

Supplementary Materials for the Paper: Hierarchical Prompt Learning for Multi-Task Learning

Anonymous CVPR submission

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1. Loss and Error Landscape Visualizations

Inspired by [1], we provide visualizations of train loss and test error surfaces. Given three classifier weights w_1 , w_2 , and w_3 (i.e., w_{rand} , w_{shr} , and w_{ind}), we set $u=w_2-w_1$ and $v=w_3-w_1-\langle w_3-w_1, \hat{u} \rangle \hat{u}$, where \hat{u} is the normalized vector $\hat{u}=u/\|u\|$. Thus, \hat{u} and \hat{v} represent an orthonormal basis in the plane which contains w_1 , w_2 , and w_3 . We can evaluate the train loss and test error of the corresponding model parameters by sampling on this plane. Note that w_{avg} is a linear combination of w_{ind} and w_{shr} , which lies in the plane defined by w_{ind} , w_{shr} , and w_{rand} .

2. Additional Visualizations of Hierarchical Task Clustering

As shown in Figure A1, we provide additional hierarchical task clustering results on Office-Home (with the split of 20%), DomainNet (with the split of 2%), and the large-scale multi-task learning benchmark (with 16-shot). Similar tasks, e.g., Real and Product on Office-Home, are clustered together.

3. Detail Results of 10 Downstream Tasks

Detail results of our HiPro and other baselines on 10 downstream classification tasks are shown in Figure A2.

References

- [1] T. Garipov, Pavel Izmailov, Dmitrii Podoprikhin, Dmitry P. Vetrov, and Andrew Gordon Wilson. Loss surfaces, mode connectivity, and fast ensembling of dnns. In *NeurIPS*, 2018.

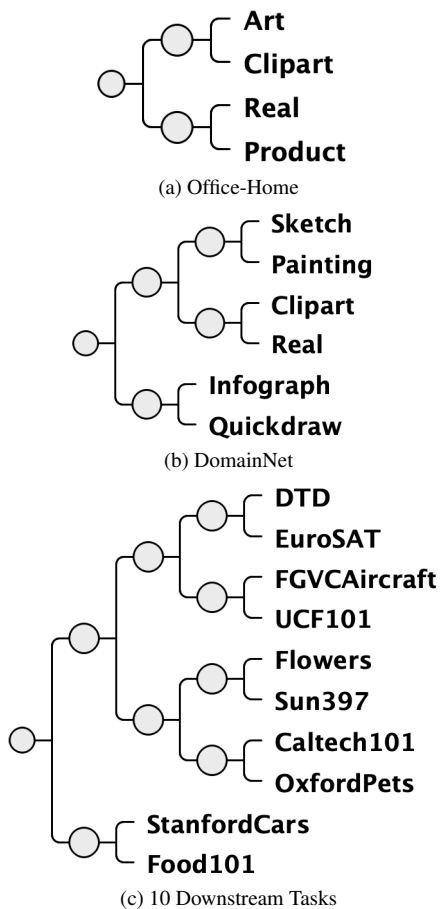


Figure A1. Hierarchical task clustering results of HiPro on Office-Home, DomainNet, and the large-scale multi-task learning benchmark.

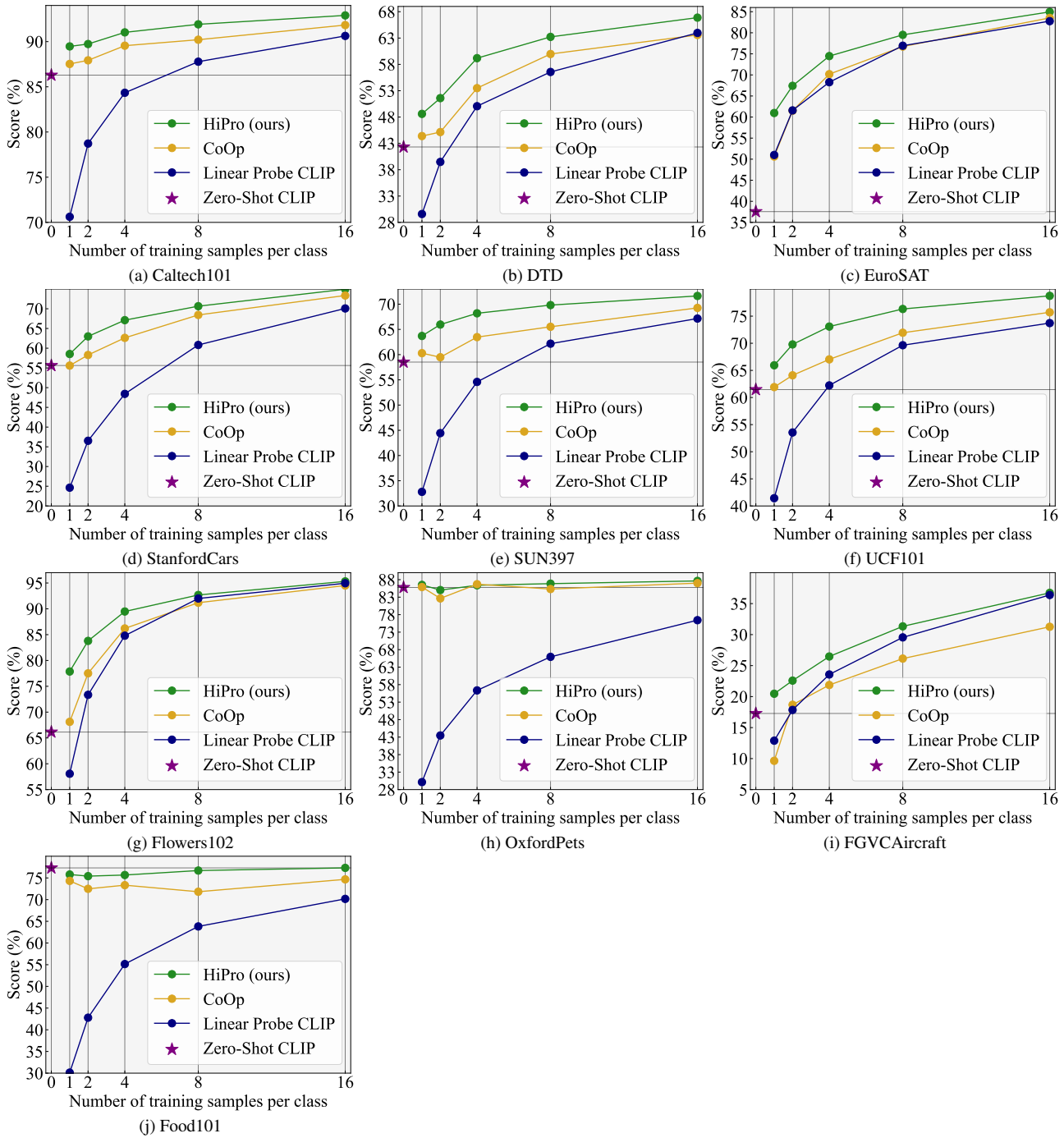


Figure A2. Detail results on 10 downstream tasks.