

S-Prompts

Learning with Pre-trained Transformers: An Occam's Razor for Domain Incremental Learning

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Problem

Exemplar-free Domain-incremental learning

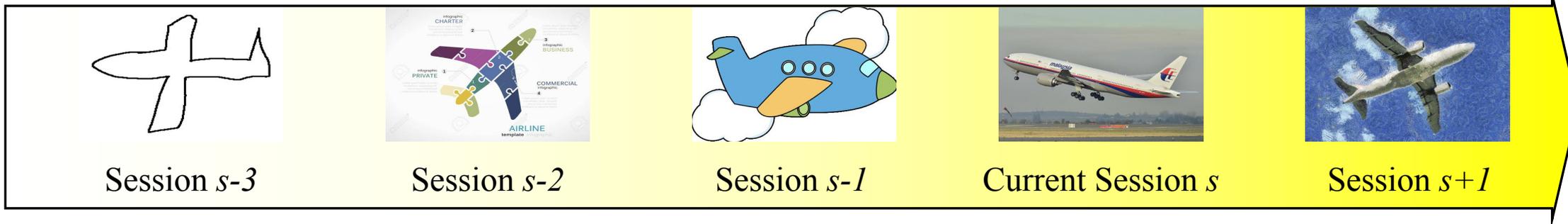
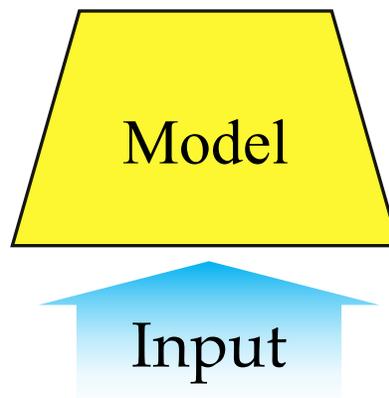
Goal: **Incrementally learning** knowledge of **various domains** in sequence.

Conditions:

Large Domain Gaps

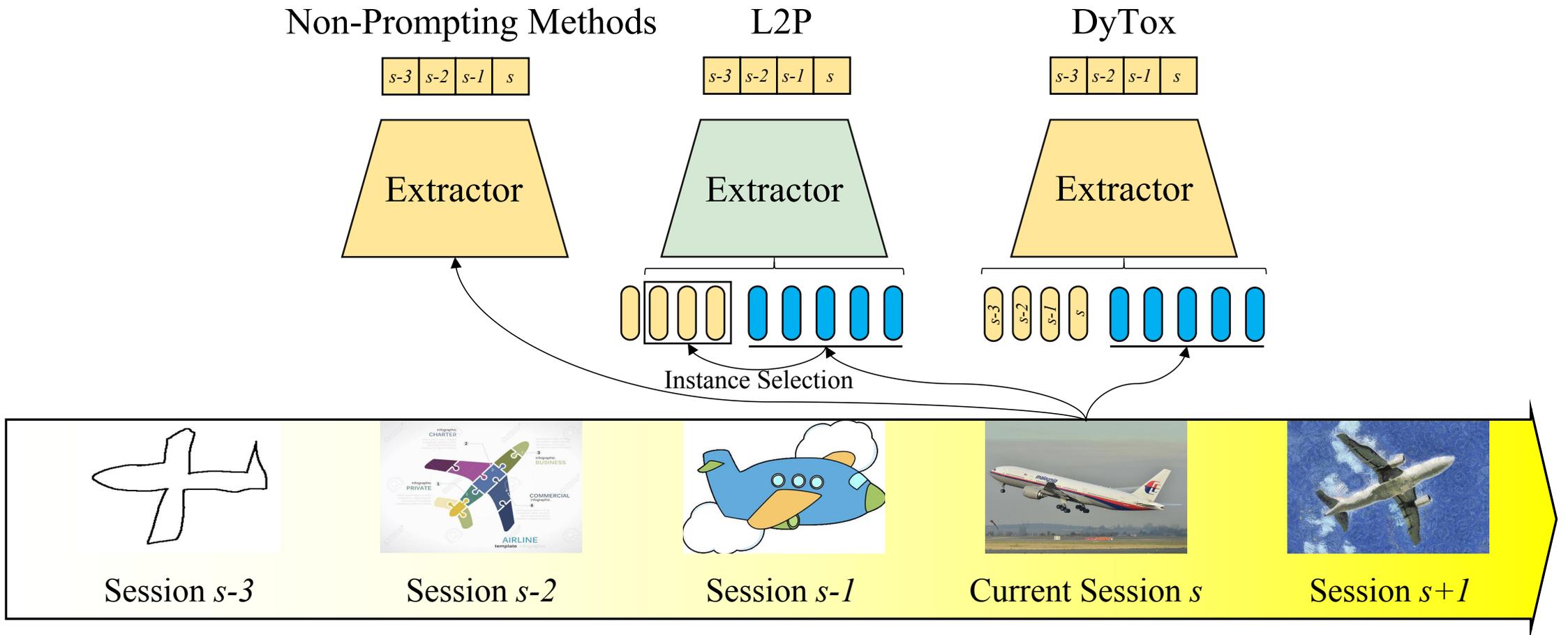
No exemplars

Data Stream



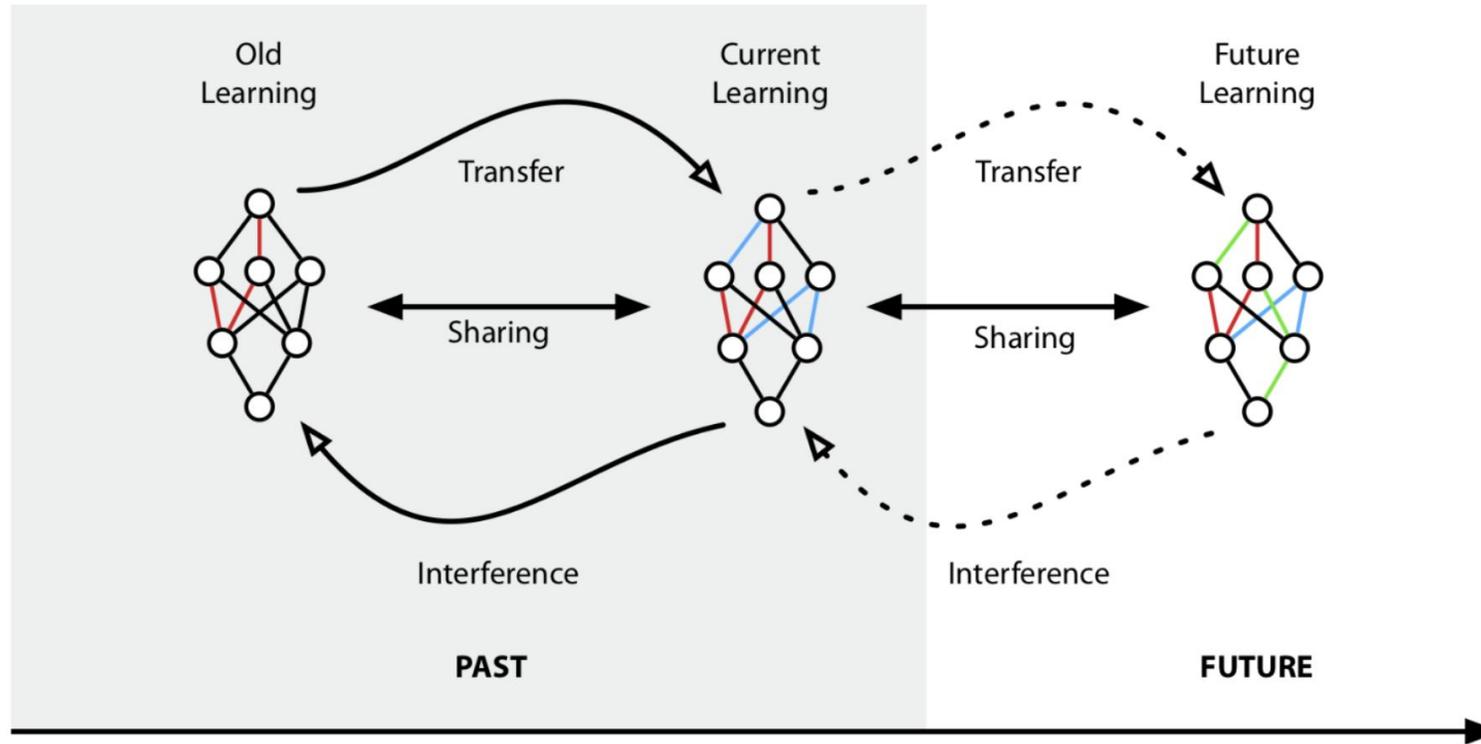
Common Learning Paradigm

Knowledge Accumulation



Common Learning Paradigm

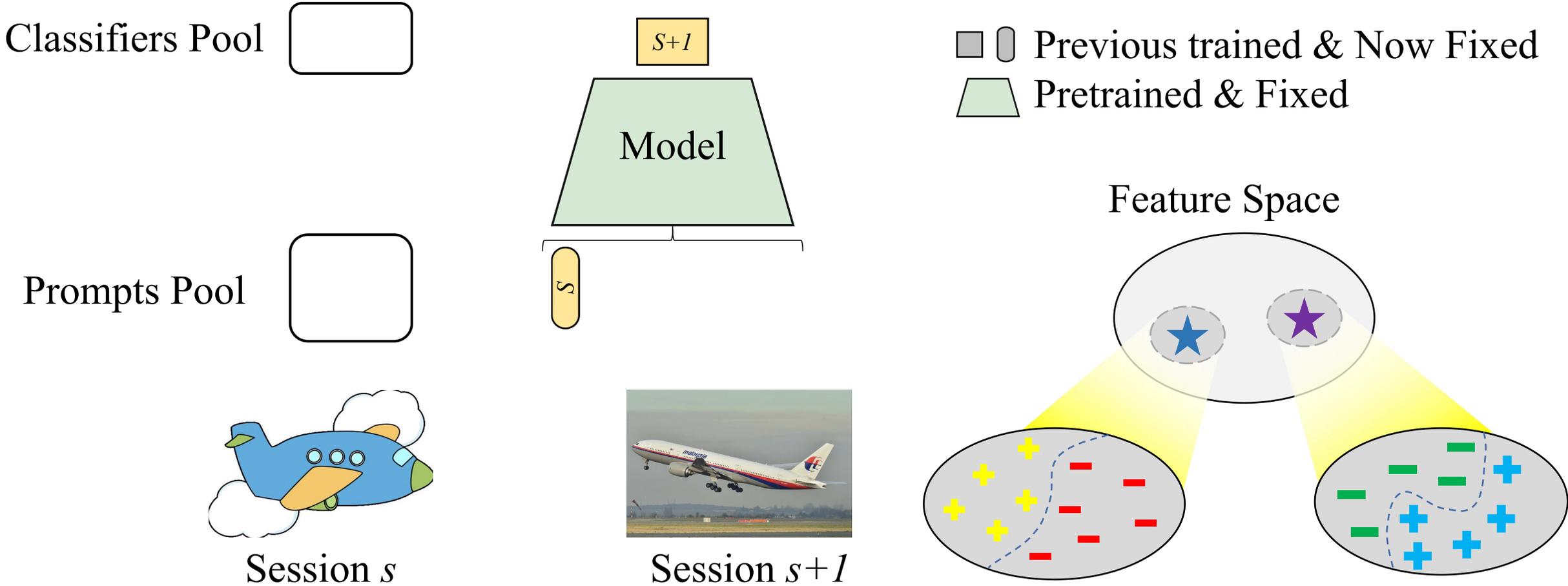
The Stability-Plasticity Dilemma



Tug-of-War: one side's gain is equivalent to the other's loss

S-Prompts paradigm

Key idea: Learning the prompts independently across domains



Experimental Results

Main Results

- Continual Deepfake Detection Benchmark [1]
Relative gain over the best of real competitors $\sim 44\%$ improvement
- CORe50 [2]
Relative gain over the best of real competitors $\sim 14\%$ improvement
- DomainNet [3]
Relative gain against the best competitor $\sim 33\%$ improvement

[1] Li, C., Huang, Z., Paudel, D.P., Wang, Y., Shahbazi, M., Hong, X. and Van Gool, L., 2022. A Continual Deepfake Detection Benchmark: Dataset, Methods, and Essentials. arXiv preprint arXiv:2205.05467.

[2] Lomonaco, V. and Maltoni, D., 2017, October. Core50: a new dataset and benchmark for continuous object recognition. In Conference on Robot Learning (pp. 17-26). PMLR.

[3] Peng, X., Bai, Q., Xia, X., Huang, Z., Saenko, K. and Wang, B., 2019. Moment matching for multi-source domain adaptation. In Proceedings of the IEEE/CVF international conference on computer vision (pp. 1406-1415).

Contributions

Contributions:

1. A **rule-breaking learning paradigm** that learns the prompts independently domain by domain to play a **win-win game for DIL**.
2. A **new paired language-image prompting scheme** to enhance CLIP's transfer learning ability.

Code link:

<https://github.com/iamwangyabin/S-Prompts>

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