# Supervising Model Attention with Human Explanations for Robust Natural Language Inference

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> We train with human explanations to create more robust NLI models, paying more attention to the words

that humans think are important.

e-SNLI human explanations:

Premise: Wet brown dog swims towards camera.

Hypothesis: A dog is sleeping in his bed.

Explanation for contradiction class: A dog cannot be sleeping while he swims.

## We see significant improvements in ID and OOD performance from

both approaches.

Performance compared to our BERT baseline:

	Dev	Test	Hard	MNLI mi	MNLI ma	ANLI	HANS
BERT baseline	90.05	89.77	79.36	72.52	72.28	31.81	56.83
Ours (extra layer)	90.40	90.09	79.96	73.03	73.10	31.47	57.85
Improvement	<b>+0.35</b> †‡	+ <b>0.32</b> †‡	<b>+0.60</b> †‡	<b>+0.51</b> †	<b>+0.82</b> †‡	-0.34	+1.02
Ours (existing attention)	90.45	90.17	80.15	73.36	73.19	31.41	58.42
Improvement	<b>+0.40</b> †‡	<b>+0.40</b> †‡	<b>+0.79</b> †‡	<b>+0.84</b> †‡	<b>+0.91</b> †‡	-0.40	<b>+1.59</b> †

### Improvements compared to prior work:

	SNLI	$\Delta$	MNLI	$\Delta$	SNLI-hard	$\Delta$	Params.
BERT Baseline	89.77		72.40		79.36		109m
LIREx-adapted	<b>90.79</b>	<b>+1.02</b> †	71.55	-0.85†	79.39	+0.03	453m
Pruthi et al-adapted.	89.99	+0.22†	73.27	+0.87†	79.90	+0.54†	109m
Ours (extra layer)	90.09	+0.35†	73.06	+0.67†	79.96	+0.60†	109m
Ours (existing attention)	90.17	+0.40†	<b>73.28</b>	<b>+0.88</b> †	<b>80.15</b>	<b>+0.79</b> †	109m

We achieve this by creating a desired attention distribution, based on the content of the e-SNLI human explanations.



to stop-words and more attention paid to nouns, verbs and adjectives.

More attention is paid to the premise, even in the layers before the supervised layer, helping to mitigate the hypothesis-only bias. We also see less attention paid



Proportion of attention to the premise:

\* Attention to premise and 1st [SEP] token

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An auxiliary loss encourages more attention to this distribution, supervising either: 1) existing attention heads, or 2) an additional attention layer.

$$a_{h_i} = \frac{\exp\left(q_{h_{CLS}}^T k_{h_i} / \sqrt{d_k}\right)}{\sum_{j=1}^n \exp\left(q_{h_{CLS}}^T k_{h_j} / \sqrt{d_k}\right)}$$

$$Loss_{Total} = Loss_{NLI} + \frac{\lambda}{H} \sum_{h=1}^{H} \sum_{i=1}^{n} (a_{h_i}$$

**Baseline attention** to premise\*: 22.86% Supervised attention to premise\*: 50.89%

Proportion of attention by Po	S tag:
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PoS Tag	12 heads	3 heads	Baseline
Noun	54.3	43.5	28.1
Verb	20.4	18.2	14.3
Adjective	8.9	8.3	5.2
Adposition	4.1	5.0	7.8
Determiner	3.4	6.0	14.3
Punctuation	0.9	7.7	14.2
Auxiliary	0.9	3.1	8.2
Other	7.1	8.2	7.9

Most attended to words:

Basel	ine	Supervise		
Words	%	Words	9	
	18.0	man	2	
а	5.2	outside	2	
is	4.0	woman	1	
are	2.6	people	1	
the	2.5	sitting	1	











