## From Alexnet to Transformers: Measuring the Non-linearity of Deep Neural Networks with Affine Optimal Transport

Quentin Bouniot<sup>1</sup> levgen Redko<sup>2</sup> Anton Mallasto<sup>3</sup> Charlotte Laclau<sup>1</sup> Karol Arndt<sup>4</sup> Oliver Struckmeier<sup>4</sup> Markus Heinonen<sup>4</sup> Ville Kyrki<sup>4</sup> Samuel Kaski<sup>4,5</sup>

<sup>1</sup>Telecom Paris <sup>2</sup>Noah's Ark Lab <sup>3</sup>Smartly.io <sup>4</sup>Aalto University <sup>5</sup>University of Manchester

#### **Motivations**

Non-linearity is at the heart of DNNs

- ► Universal function approximators thanks to non-linearity.
- Mainly introduced through activation functions which are the common ingredients between architectures.

#### No such notion of quantifying non-linearity exists in the literature.

► Research mainly focus on quantifying expressive power of DNNs.

Goal: Measure non-linearity of activation functions from data distribution

#### General idea

Measure non-linearity as lack of linearity through Optimal Transport (OT)

- ► We know the closed-form solution of the OT problem for random variables (RVs) following normal distributions.
- ► For any RVs X and Y, if Y = TX with T Positive Semi-Definite (PSD) matrix, then the solution of OT problem is exactly the one of their normal approximations (N<sub>X</sub> ~ N(µ(X), Σ(X)) and N<sub>Y</sub> ~ N(µ(Y), Σ(Y))).
- ► We obtain an **upper bound** on the difference of the two OT problems.
- ► We can define the **affinity score** using this bound.

#### **Affinity Score**



•  $\rho_{\text{aff}}$  describes how much Y differs from being a PSD affine transformation of X.

▶ 
$$0 \le \rho_{\text{aff}}(X, Y) \le 1$$
, and  $\rho_{\text{aff}}(X, Y) = 1 \Leftrightarrow Y = T_{\text{aff}}X$ .

Bouniot, Redko, Mallasto, Laclau, Arndt, Struckmeier, Heinonen, Kyrki, Kaski

#### **ReLU** example

Affinity scores throughout the input domain of ReLU



► Affinity scores will vary depending on the input domain considered.

 $\blacktriangleright\,$  For ReLU, high  $\rho_{\rm aff}$  values in the linear part of the transformation.

#### Non-linearity signature



#### **Throughout DNNs Architectures**



- ► Affinity scores statistics and Accuracy (in red) throughout DNNs architectures.
- ▶ Before ViTs: max and median values are increasing, also gap between min and max.
- Within ViTs: Trend of decreasing min values

#### Take-Home Message

# From Alexnet to Transformers: Measuring the Non-linearity of Deep Neural Networks with Affine Optimal ${\sf Transport}^1$

- First theoretical sound tool to measure non-linearity in DNNs
- ✓ Different developments in Deep Learning can be understood through the prism of non-linearity
- ✓ Variety of potential applications

<sup>&</sup>lt;sup>1</sup>Quentin Bouniot et al. "From Alexnet to Transformers: Measuring the Non-linearity of Deep Neural Networks with Affine Optimal Transport". In: arXiv preprint arXiv:2310.11439 (2023).

## Thank you for listening !

### Do not hesitate to contact us if you have questions.

[1] Quentin Bouniot et al. "From Alexnet to Transformers: Measuring the Non-linearity of Deep Neural Networks with Affine Optimal Transport". In: *arXiv preprint arXiv:2310.11439* (2023).