Focus Directions Make Your Language Models Pay More Attention to Relevant Contexts



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1. LLMs are prone to distracted by irrelevant context. Why?

1.1 Identify contextual heads

Contextual Scoring: A metric that quantifies the degree of attention allocated to specific segments of the input (e.g., relevant contexts) during response generation.

Contextual Heads: The top-k attention heads ranked by contextual score (to relevant contexts)

1.2 Properties of contextual heads

Contextual heads are sparse, located in middle and late layers

Correct

More

Wrong

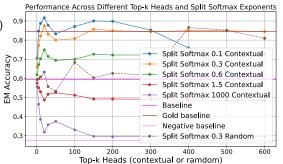
Less

	contexts																					
									Lla	am	a-3	.2-	3B-	-Ins	stru	ıct						3
Layer lds	6 4 2 0																				- 0.5-1	
	12 10 8																				- 0.2-0.5	2
	87																				- 0.1-0.2	Ć
	24 22																				- 0-0,1	C

0 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23

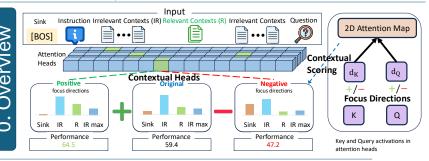
1.3 Modifying attention on contextual heads

- Method: Split softmax (<1 increase attention, > 1 decrease)
- Increase attention to relevant contexts: performance ↑
- **Decrease** attention to relevant contexts: performance ↓
- · Non contextual heads have minimum such effects



Contextual heads controls the overall attention of LLMs

Modifying attention on contextual heads could make performance better than gold baseline (relevant context only), or close to the negative baseline (irrelevant contexts only)



2. Focus directions move attention from sink to relevant contexts

Motivation: Can contextual heads figure out the relevant contexts by themselves?

Locating focus directions: simply train d_K and d_Q to maximize the contextual score for the relevant contexts.

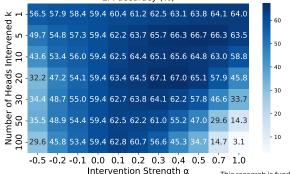
Main findings:

LLM Response

Focus to relevant

Focus direction	Relevant contexts	Sink
Positive	More attention	Less attention
Negative	Less attention	More attention

 Focus directions only help mitigate distraction on contextual heads. EM accuracy (%)



3. Focus directions mitigate poor task alignment

Benchmark: HELMET (5 categories, 16 tasks used)

Main findings:

- Focus directions help for the long context tasks that LLM could do well in the short context.
- Most of the tasks could be improved by either positive or negative focus direction.
- Focus direction improves the overall performance of poorly aligned LLMs. (e.g., base vs. instruct, inconsistent sink score for the same context length)

Model	Recall	RAG	Re-ranking	ICL	Long QA	Overall Average	Model	Recall	RAG	Re-ranking	ICL	Long QA	Overall Average	
Llama-3.2							Llama-3.2	2-3B						
200.2	66.00	54.96	29.22	82.20		58.10	200.2	55.50	50.38	6.83	85.20		49.48	
100.2	73.81	56.58	26.73	83.00		60.03	100.2	64.56	53.96	6.24	86.20		52.74	
20.0.2	81.50	58.75	25.37	80.20		61.46	20.0.2	66.31	56.46	7.08	85.40		53.81	
10.0.2	82.00	58.54	26.16	80.60		61.83	10.0.2	65.69	55.83	9.27	85.40		54.05	
baseline	78.88	58.83	26.10	82.20		61.50	baseline	65.50	54.83	7.29	86.20		53.46	
Llama-3.2	2-3B-Inst	ruct					Llama-3.2		ruct					
200.2	73.00	58.04	13.68	78.80	27.32	50.17	20,-0.2	56.38	56.75	3.77	83.80	28.64	45.87	
100.2	79.12	60.21	13.32	79.40	26.66	51.74	100.2	61.06	58.33	2.44	85.00	30.38	47.44	
20.0.2	83.50	60.25	20.58	80.60	26.09	54.20	20.0.2	65.81	59.21	2.72	83.40	28.23	47.87	
10.0.2	83.69	62.08	20.77	80.40	25.94	54.58	10.0.2	64.25	59.79	3.10	84.20	26.80	47.63	
baseline	84.38	63.00	17.13	80.20	26.78	54.30	baseline	64.12	59.96	3.77	85.00	31.13	48.80	
Owen2.5	-7B						Owen2.5	-7B						
200.2	94.56	53.50	22.86	77.60	-	62.13	200.2	42.06	41.96	1.30	77.40	-	40.68	
100.2	95.31	54.58	24.84	78.40	-	63.28	100.2	45.00	43.42	2.64	77.40	-	42.11	
20.0.2	95.88	54.04	23.25	79.40	-	63.14	20.0.2	46.56	43.08	1.19	77.60	-	42.11	
10.0.2	95.50	54.08	23.11	80.00	-	63.17	10.0.2	46.56	43.62	1.07	78.80	-	42.51	
baseline	96.00	54.21	23.15	79.60	-	63.24	baseline	45.19	44.12	1.88	78.00	-	42.30	
Owen2.5	wen2,5-7B-Instruct						Owen2.5-7B-Instruct							
200.2	94.38	55.87	35.88	78.40	33.73	59.65	200.2	46.31	43.96	11.92	78.80	22.07	40.61	
100.2	95,38	56.54	35.78	78.40	33,63	59.94	100.2	46.38	44.42		78.60	21.66	40.65	
20.0.2	95.44	58.50	36.75	78.40	33,45	60.51	20.0.2	51.38	46.88	10.28	78.20	22.95	41.94	
10.0.2	95.44	58.79	35.85	78.20	32.61	60.18	10.0.2	49.25	47.79	11.66	78.20	23.94	42.17	
baseline	95.25	57.71	36.56	77.40	31.92	59.77	baseline	47.88	46.54	11.88	78.60	22.43	41.46	
							Ministral-8B-Instruct-2410							
200.2	94.62	61.79	31.31	77.20	33.59	59.70	200.2	30.56	46.17	0.00	80.60	21.41	35.75	
100.2	94.56	62.17	29.74	78.80	33.17	59.69	100.2	30.06	46.04	0.00	80.00	20.62	35.34	
20.0.2	93.81	63.46	38.86	79.40	29.00	60.91	20.0.2	30.88	47.12	0.00	81.80	19.98	35.96	
10.0.2	93.81	63.87	36.69	79.60	28.74	60.54	10.0.2	31.19	46.79	0.00	82.80	19.49	36.05	
baseline	94.75	63.58	33.68	79.00	31.56	60.51	baseline	30.62	47.17	0.00	81.40	21.40	36.12	

Table 2: Results of HELMET benchmark under 32k (left) and 64k (right) context. Green indicates better than the baseline; red indicates worse than the baseline.