

From Contrastive to Abductive Explanations and Back Again

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Motivation

Ongoing ML Revolution



DeepMind

AlphaGo

AlphaGo Zero & Alpha Zero

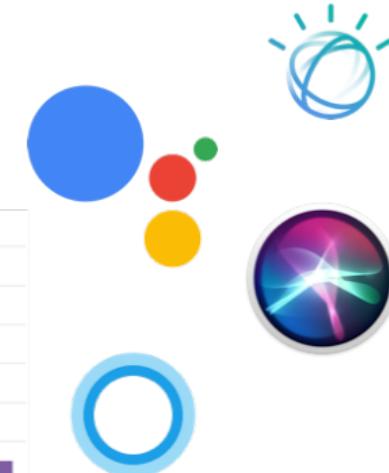
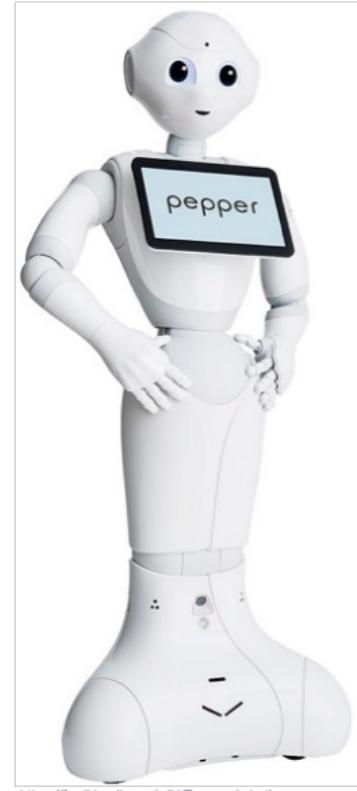
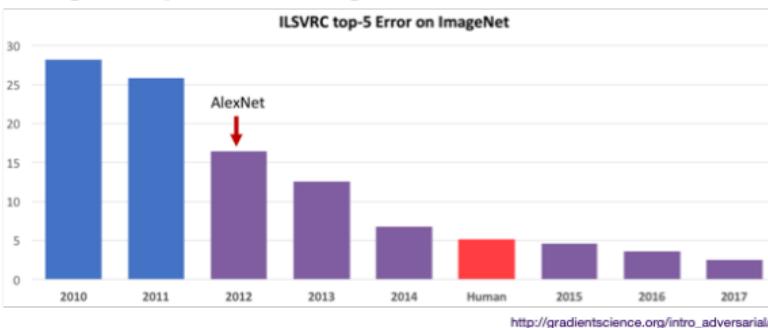


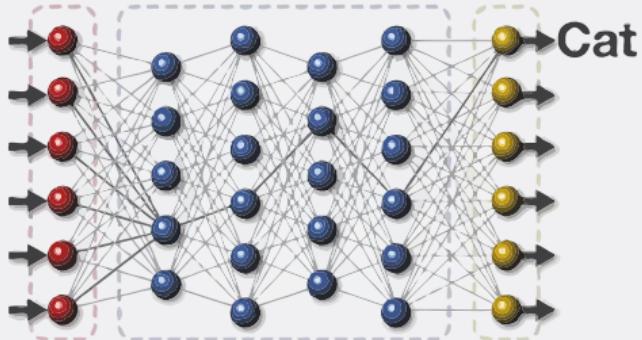
Image & Speech Recognition



And yet...

	A parrot	Machine learning algorithm
Learns random phrases		
Doesn't understand s**t about what it learns		
Occasionally speaks nonsense		

Machine Learning System



This is a cat.

Current Explanation

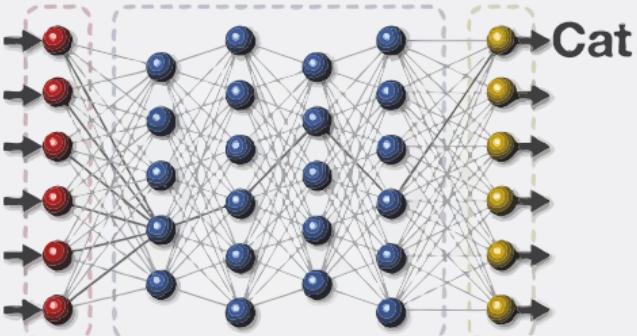
This is a cat:

- It has fur, whiskers, and claws.
- It has this feature:



XAI Explanation

Machine Learning System



Which features?

Why cat?

This is a cat.

Explain?!

Current Explanation

This is a cat:

- It has fur, whiskers, and claws.
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XAI Explanation

Formal explanations

classifier $\tau : \mathbb{F} \rightarrow \mathcal{K}$, **instance** v **s.t.** $\tau(v) = c$

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abductive explanation X

$$\forall(x \in \mathbb{F}) \cdot \bigwedge_{j \in X} (x_j = v_j) \rightarrow (\tau(x) = c)$$

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abductive explanation x

“why?”

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abductive explanation \mathcal{X}

"why?"

$$\forall(x \in \mathbb{F}) \cdot \bigwedge_{j \in \mathcal{X}} (x_j = v_j) \rightarrow (\tau(x) = c)$$

contrastive explanation \mathcal{Y}

$$\exists(x \in \mathbb{F}) \cdot \bigwedge_{j \notin \mathcal{Y}} (x_j = v_j) \wedge (\tau(x) \neq c)$$

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abductive explanation x

"why?"

$$\forall(x \in \mathbb{F}) \cdot \bigwedge_{j \in x} (x_j = v_j) \rightarrow (\tau(x) = c)$$

contrastive explanation y

"why not?"

$$\exists(x \in \mathbb{F}) \cdot \bigwedge_{j \notin y} (x_j = v_j) \wedge (\tau(x) \neq c)$$

this work!

$$\mathbb{F} = \{0, 1, 2\}^5 \quad \mathcal{K} = \{\ominus, \oplus\}$$

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$R_1:$	ELSE IF	$x_3 \neq 1$	THEN	\oplus
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observe $\tau(1, 1, 1, 1, 1) = \ominus$



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Axps X = { {1, 2}, {3} }

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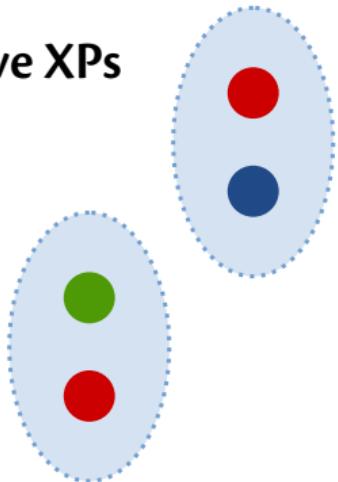
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$$\text{AXps } \mathbb{X} = \{\{1, 2\}, \{3\}\}$$

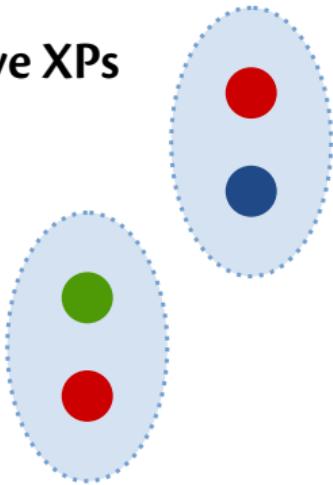
$$\text{CXps } \mathbb{Y} = \{\{1, 3\}, \{2, 3\}\}$$

Abductive XPs

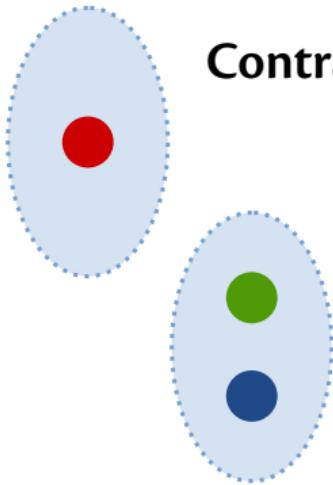


Minimal hitting set duality

Abductive XPs



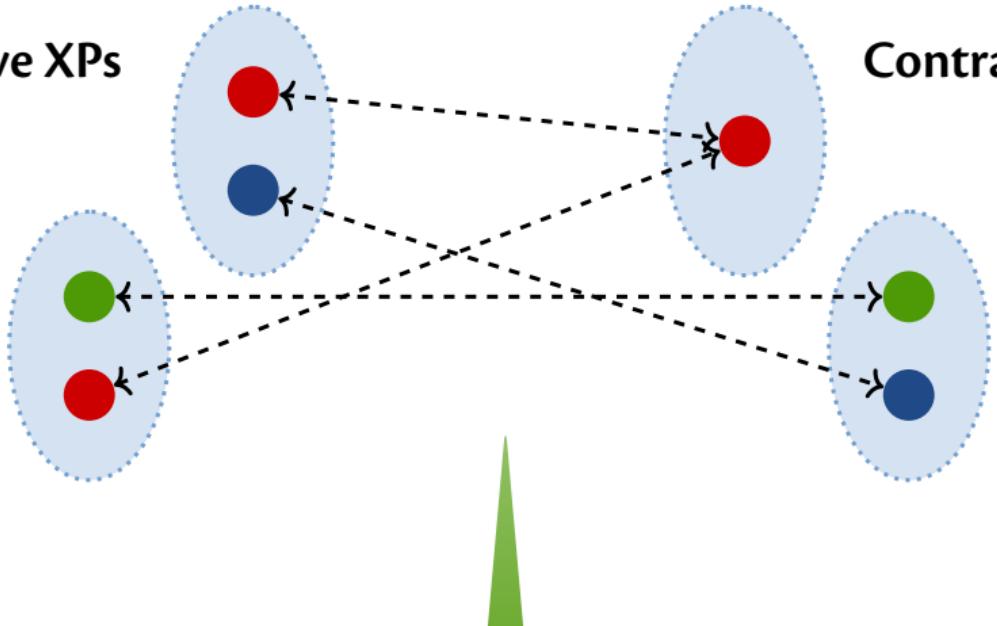
Contrastive XPs



Minimal hitting set duality

Abductive XPs

Contrastive XPs



AXps are minimal hitting sets of CXps, and vice versa

CXp computation

CXp computation – example

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$$\exists(x \in \mathbb{F}). \bigwedge_{j \notin \mathcal{Y}} (x_j = v_j) \wedge (\tau(x) \neq c)$$

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CXp $\mathcal{Y} = \{2, 3\}$

$$\exists(x \in \mathbb{F}). \bigwedge_{j \notin \mathcal{Y}} (x_j = v_j) \wedge (\tau(x) \neq c)$$

Explanation Enumeration

Function $\text{XPENUM}(\tau, \mathbf{v}, c)$

Input: τ : ML model, \mathbf{v} : Input instance, $c = \tau(\mathbf{v})$: Prediction

```
1    $\mathcal{K} = (\mathcal{N}, \mathcal{P}) \leftarrow (\emptyset, \emptyset)$                                 // Block AXps & CXps
2   while true:
3      $(st_\lambda, \lambda) \leftarrow \text{FindMHS}(\mathcal{P}, \mathcal{N})$                       // MHS of  $\mathcal{P}$  s.t.  $\mathcal{N}$ 
4     if  $\neg st_\lambda$ : break
5      $st_{c'} \leftarrow \text{SAT}(\bigwedge_{j \in \lambda} (x_j = v_j) \wedge \tau(\mathbf{x}) \neq c)$ 
6     if  $\neg st_{c'}$ :                                                               // entailment holds
7       ReportAXp( $\lambda$ )
8        $\mathcal{N} \leftarrow \mathcal{N} \cup \bigvee_{j \in \lambda} (x_j \neq v_j)$ 
9     else:
10       $\mu \leftarrow \text{ExtractCXp}(\tau, \mathbf{v}, c, \mathcal{P})$ 
11      ReportCXp( $\mu$ )
12       $\mathcal{P} \leftarrow \mathcal{P} \cup \bigvee_{j \in \mu} (x_j = v_j)$ 
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implicit hitting set enumeration!

see paper for details

Conclusions

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 - similar to abductive explanations

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Conclusions

- formal definition of contrastive explanations
 - similar to abductive explanations
 - minimal hitting set duality between CXps and AXps
 - explanation enumeration algorithms
 - solving membership problems
 - experimental results
 - XP enumeration
 - CXp enumeration – *helps to debug SHAP*
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- proved helpful in several papers!

Questions?

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