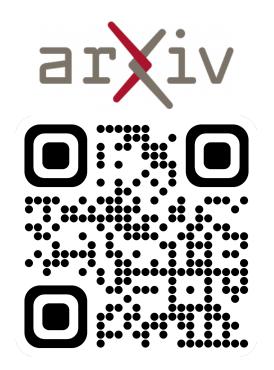




Improving Zero-shot Generalization of Learned Prompts via Unsupervised Knowledge Distillation

*Marco Mistretta, *Alberto Baldrati, Marco Bertini, Andrew D. Bagdanov



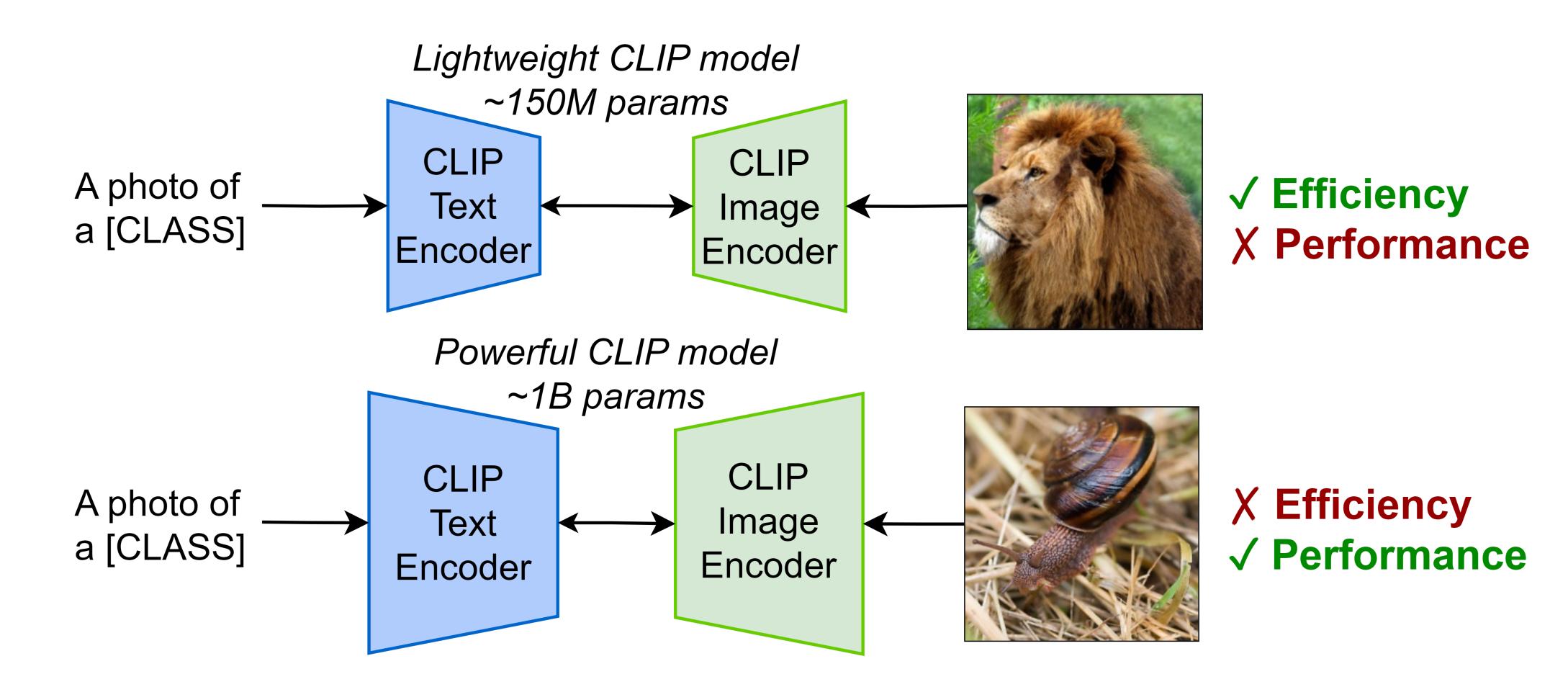


Presented by Andrew D. Bagdanov





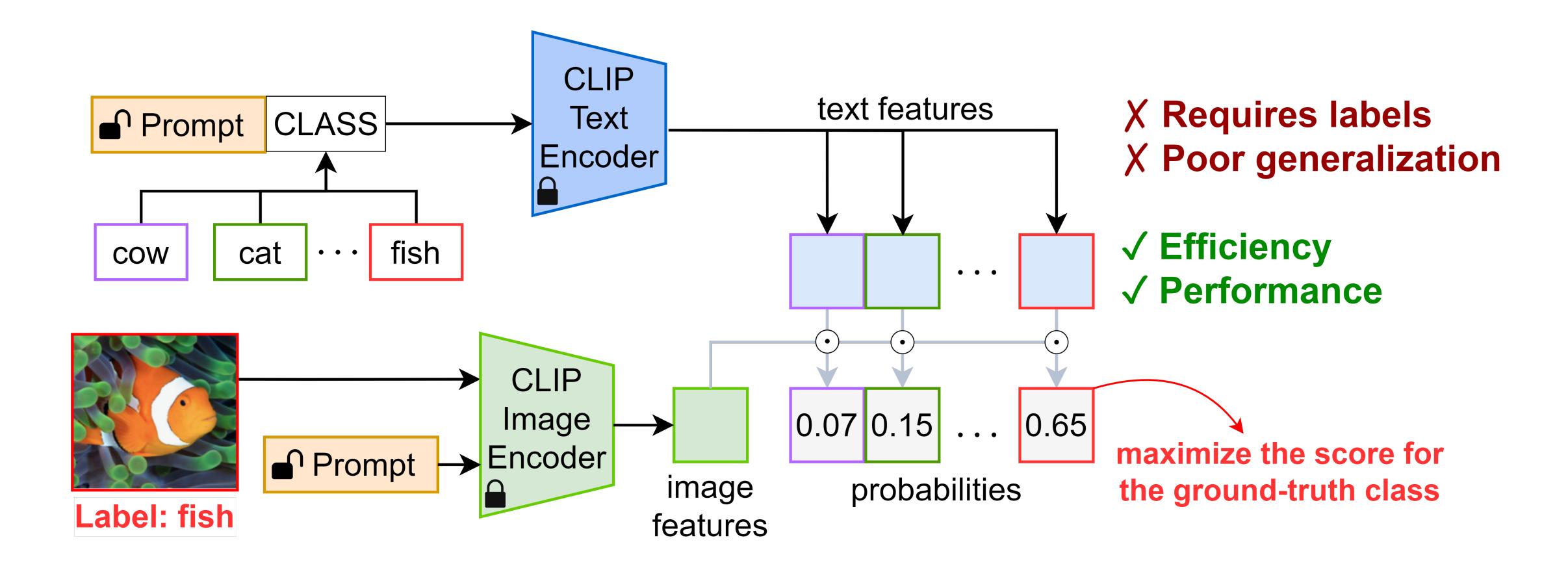
Motivation



Larger CLIPs are powerful but are not efficient. Fine-tuning smaller CLIPs is too expensive.

Existing approaches

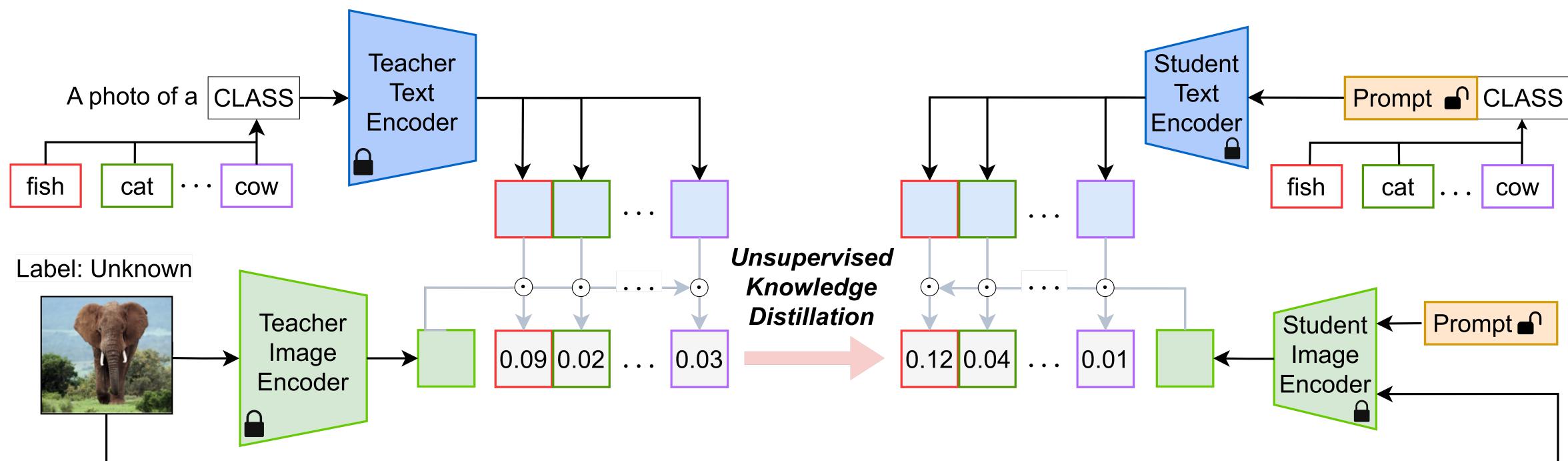
Prompt Learning is an efficient solution, but the best performing SOTA techniques require ground truth labels.



2

Knowledge Distillation for Prompt Learning (KDPL)

KDPL can be integrated with any existing prompt learning technique and eliminates the need for labels!



Note that the teacher is frozen and only needed at training time! Furthermore, distilling at logit level makes KDPL completely architecture independent!





































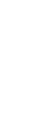
















3

Cross Domain Evaluation

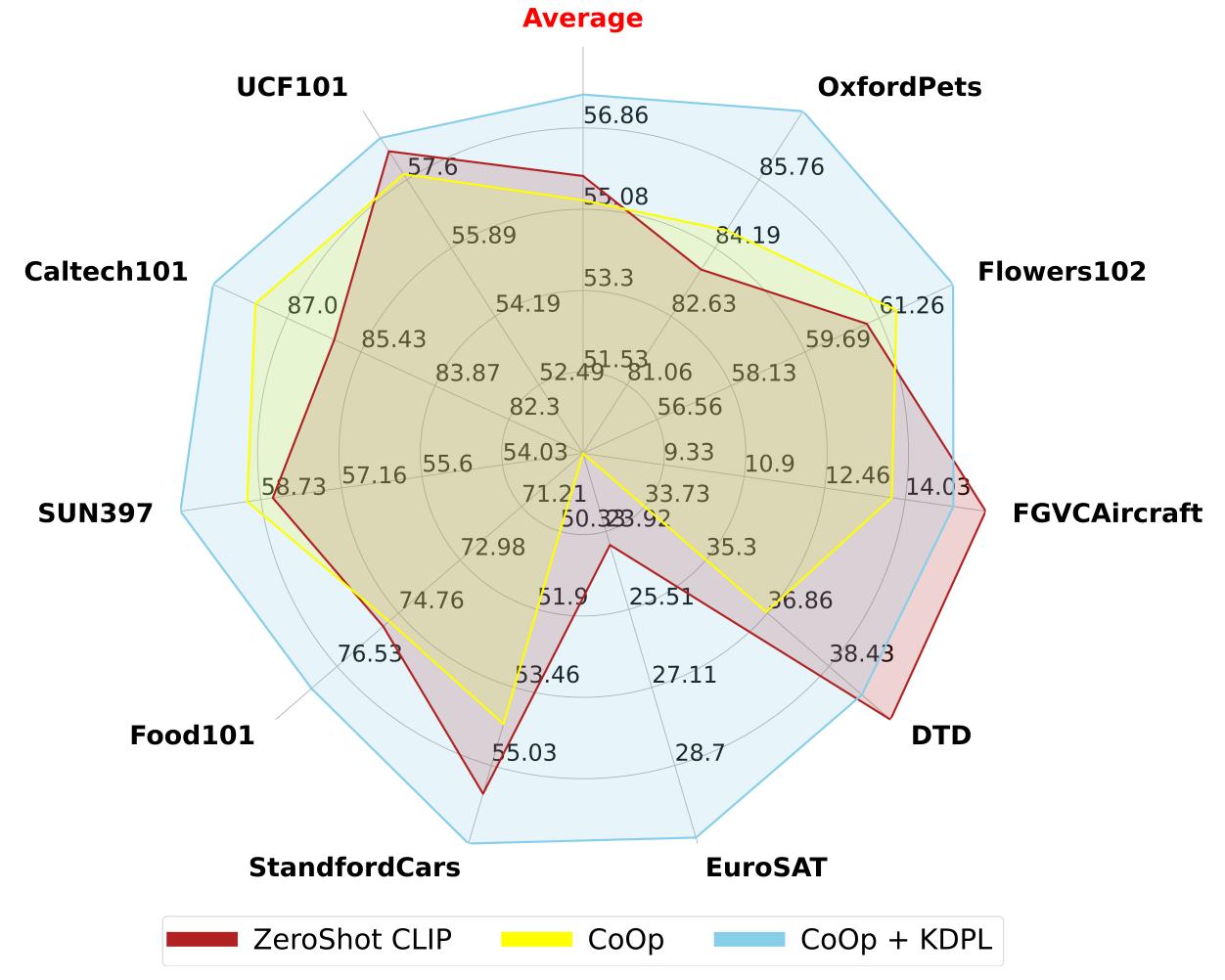
Train 16-shots on ImageNet and evaluate on ImageNet -V2/-Sketch/-A/-R.

		Source	Target				
Backbone	Method	ImageNet	-V2	-S	-A	-R	Average
RN50	CLIP (student)	58.20	51.50	33.30	21.70	56.00	40.63
	CoOp	62.40	55.17	33.70	23.13	56.20	42.05
	$\mathrm{CoOp}+\mathrm{KDPL}$	62.73	55.37	35.20	23.27	57.77	42.90
	CoCoOp	63.07	55.53	34.77	23.73	59.47	43.38
	m CoCoOp+KDPL	62.70	55.60	35.30	23.43	57.90	43.06
ViT-B/32	CLIP (student)	62.00	54.70	40.80	29.60	66.00	47.78
	CoOp	66.33	58.30	41.40	31.47	65.87	49.26
	$\mathrm{CoOp}+\mathrm{KDPL}$	65.97	58.10	42.50	31.63	67.37	49.90
	VPT	64.97	56.73	41.27	27.00	66.50	47.88
	$\mathrm{VPT}+\mathrm{KDPL}$	65.10	57.37	41.67	27.77	67.47	48.57
	MaPLe	66.80	58.53	42.23	30.13	66.40	49.32
	MaPLe + KDPL	66.50	58.47	42.77	29.87	67.70	49.70
	PromptSRC	66.33	58.70	42.97	32.07	68.93	50.67
	$\operatorname{PromptSRC} + \operatorname{KDPL}$	66.27	58.60	43.07	31.70	68.83	50.55
ViT-H/14	CLIP (teacher)	82.80	76.60	71.10	71.10	91.30	77.53



Cross Dataset Evaluation

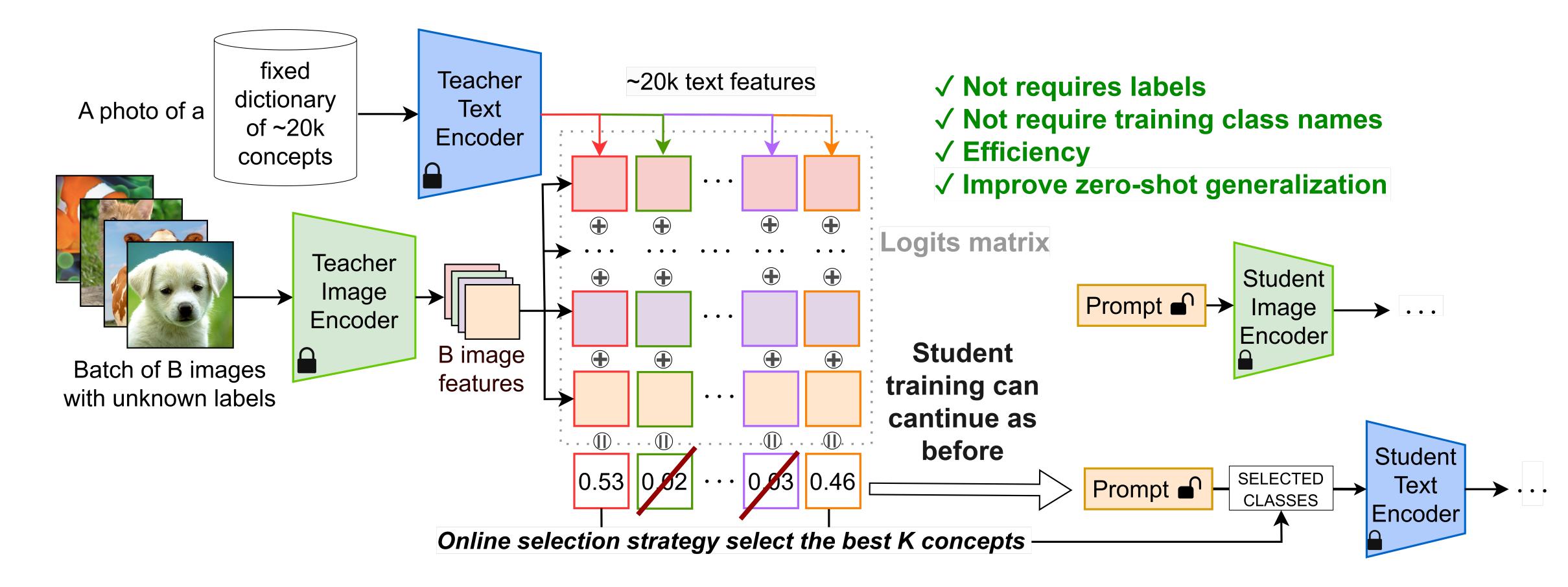
Train 16 shots on ImageNet and evaluate on 10 different benchmark datasets.





Class-Agnostic Approach: KDPL-CA

What can we do when we know neither the labels nor the *training class names*?



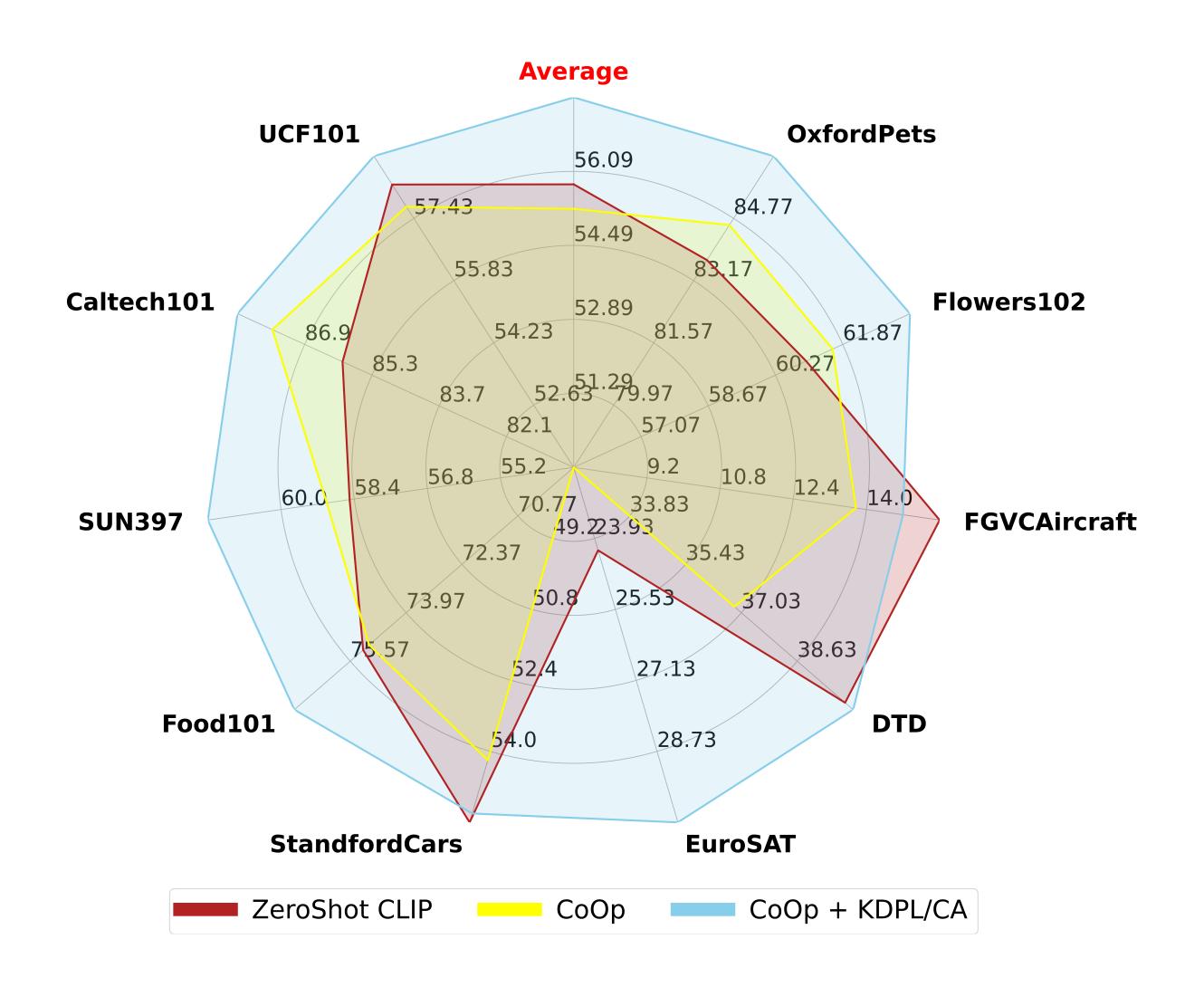
See section 3.3 of the paper for more details on the Online concepts selection strategy.





KDPL-CA improves zero-shot generalization

Train 16 shots on ImageNet and evaluate on 10 different benchmark datasets.

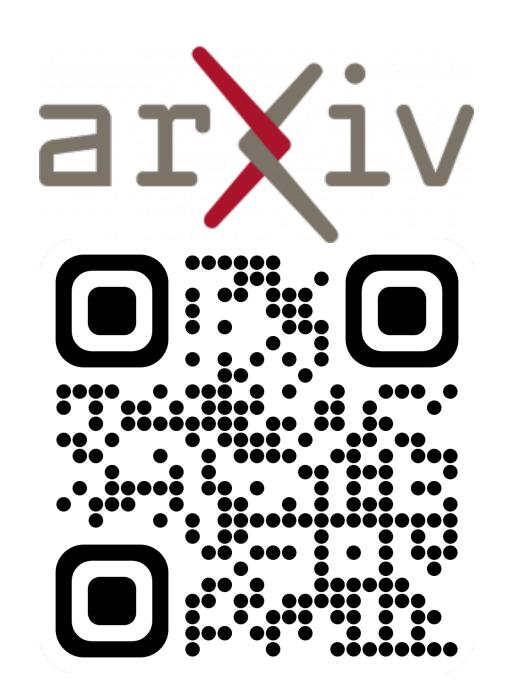




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Thank you for your attention!











Presented by Andrew D. Bagdanov