Simple Self-Distillation Learning for Noisy Image Classification

ICIP2023, Kuala Lumpur, Malaysia

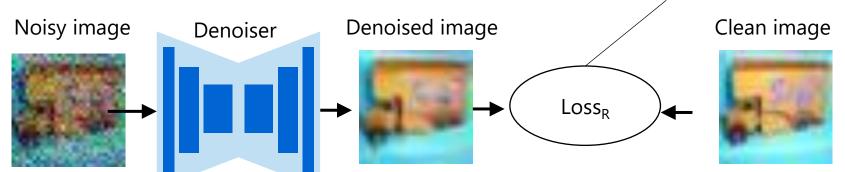
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Introduction: Image Denoising for Noisy Image Classification

Neural network denoiser using clean image

- ✓ Higher performance than traditional methods
- × Using clean image (**NOT available** for most practical uses)

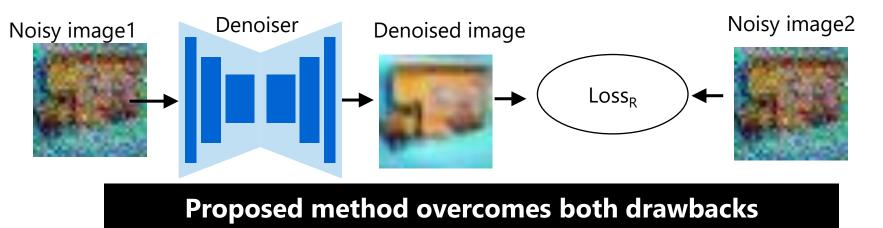


Reconstruction loss

(e.g. mean-squared error)

•Training strategy using only **noisy** image (Noise2Noise_[Lehtinen+,ICML2018])

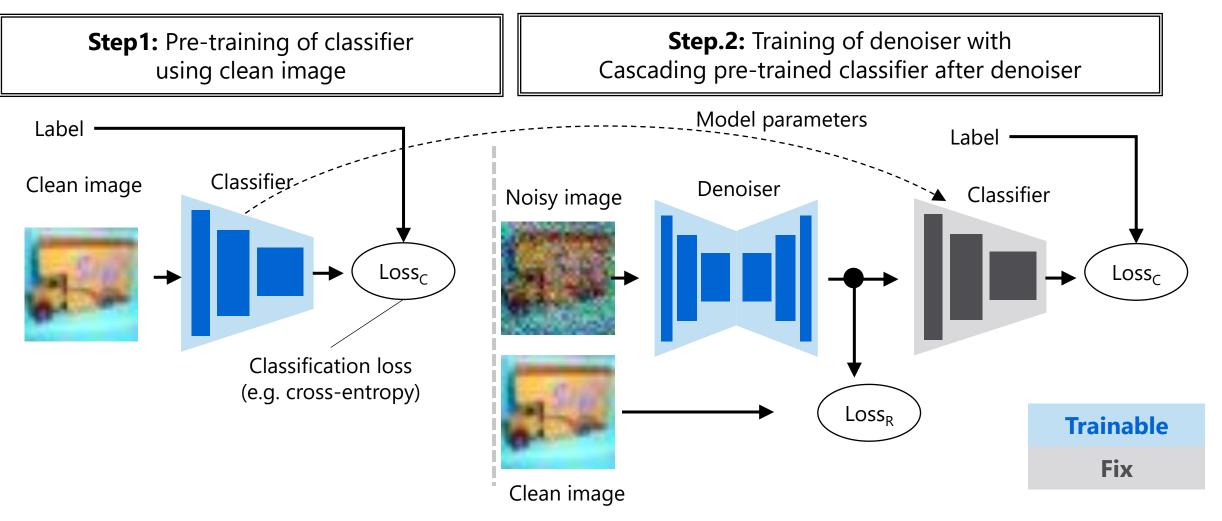
- ✓ Comparable performance to methods using clean image
- × Blurry image (**NOT suitable** for image classification)



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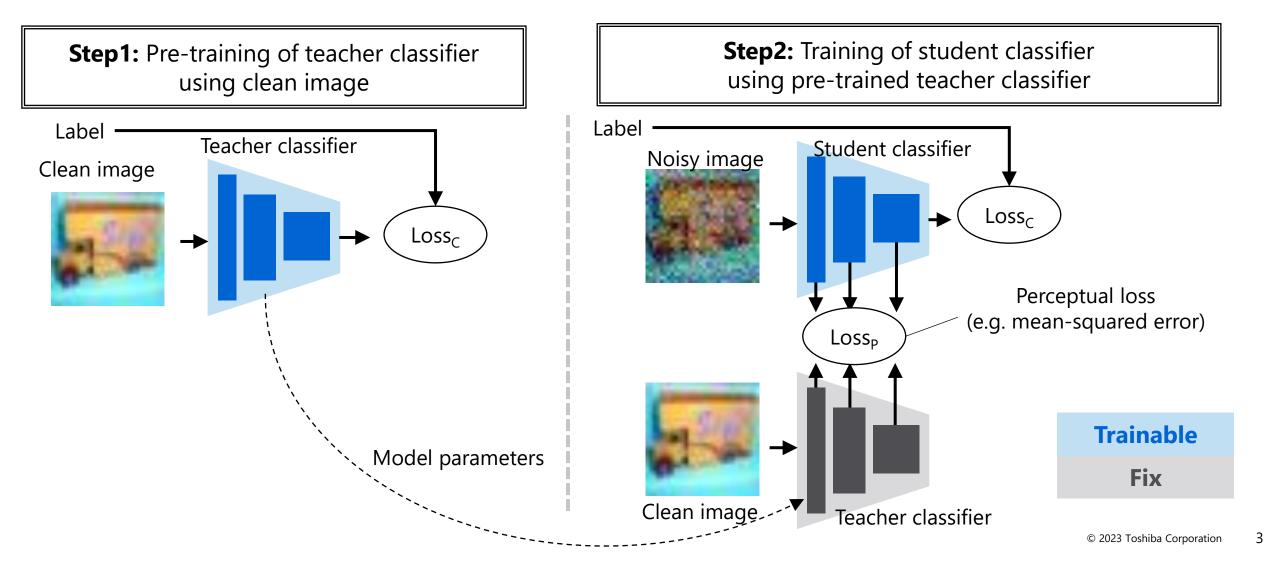
Prior Work (1): Joint Training_[Liu+, IJCAI2018]

Training denoiser based on both image quality and performance on downstream task



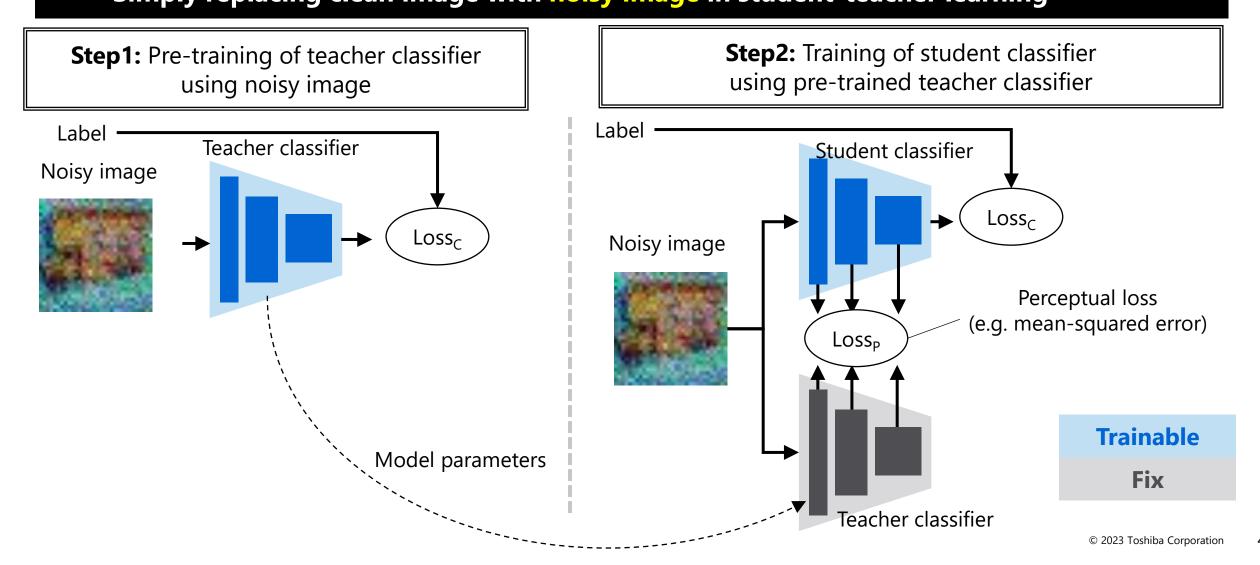
Prior Work (2): Student Teacher Learning_[Gnanasambandam+, ECCV2020]

More direct way without denoiser inspired by knowledge distillation

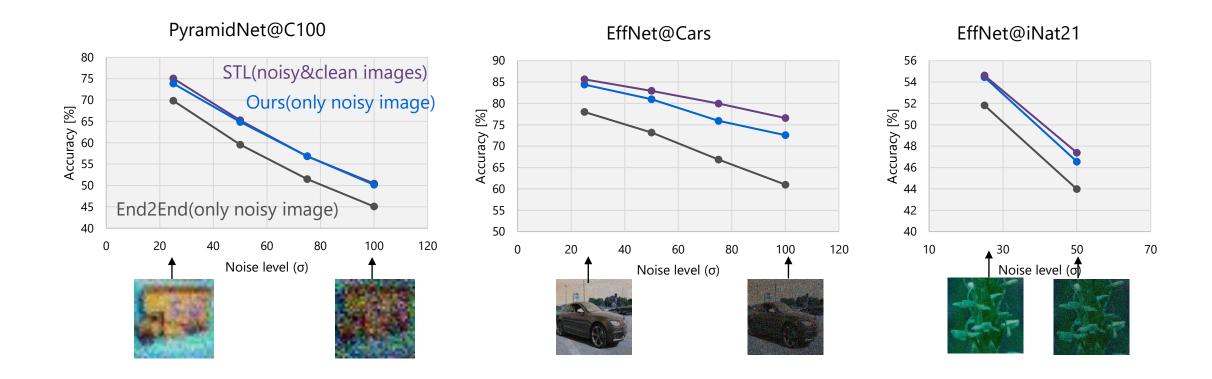


Proposed Method: Simple Self-Distillation Learning

Leveraging property of feature extractor that naturally removes unnecessary features
Simply replacing clean image with noisy image in student-teacher learning

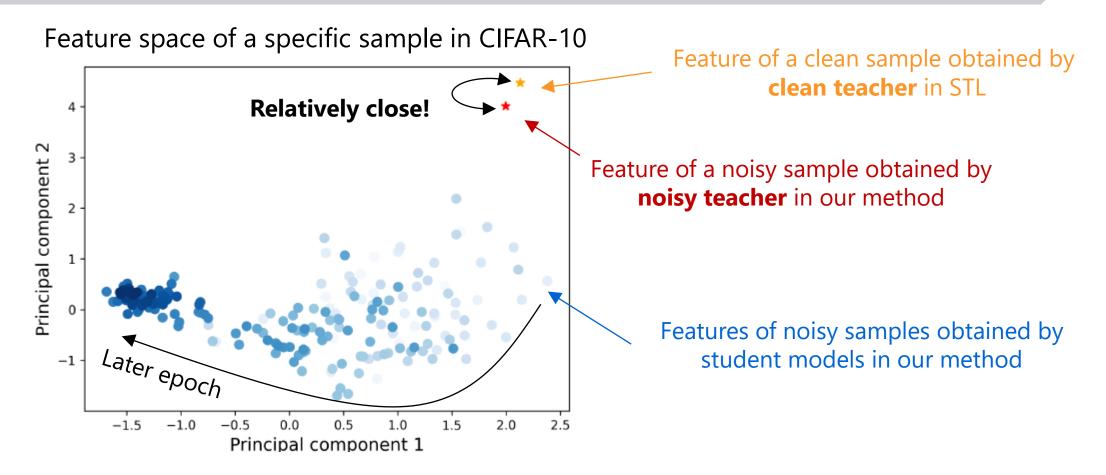


Results



Proposed method achieves comparable performance to STL even without clean images

Qualitative Analysis



Quality of noisy teacher is comparable to that of clean teacher

Summary

Our method: simple self-distillation learning

- Without explicit image denoising
- Without clean image
- •Results on noisy image classification task
 - Comparable performance to prior method using clean image
- •Future work
 - Theoretical analysis
 - Investigation on other tasks