Unsupervised Discriminative Learning of Sounds for Audio Event Classification

Sascha Hornauer Ke Li Stella X. Yu

UC Berkeley / ICSI

Shabnam Ghaffarzadegan Liu Ren

Robert Bosch LLC







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State-of-the-art Relies on ImageNet Pre-Training



Andrey Guzhov, Federico Raue, Jorn Hees, and Andreas Dengel, "Esresnet: Environmental sound classification based on visual domain models," arXiv preprint arXiv:2004.07301, 2020.

Subsequent Classifier Fine-Tuning on Audio for Quick Convergence



Pro: Quicker Accuracy Gain in Early Epochs Than No Pretraining



Quick User Data Adaptation Is Useful for Edge Devices



Con: ImageNet PreTraining Needs Large, Image-Parsing Nets



 Image data requires layers with many parameters

Our Idea: Pretrain on Sound Data Directly without Supervision



Using Audio Data Alone, Faster Pre-Training Performs On Par



Details of the Sound Encoding



Frequency Bins are Distributed Along Channel Dimension



Each resulting instance has three Channels, containing different parts of the frequency spectrum.

Schema Exactly as in ESResnet for Direct Comparison

Details of the Unsupervised Embedding for Pretraining





Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition, 2018

Train Instance Discrimination on Spectrograms



Zhirong Wu, Yuanjun Xiong, Stella X Yu, and Dahua Lin, "Unsupervised feature learning via non-parametric instance discrimination," in Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition, 2018

Training on Sound Classification Task Yields Fast Improvement

Datasets:

- a) ESC10
- b) DCASE2013
- c) ESC50
- d) US8K

Pre-Training with NPID unsupervised on all Datasets.

Downstream training follows official train/val folds.

Results are averages over all Folds



In Summary: Quick Performance Gain for Training Arbitrary Networks on the Edge

