

OPT-R: Exploring the Role of Explanations in Finetuning and Prompting for Reasoning Skills of Large Language Models

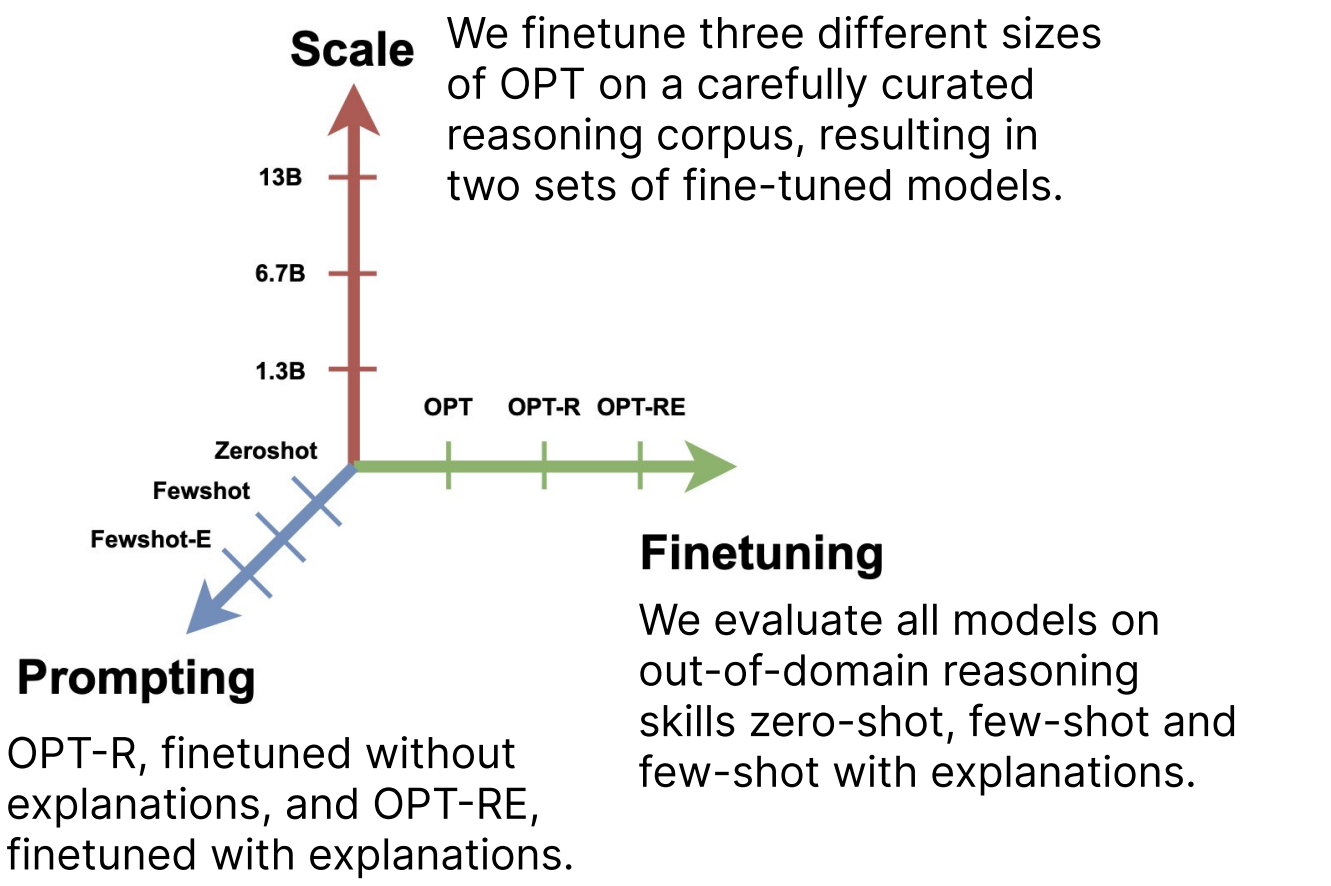
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Introduction

We investigate the reasoning capabilities of LLMs, focusing on the OPT models. We ablate across the following dimensions.



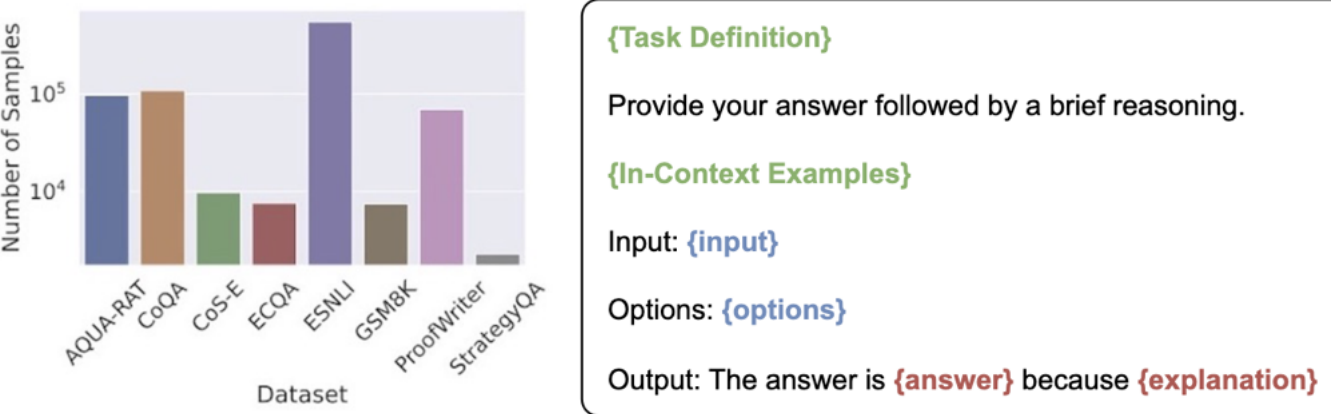
This results in a comprehensive grid of 27 configurations and 6,156 test evaluations.

Skill Analysis

Classification accuracy achieved by different models as a function of the reasoning skill and few-shot prompting method employed. The cells are shaded to according to score, with the highest bolded. Results are grouped by which model performs best.

Skill	OPT		OPT-R		OPT-RE	
	Fewshot	Fewshot-E	Fewshot	Fewshot-E	Fewshot	Fewshot-E
Numerical	39.9	49.7	65.1	65.3	64.7	64.8
Analogical	51.9	46.2	63.3	62.5	60.7	60.9
Objects	53.5	55.1	61.4	63.8	60.0	59.7
Social Interactions	33.6	34.7	43.8	42.3	40.2	40.0
Textual Entailment	43.3	42.0	47.1	47.3	51.9	51.2
Grammatical	54.4	55.1	61.2	60.0	62.0	63.1
Multihop	36.6	31.7	38.9	39.9	39.5	37.0
Symbols	44.2	47.2	51.7	51.8	51.9	52.4
Spatial	44.1	47.1	49.8	51.8	49.6	49.2
Social Situations	46.3	46.6	53.2	53.2	51.9	52.3
Counting	19.6	20.0	13.5	12.7	29.8	32.9
Physical	35.8	40.6	36.9	38.8	48.1	50.0
Logical	31.7	33.4	33.7	34.1	36.9	38.4
Temporal	50.7	49.7	43.4	46.5	48.5	38.5
Argument	55.8	60.1	46.3	45.9	48.6	48.8
TE - Deductive	33.7	38.3	27.9	30.1	29.0	29.9
Relational	47.4	51.1	47.6	47.9	44.8	44.6
Commonsense	35.0	31.8	29.8	29.5	28.5	29.2
TE - Analogical	16.3	18.7	18.6	20.7	18.7	18.1
Abductive	33.9	36.1	36.9	34.4	34.2	35.3
Ethics	26.8	25.8	26.5	25.9	26.2	27.6
Deductive	39.4	40.4	39.4	40.4	40.0	41.1
Causal	50.2	50.6	49.1	48.9	50.1	50.5
Scientific	23.4	23.3	24.3	24.5	25.0	24.5
Numerical Commonsense	59.5	59.2	59.0	59.0	59.2	59.4
Strings	60.7	60.7	61.1	61.2	60.7	60.7

Finetuning Corpus

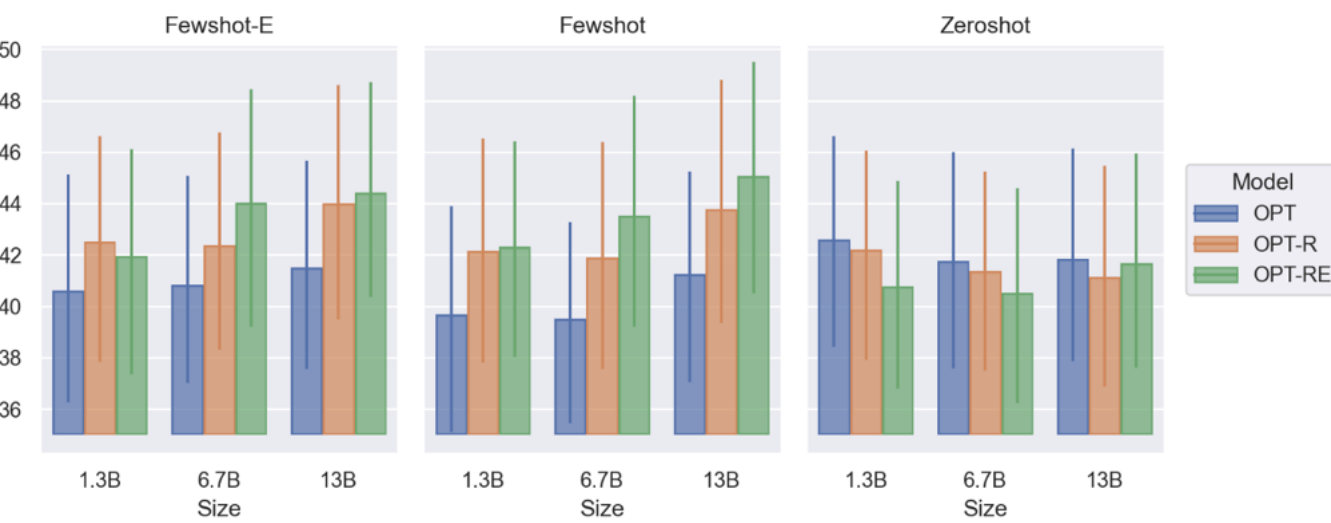


The finetuning corpus is comprised of 8 reasoning datasets with explanations that are either free-form or step-by-step. We format all datasets using the above prompt and generate two training corpuses, one with explanations for OPT-RE and one without explanations for OPT-R.

Evaluation

We evaluate OPT-R on 26 reasoning skills, sampled from a held-out set of 57 Super-NaturalInstructions tasks. Performance is measured by calculating the most likely output from a set of candidates (i.e. rank classification).

Explanations had a small effect on performance as we incorporate it during finetuning, prompting and as model size increases.



Further Analysis

Model	Std(F-FE)	Avg(F)	Avg(FE)
OPT	2.31	40.68	41.82
OPT-R	0.84	43.44	43.68
OPT-RE	0.78	44.49	44.86

Explanations during prompting does not significantly impact finetuned models on reasoning datasets but makes a difference for vanilla OPT.

Skill	OPT	OPT-R	OPT-RE
Numerical	44.8	65.2*	64.7*
Analogical	49.0	62.9*	60.8*
Counting	19.8	13.1	31.3*
Physical	38.2	37.8	49.1*
Entailment	42.6	47.2	51.6*
Social Int	34.1	43.0*	40.1
Objects	54.3	62.6*	59.9*

Reasoning skills where either OPT-RE or OPT-R are **significantly better than the vanilla OPT**. Explanations help Counting, Physical Reasoning and Entailment.

Skill	OPT	OPT-R	OPT-RE
Argument	57.9	46.1 ⁻	48.7 ⁻
TE - Deductive	36.0	29.0 ⁻	29.4 ⁻
Commonsense	33.4	29.7	28.8 ⁻

Reasoning skill where **OPT performs significantly better** than either OPT-R or OPT-RE

[^] Significance is measured by Welch's t-test ($p < 0.05$) denoted by the * symbol.

