



MIT-IBM

Watson AI Lab

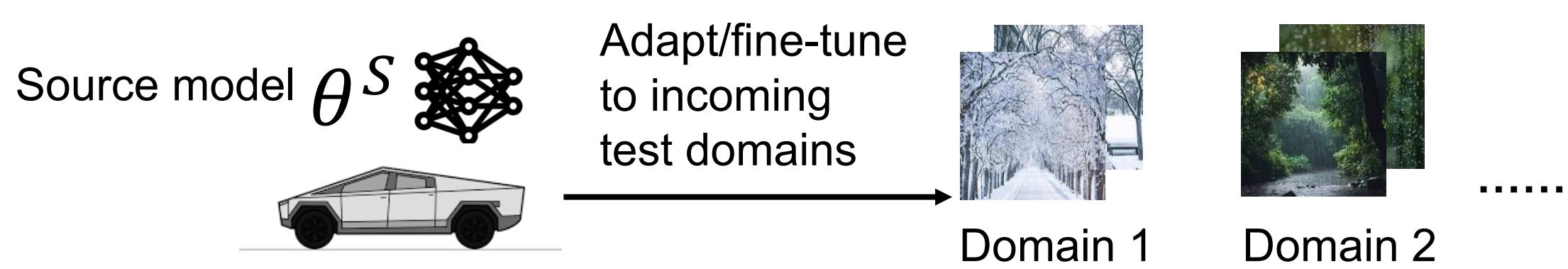
BATCLIP: Bimodal Online Test-Time Adaptation for CLIP

Sarthak Kumar Maharana¹, Baoming Zhang¹, Leonid Karlinsky², Rogério Schmidt Feris², Yunhui Guo¹UT Dallas¹, MIT-IBM Watson AI Lab²

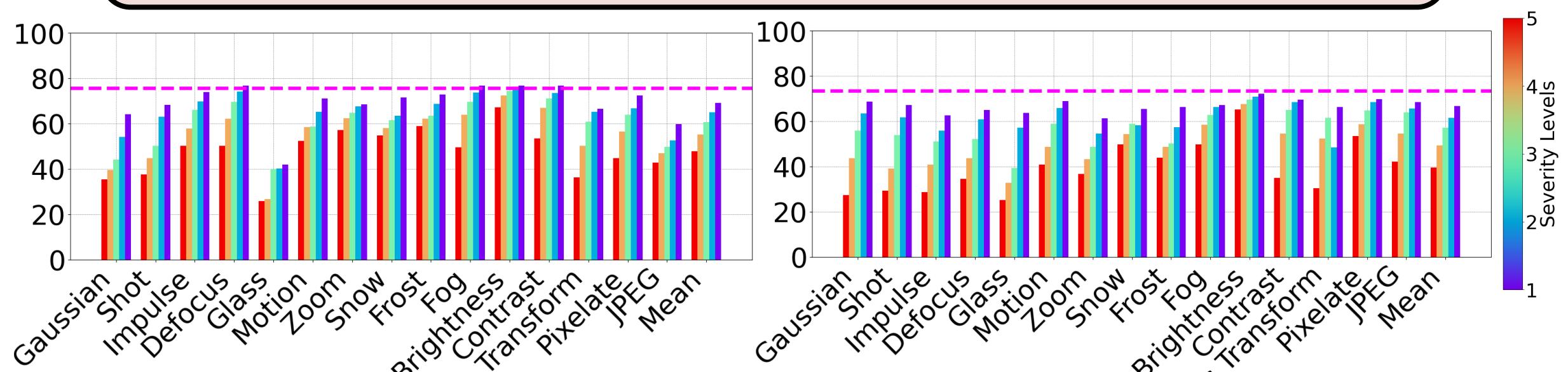
ICCV OCT 19-23, 2025 HONOLULU HAWAII

Problem Overview and Motivation

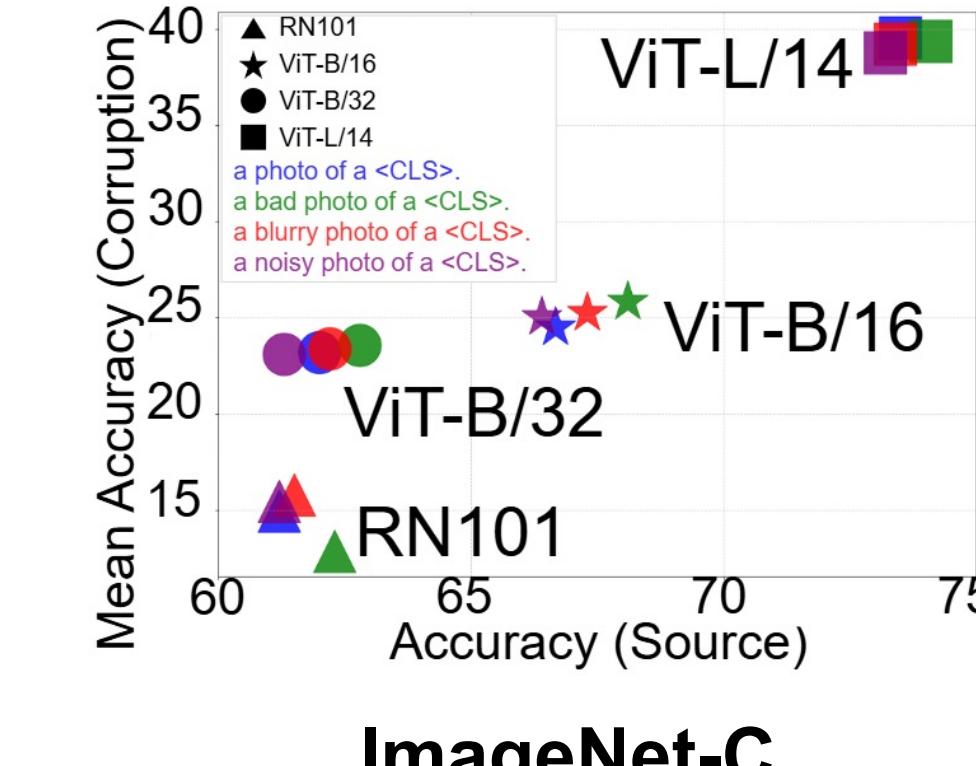
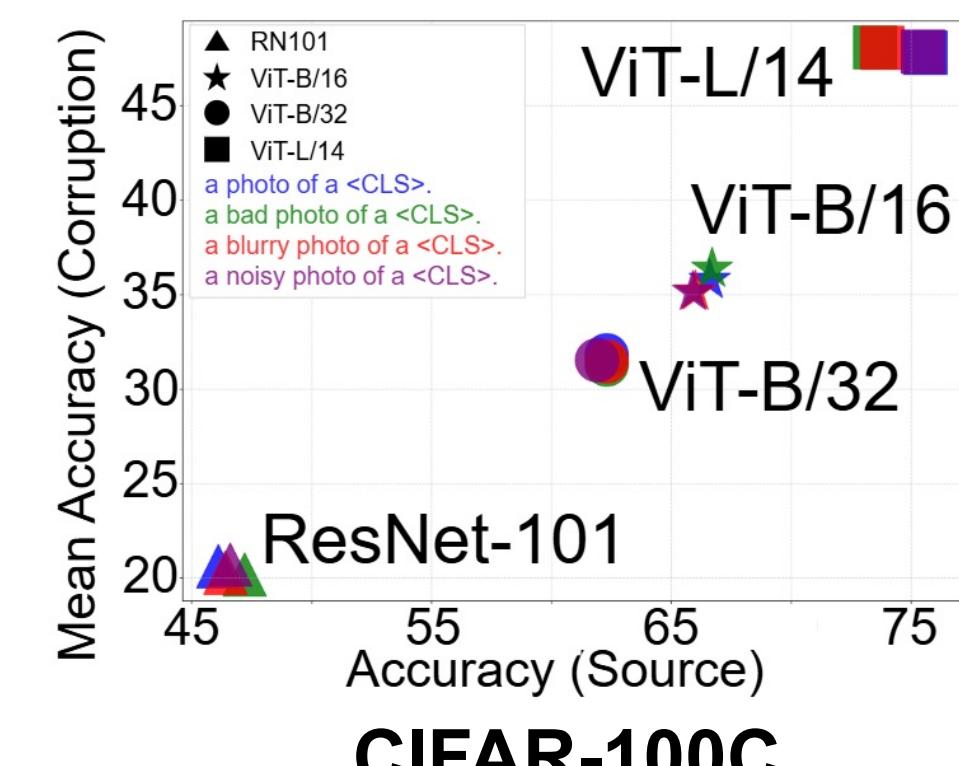
- **Online test-time adaptation (TTA)** involves pre-trained/source model adaptation to incoming **unlabeled test data** to minimize the source-target domain distribution gap.
- *Single forward pass* to preserve privacy.
- No access to the pre-training/source dataset.



Are zero-shot CLIP features transferable to “new” domain shifts/corruptions? NO!



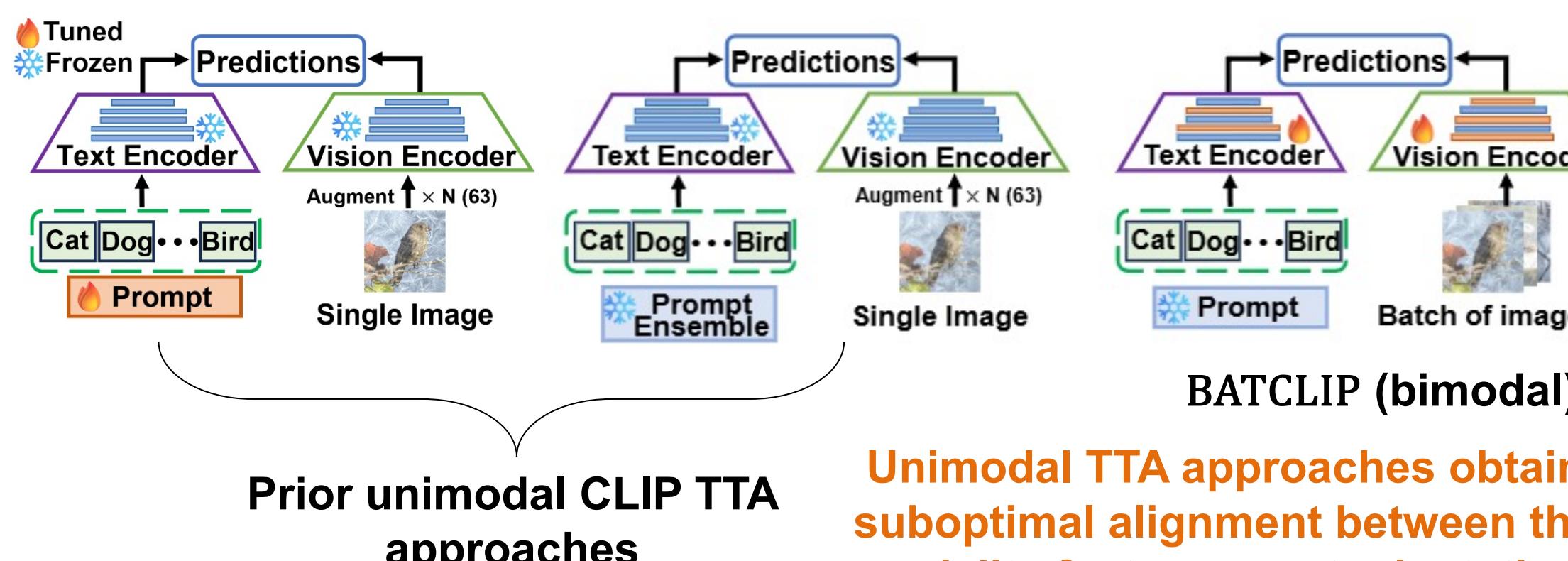
Across backbones, CLIP is very sensitive to severe visual corruptions (up, ViT-L/14); at test-time, “relevant” prompting doesn’t help (bottom). So, there’s a need for minimal adaptation as text and visual features are independent.



*Leonid Karlinsky is with Xero.

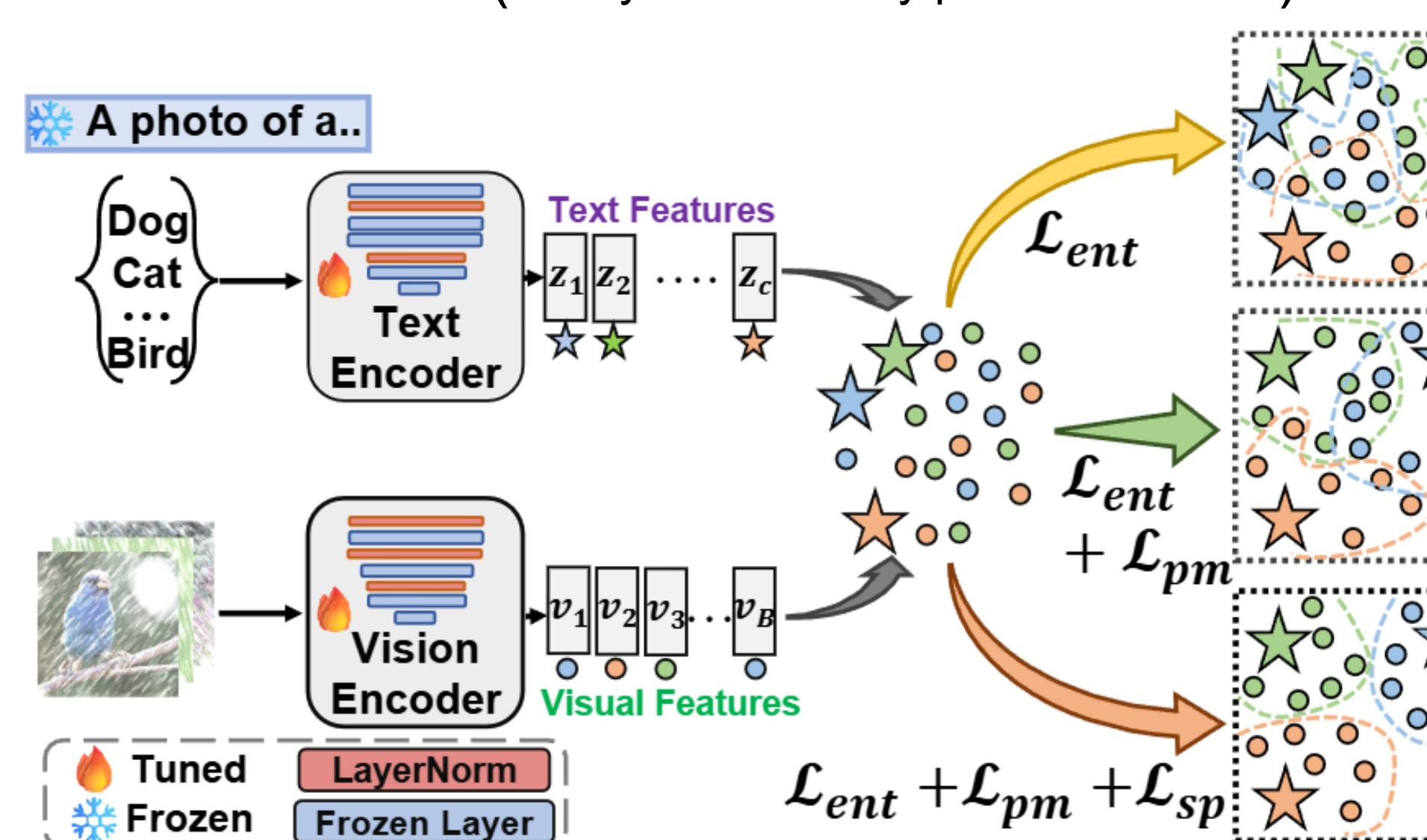
Proposed Methodology

- Existing online TTA approaches using CLIP vs ours.



Unimodal TTA approaches obtain suboptimal alignment between the modality features post adaptation.

- Ours for **online CLIP** (or any contrastively pre-trained VLM) TTA



$$\mathcal{L}_{ent} = - \sum_c p(l_c) \log p(l_c)$$

Entropy minimization

$$\mathcal{L}_{sp} = \sum_{l \in C} \sum_{c \in C} \mathbf{1}[l \neq c] (1 - \cos(\bar{v}_c, \bar{v}_l))$$

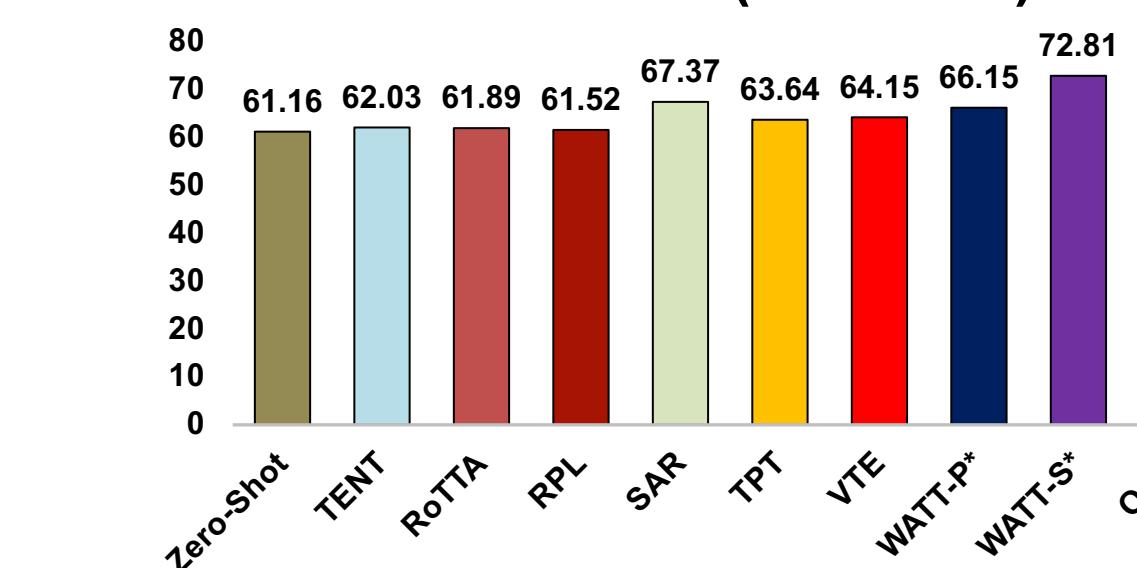
Maximize inter-class separability of prototypes to obtain good on-the-fly decision boundaries and generalization.

Overall, BATCLIP maximizes vision-text alignment and inter-class separability for online TTA under severe domain shifts.

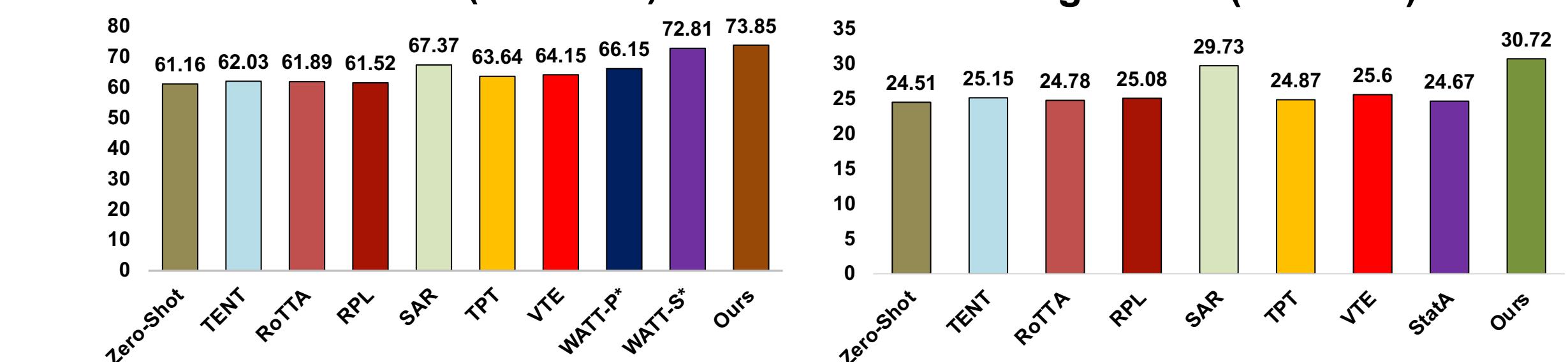
Results and Discussions

- Mean accuracy (%) across 15 domains/tasks of image corruptions.

CIFAR-10C (ViT-B/16)

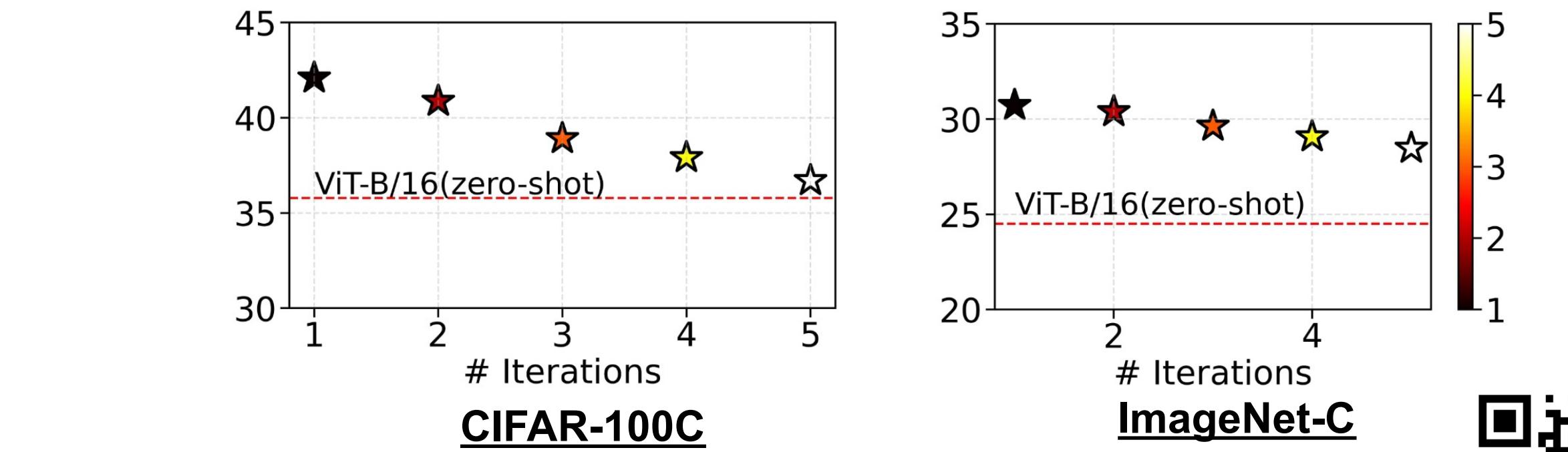
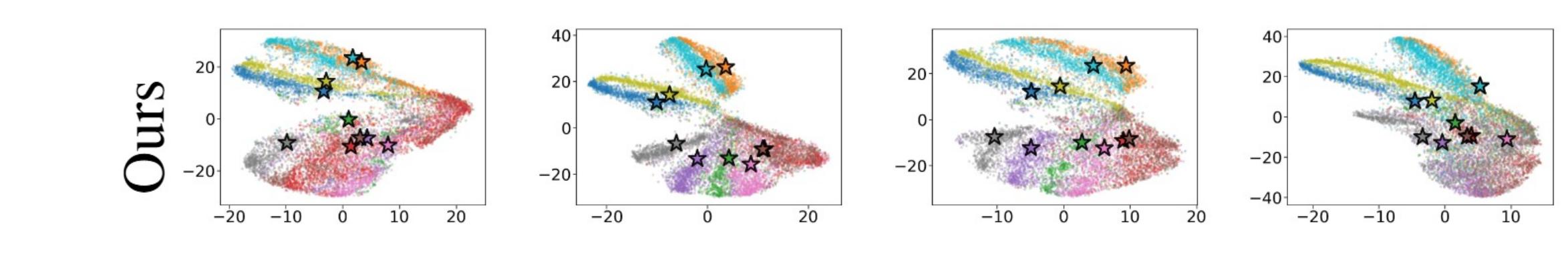
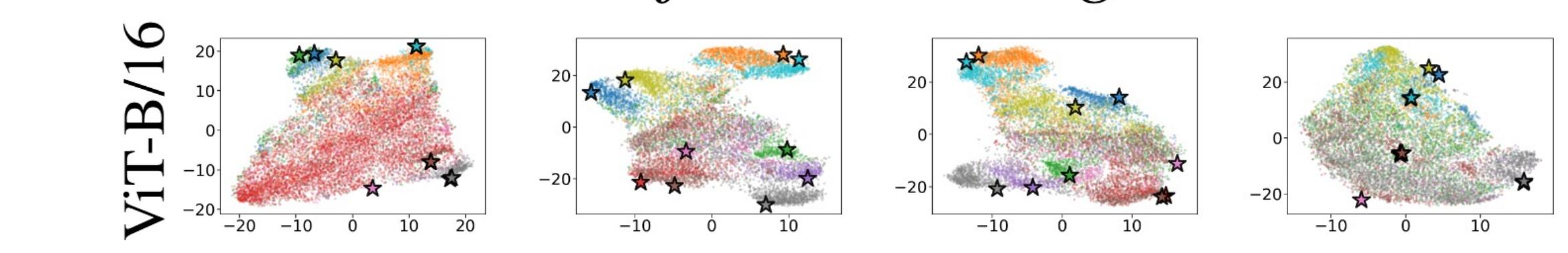


ImageNet-C (ViT-B/16)



BATCLIP yields more discriminative visual features that exhibit stronger alignment with their corresponding text features – with just one adaptation step.

Gaussian Defocus Fog Pixelate



Adaptation for multiple iterations on a single test batch >> zero-shot CLIP.

For more results on complex domains including lighting conditions, camera types →

