



Understanding Imbalanced Semantic Segmentation Through Neural Collapse

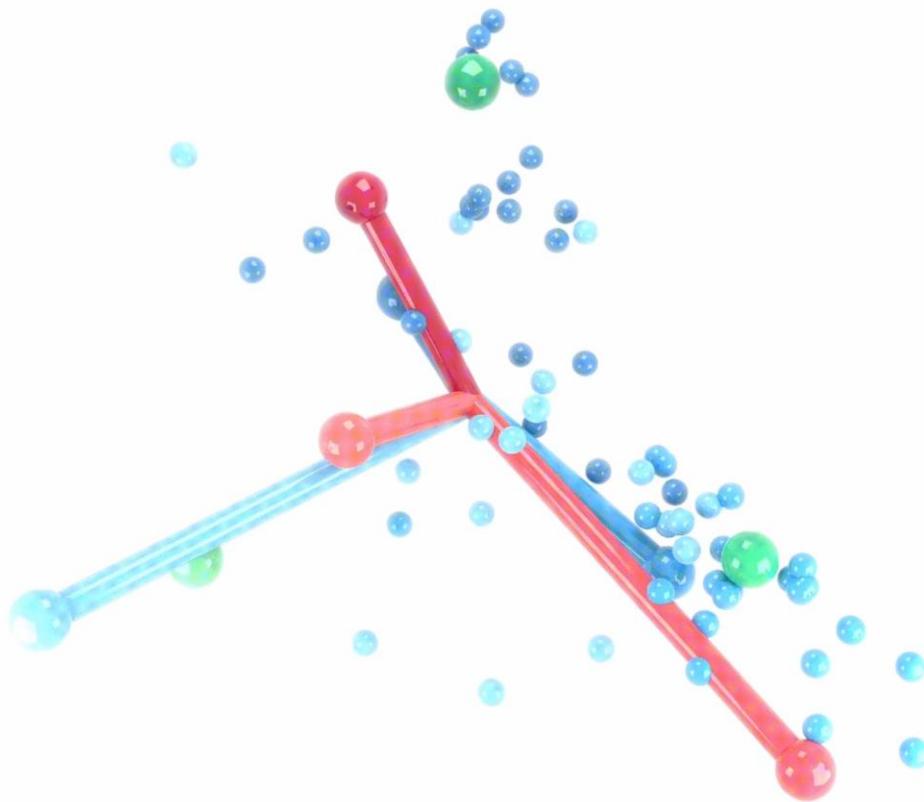
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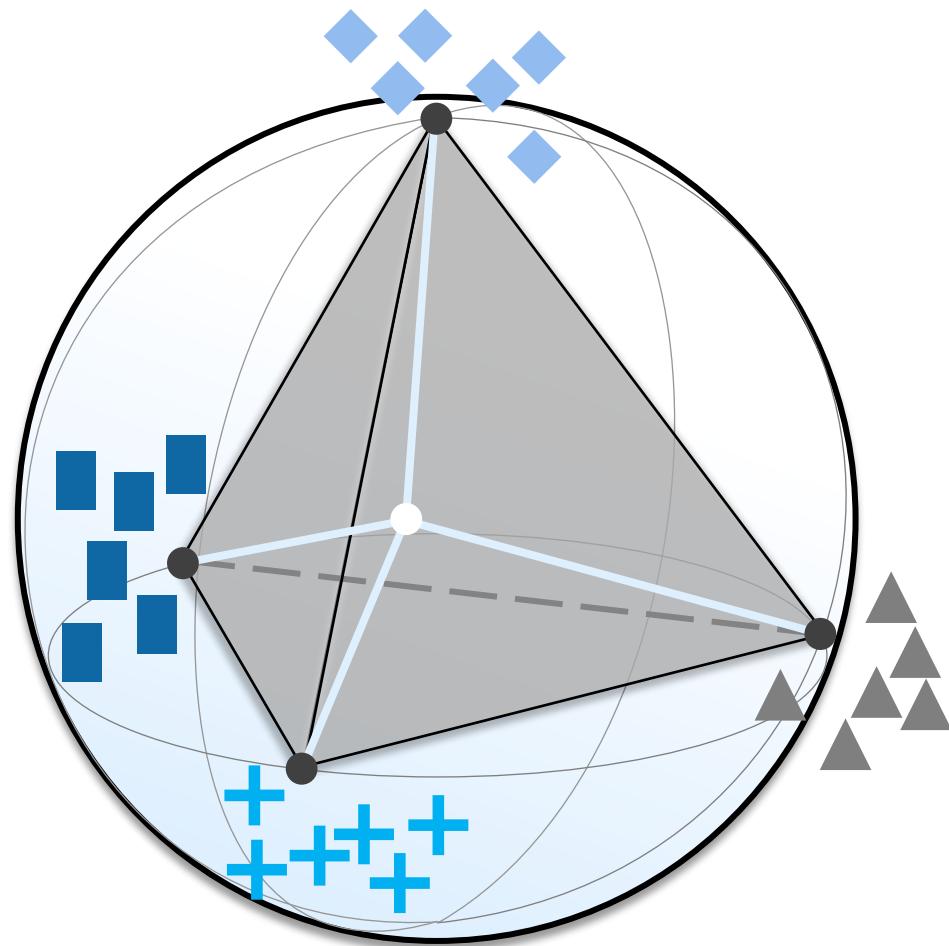


Red ball-and-sticks: Linear classifiers
Blue ball-and-sticks: Class-means (centers)
Small blue spheres: Last-layer features
Green spheres: Simplex equiangular tight frame (ETF)

Neural Collapse: As training progresses

Linear classifiers
Class-means All collapse to **Simplex ETF**
Last-layer features

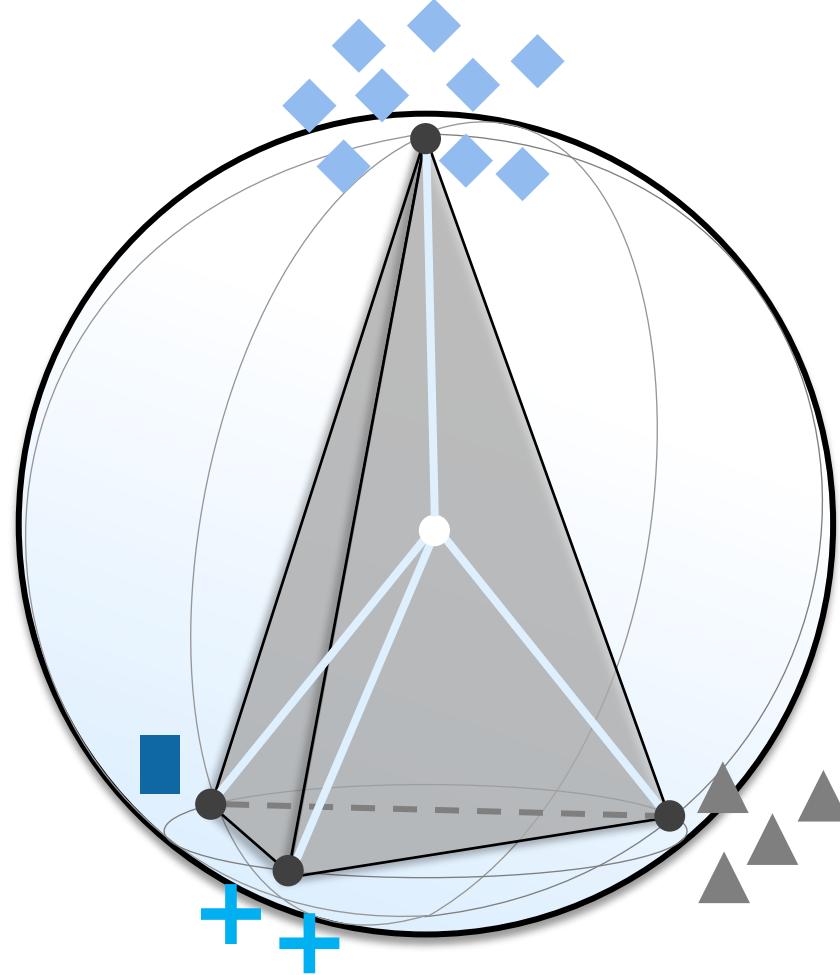
Classification



normally **balanced distribution** &
relatively unique among instances

→ easier to reach the **equiangular** and **maximal separated** structure (ETF) of neural collapse

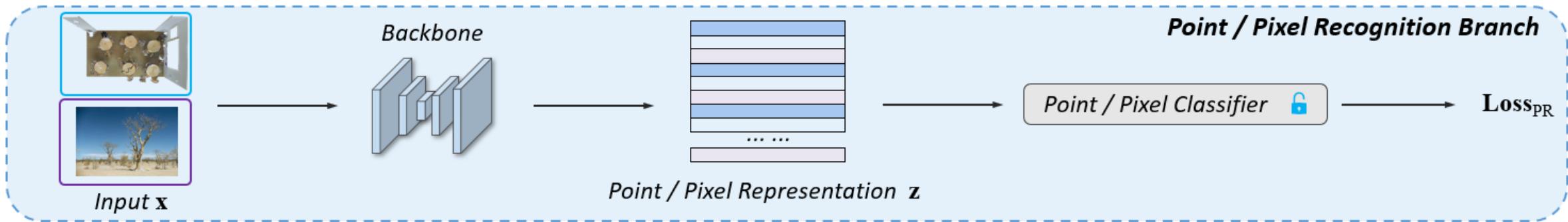
Semantic Segmentation



naturally brings **contextual correlation** &
imbalanced distribution (point/pixel)

⇒ break the **equiangular** and **maximal separated** structure (ETF) of neural collapse

Conventional point cloud & image semantic segmentation framework



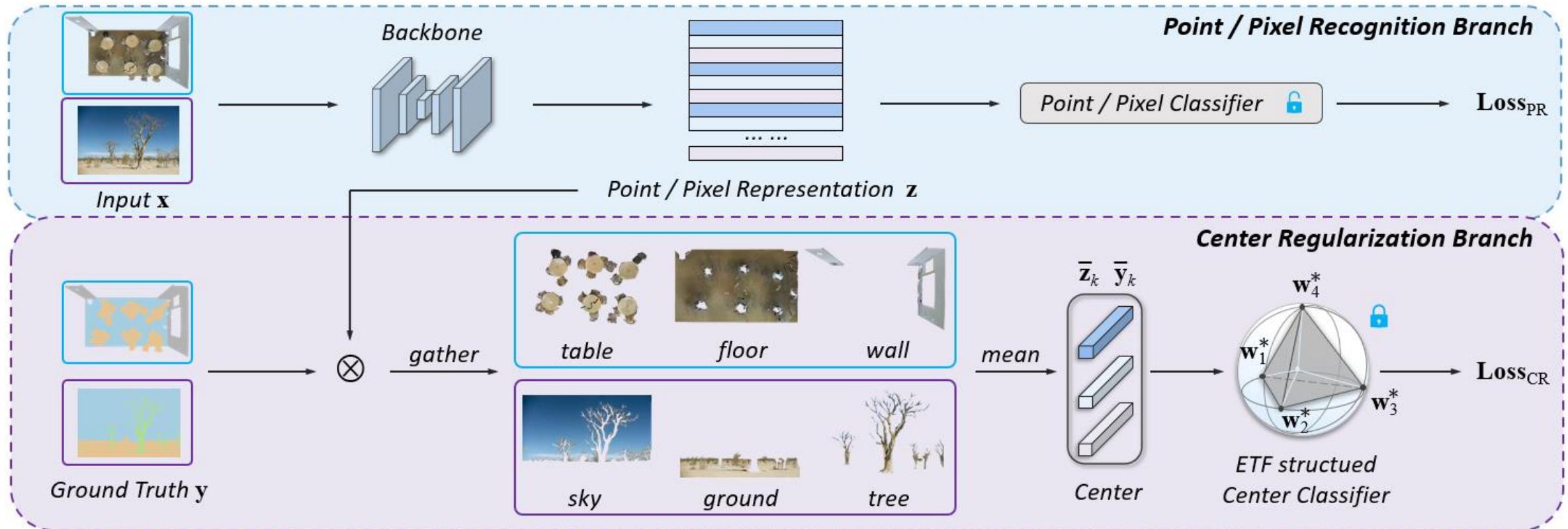
Feature centers and classifiers converge to a *non-equianangular* and *non-maximal separated* structure

However, for imbalanced semantic segmentation:

This asymmetric structure is detrimental to discrimination among the tailed/minor classes.



CeCo (Ours)



We introduce a **center collapse (CeCo) loss** to regularize the feature centers to close the **equiangular and maximal separated ETF structure**.

ScanNet200 3D Semantic Label Benchmark

This table lists the benchmark results for the ScanNet200 3D semantic label scenario.

Column Sorting: Column Sort Alphabetically ▾

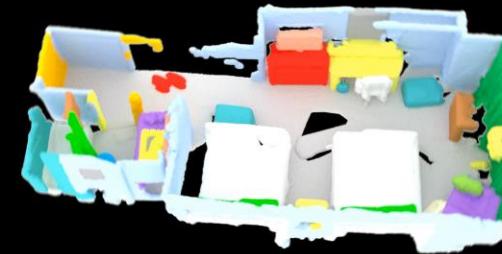
Method	Info	avg iou	head iou	common iou	tail iou	alarm clock	armchair	backpack	bag	ball	bar	basket	bathroom cabinet	bathroo
CeCo	: Understanding Imbalanced Semantic Segmentation Through Neural Collapse.	0.340 1	0.551 1	0.247 1	0.181 1			0.475 2	0.057 4	0.142 3	0.000 1	0.000 1	0.000 1	
Rank 1st														
LGround	David Rozenberszki, Or Litany, Angela Dai: Language-Grounded Indoor 3D Semantic Segmentation in the Wild. arXiv	0.272 2	0.485 2	0.184 2	0.106 2			0.476 1	0.077 2	0.218 1	0.000 1	0.000 1	0.000 1	
Minkowski 34D	C. Choy, J. Gwak, S. Savarese: 4D Spatio-Temporal ConvNets: Minkowski Convolutional Neural Networks. CVPR 2019	0.253 3	0.463 3	0.154 4	0.102 3			0.381 4	0.084 1	0.134 4	0.000 1	0.000 1	0.000 1	
CSC-Pretrain	Ji Hou, Benjamin Graham, Matthias Nießner, Saining Xie: Exploring Data-Efficient 3D Scene Understanding with Contrastive Scene Contexts. CVPR 2021	0.249 4	0.455 4	0.171 3	0.079 4			0.418 3	0.059 3	0.186 2	0.000 1	0.000 1	0.000 1	

Improve a lot (+6.8% mIoU)

Org.



GT



CE



CeCo



Org.



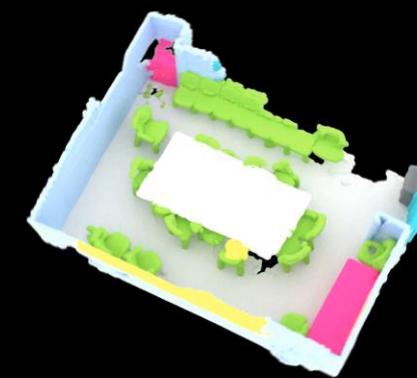
GT



CE



CeCo



2D Visualization Example:



2D Visualization Example:





Thanks for your watching!

For more details, please visit:

Our paper



Our code

