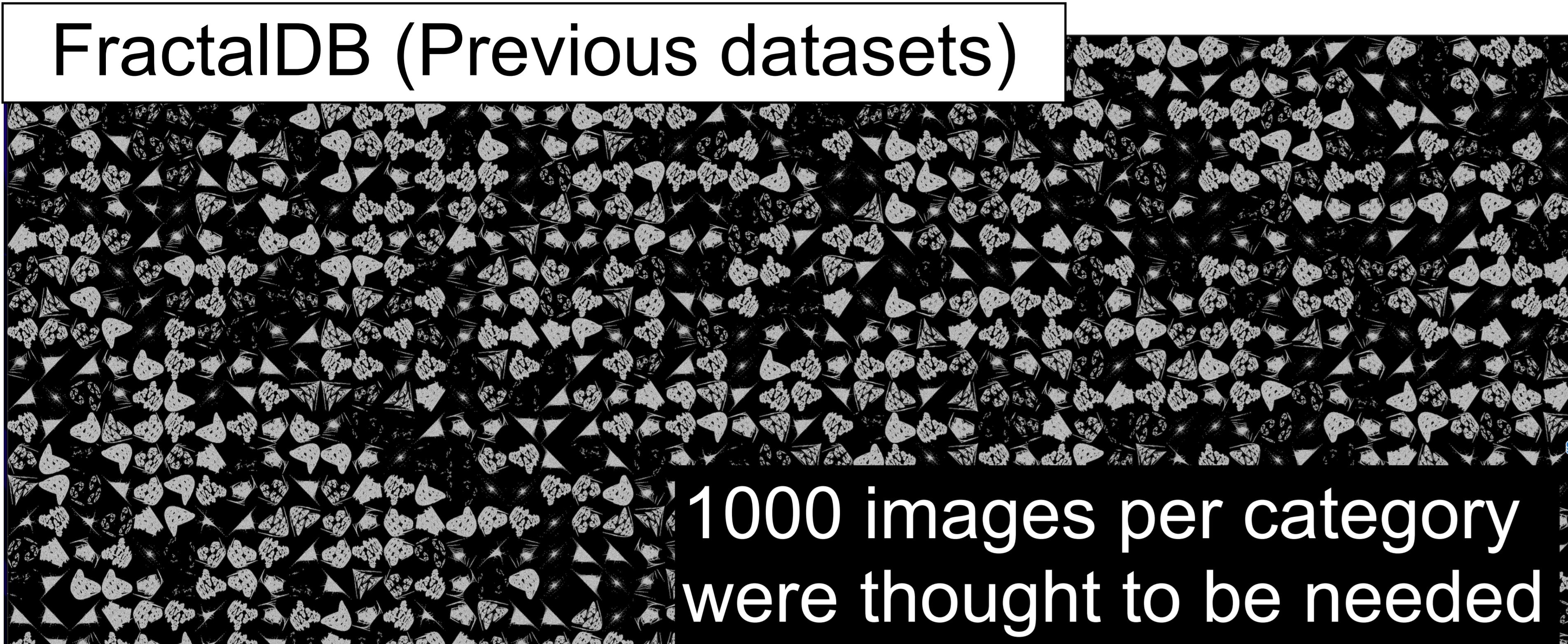


Vision Transformer (ViT) can be pre-trained with 1k synthesized images!

Number of Images required for ViT pre-training :

JFT (300M), ImageNet-1k/21k (1.28M / 14M), FractalDB-1k/21k (1M/21M), OFDB-1k/21k (1k/21k)

FractalDB (Previous datasets)



Contribution

Same level of accuracy

GPU hours 78% faster

High performance with limited data pre-training
(For ViT)

2D-OFDB



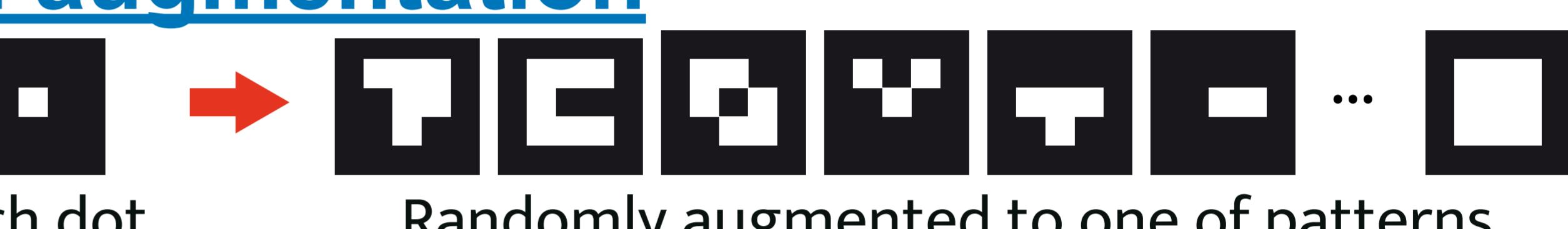
-Proposal methods

Why can pre-train with 1 image per category? Update

FractalDB generates 1000 images from a single fractal image ↓

Minibatch data augmentation to perform essentially the same conversion

Pattern augmentation



Experiment : Verification performance of Data augmentation

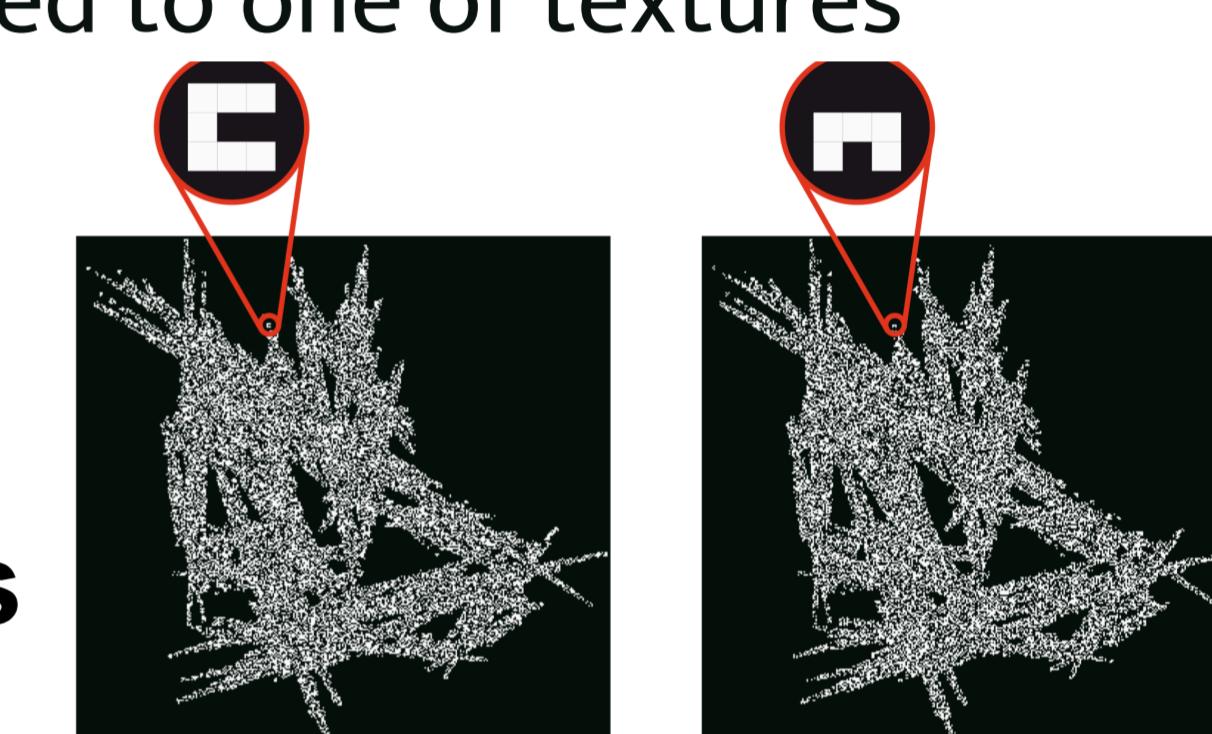
Texture augmentation



IFS's Parameter variation ×25



Possible to generate a variety Of Fractals with similar shapes



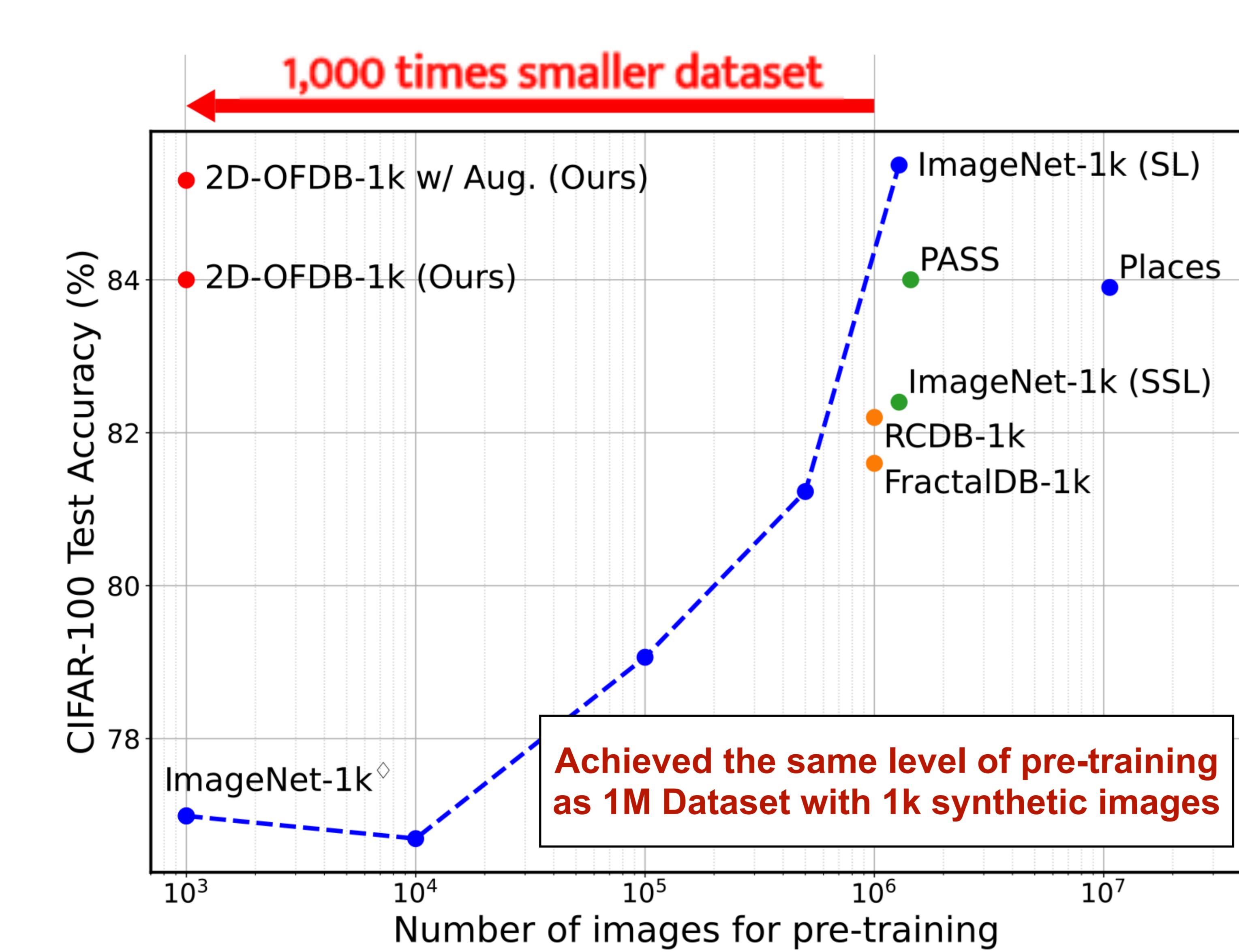
The same accuracy can be achieved with a single image, and the patch augmentation can be further improved.

Experiment

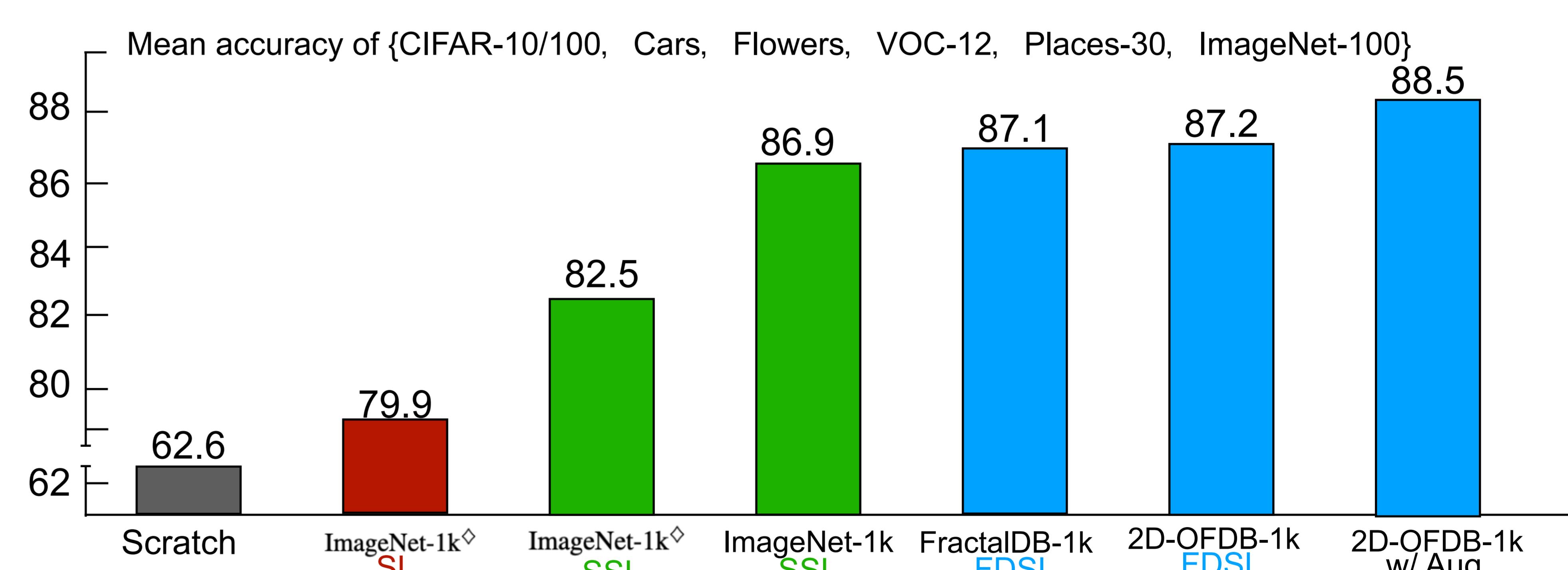
Experimental setting

Verify pre-training effectiveness of datasets : Transition training using pre-trained network weights as initial values

Evaluated by test performance on the dataset for transfer learning, data augmentation use DeiT setting.



Comparison of pre-training effect with 1k category dataset



Accuracy close to 1,000,000 images despite 1,000 images of data

Comparison of pre-training effectiveness with 21k category data sets

Pre-training	#Img	Type	ViT-T	ViT-B	GPU hours	Batch	#Iterations
Scratch	–	–	72.6	79.8	–	–	–
ImageNet-21k	14M	SL	74.1	81.8	3,657	8,192	300k
FractalDB-21k	21M	FDSL	73.0	81.8	5,120	8,192	300k
ImageNet-21k [△]	21k	SL	71.0	81.1	1,132	1,024	300k
2D-OFDB-21k	21k	FDSL	73.8	82.2	1,088	1,024	300k

Higher accuracy with smaller amount of data than IDMM with synthetic images

Scaling Data Improves Accuracy ViT-B exceeds ImageNet-21k Pre-training time reduced by 78%

Enables ViT pre-training for anyone with limited data and computational resources