

***ForAug: Recombining Foregrounds and Backgrounds to Improve Vision
Transformer Training with Bias Mitigation
- Supplementary Material -***

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Anonymous ICCV submission

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Paper ID 6426 - Supplementary

Abstract

This is the supplementary material for the paper: ForAug: Recombining Foregrounds and Backgrounds to Improve Vision Transformer Training with Bias Mitigation

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A. Training Setup

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On ImageNet we use the same training setup as [1] and [2] without pretraining. As our focus is on evaluating the changes in accuracy due to *ForAug/ForNet*, like [1], we stick to one set of hyperparameters for all models. We list the settings used for training on ImageNet and *ForNet* in Table 1 and the ones used for finetuning those weights on the downstream datasets in Table 2.

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Parameter	Value
Image Resolution	224 × 224
Epochs	300
Learning Rate	3e-3
Learning Rate Schedule	cosine decay
Batch Size	2048
Warmup Schedule	linear
Warmup Epochs	3
Weight Decay	0.02
Label Smoothing	0.1
Optimizer	Lamb [3]
Data Augmentation Policy	3-Augment [2]

Table 1. Training setup for our ImageNet and *ForNet* training.

Dataset	Batch Size	Epochs	Learning Rate
Aircraft	512	500	3e-4
Cars	1024	500	3e-4
Flowers	256	500	3e-4
Food	2048	100	3e-4
Pets	512	500	3e-4

Table 2. Training setup for finetuning on different downstream datasets. Other settings are the same as in Table 1.

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B. Infill Model Comparison

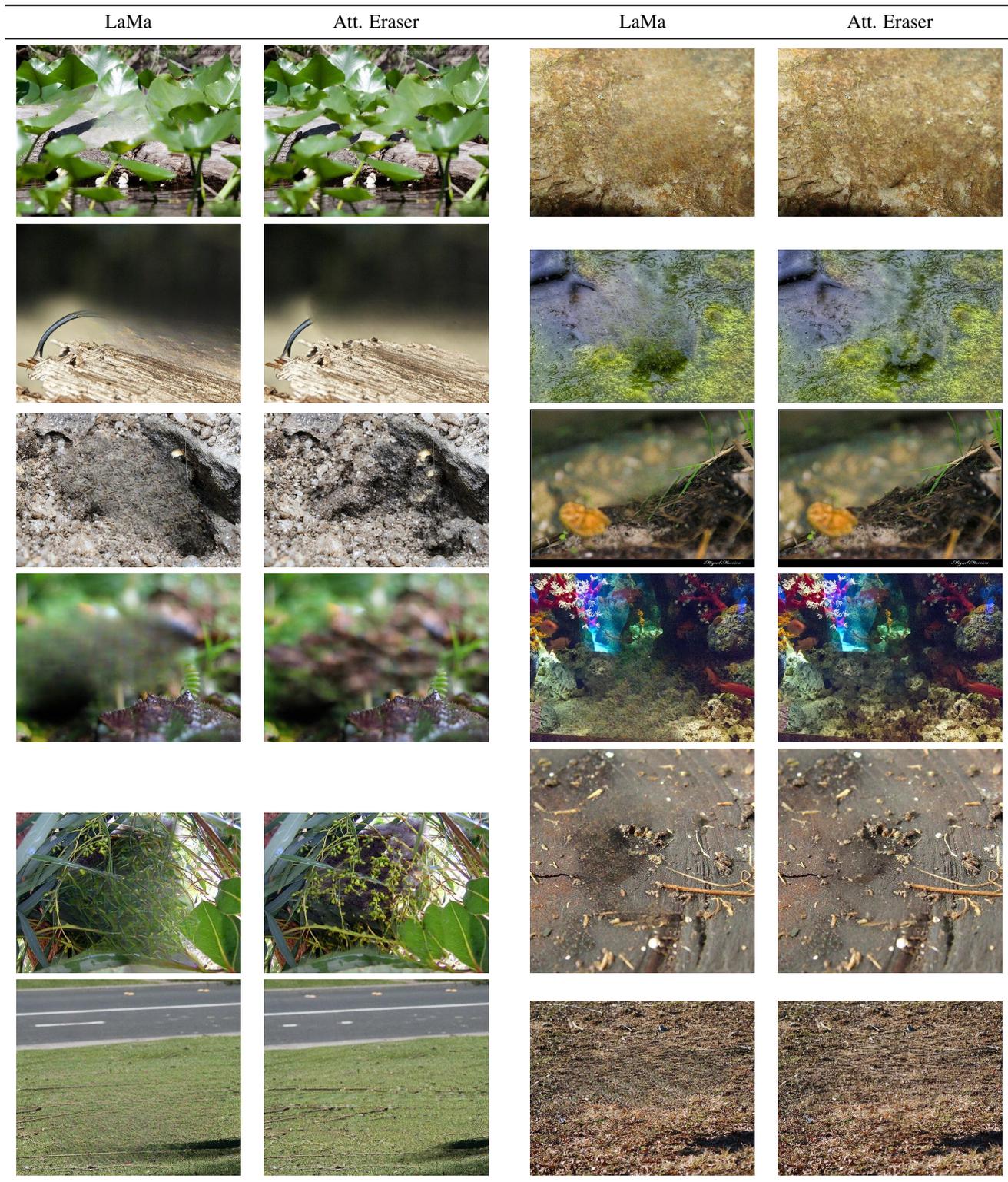


Table 3. Example infills of LaMa and Attentive Eraser.

013 **C. Images with High Infill Ratio**

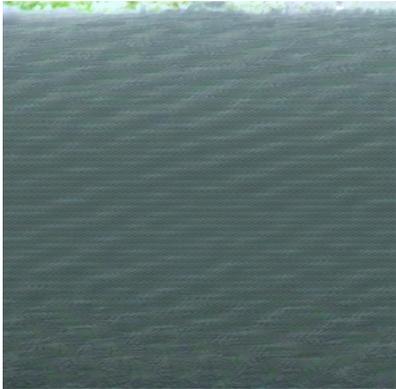
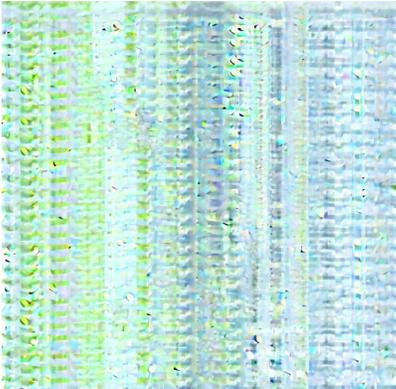
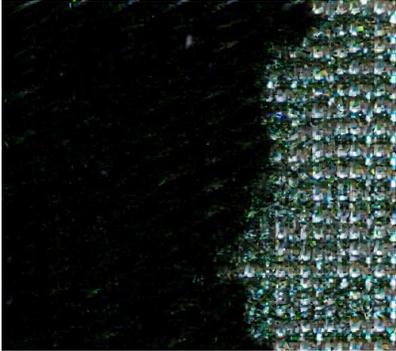
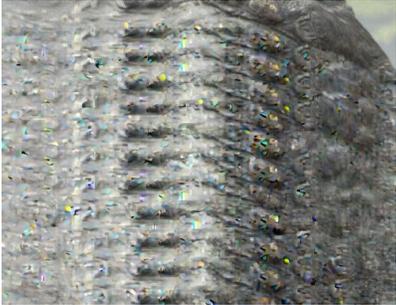
Infill Ratio	LaMa	Att. Eraser
93.7		
95.7		
83.7		
88.2		

Table 4. Example infills with a large relative foreground area size that is infilled (infill ratio).

014 **References**

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