Hyperparameters

Type	Hyperparameters	Values
Model	Layers	6
	Hidden size	768
	FFN inner hidden size	3072
	Attention Heads	12
	Patch size	16×16
	Input resolution	224×224
Training	Epochs	10
	Total steps	50040
	Batch size	256
	Optimizer	AdamW
	AdamW ε	1e-06
	AdamW β	(0.9,0.98)
	Weight decay	0.01
	Base learning rate	1e-4
	Learning rate schedule	Cosine
	Warmup steps	5004 (10% of total steps)
	Hardware	1 × RTX 4090 24GB
Augmentations	Horizontal flipping prob.	0.5
	RandomResizeCrop range	[0.08, 1.0]

Table 1: Hyperparameters used for distilling a Data2Vec2 image model.

Type	Hyperparameters	ImageNet		CIFAR10		CIFAR100	
		Finetune	Linear probe	Finetune	Linear probe	Finetune	Linear probe
Training	Epochs	15					
	Batch size	256					
	Optimizer	AdamW					
	AdamW ε	1e-8					
	AdamW β	(0.9, 0.999)					
	Weight decay	0.01					
	Base learning rate	1e-3					
	Layer Decay	0.81					
	Learning rate schedule	Cosine					
	Warmup steps	10% of total steps					
	Hardware	1 × RTX 4090 24GB					
Mixup [1]/**Cutmix** [2]	Mixup prob.	0.8					
	Cutmix prob.	1.0					
	Prob.	0.9					
	Switch prob.	0.5					
	Label smooting	0.1					
RandAugment [3]	Magintude	9					
	Magnitude std.	0.5					
	Magnitude inc.	1					
	# ops	2					
RandomErase [4]	Prob.	0.25					
	Mode	pixel					
	# erase	1					

Table 2: Hyperparameters used for the ImageNet-1K [5], CIFAR10 [6], and CIFAR100 [6] of the distilled Data2Vec2 image model. We refer to the respective papers for details on the augmentation techniques [1]–[4].

Type	Hyperparameters	MNLI	QNLI	RTE	MRPC	QQP	STS-B	CoLA	SST
Training	Epochs	15							
	Batch size	256							
	Optimizer	AdamW							
	AdamW ε	1e-8							
	AdamW β	(0.9, 0.999)							
	Weight decay	0.01							
	Base learning rate	1e-3							
	Layer Decay	0.81							
	Learning rate schedule	Cosine							
	Warmup steps	10% of total steps							
	Metric	Accuracy	Accuracy	Accuracy	F1	F1	Spearman	Accuracy	Accuracy
	Hardware	1 × RTX 4090 24GB							

Table 3: Hyperparameters for the GLUE [7] benchmark tasks of the distilled Data2Vec2 image model.

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