



Point MixSwap: Attentional Point Cloud Mixing via Swapping Matched Structural Division

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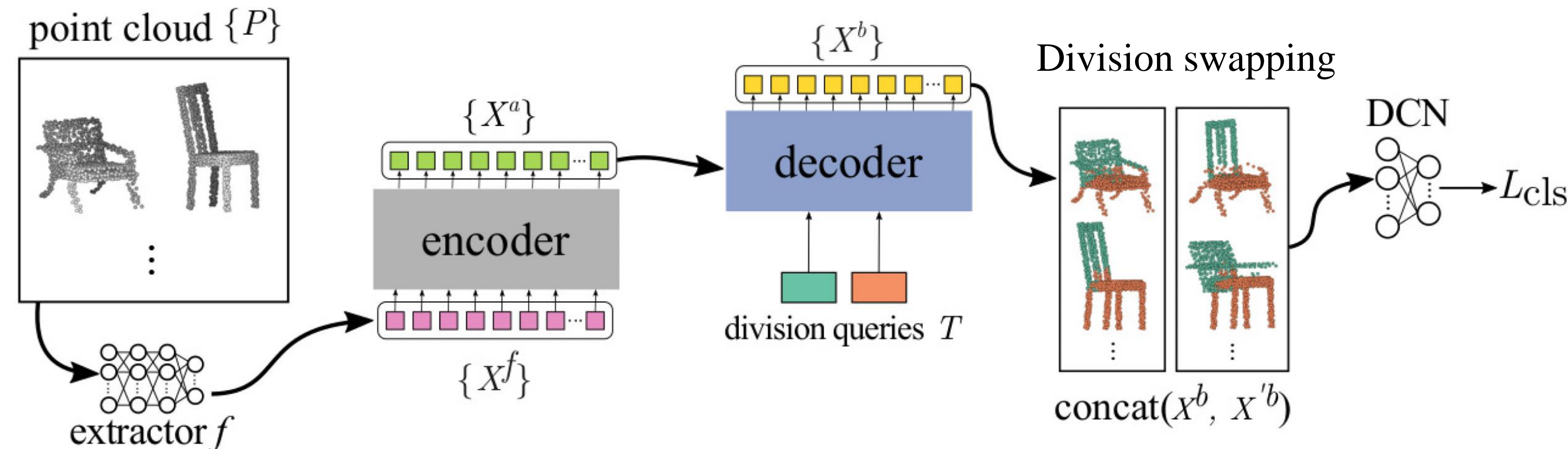
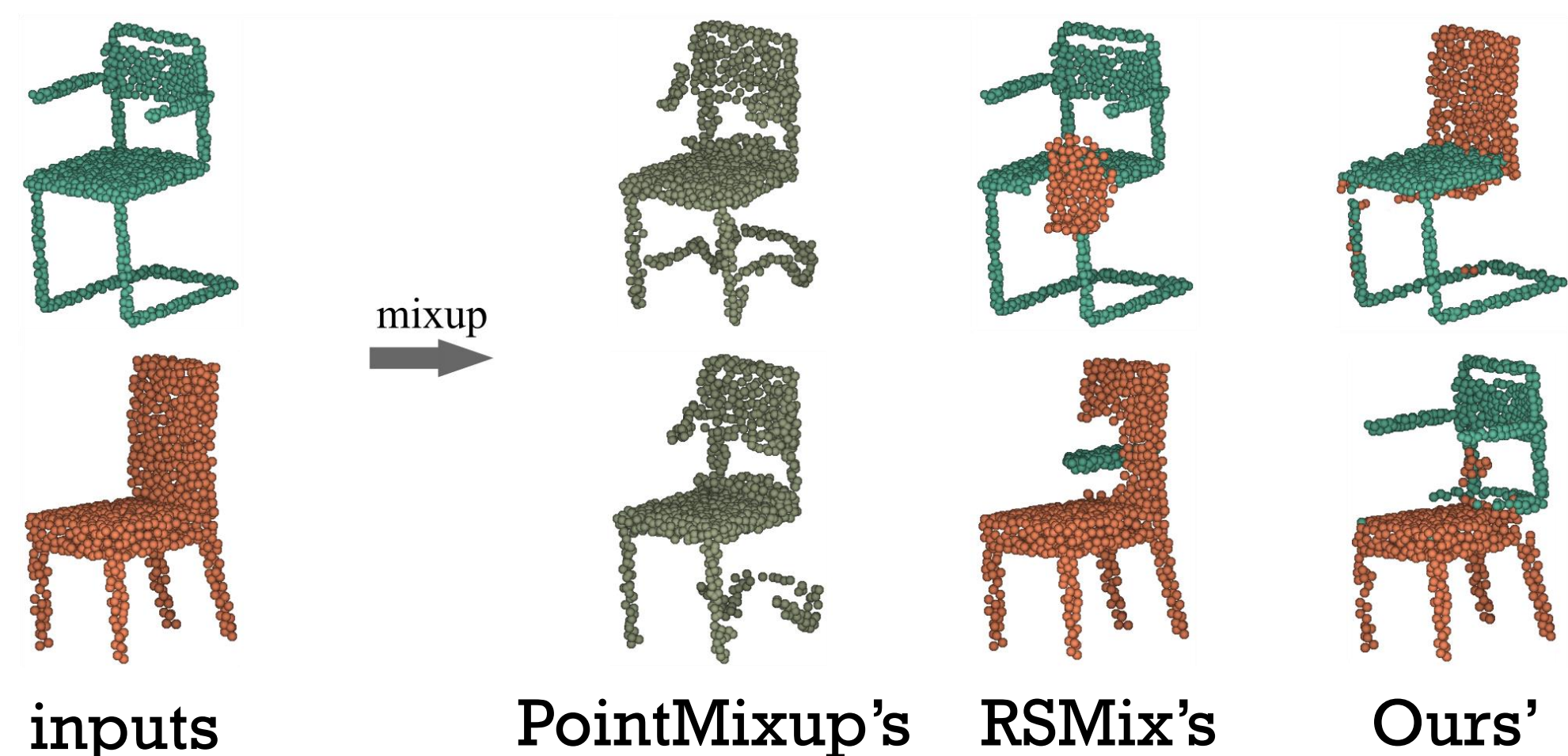


Contributions:

- Introduce an effective technique for synthesizing diverse and realistic point clouds by **swapping matched structural division**, without any part-segmentation labels
 - Develop a novel encoder-decoder structure to decompose point cloud with cross-cloud correspondence
- Mix the point clouds to generate augmented data utilizing division queries, leading to significant improvement on point cloud classification task
 - Augment the point clouds by swapping matched divisions cross different clouds
- Evaluated on both **synthetic** and **real-world** datasets
 - ModelNet10, ModelNet40, ScanObjectNN
 - Experiment with all data (100%) and reduced data (20%)

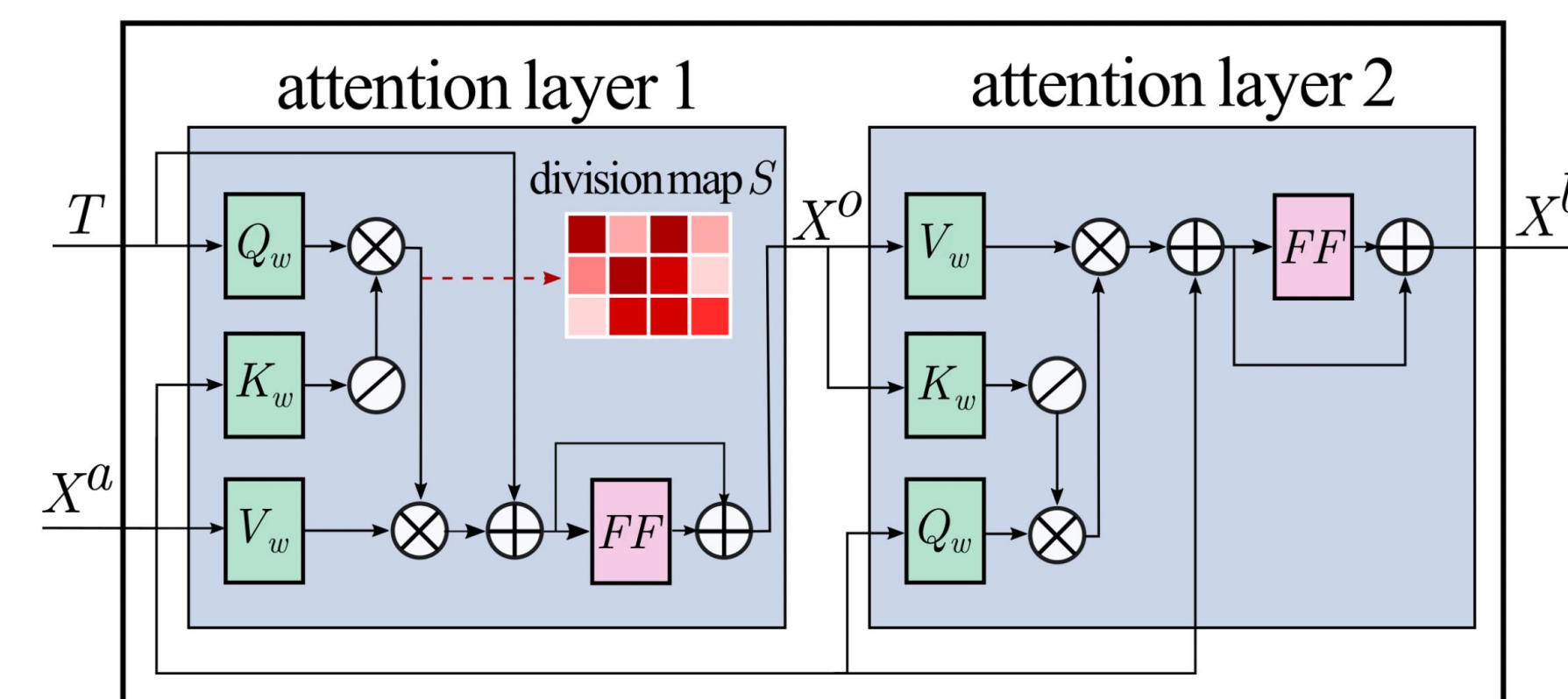
Observations:

- Existing point cloud mixup methods does not consider the **semantic structural information** while performing mixup:
 - We generalize the encoder-decoder model to explores **inter-cloud division correspondence** for division swapping
 - Through the learned division queries, we can divide the point cloud into **R divisions** and synthesize the new point clouds



Designed encoder-decoder architecture:

- The learnable division queries jointly decompose point cloud into R disjoint subsets, for further division swapping
- The query and key-value pairs will be switched in the second layer and jointly produce point-specific features



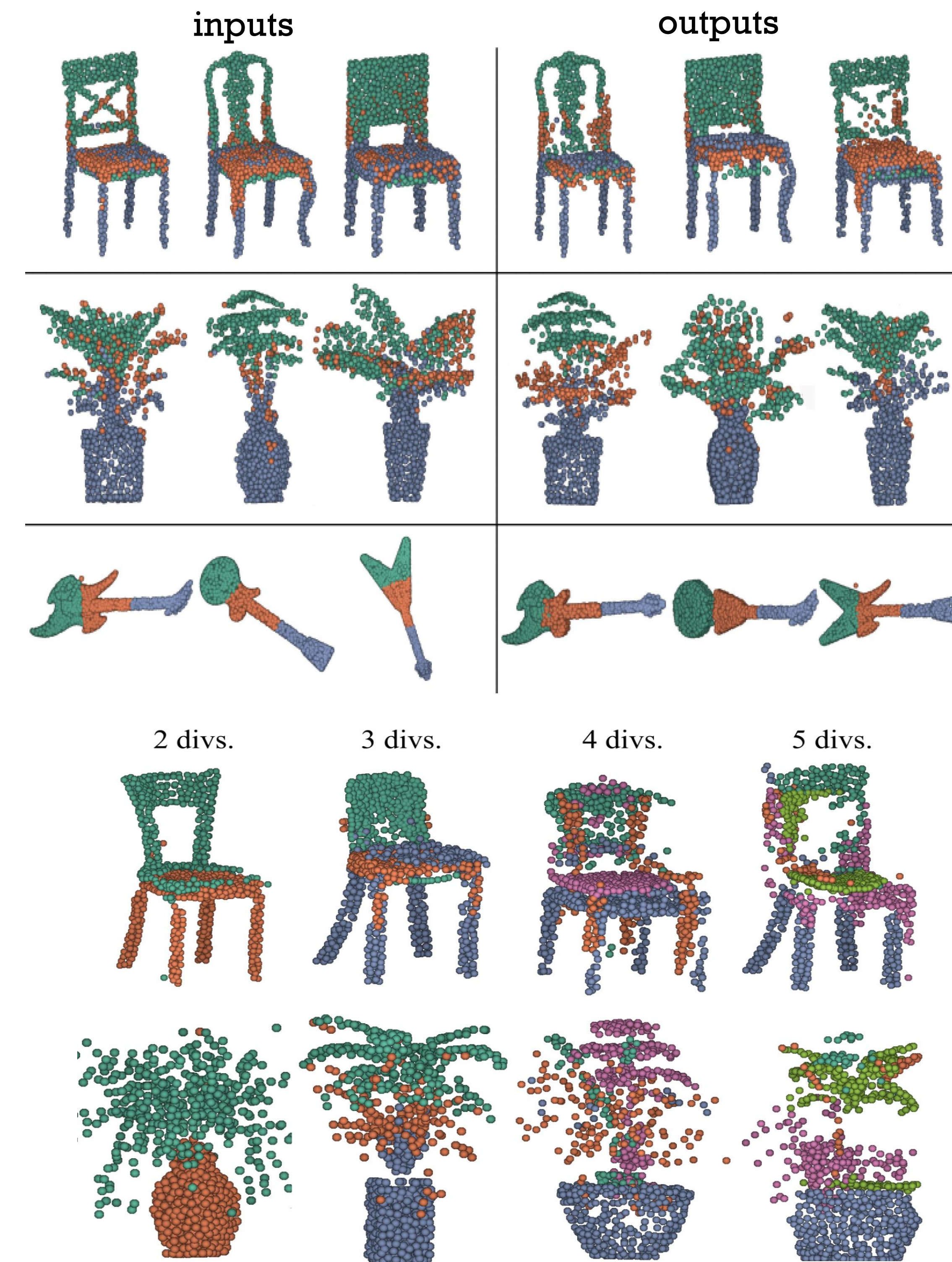
Quantitative results:

- Different SoTA methods and training data usage are compared

Method	Rate 20%		Rate 100%	
	M40	M10	M40	M10
DGCNN	87.5	93.2	92.6	94.8
DGCNN + PointMixup [3]	89.0	93.8	93.1	95.1
DGCNN + PointAugment [12]	88.6	92.8	93.4	95.2
DGCNN + RSMix [11]	90.1	93.7	93.5	95.9
DGCNN + PointWOLF [9]	89.3	93.5	93.2	95.1
DGCNN + Ours	91.3	94.6	93.5	96.0

Divisions	Level	M40	M10	SON
2	Input	91.0	94.6	75.9
	Feature	91.1	94.7	76.2
3	Input	91.2	94.5	76.1
	Feature	91.3	94.6	76.3
4	Input	91.0	94.4	75.7
	Feature	91.2	94.6	76.1
5	Input	91.0	94.3	75.5
	Feature	91.2	94.6	76.0

Visualization:



Source code available:

