# A (MORE) GENERAL FRAMEWORK FOR EQUIVARIANT NEURAL NETWORKS

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### GOAL: WE WANT TO (MATHEMATICALLY) UNIFY ML MODELS IN ORDER TO SHOW SIMILARITIES AND DIFFERENCES BETWEEN THEM, AND POTENTIALLY CHARACTERISE THESE FEATURES USING SYMMETRY ARGUMENTS

### EXAMPLE: CONVOLUTIONAL LAYERS = LINEARITY + TRANSLATIONAL SYMMETRY

#### **OUTLINE:**



#### **REPRESENTING DATA AND MODEL**

TRANSFORMING DATA	EQUIVARIANT MODEL	TRANSFORMING DATA
$\mathcal{F} \ni f: X \to \mathbb{R}^n$	$\Phi_\omega:\mathcal{F} o\mathcal{F}$	$\mathcal{F} \ni f : X \to \mathbb{R}^n$

W/ GROUP ACTION  $(g,f)\mapsto g\cdot f$ 

SATISFYING EQUIVARIANCE  $\Phi[g \cdot f] = g \cdot \Phi[f]$ 

W/ GROUP ACTION  $(g,f)\mapsto g\cdot f$ 

**EQUIVARIANT ARCHITECTURES** 



#### ANSATZ' FOR OTHER EQUIVARIANT MODELS

$$\begin{split} & \overset{\mathbf{G}-\mathsf{CNN}}{\longrightarrow} \quad \overset{\mathbf{G}-\mathsf{CNN}}{\longrightarrow} \quad \overset{\mathbf{G}}{\longrightarrow} \quad \overset{\mathbf{EQUIVARIANT IF}}{\longrightarrow} \quad \overset{\mathbf{F}(g,g') \in \mathbb{R}^{n \times m}}{\longrightarrow} \quad \overset{\mathbf{F}(g,g') \in \mathbb{R}^{n \times m}} \quad \overset{\mathbf{F}(g,g') = \kappa(g^{-1}g')}{\longleftarrow} \quad \overset{\mathbf{F}(g,g') =$$

[1] Cohen, Taco, and Welling, Max. "Group Equivariant Convolutional Neural Networks", International Conference on Machine Learning, 2016.
[2] Hutchinson, Michael J., Charline Le Lan, Sheheryar Zaidi, Emilien Dupont, Yee Whye Teh and Hyunjik Kim. "LieTransformer: Equivariant self-attention for Lie Groups." International Conference on Machine Learning, 2020.



#### **EQUIVARIANCE CONDITION**



## $\omega(f, g, g')$ is a function of $f(g), f(g'), g^{-1}g' \dots$

CONCLUSION 1: A TRANSFORMER-LIKE DEPENDENCE ARISES NATURALLY FROM EQUIVARIANCE, WITH RELATIVE POSITIONAL ENCODING ON THE TRANSFORMER SIDE DIRECTLY CORRESPONDING TO THE KERNEL OF CONVOLUTIONAL NETWORKS.

#### **RELATION TO OTHER ARCHITECTURES**

$$\Phi_{\omega}[f](g) = \int_{G} \omega(f, g, g') f(g') dg'$$



CONCLUSION 2: THE PROPOSED FRAMEWORK IS FLEXIBLE, GENERALISES NOT ONLY WELL-KNOWN PRE-EXISTING EQUIVARIANT ARCHITECTURES BUT CAN IN FACT DESCRIBE ANY EQUIVARIANT LAYER/OPERATOR.

## $\omega(f,g,g')$ - form a space of sections of some equivariant bundle

- IS AN INDUCED REPRESENTATION OF SOME SUBRGOUP H OF G

# - IS ASSOCIATED TO AN EQUIVALENCE CLASS OF MAPS WHICH ALL RESULT IN THE SAME OPERATOR

CONCLUSION 3: WE HAVE SEVERAL WAYS OF CHARACTERISING THE SPACE OF OPERATORS. ANALYSIS THESE SPACES FURTHER MIGHT GIVE US INSIGHTS IN HOW LARGE THIS SPACE IS, ITS TOPOLOGY ETC. CONCLUSION 1: A TRANSFORMER-LIKE DEPENDENCE ARISES NATURALLY FROM EQUIVARIANCE, WITH RELATIVE POSITIONAL ENCODING ON THE TRANSFORMER SIDE DIRECTLY CORRESPONDING TO THE KERNEL OF CONVOLUTIONAL NETWORKS.

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## **THANK YOU!**

## **QUESTIONS?**