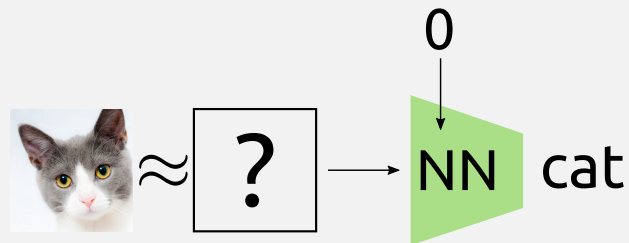
Marc Fischer, Mislav Balunović, Dana Drachler-Cohen, Timon Gehr, Ce Zhang, Martin Vechev

differencing neural networks

[Pei et al., 2017]

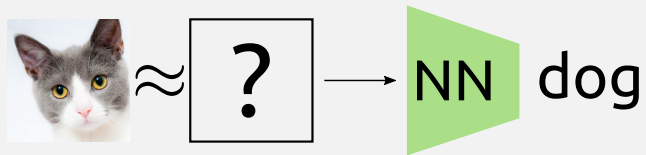


finding inputs that deactivates neurons



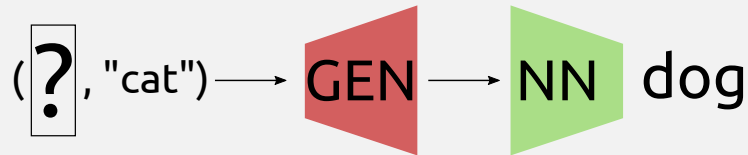
finding adversarial examples

[Szegedy et al., 2013]



finding adversarial examples using a generator

[Song et al., 2018]



DL2 Deep Learning with Differentiable Logic

differencing neural networks

[Pei et al., 2017]

```
find i[32, 32, 3]
where i in [0, 1],
      class(NN1(i)) = dog,
      class(NN2(i)) = cat,
      ||i - image||2 < 2
```

finding inputs that deactivates neurons

```
find i[32, 32, 3]
where i in [0, 1],
      NN(i).l3[17] = 0,
      class(NN(i)) = cat,
      ||i - image||1 < 100
```

finding adversarial examples

[Szegedy et al., 2013]

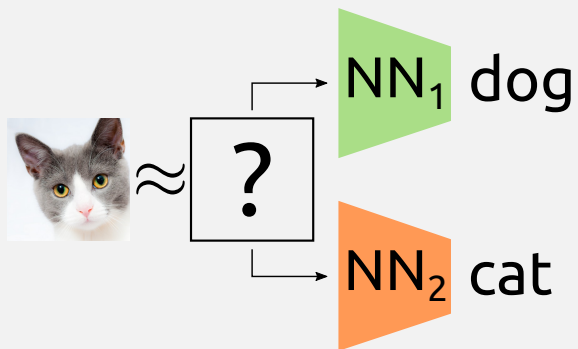
```
find i[224, 224, 3]
where i in [0, 1],
      class(NN1(i)) = dog,
      ||i - image||∞ < 25
```

finding adversarial examples

using a generator [Song et al., 2018]

```
find i[100]
where i in [-1, 1],
      class(NN(GEN(i, cat))) = dog
return GEN(i, cat)
```

differencing neural networks



DL2 Deep Learning with
Differentiable Logic

```
find i[32, 32, 3]  
where i in [0, 1],  
      class(NN1(i)) = dog,  
      class(NN2(i)) = cat,  
       $\|i - \text{image}\|_2 < 2$ 
```

differencing neural networks

[Pei et al., 2017]

```
find i[32, 32, 3]
where i in [0, 1],
      class(NN1(i)) = dog,
      class(NN2(i)) = cat,
      ||i - image||2 < 2
```

finding inputs that deactivates neurons

```
find i[32, 32, 3]
where i in [0, 1],
      NN(i).l3[17] = 0,
      class(NN(i)) = cat,
      ||i - image||1 < 100
```

finding adversarial examples

[Szegedy et al., 2013]

```
find i[224, 224, 3]
where i in [0, 1],
      class(NN1(i)) = dog,
      ||i - image||∞ < 25
```

finding adversarial examples

using a generator [Song et al., 2018]

```
find i[100]
where i in [-1, 1],
      class(NN(GEN(i, cat))) = dog
return GEN(i, cat)
```

differencing neural networks

[Pei et al., 2017]

```
find i[32, 32, 3]
where i in [0, 1],
    class(NN1(i)) = dog,
    class(NN2(i)) = cat,
    ||i - image||2 < 2,
    NN1(i).p[dog] > 0.8,
    NN1(i).p[cat] < 0.1
```

finding adversarial examples

[Szegedy et al., 2013]

```
find i[224, 224, 3]
where i in [0, 1],
    class(NN1(i)) = dog,
    ||i - image||∞ < 25,
    ||i - image||∞ > 5
```

finding inputs that deactivates neurons

```
find i[32, 32, 3]
where i in [0, 1],
    NN(i).l3[17] = 0,
    class(NN(i)) = cat,
    ||i - image||1 < 100,
    i[:8, :8, :] = image[:8, :8, :]
```

finding adversarial examples

using a generator [Song et al., 2018]

```
find i[100]
where i in [-1, 1],
    class(NN1(GEN(i, cat))) = dog,
    class(NN2(GEN(i, cat))) = car
return GEN(i, cat)
```

DL2 Querying

query

```
find i[100]
where i in [-1, 1],
class(NN(GEN(i, cat))) = dog
return GEN(i, cat)
```

logical formula φ

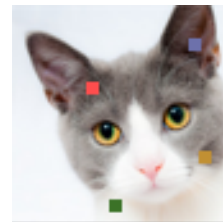
$$\varphi := \left(\bigwedge_{j=1}^{100} (-1 \leq i_j \wedge i_j \leq 1) \right) \wedge \left(\bigwedge_{\substack{k \in \text{classes} \\ k \neq \text{dog}}} \text{logit}_{\text{NN}}(\text{GEN}(i, \text{cat}))_k < \text{logit}_{\text{NN}}(\text{GEN}(i, \text{cat}))_{\text{dog}} \right)$$

differentiable loss

$$\mathcal{L}(\varphi) \geq 0$$

$$\arg \min_{i \in [-1, 1]^{100}} \sum_{\substack{k \in \text{classes} \\ k \neq \text{dog}}} \max(\text{logit}_{\text{NN}}(\text{GEN}(i, \text{cat}))_k - \text{logit}_{\text{NN}}(\text{GEN}(i, \text{cat}))_{\text{dog}}, 0)$$

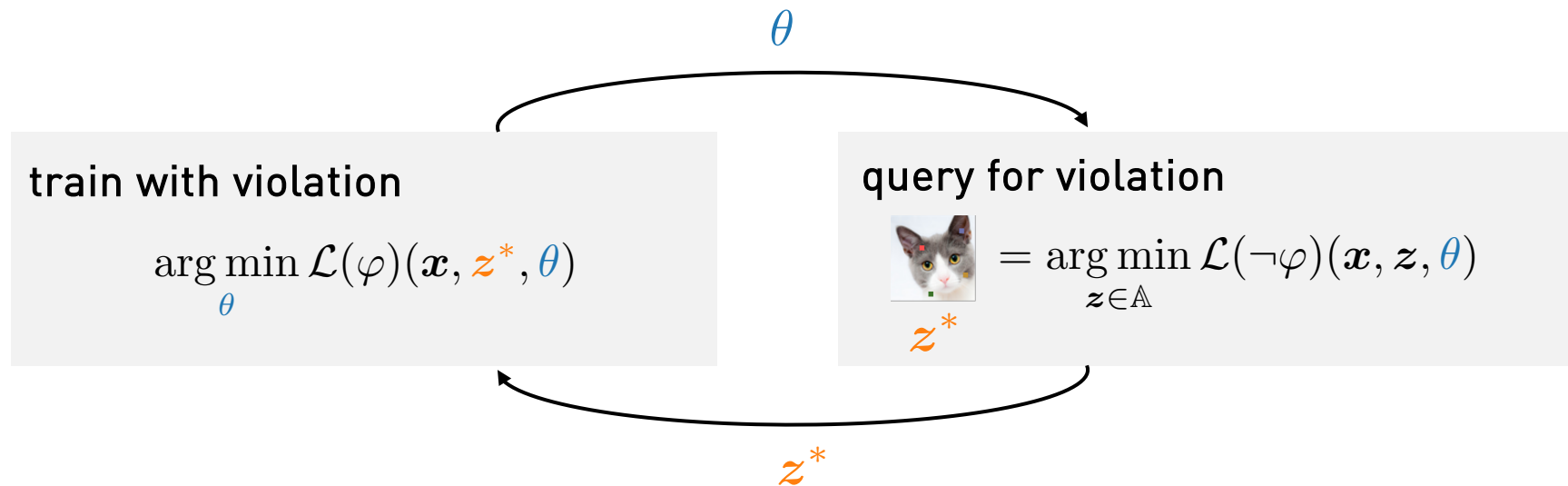
minimize loss



Theorem: $\mathcal{L}(\varphi) = 0$ if and only if φ is satisfied

DL2 Training

Goal: φ holds for all inputs

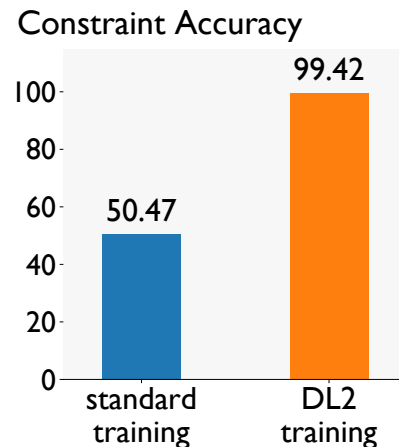
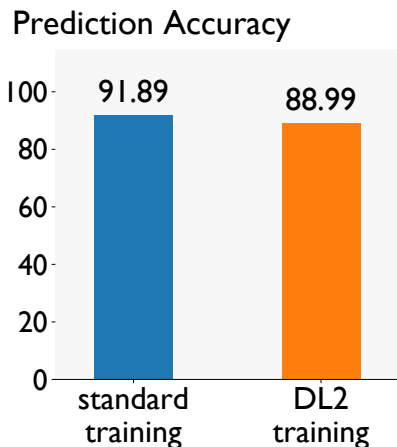


generalizes adversarial robustness training
generalizes previous work for training with constraints
applicable to supervised, semi-supervised and unsupervised training

Supervised Training Example

“A car should be considered more similar to a truck than a dog.”

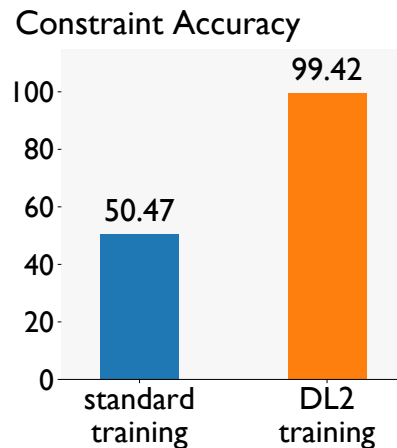
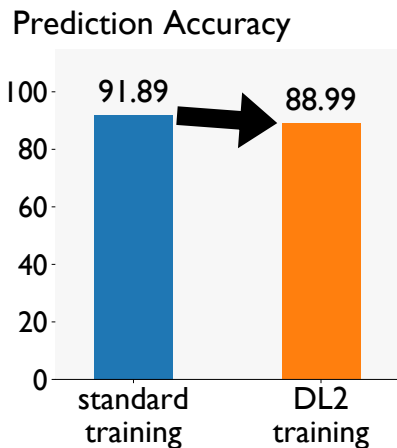
$$\forall \mathbf{z} \in B_\epsilon(\mathbf{x}) \cap [0, 1]^d. y = \text{car} \implies \text{logit}_\theta(\mathbf{z})_{\text{truck}} > \text{logit}_\theta(\mathbf{z})_{\text{dog}} + \delta$$



Supervised Training Example

“A car should be considered more similar to a truck than a dog.”

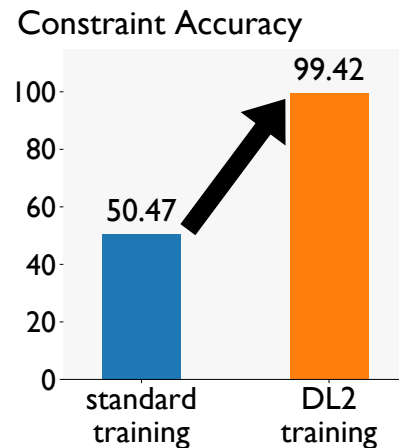
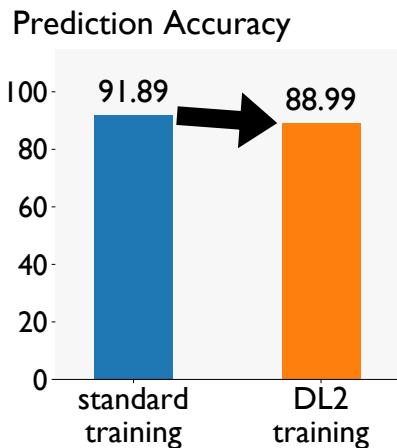
$$\forall \mathbf{z} \in B_\epsilon(\mathbf{x}) \cap [0, 1]^d. y = \text{car} \implies \text{logit}_\theta(\mathbf{z})_{\text{truck}} > \text{logit}_\theta(\mathbf{z})_{\text{dog}} + \delta$$



Supervised Training Example

“A car should be considered more similar to a truck than a dog.”

$$\forall \mathbf{z} \in B_\epsilon(\mathbf{x}) \cap [0, 1]^d. y = \text{car} \implies \text{logit}_\theta(\mathbf{z})_{\text{truck}} > \text{logit}_\theta(\mathbf{z})_{\text{dog}} + \delta$$

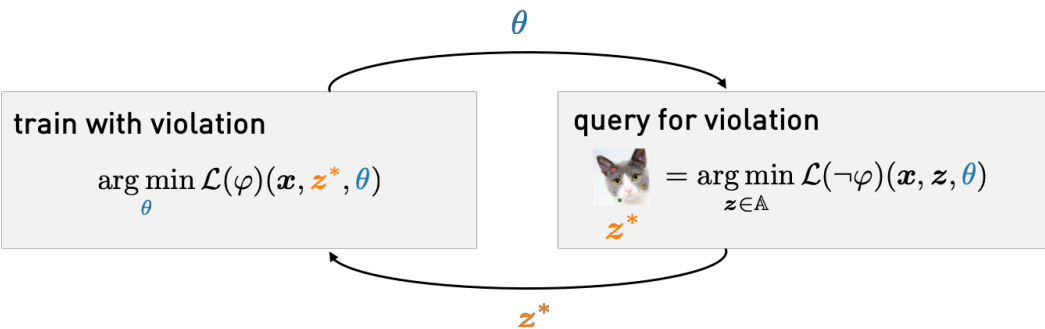


DL2: Training and Querying Neural Networks with Logic

Querying

```
find i[32, 32, 3]
where i in [0, 1],
      class(NN1(i)) = dog,
      class(NN2(i)) = cat,
      ||i - image||2 < 2,
      NN1(i).p[7] > 0.8
      NN1(i).p[1] < 0.1
```

Training



generalizes adversarial robustness training
generalizes previous work for training with constraints
applicable to supervised, semi-supervised and unsupervised training

github.com/eth-sri/dl2