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MoA+: Mixture of Autoencoders with Various Concentrations for Enhanced Image Clustering

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In this paper, we consider the improvement of the vanilla Mixture of Experts model [1] in the unsupervised image clustering task by introducing an unbiased loss function based on the idea of a mixture of various concentrations. This addresses expert dominance, enhancing training balance and clustering performance.

The model is tested on the MNIST dataset, a benchmark of N grayscale handwritten digit images X_i , i = 1..N, treated as a mixture with M = 2 components.

In the MoA+ framework, each expert is a convolutional autoencoder (CAE), similar to simplified U-Net or SegNet, where the k-th CAE reconstructs image X_i into $v_{i,k}$ with MSE error. A convolutional gate network assigns probabilities $p_{i,k}$, clustering images by selecting the most suitable CAE.

To prevent expert imbalance, MoA+ uses the Q-loss, inspired by ST-MoEs z-loss [2].

$$Q\text{-loss} = \sqrt{\sum_{k=1}^{M} \left(\sum_{i=1}^{N} p_{i,k} \cdot \text{MSE}(v_{i,k}, X_i)\right)^2}$$

This regularizes the gate network, ensuring all experts contribute effectively to the clustering process.

Compared under identical conditions, the standard MoA achieved a Normalized Mutual Information (NMI) of ~ 0.08 , while MoA+ scored ~ 0.8 , showing significantly better clustering performance.

References

- [1] Jacobs, R. A., Jordan, M. I., Nowlan, S. J., & Hinton, G. E. (1991). Adaptive Mixtures of Local Experts. Neural Computation, 3(1), 79–87. https://direct.mit.edu/neco/article/3/1/79/5560/Adaptive-Mixtures-of-Local-Experts.
- [2] Author, A. B. (2022). ST-MoE: Designing Stable and Transferable Sparse Expert Models. arXiv preprint arXiv:2202.08906. https://arxiv.org/abs/2202.08906.

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