Semantically Equivalent Adversarial Rules for Debugging NLP Models

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NLP / ML models are getting smarter: VQA





NLP / ML models are getting smarter: MC (SQuAD)





Oversensitivity in images



"panda"
57.7% confidence

"gibbon" 99.3% confidence



Adversarial examples





What about text?







What about text?



















Adversarial Rules





Generalizing adversaries

















Adversarial Rules





Semantically Equivalent Adversary (SEA)

Ingredients

0





Revisiting adversaries



 $\max_{x'} S(x,x') > \gamma s.t. SEA(x,x') =$

Semantic Similarity: Paraphrasing



Finding an adversary





Semantically Equivalent Adversarial Rules (SEARs)

From SEAs to Rules





From SEAs to Rules



Semantically Equivalent Adversarial Rules (SEARS)

- O High Adversary Count
- Non-Redundancy





Examples: VQA Visual7a-Telling [Zhu et al 2016]

SEAR	Questions / SEAs	f(x)	Flips
WP VBZ \rightarrow WP's	What has What's been cut?	Cake Pizza	3.3%
What NOUN → Which NOUN	What Which kind of floor is it?	Wood Marble	3.9%
color \rightarrow colour	What color colour is the tray?	Pink Green	2.2%
ADV is → ADV's	Where is Where's the jet?	Sky Airport	2.1%

Examples: Machine Comprehension BiDAF [Seo et al 2017]

SEAR	Questions / SEAs	f(x)	Flips
What VBZ → What's	What is What's the NASUWT?	Trade union Teachers in Wales	2%
What NOUN → Which NOUN	What resource Which resource was mined in the Newcastle area?	coal wool	1%
What VERB → So what VERB	What was So what was Ghandi's work called?	Satyagraha Civil Disobedience	2%
What VBD→ And what VBD	What was And what was Kenneth Swezey's job?	journalist sleep	2%

Examples: Movie Review Sentiment Analysis

FastText [Joulin et al 2016]

SEAR	Reviews / SEAs	f(x)	Flips
movie → film	Yeah, the <i>movie</i> film pretty much sucked .	Neg Pos	2%
film → movie	Excellent <i>film</i>-movie .	Pos Neg	1%
is → was	Ray Charles is was legendary .	Pos Neg	4%
this → that	Now this that is a movie I really dislike .	Neg Pos	1%

Experiments

1. SEAs vs Humans



How often can SEAs be produced?



Humans produce different adversaries:





Soul What kind of meat is on the boy's plate?



How many suitcases?re on the shelf?

Pholography and directing were on point. Also great directing and photography

2. SEARs vs Experts

Part 1: experts come up with rules

Individual predictions Rules

Try different rules	Results		
List of POS tags Replace first instance of: What NOUN With:	replace(What NOUN, Which NOUN) Save rule		
Which NOUN Submit	Mistake examples (click images to see them in more detail)		
Saved Rules			
replace(<mark>Who is, Who's</mark>) × replace(<mark>color, colour</mark>) ×		low. ed. green.	
	Q: What food item is above the burger Q: Which for ? Answer: (a) Fries. b) Chips. c) Cole slaw. d) Ketchup. (b) Chips. c) Ketchup.		

Part 2: experts evaluate our SEARs

Rules to evaluate	Results		
List of POS tags Please look at the rule results on the circle The current rule:	replace(<mark>What N</mark>	NOUN, Which NOUN)	
repla Does the c No Yes	Image	Original	After rule
Progress 1 of 20.		Q: What color are the pots ? Answer: a) Silver. b) Black. c) White. d) Gold.	Q: Which color are the pots ? Answer: a) Silver. b) Black. c) White. d) Gold.
		Q: What color is the lampshade ? Answer: (a) A light yellow. (b) A bright red. (c) A subtle green. (d) A vivid orange.	Q: Which color is the lampshade ? Answer: a) A light yellow. b) A bright red. c) A subtle green. d) A vivid orange.
	A Mas	Q: What animal is running in the background ? Answer: (a) A dog.) (b) A horse. (c) A llama. (d) A kangaroo.	Q: Which animal is running in the background ? Answer: a) A dog. (b) A horse. c) A llama. d) A kangaroo.
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Results

% correct predictions flipped



Time (minutes)



3. Fixing bugs



Results





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Semantic scoring is still a research problem...

SEAR	Questions / SEAs	f(x)
	What is on in the background?	A building Mountains
on → in	What is on in?	Lights The television
	Where are is the water bottles?	Table Vending machine
VBP → is	Where are is the people gathered?	Living room Kitchen
	What is on the backbround?	A building Mountains
VERB on \rightarrow VERB	Where are the planes parked on?	Concrete Landing strip



Problem: not comparable across instances



Examples: VQA

SEAR	Questions / SEAs	f(x)	Flips
WP VBZ \rightarrow WP's	What has What's been cut?	Cake Pizza	
	Who is Who's holding the baby?	Woman Man	3.3%
What NOUN \rightarrow	What Which kind of floor is it?	Wood Marble	
Which NOUN	What Which color is the jet?	Gray White	3.9%
$color \rightarrow colour$	What color colour is the tray?	Pink Green	2 20/
	What color colour is the jet?	Gray Blue	2.2%
ADV is → ADV's	Where is Where's the jet?	Sky Airport	2 10/
	How is How's the desk?	Messy Empty	2.1%

Examples: Movie Review Sentiment Analysis

FastText [Joulin et al 2016]

SEAR	Reviews / SEAs	f(x)	Flips	
movie → film	Yeah, the movie film pretty much sucked .	Neg Pos	2%	
	This is not <i>movie</i> film making .	Neg Pos		
film → movie	Excellent film movie .	Pos Neg	10/	
	I'll give this film movie 10 out of 10 !	Pos Neg	1%	
is → was	Ray Charles is was legendary .	Pos Neg	40/	
	It is was a really good show to watch .	Pos Neg	4%	
this → <mark>that</mark>	Now this that is a movie I really dislike .	Neg Pos	10/	
	The camera really likes her in this that movie.	Pos Neg	1%	

 $SEA(x,x') = 1 [| S(x,x') > \gamma \land f(x) \neq f(x')]$

$\max_{x'} S(x,x') > \gamma s.t. SEA(x,x') =$