

ScoNe: Benchmarking Negation Reasoning in Language Models With Fine-Tuning and In-Context Learning

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База данных ScoNe-NLI

Scoped Negation Natural Language Inference – расширенная база Monotonicity NLI

- 1202 набора контрастов
- Набор контрастов –
 - 0/1/2 отрицания
 - Наличие/отсутствие отрицания в сфере действия для таргета
 - Метка NLI зависит от наличия/отсутствия следствия

Split	Premise	Rel.	Hypothesis	Examples
No negation	The cowboy fell off a horse at the competition	⊑	The cowboy fell off a racehorse at the competition	1,202
One Not Scoped	The cowboy did not fear anything, until he fell off a horse at the competition	⊑	The cowboy did not fear anything, until he fell off a racehorse at the competition	1,202
Two Not Scoped	The cowboy, who was not very old, was not proud that he fell off a horse at the competition	⊑	The cowboy, who was not very old, was not proud that he fell off a racehorse at the competition	1,202
Two Scoped	There is no way that the cowboy did not fall off a horse at the competition	⊑	There is no way that the cowboy did not fall off a racehorse at the competition	1,202
One Scoped	The cowboy did not fall off a horse at the competition	⊑	The cowboy did not fall off a racehorse at the competition	1,202
One Scoped, One not Scoped	The cowboy did not fall off a horse, but the competition was not too important	⊑	The cowboy did not fall off a racehorse, but the competition was not too important	1,202

(a) A six-example contrast set from ScoNe-NLI.

Fine-Tuning DeBERTa на ScoNe-NLI

Использовались предобученные на MNLI, Fever-NLI и Adversarial-NLI DeBERTa-v3-base и RoBERTa

Fine-tuning Datasets	No Negation	One Not Scoped	Two Not Scoped	Two Scoped	One Scoped	One Scoped, One not Scoped
MAF-NLI	82.0	86.0	81.5	91.0	5.0	5.0
MAF-NLI+ MoNLI (Geiger et al., 2020)	96.2	87.5	99.5	8.9	100.0	100.0
MAF-NLI+ MED (Yanaka et al., 2020)	84.8	83.5	82.0	58.9	99.5	97.0
MAF-NLI+ Neg-NLI (Hossain et al., 2020)	91.3	88.5	83.0	70.4	37.0	29.0
MAF-NLI+ MoNLI + ScoNe-NLI	100.0	100.0	100.0	100.0	100.0	100.0

Table 2: DeBERTa fine-tuning results on ScoNe-NLI. MAF-NLI stands for on MNLI, ANLI, and Fever-NLI.

Fine-Tuning RoBERTa на ScoNe-NLI

Использовались предобученные на MNLI, Fever-NLI и Adversarial-NLI DeBERTa-v3-base и RoBERTa

B RoBERTa Results

Fine-tuning Datasets	No Negation	One Not Scoped	Two Not Scoped	Two Scoped	One Scoped	One Scoped, One not Scoped
MAF-NLI	96.5	97.0	97.0	96.5	3.0	5.0
MAF-NLI+ MoNLI (Geiger et al., 2020)	85.4	100.0	100.0	4.5	100.0	100.0
MAF-NLI+ MED (Yanaka et al., 2020)	85.1	92.0	89.5	44.6	85.5	81.5
MAF-NLI+ Neg-NLI (Hossain et al., 2020)	93.1	97.5	93.0	73.2	20.5	17.5
MAF-NLI+ MoNLI + ScoNe-NLI	100.0	100.0	100.0	100.0	100.0	100.0

Table 6: RoBERTa fine-tuning results on ScoNe-NLI. MAF-NLI stands for on MNLI, ANLI, and Fever-NLI.

Обучение Instruct-GPT в контексте на ScoNe-NLI

Если ответ содержит "yes" в любом виде, то понимаем как entailment, в противном случае – neutral.

Использовалось шесть типов промптов.

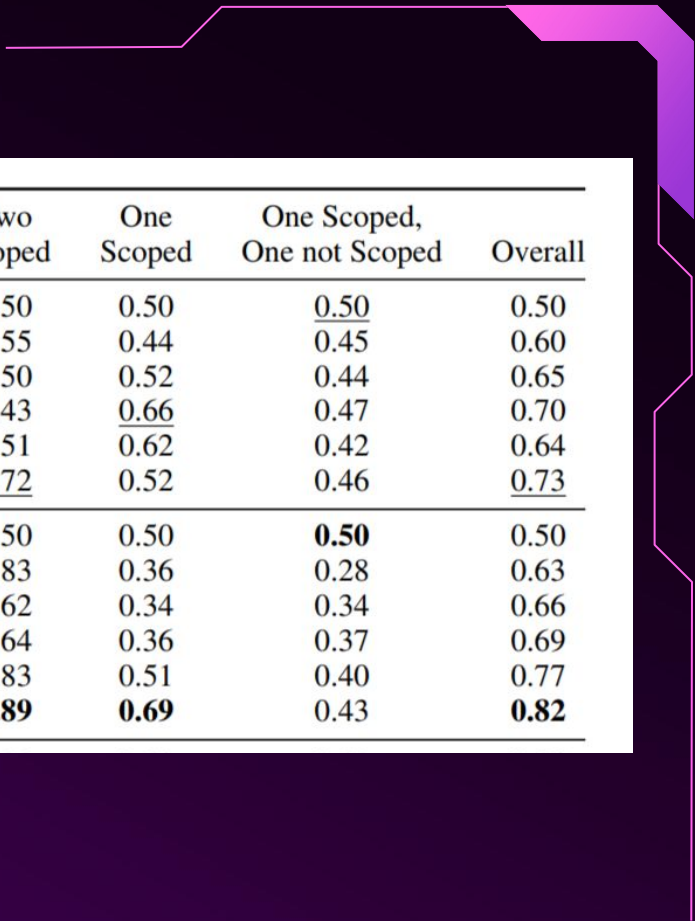
Zero-shot: сразу таргет

Few-shot: 4 примера задания без оценки ответа + таргет

Conditional Q	Is it true that if Premise , then Hypothesis ?
Hypothesis Q	Assume that Premise . Is it then definitely true that Hypothesis ? Answer yes or no.
Conditional Truth	If Premise , then Hypothesis . Is this true?
Brown et al.	P: Premise \n Q: Hypothesis \n Yes, No, or Maybe?
Structured	P: Premise \n H: Hypothesis \n L:
Reasoning	Logical and commonsense reasoning exam.\n\n Explain your reasoning in detail, then answer with Yes or No. Your answers should follow this 4-line format:\n\n Premise: <a tricky logical statement about the world>.\n\n Question: <question requiring logical deduction>.\n\n Reasoning: <an explanation of what you understand about the possible scenarios>.\n\n Answer: <Yes or No>.\n\n Premise: Premise \n\n Question: Hypothesis \n\n Reasoning: Let's think logically step by step. The premise basically tells us that

		No Negation	One Not Scoped	Two Not scoped	Two Scoped	One Scoped	One Scoped, One not Scoped	Overall
Zero-shot	Structured	0.50	0.50	0.50	0.50	0.50	0.50	0.50
	Brown et al.	0.69	0.60	0.59	0.55	0.50	0.48	0.57
	Conditional Q	0.76	0.55	0.65	0.50	0.50	0.50	0.58
	Conditional Truth	0.76	0.64	0.66	0.60	0.50	<u>0.57</u>	0.62
	Hypothesis Q	0.80	<u>0.83</u>	<u>0.86</u>	<u>0.62</u>	0.45	<u>0.40</u>	<u>0.66</u>
	Reasoning	<u>0.85</u>	0.70	0.68	<u>0.62</u>	<u>0.57</u>	0.56	<u>0.66</u>
Few-shot	Structured	0.50	0.50	0.50	0.50	0.50	0.50	0.50
	Brown et al.	0.82	0.75	0.78	0.72	0.35	0.29	0.62
	Conditional Q	0.92	0.82	0.78	0.52	0.36	0.32	0.62
	Conditional Truth	0.92	0.89	0.88	0.59	0.36	0.37	0.67
	Hypothesis Q	0.99	0.91	0.92	0.68	0.38	0.40	0.72
	Reasoning	0.73	0.85	0.78	0.62	0.74	0.54	0.71

Table 7: In-context learning results for GPT-3 (davinci-002 engine).



		No Negation	One Not Scoped	Two Not scoped	Two Scoped	One Scoped	One Scoped, One not Scoped	Overall
Zero-shot	Structured	0.50	0.50	0.50	0.50	0.50	<u>0.50</u>	0.50
	Brown et al.	0.74	0.70	0.74	0.55	0.44	0.45	0.60
	Conditional Q	0.79	0.84	0.80	0.50	0.52	0.44	0.65
	Conditional Truth	<u>0.98</u>	0.86	0.80	0.43	<u>0.66</u>	0.47	0.70
	Hypothesis Q	0.69	<u>0.90</u>	0.70	0.51	0.62	0.42	0.64
	Reasoning	0.90	<u>0.88</u>	<u>0.94</u>	<u>0.72</u>	0.52	0.46	<u>0.73</u>
Few-shot	Structured	0.50	0.50	0.50	0.50	0.50	0.50	0.50
	Brown et al.	0.86	0.66	0.80	0.83	0.36	0.28	0.63
	Conditional Q	0.92	0.85	0.90	0.62	0.34	0.34	0.66
	Conditional Truth	0.94	0.90	0.94	0.64	0.36	0.37	0.69
	Hypothesis Q	0.98	0.96	0.94	0.83	0.51	0.40	0.77
	Reasoning	0.99	0.97	0.98	0.89	0.69	0.43	0.82





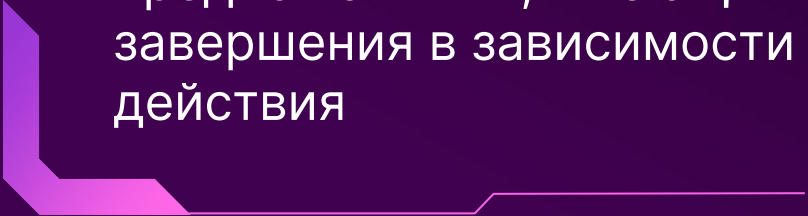
ScoNe-NLG

Гипотеза:

InstructGPT может корректно рассуждать об отрицании при оценке примеров, созданных с учетом задачи, на которую её обучали

Датасет:

ScoNe-NLG - это датасет для генерации текста, который содержит 74 тройки примеров с недописанными предложениями, имеющими различные логические завершения в зависимости от наличия отрицания и сферы его действия



E.13 ScoNe-NLG Prompts

In the zero-shot condition, models are simply prompted with the ScoNe-NLG examples. In the few-shot condition, the test example is preceded with a fixed set of four demonstrations, separated by double newlines. The examples are as follows:

Prompt example

Glen is not a fan of learning math. When he sees that his new high school requires that he take a geometry course, he is not pleased.\n

\n

I saw John take his BMW to the store the other day, so when Suzy asked me if John owns a car, I said yes.\n

\n

I've seen John with a dog that isn't very cute, so when Suzy asked me if John owns a pet, I said yes.\n

\n

I recently confirmed that John is not allergic to any shellfish. So it makes sense that when we served shrimp



Обучение в контексте на ScoNe-NLG

Анализ ответов экспертами вручную на адекватность и связность, согласованность - 216/222 случаях в zero-shot, Fleiss kappa 0,84 и в 220/222 случаях в few-shot, Fleiss kappa 0,91.

Zero-shot: 92% успешности

Few-shot: 95% успешности



Возможные алгоритмы интерпретации

SCOPE-BOOL(*p*, *h*)

```
1 lexrel ← GET-LEXREL(p, h)
2 neg1 ← FIRST-SCOPE(p, h)
3 neg2 ← SECOND-SCOPE(p, h)
4 if (neg1 ⊕ neg2):
5   return REVERSE(lexrel)
6 return lexrel
```

(a) An interpretable program that solves ScoNe-NLI by computing two Boolean variables that encode whether the first and second negation scope and reversing entailment if exactly one is true.

SCOPE-COUNT(*p*, *h*)

```
1 lexrel ← GET-LEXREL(p, h)
2 count ← COUNT-SCOPED(p, h)
3 if count == 1:
4   return REVERSE(lexrel)
5 return lexrel
```

(b) An interpretable program that solves ScoNe-NLI by counting the scoped negations and reversing entailment if there is exactly one.

IGNORE-SCOPE(*p*, *h*)

```
1 lexrel ← GET-LEXREL(p, h)
2 count ← COUNT-NEG(p, h)
3 if count == 1:
4   return REVERSE(lexrel)
5 return lexrel
```

(c) A flawed heuristic program: we count the negations and reverse entailment if there is a single negation, which is equivalent to ignoring the scope of negation.

IGNORE-NEGATION(*p*, *h*)

```
1 lexrel ← GET-LEXREL(p, h)
2 return lexrel
```

(d) A flawed heuristic program for ScoNe-NLI that outputs the lexical relation and ignores negation entirely.



Ограничения

- Англоязычность
 - Только лексический entailment
 - ScoNe может наследовать проблемы датасетов-предшественников
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