

Language Models in Cryptanalysis

How do we crack the long ciphers?

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1. The papers

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X-former Elucidator: Reviving Efficient Attention for Long Context Language Modeling

Miao, et al., 2024

Transformers are RNNs: Fast Autoregressive Transformers with Linear Attention

Katharopoulos, et al., 2020

Long-Short Transformer: Efficient Transformers for Language and Vision

Zhu, et al., 2021

Rethinking Attention with Performers

Choromanski, et al., 2020

Linformer: Self-Attention with Linear Complexity

Wang, et al., 2020

1.1 Why these papers?

Efficient computation during training

- Causal LM inference is fast

Best methods from comparative study

- We don't have time for them all
- Some are more inference focused

Summary

- We care about efficiency during training
- Inference for long cipher struggle due to lack of generalization on long ciphers

2. Recap from last time

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Homophonic Substitution Ciphers

- 1:>0 mappings
- English without spaces & punctuation

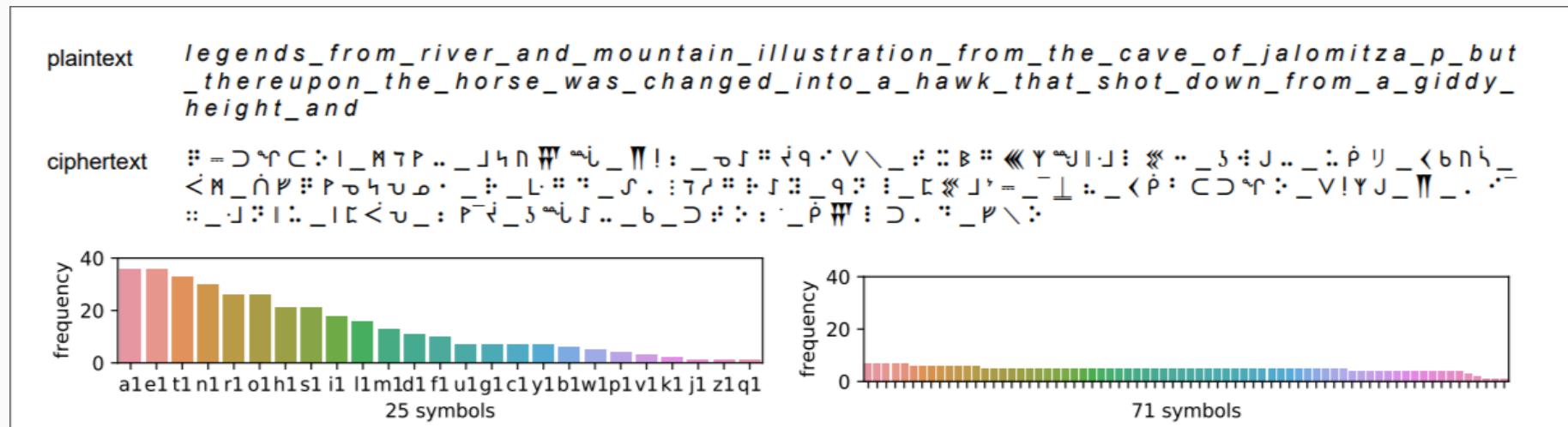


Figure 1: Example of a homophonic substitution cipher. (Kambhatla et al., Findings 2023)

2. Recap from last time

Causal LM

- Learns both cipher & plaintext
- Reads left to right

Seq2seq

- Learns plaintext only
- Bidirectional

Both suffer from $O(N^2)$ attention 😞

#keys	Model	Max Len.	
		400	700
30-45	Seq-to-Seq	72.30	fail
	PrefixLM	54.73	69.50
	CausalLM (tgt)	29.99	37.20
	CausalLM	0.40	0.21
40-65	PrefixLM	69.50	54.73
	CausalLM (tgt)	29.99	37.20
	CausalLM	0.83	0.80
30-85	PrefixLM	70.52	71.82
	CausalLM (tgt)	42.05	42.69
	CausalLM	2.25	2.19

Figure 2: SER on synthetic HS ciphers.
(Kambhatla et al., Findings 2023)

3. Standard Attention Computation

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Rows (Queries)

- Token we are looking for

Columns (Keys)

- Token we are looking at

Values (Cells)

- Attention Score

Notice: $(N \times N) = N^2$

Cipher: X Y Z

$$X \rightarrow [Y, Z]$$

$$Y \rightarrow [X, Z]$$

$$Z \rightarrow [X, Y]$$

	X	Y	Z
X	$X \rightarrow X$	$X \rightarrow Y$	$X \rightarrow Z$
Y	$Y \rightarrow X$	$Y \rightarrow Y$	$Y \rightarrow Z$
Z	$Z \rightarrow X$	$Z \rightarrow Y$	$Z \rightarrow Z$

Table 1: Attention Weight Matrix ($N \times N$)

3. Standard Attention Computation

Why is $O(N^2)$ Ineffecient?

Many attention heads

- Each head has its own matrix

Many forward passes

- Each matrix recomputed at each pass

Causal LM as an example

- 12 layers \times 12 heads =
144 attention heads

**Ciphers with 1000s of characters
does not scale well 😭**

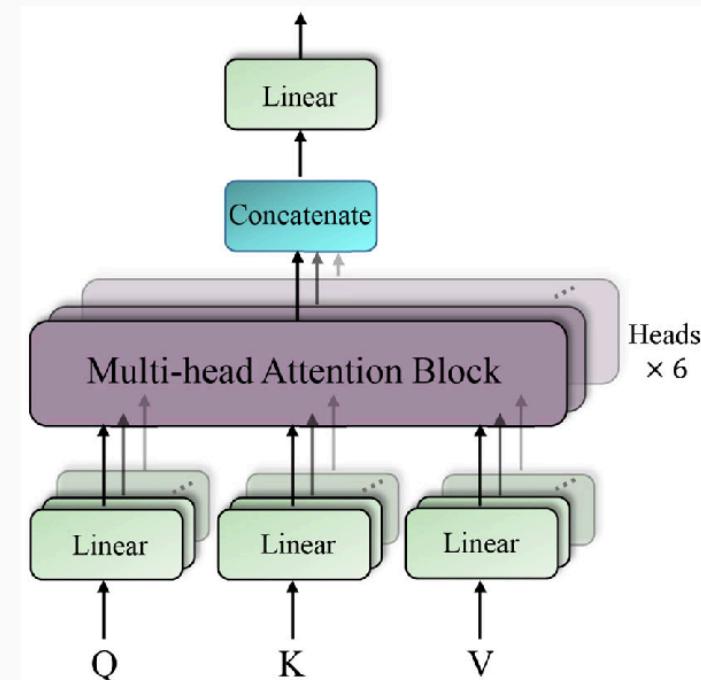


Figure 3: Example of attention with multiple heads. (Yuan et al., 2022)

4. Questions?
