# Your Transformer is Secretly Linear

8th June 2024

#### **Transformers**

- 1. Large number of parameters →Subject to overfitting
- 2. As size grows → More hardware required for inference (bottleneck)
- 3. Hard to visualize/understand all the layers and importance (as opposed to CNN's where several hierarchical features can be understood)

### Efficient transformers

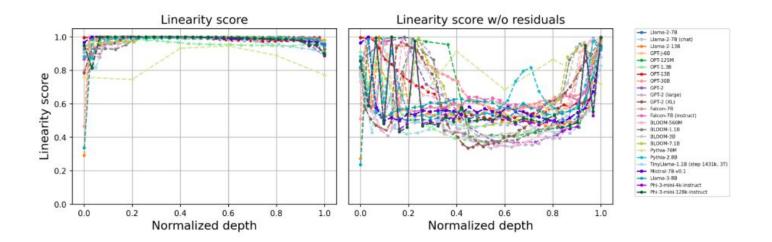
1. <a href="https://arxiv.org/pdf/2009.06732">https://arxiv.org/pdf/2009.06732</a> (Efficient Transformers: A Survey)

# Key concepts

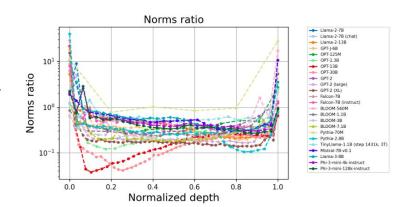
- 1. Embedding transformations can be thought of as change of the values of input embeddings as they go through the different transformer layers
- 2. Residual component refers to the skip connections that are present in transformers i.e. Output = f(input)+Input
  - a. Introduced originally to solve the problem of vanishing gradients
  - b. Keep the input context intact
- 3. Output norm is the magnitude of the output from a layer
- 4. Linearity score measured by Procrustes similarity

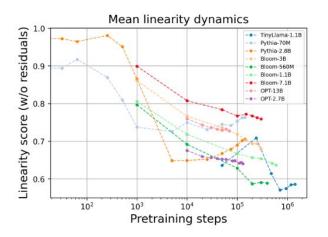
# Main motivation (make the model smaller)

- 1. Understand and quantify this linearity
- 2. Depth pruning
- 3. Cosine-similarity-based regularization, aimed at reducing layer linearity (having more information / layer )



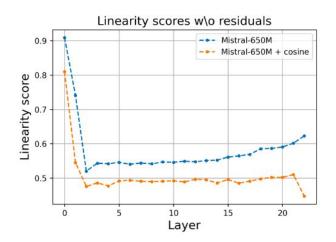
- Normalized depth → Layer index / Total number of layers
- 2. Linearity score w/o residuals is non-linear since the transformation function can be highly non-linear.. Output = f(input), where as with residuals Output = f(input)+Input the linearity score is ~0.99
- 3. Norms ratio = f(input) / Output

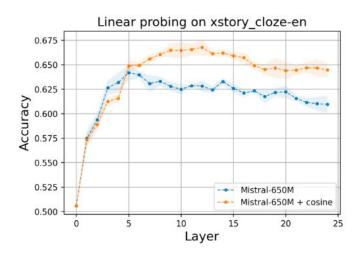




Model Name	Super_Glue/MultiRC	Super_Glue/BoolQ	Super_Glue/CB	Reward Modeling
OPT-125M	$0.085 \pm 0.008$	$0.217 \pm 0.038$	$0.048 \pm 0.009$	$0.060 \pm 0.008$
OPT-1.3B	$0.055 \pm 0.021$	$0.382 \pm 0.004$	$0.088 \pm 0.010$	$0.062 \pm 0.007$
OPT-2.7B	$0.061 \pm 0.025$	$0.356 \pm 0.005$	$0.066 \pm 0.029$	$0.054 \pm 0.003$
Llama2-7B	$0.141 \pm 0.006$	$0.051 \pm 0.024$	$0.081 \pm 0.070$	$0.194 \pm 0.027$
GPT2	$0.085 \pm 0.021$	$0.048 \pm 0.016$	$0.004 \pm 0.003$	$0.092 \pm 0.013$
GPT2-Large	$0.049 \pm 0.003$	$0.023 \pm 0.008$	$0.025 \pm 0.014$	$0.085 \pm 0.008$
GPT2-XL	$0.040 \pm 0.007$	$0.037 \pm 0.007$	$0.028 \pm 0.019$	$0.038 \pm 0.008$

Table 1: Delta of linearity score w/o residuals after fine-tuning various tasks. Note that all values are strictly positive, which means that linearity always increases during fine-tuning.





XStoryCloze consists of the professionally translated version of the English StoryCloze dataset (Spring 2016 version) to 10 non-English languages. This dataset is intended to be used for evaluating the zero- and few-shot learning capabilities of multiingual language models. This dataset is released by Meta AI.

