Llama3 & DeepSeek v3 (part 2)

Architecture | Inference



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1. Llama3 Arch

3. Benchmarking out of box



2. DeepSeekV3 Arch

4. Notebook Walkthrough





Herd of models including 405B LM, 70B, 8B, 1B versions and also Llama Guard 3 for input/ output safety

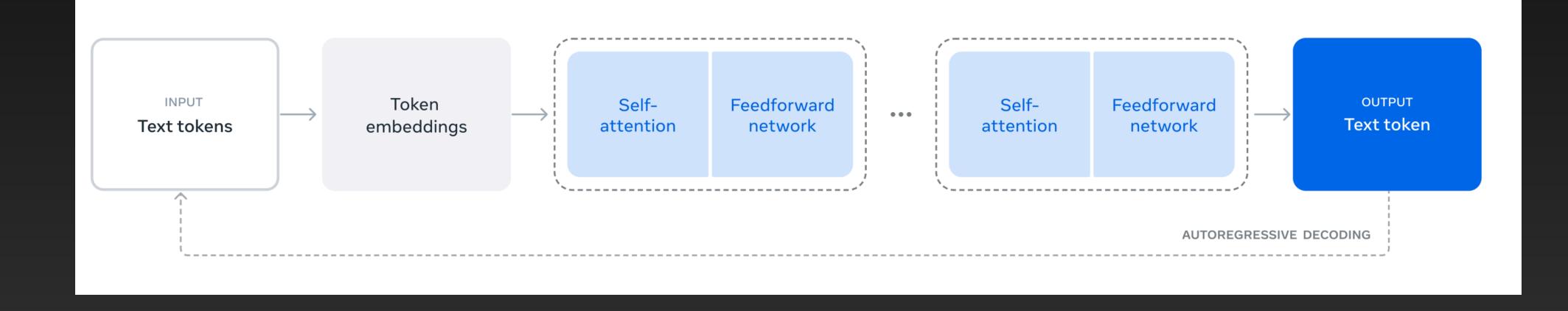
Reference: <u>https://arxiv.org/pdf/2407.21783</u>

Lama3 Herd

Lama 3 Herd of Models

	Finetuned	Multilingual	Long context	Tool use	Release
Llama 3 8B	×	X ¹	×	×	April 2024
Llama 3 8B Instruct		×	×	×	April 2024
Llama 3 70B	×	\mathbf{X}^{1}	×	×	April 2024
Llama 3 70B Instruct		×	×	×	April 2024
Llama $3.1 8B$	×			×	July 2024
Llama 3.1 8B Instruct				√	July 2024
Llama $3.1~70B$	×			×	July 2024
Llama 3.1 70B Instruct				 Image: A second s	July 2024
Llama $3.1 405B$	×			×	July 2024
Llama 3.1 405B Instruct		✓	✓		July 2024

Reference: <u>https://arxiv.org/pdf/2407.21783</u>



Tokenization

def print_tokens_with_ids(txt): tokens = tokenizer.tokenize(txt, add_special_tokens=False) token_ids = tokenizer.encode(txt, add_special_tokens=False) print(list(zip(tokens, token_ids)))

Based on the information provided, rewrite the sentence by changing its tense fro

She played the piano beautifully for hours and then stopped as it was midnight.<

11 11 11

print_tokens_with_ids(prompt)

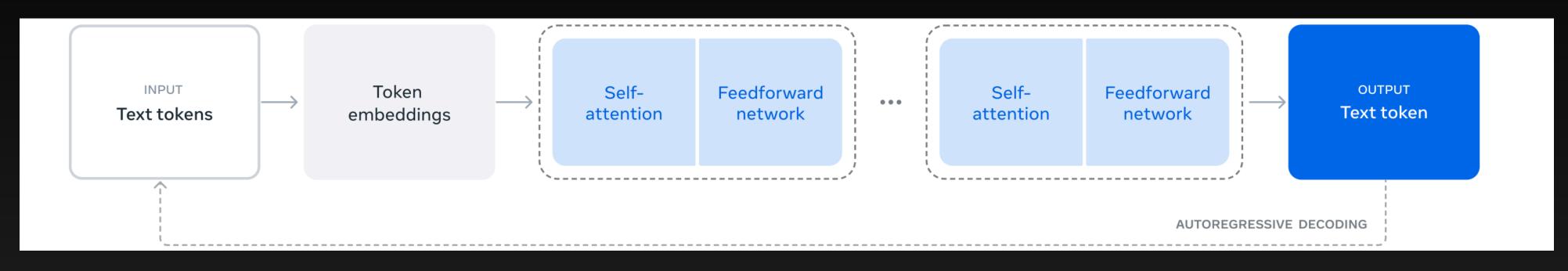
```
# Token and Token ID
```

[128000, 1	128000,	128006,	9125,	128007,	271,	29815,	389,	279,	
	2038,	3984,	11,	18622,	279,	11914,	555,	10223,	1202,
	43787,	505,	3347,	311,	3938,	13,	128009,	128006,	882,
	128007,	271,	8100,	6476,	279,	27374,	32719,	369,	4207,
	323,	1243,	10717,	439,	433,	574,	33433,	13,	128009,
	128006,	78191,	128007,	271]					

prompt = """<|begin_of_text|><|start_header_id|>system<|end_header_id|>

```
> [('<|begin_of_text|>', 128000), ('<|start_header_id|>', 128006), ('system', 912
```

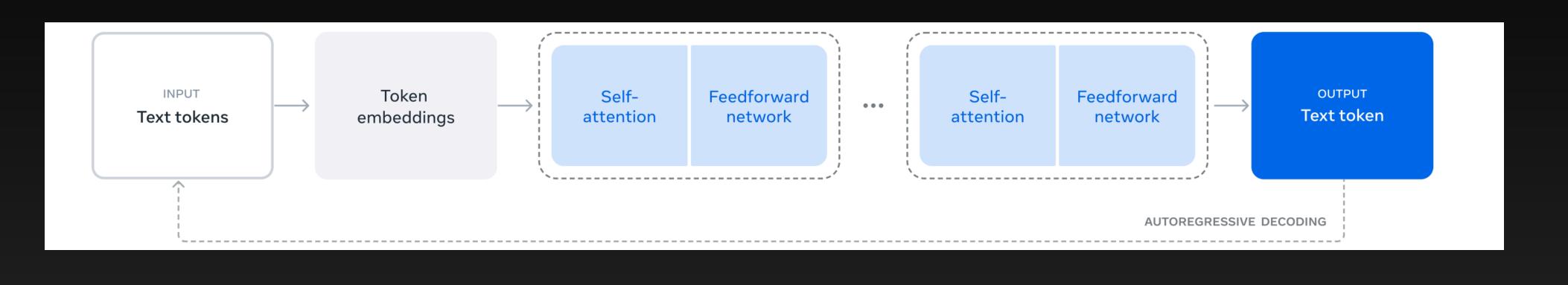


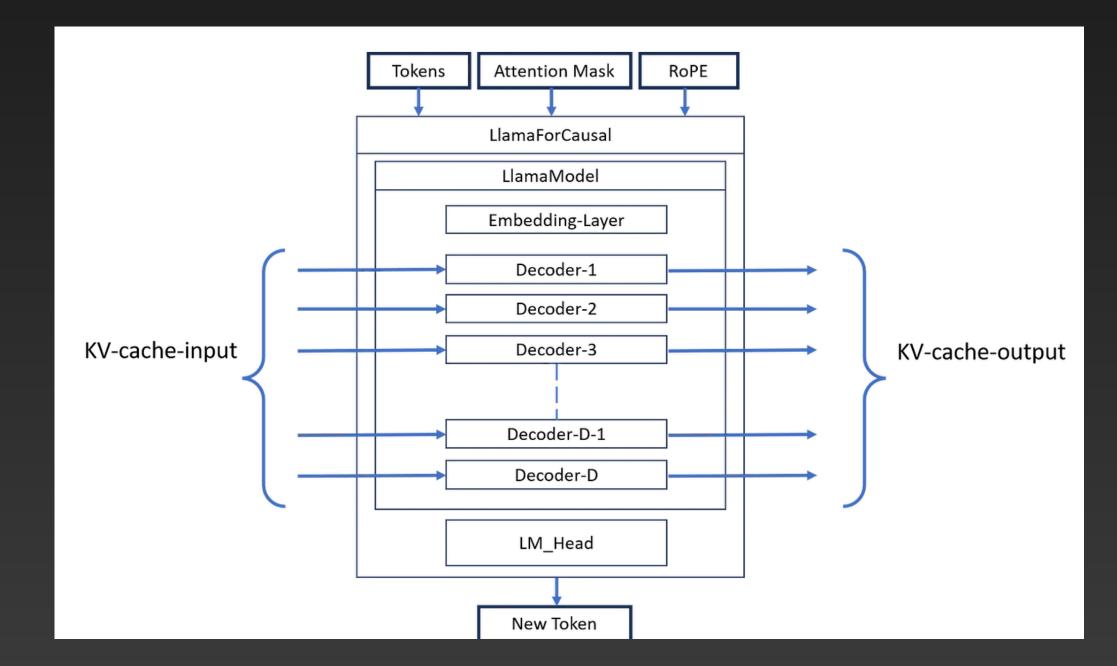


Assume that Llama4 is trained on 40T tokens of data. It has a context window of 256k tokens and has a tokenizer vocab size of 108k tokens and each token has a token embedding size of 4096. What is the number of classes present in the classifier of the LM head to generate the next token in auto-regressive decoding? 256k **|a)** b) **40T** 108k 128k 4096

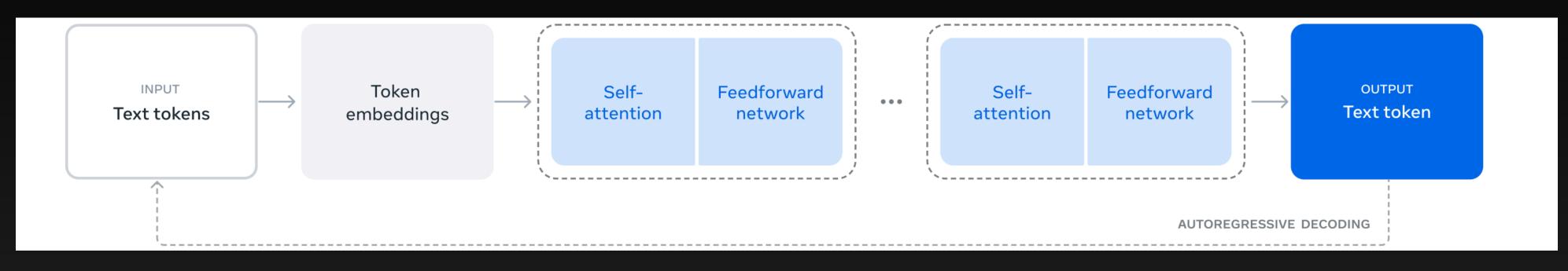
ď

e)





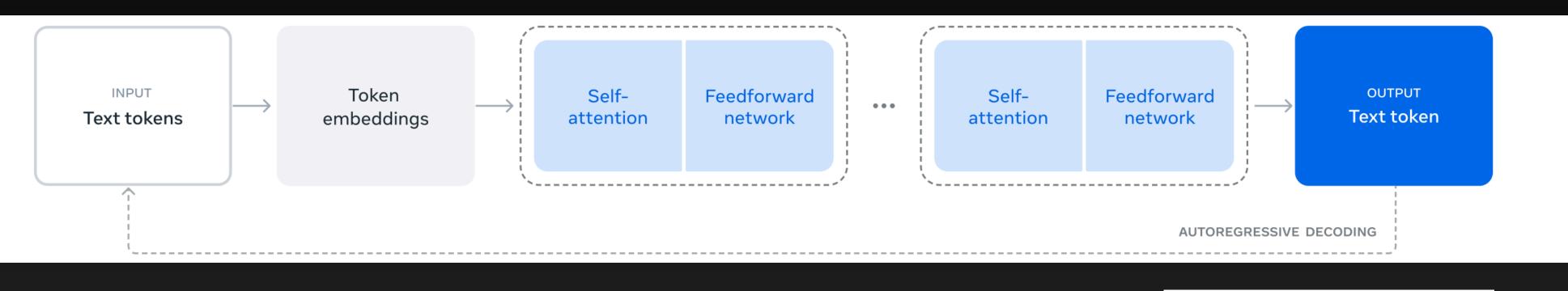


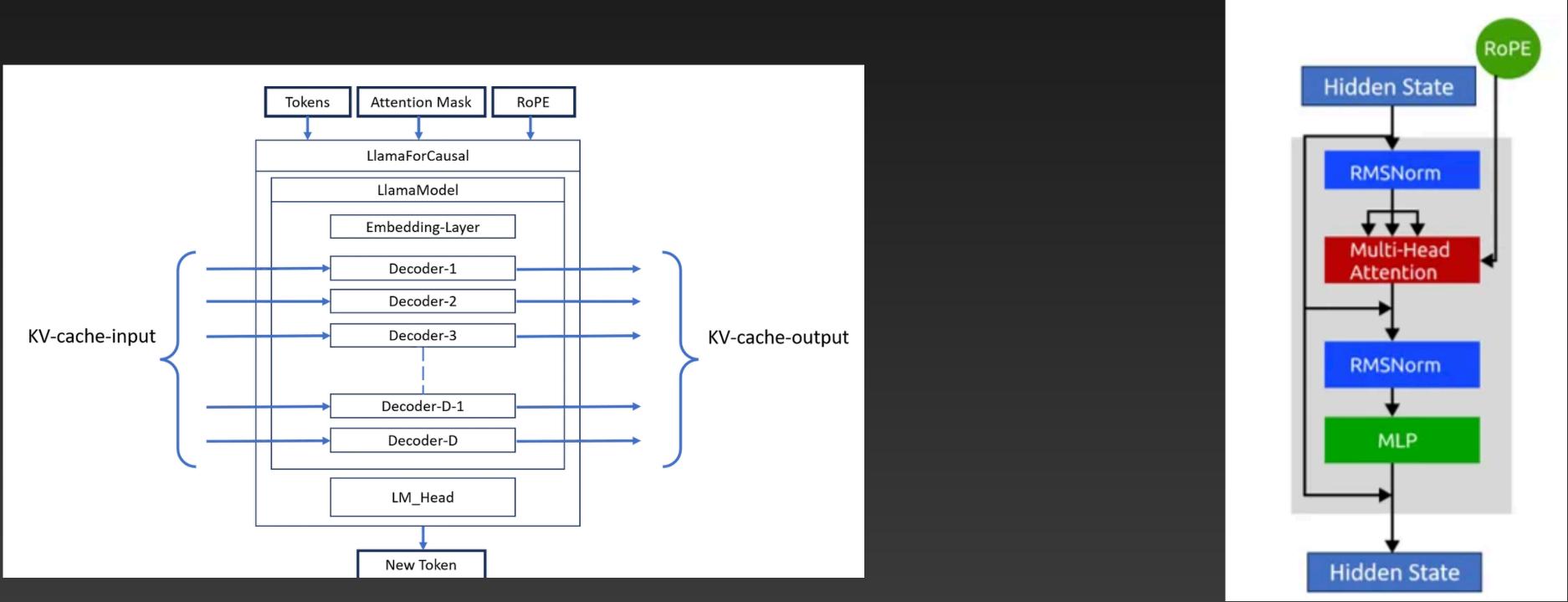


Assume that input sentence has 100 words and tokenized into 150 tokens. The 150 tokens are now assigned a token embedding and passed through 64 decoder blocks of Llama model. At the very end, a new token is also generated. How many tokens exist right after the 64 decoder blocks and how many new tokens are generated?

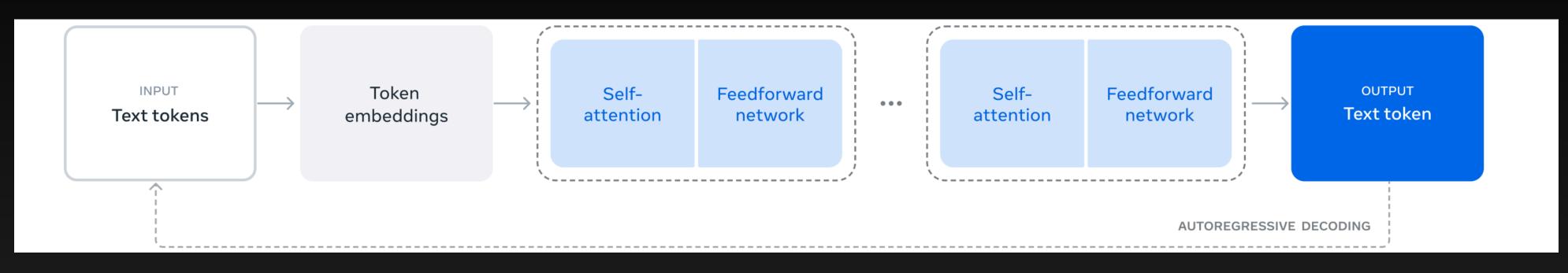
a) b) c) d) e)

150,64 64,1 1,1 150,1 150,150





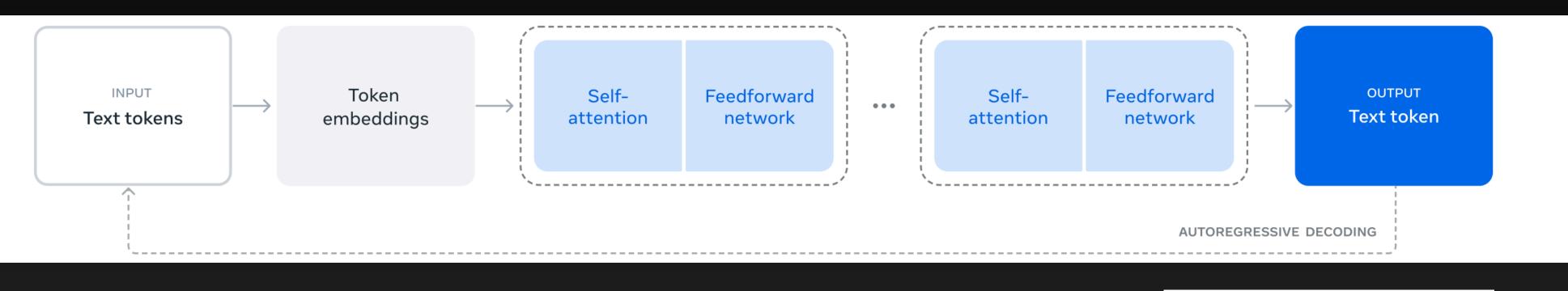


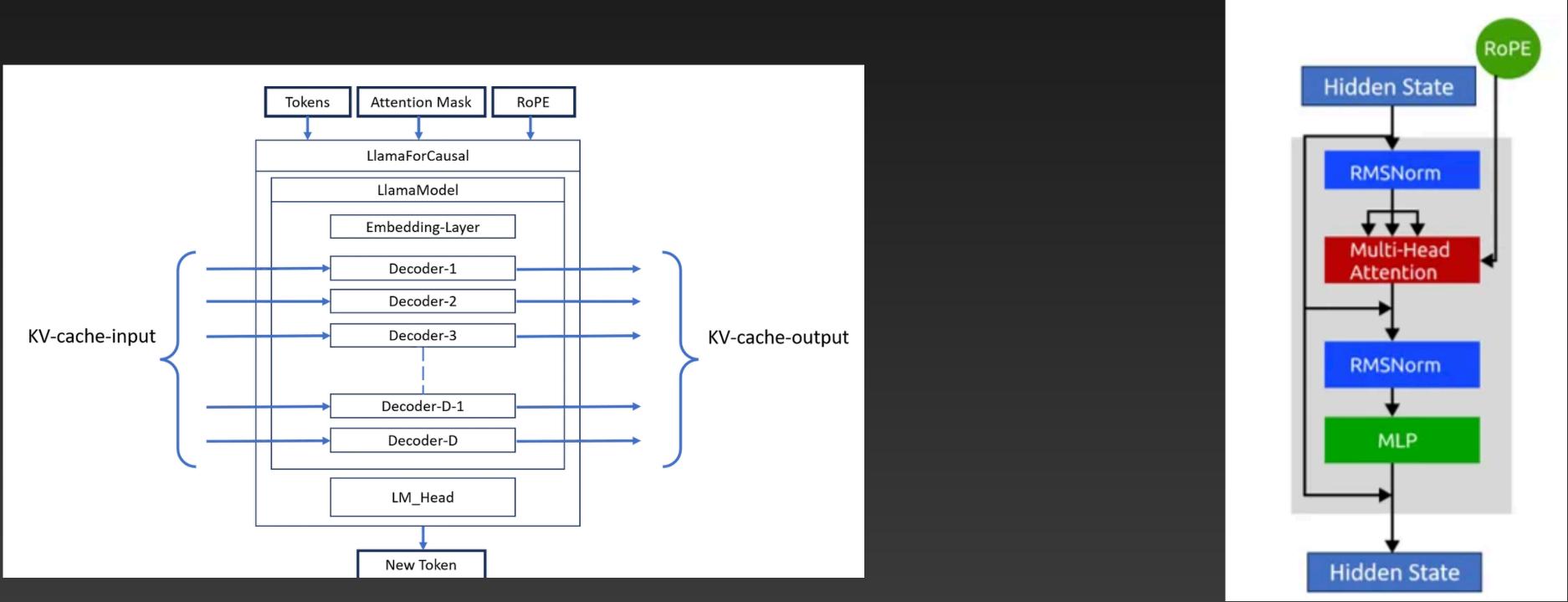


Assume that each word in a sentence corresponds to a token (for simplicity). Consider the sentence: "The sun rises". Passing this into Ilama produces the output token as "in". How many tokens will be passed in to the next step of autoregressive decoding and what will the expected output?

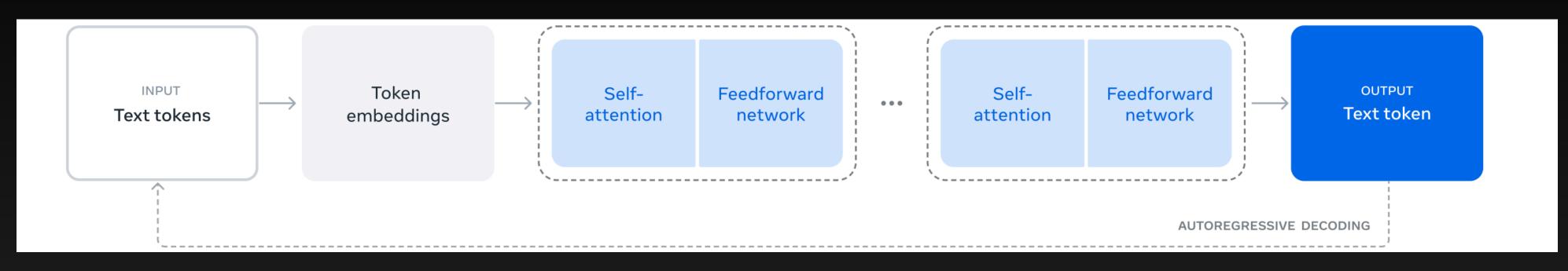
a) b) c) d)

- 3,"east"
- 4,"east"
- 3,"the
- 4, "the"









What is the role of KV-ca
To compute attentio
To cache all the m
To speed up computation in
To cache intermediate query, t
deo
All

a)

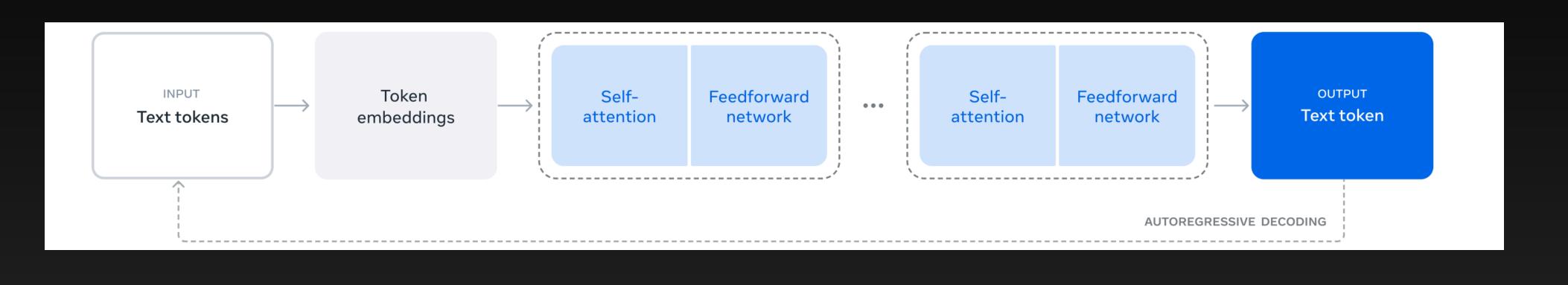
 $|b\rangle$

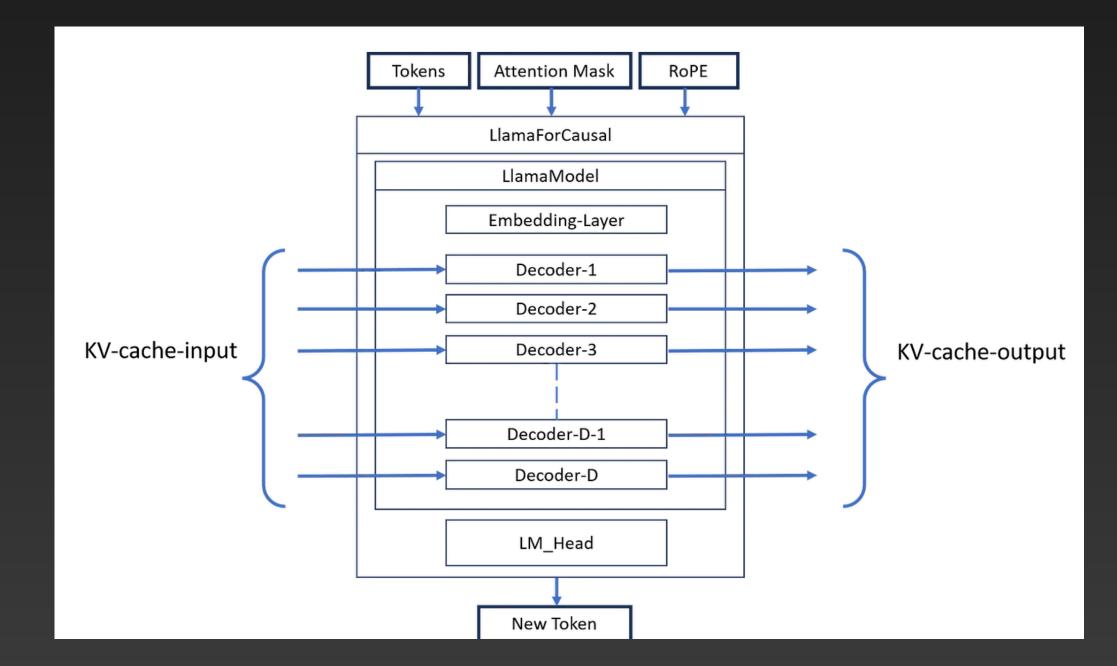
c)

d)

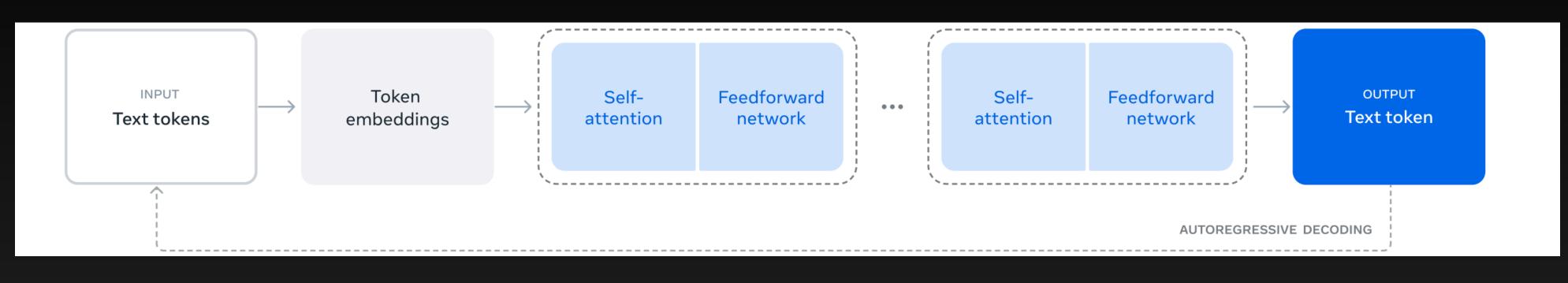
e)

- The in the Llama architecture? In using query, keys and values hodel parameters in memory the auto-regressive decoding process token embeddings to be used in the next coding phase
- of the above





ICE #5 | LM head

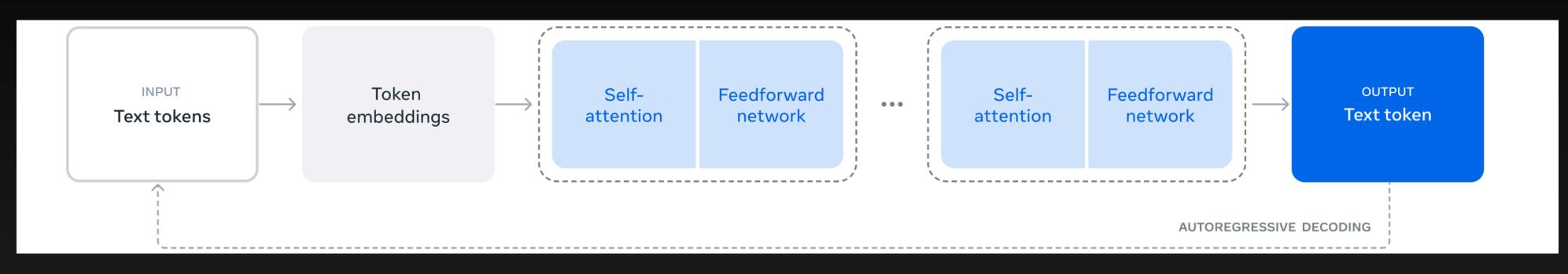


a) b) c) d) What is the composition of LM head? Linear transformation + softmax activation Non-linear transformation + relu activation Non-linear transformation + softmax activation Linear transformation + relu activation

Context Window: 128k tokens Vocab size: 128k tokens **Training data:** 15T tokens **Decoder blocks: 32 Positional Embedding:** RoPE







a) b)

c)

Why is increasing the input context window been a challenge for LLM models. Only recently have LLM models increased it from 4000 tokens to 100k tokens

Compute increases linearly with context window size

Compute increases quadratically with context window size

Compute increases exponentially with context window size

Lama2vs Lama3

7 times larger pre-train data set. 15 Trillion **Tokens** of data ~ 150 million books High-quality filters to filter out bad data in training - Use Llama2 Better "data mix" - Trivia, STEM, coding, historical knowledge

Larger model means better performance (8B vs **70B**) But more data = better performance (also avoids over-fitting). Log-linear improvement from 200B to 15T tokens

Lama3 vs DeepSeek

	Llama3	DeepSeek V3		
Parameters	405b	405b with 37b active at inference		
Architecture	Traditional Transformers (Decoder)	Transformer with MOE and MLA		
Context Length	128k tokens	128k tokens		
Post Training	Instruct FT	Instruct FT		
RL	DPO	DPO		

Lama3 Benchmarks

Category	Benchmark	Llama 3 8B	Gemma 2 9B	Mistral 7B	Llama 3 70B	Mixtral 8x22B	GPT 3.5 Turbo	Llama 3 405B	Nemotron 4 340B	GPT-4 (0125)	GPT-40	Claude 3.5 Sonnet
	MMLU (5-shot)	69.4	72.3	61.1	83.6	76.9	70.7	87.3	82.6	85.1	89.1	89.9
General	MMLU (0-shot, CoT)	73.0	$72.3^{ riangle}$	60.5	86.0	79.9	69.8	88.6	78.7^{\triangleleft}	85.4	88.7	88.3
General	MMLU-Pro (5-shot, Cot)	48.3	—	36.9	66.4	56.3	49.2	73.3	62.7	64.8	74.0	77.0
	IFEval	80.4	73.6	57.6	87.5	72.7	69.9	88.6	85.1	84.3	85.6	88.0
Code	HumanEval (0-shot)	72.6	54.3	40.2	80.5	75.6	68.0	89.0	73.2	86.6	90.2	92.0
Code	$MBPP \ EvalPlus \ {\rm (0-shot)}$	72.8	71.7	49.5	86.0	78.6	82.0	88.6	72.8	83.6	87.8	90.5
Math	GSM8K (8-shot, Cot)	84.5	76.7	53.2	95.1	88.2	81.6	96.8	92.3^{\diamondsuit}	94.2	96.1	96.4^{\diamondsuit}
Math	MATH (0-shot, Cot)	51.9	44.3	13.0	68.0	54.1	43.1	73.8	41.1	64.5	76.6	71.1
Reasoning	ARC Challenge (0-shot)	83.4	87.6	74.2	94.8	88.7	83.7	96.9	94.6	96.4	96.7	96.7
Reasoning	GPQA (0-shot, CoT)	32.8	—	28.8	46.7	33.3	30.8	51.1	_	41.4	53.6	59.4
Tool use	BFCL	76.1	_	60.4	84.8	-	85.9	88.5	86.5	88.3	80.5	90.2
iooiuse	Nexus	38.5	30.0	24.7	56.7	48.5	37.2	58.7	_	50.3	56.1	45.7
	ZeroSCROLLS/QuALITY	81.0	_	_	90.5	—	_	95.2	_	95.2	90.5	90.5
Long context	InfiniteBench/En.MC	65.1	—	—	78.2	-	—	83.4	—	72.1	82.5	—
	$\rm NIH/Multi-needle$	98.8	—	—	97.5	—	—	98.1	—	100.0	100.0	90.8
Multilingual	MGSM (0-shot, CoT)	68.9	53.2	29.9	86.9	71.1	51.4	91.6	_	85.9	90.5	91.6

Lama 3 Instruct Performance

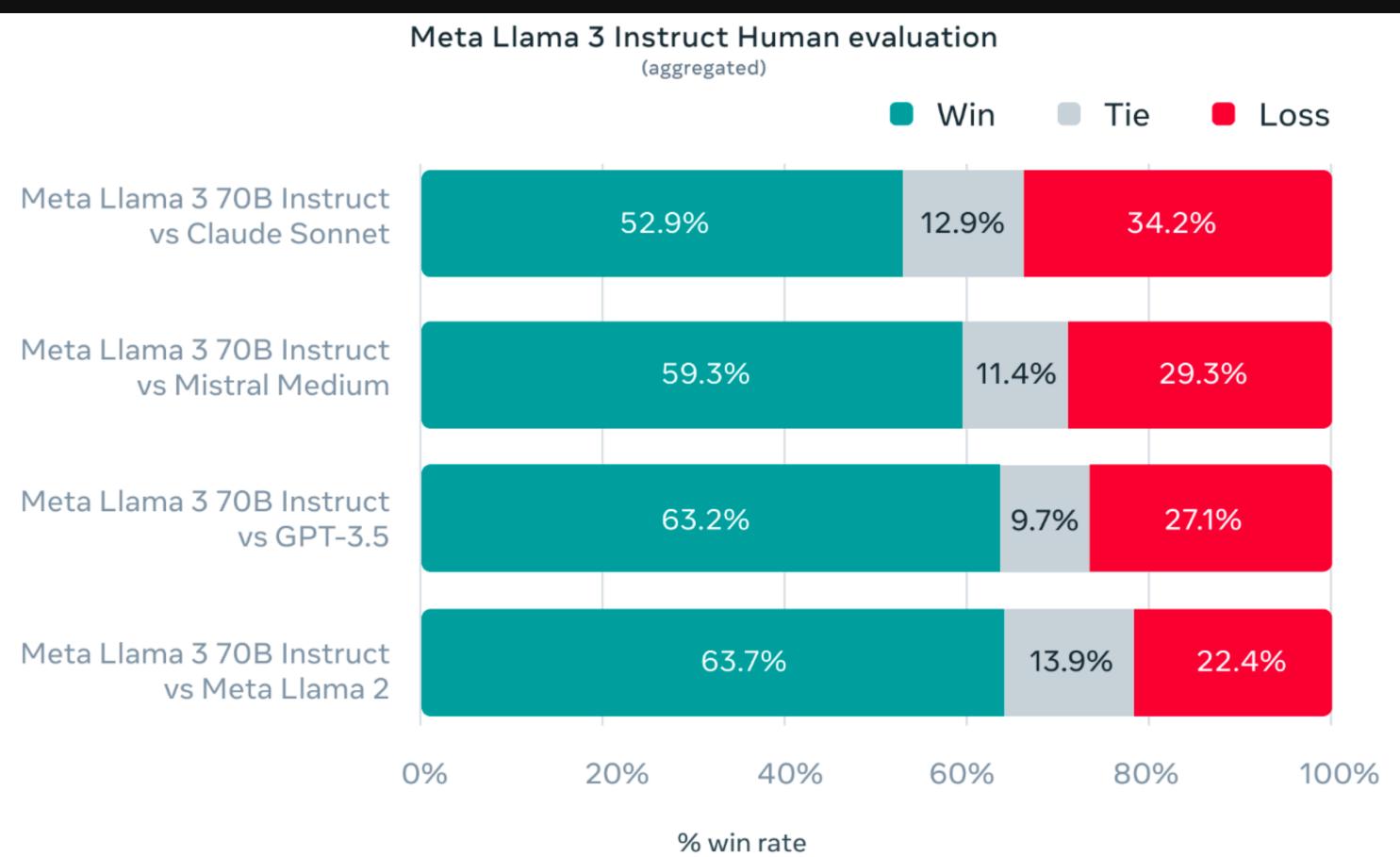
Meta Llama 3 Instruct model performance

	Meta Llama 3 8B	Gemma 7B - It Measured	Mistral 7B Instruct Measured
MMLU 5-shot	68.4	53.3	58.4
GPQA 0-shot	34.2	21.4	26.3
HumanEval 0-shot	62.2	30.5	36.6
GSM-8K 8-shot, CoT	79.6	30.6	39.9
MATH 4-shot, CoT	30.0	12.2	11.0

Reference: <u>https://ai.meta.com/blog/meta-llama-3/</u>

	Meta	Gemini	Claude 3
	Llama 3	Pro 1.5	Sonnet
	70B	Published	Published
MMLU 5-shot	82.0	81.9	79.0
GPQA	39.5	41.5	38.5
0-shot		сот	Сот
HumanEval 0-shot	81.7	71.9	73.0
GSM-8K	93.0	91.7	92.3
8-shot, CoT		11-shot	O-shot
MATH 4-shot, CoT	50.4	58.5 Minerva prompt	40.5

Lama 3 Instruct Performance



Reference: <u>https://ai.meta.com/blog/meta-llama-3/</u>

