

# TetraSphere

## A Neural Descriptor for O(3)-Invariant Point Cloud Analysis

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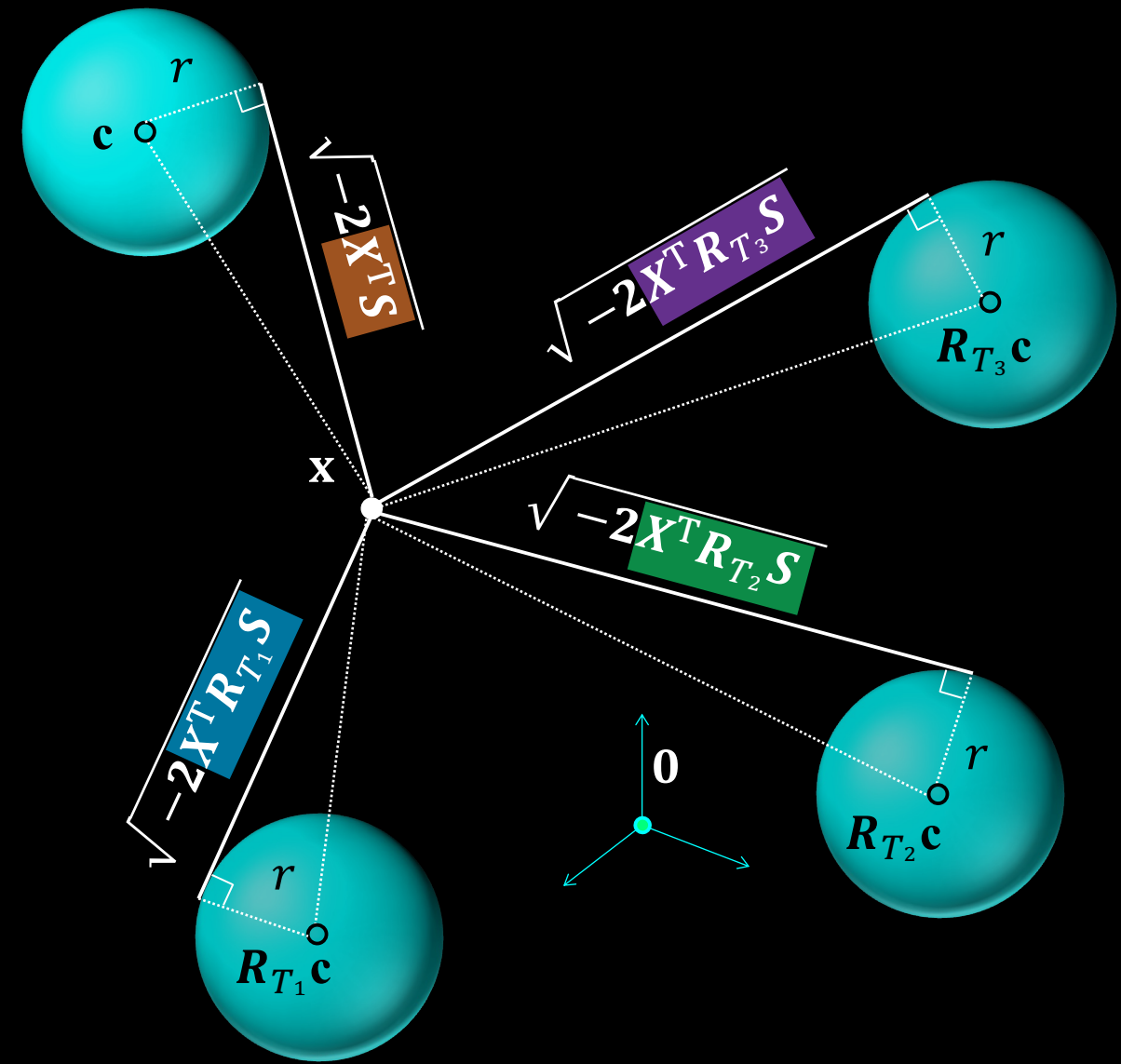


### Motivation

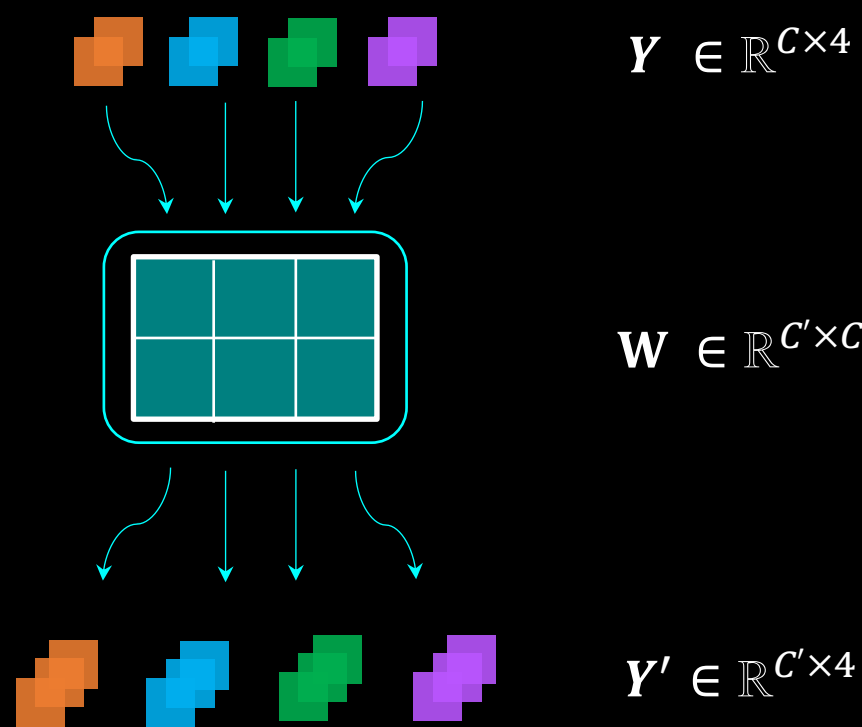
- Geometric learning on 3D point clouds with NNs
- Permutation and O(3)-equivariance  $F(RX) = \rho(R) F(X)$
- Requirement for natural science applications
- Invariant predictions if desired  $F(RX) = F(X)$
- Eliminates the need for data augmentation

### Background

- Two sorts of O(n)-equivariant neurons
- Steerable 3D spherical neurons [Melnyk et al., ICML'22] never used for end-to-end learning previously



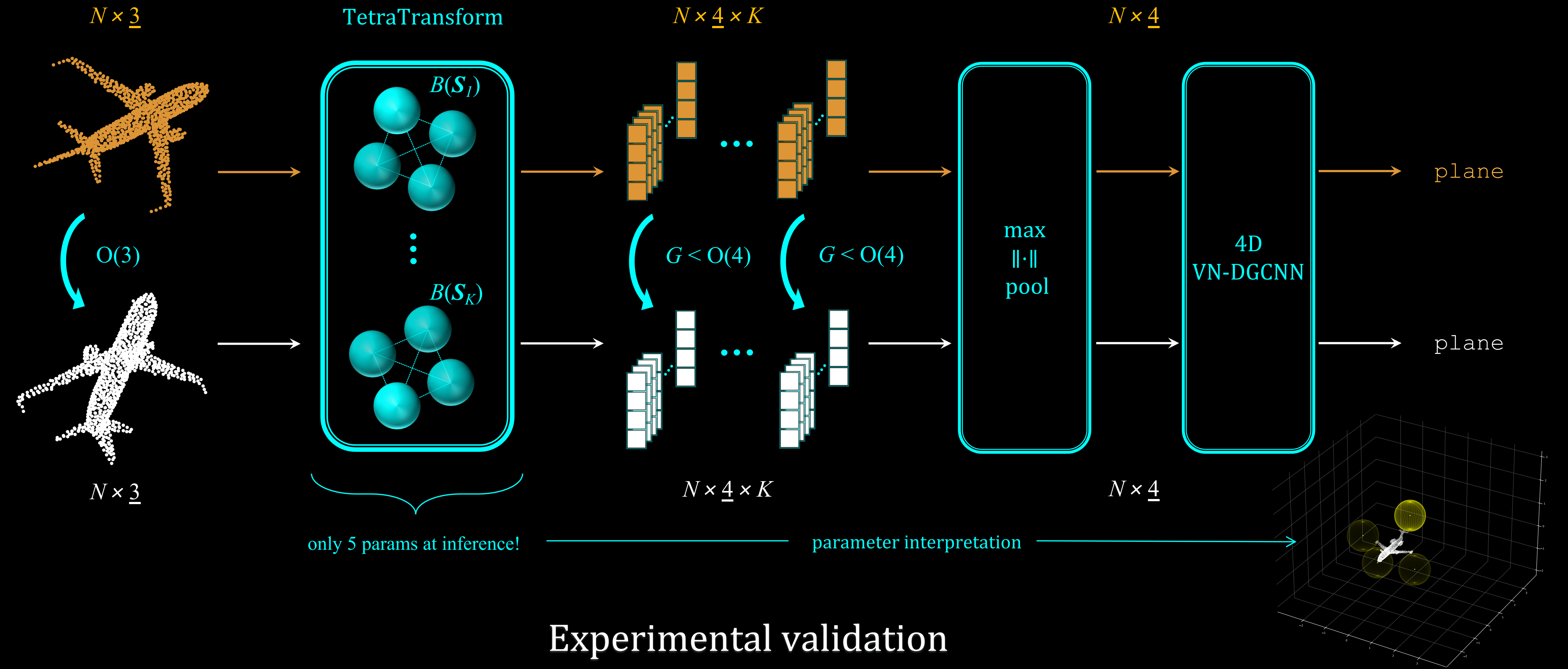
$$\begin{aligned} X &= (x, -1, -\frac{1}{2}\|x\|^2) && \in \mathbb{R}^5 \\ S &= (c, \frac{1}{2}(\|c\|^2 - r^2), 1) && \in \mathbb{R}^5 \\ RS &= (Rc, \frac{1}{2}(\|c\|^2 - r^2), 1) && \in \mathbb{R}^5 \\ B(S)X &= [\text{orange}, \text{blue}, \text{green}, \text{purple}]^T && \in \mathbb{R}^4 \end{aligned}$$



- Vector neurons [Deng et al., ICCV'21]

### Method

- TetraSphere := steerable 3D spherical neurons + equivariant pooling + 4D VN-DGCNN



### Experimental validation

- SOTA classification results
- Randomly rotated **ScanObjectNN** real-world scans
- $\Delta=0.7-1.3\%$  better than the previous best
- Decent on synthetic **ModelNet40** and **ShapeNet** (segmentation)
- TetraSphere > 4D VN-DGCNN > 3D VN-DGCNN

