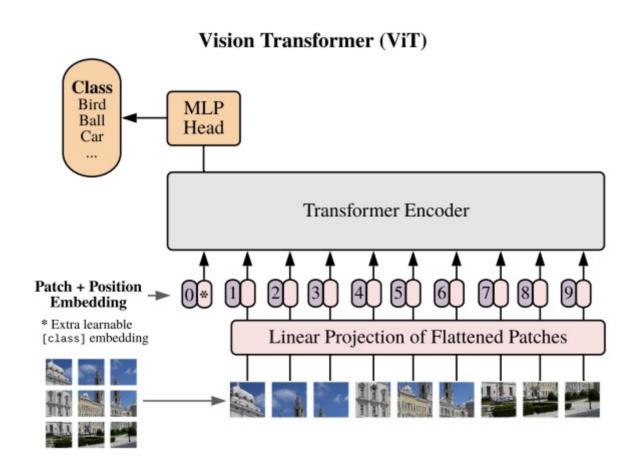
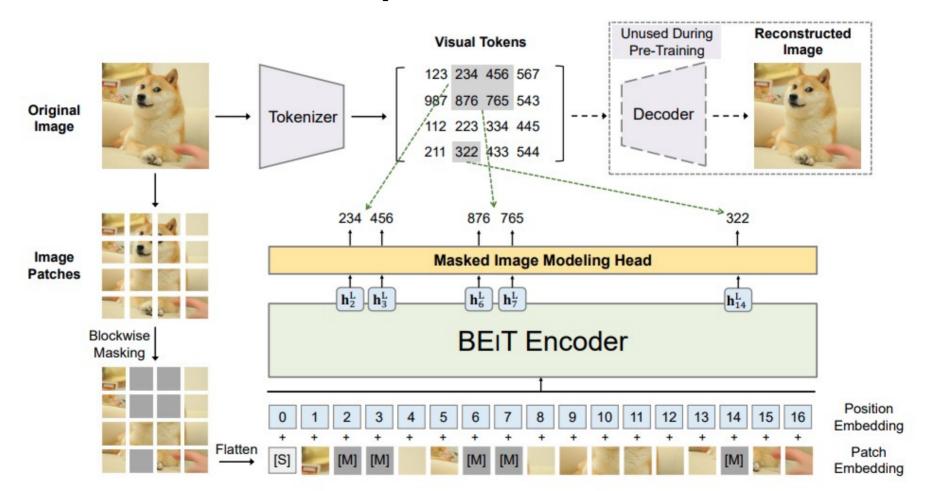
Are Large-scale Datasets Necessary for Self-Supervised Pretraining?

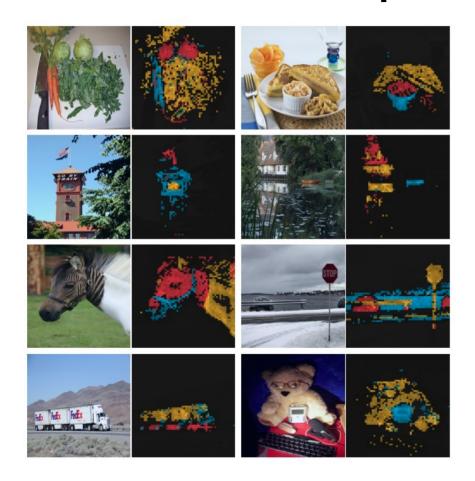
Prerequisites: Vision Transformer

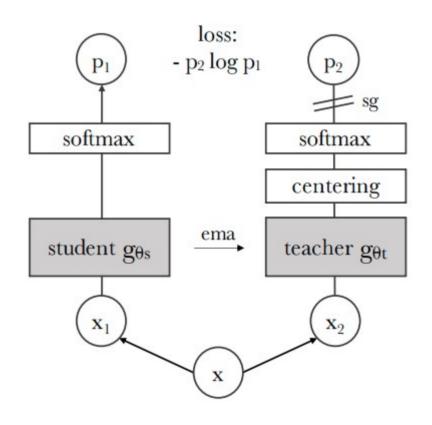


Prerequisites: BEiT



Prerequisites: DINO





: mask token : input tokens -X-: encoder output : decoded masks masked shallow deep ViT image ViT modelling ! encoder decoder loss InfoNCE average pooling loss masked shallow deep ViT image ViT encoder modelling decoder loss

SplitMask

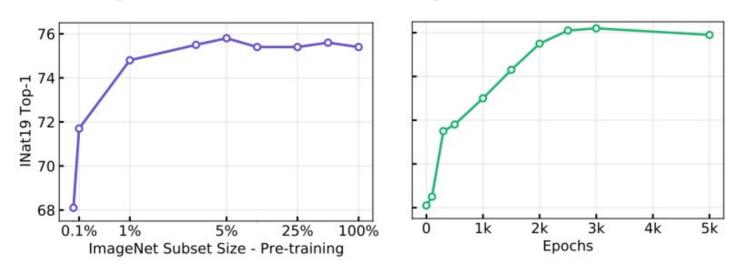
$$\ell(\mathbf{x}_a) = \frac{\exp(\mathbf{x}_a^{\top} \mathbf{x}_b / \tau)}{\sum_{\mathbf{y} \in \{\mathbf{x}_b\} \cup \mathcal{N}} \exp(\mathbf{x}_a^{\top} \mathbf{y} / \tau)},$$

SplitMask: Tokenizers

Table 2. Ablation study on the effect of different tokenization methods. We compare the DALL-E tokenizer originally used in BEiT with patch level techniques: random projection, random patches and k-means clustering. We observe that the DALL-E tokenizer can be effectively replaced by simpler methods that do not require training on a large dataset.

	DALL-E	Rand. Proj.	Rand. Patches	K-Means		
iNat19	75.2	75.2	75.3	75.0		

SplitMask: Expreriments



Method	IMNet 1% epochs: 30k	IMNet 10% epochs: 3k	IMNet Full epochs: 300	COCO epochs: 3k		
Supervised	71.6	75.0	75.8	_		
DINO [18]	70.1	73.1	78.4	71.9		
BEiT [24]	74.1	74.5	75.2	74.4		
SplitMask	74.8	75.4	75.4	76.3		

COCO detection and instance segmentation performance

Method	Backbone	Pre	AP^b	AP^b_{50}	AP^b_{75}	AP^m	AP_{50}^m	AP^m_{75}		
		Supervised	IMNet	COCO		60.73				
Random Initialization		X	X	X	38.3	60.1	41.4	35.6	57.1	37.7
Random Initialization†		X	X	X	42.8	64.5	45.6	39.1	61.5	41.7
DeiT [50]	ViT-S	/	/	X	44.2	66.6	47.9	40.1	63.2	42.7
BEiT [24]	V11-3	X	/	X	44.5	66.2	48.8	40.3	63.2	43.1
DINO [18]		X	X	/	43.7	65.5	47.7	39.6	62.3	42.3
BEiT		X	X	/	44.7	66.3	48.8	40.2	63.1	43.2
SplitMask		X	X	1	45.3	66.9	49.4	40.6	63.6	43.5
Random Initialization		X	X	X	40.7	62.7	44.2	37.1	59.1	39.4
Random Initialization†		X	X	X	43.0	64.2	46.9	38.8	61.3	41.6
DeiT [50]	VETD	/	1	X	45.5	67.9	49.2	41.0	64.6	43.8
BEiT [24]	ViT-B	X	/	X	46.3	67.6	50.6	41.6	64.5	44.9
DINO [18]		X	X	/	43.1	64.4	46.9	38.9	61.4	41.4
BEiT		X	X	1	46.7	67.7	51.2	41.8	65.0	44.6
SplitMask		X	X	1	46.8	67.9	51.5	42.1	65.3	45.1

Pretraining on target dataset

Method	Backbone	Supervised	Data	Used	iNat-18	iNat-19	Food 101	Cars	Clipart	Painting	Sketch
		pre-training	IMNet	Target	437k	265k	75k	8k	34k	52k	49k
Liu et al. [67] [∓]	CVT-13	X	X	/	_		_		60.6	55.2	57.6
	ResNet-50	X	X	1	-	-	-	-	63.9	53.5	59.6
Random Init.		Х	X	/	59.6	67.5	84.7	35.3	41.0	38.4	37.2
DeiT [50]		/	/	1	69.9	75.8	91.5	92.2	79.6	74.2	72.5
BEiT [24]	ViT-S	X	1	1	68.1	75.2	90.5	92.4	75.3	68.7	68.5
BEiT		X	X	/	68.8	76.1	90.7	92.7	-	69.0	-
SplitMask		X	X	/	70.1	76.3	91.5	92.8	<u>78.3</u>	<u>69.2</u>	70.7
Random Init.		Х	X	/	59.6	68.1	83.3	36.9	41.9	37.6	34.9
DeiT [50]		/	1	/	73.2	77.7	91.9	92.1	80.0	73.8	72.6
BEiT [24]	ViT-B	X	1	1	71.6	78.6	91.0	93.9	78.0	71.5	71.4
BEiT		X	X	/	72.4	79.3	91.7	92.7	-	70.7	-
SplitMask		X	X	/	74.6	80.4	91.2	93.1	<u>79.3</u>	<u>72.0</u>	<u>72.1</u>