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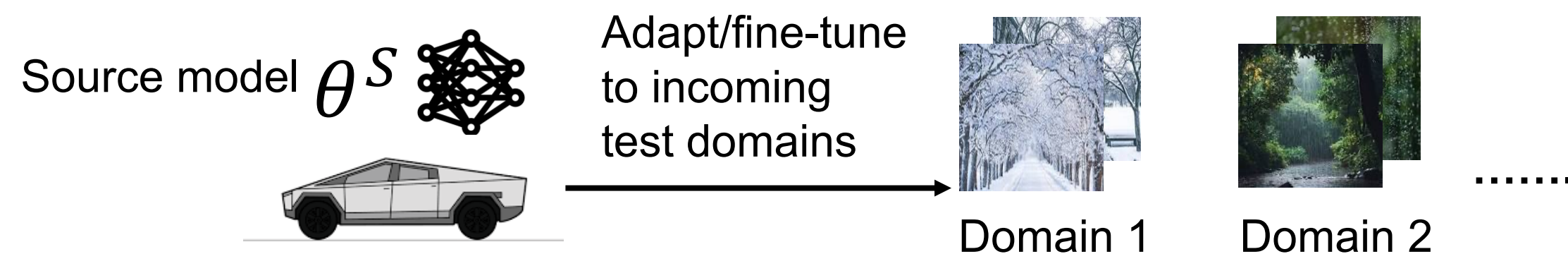
BATCLIP: Bimodal Online Test-Time Adaptation for CLIP

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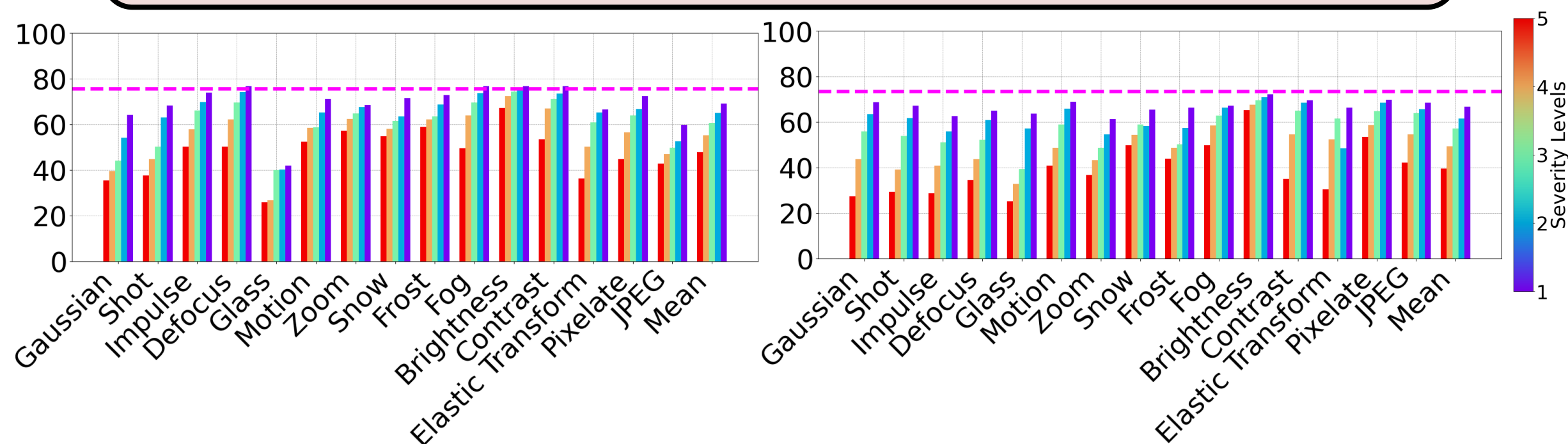


Problem Overview and Motivation

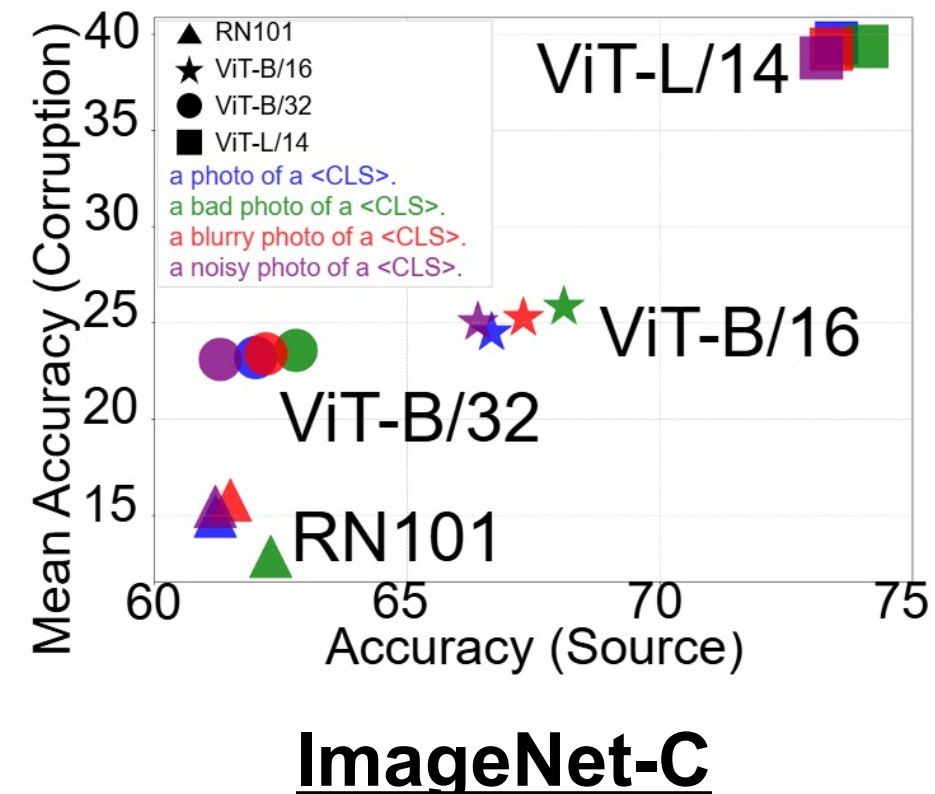
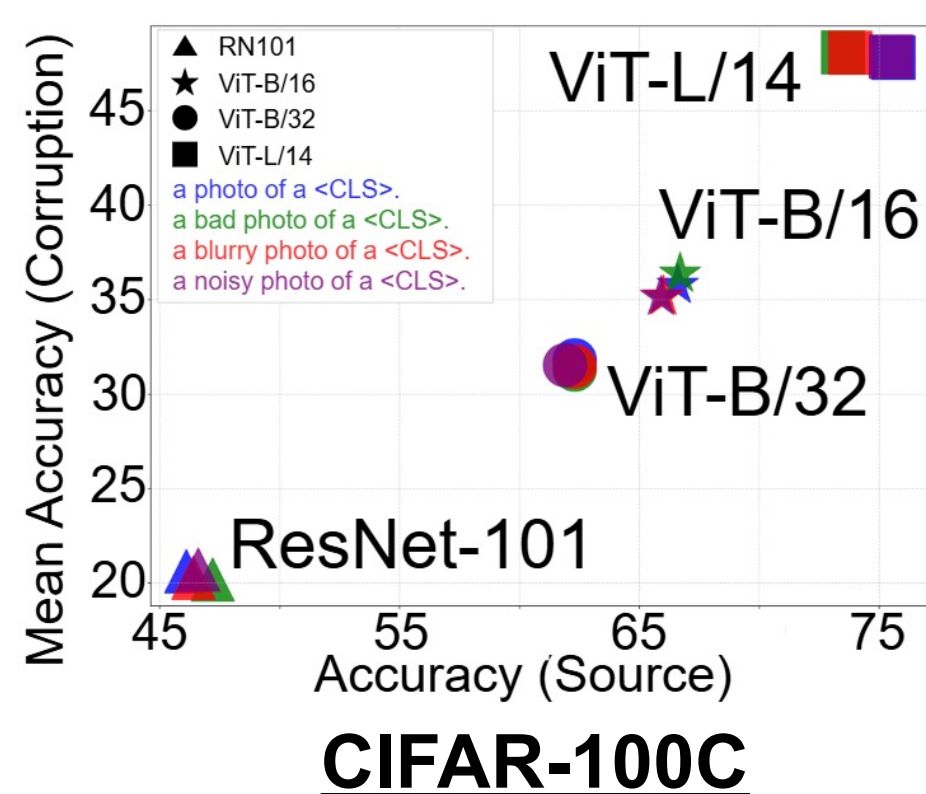
- **Online test-time adaptation (TTA)** involves pre-trained/source model adaptation to incoming **unlabeled test data** to minimize the source-target domain distribution gap.
- *Single forward pass* to preserve privacy.
- No access to the pre-training/source dataset.



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

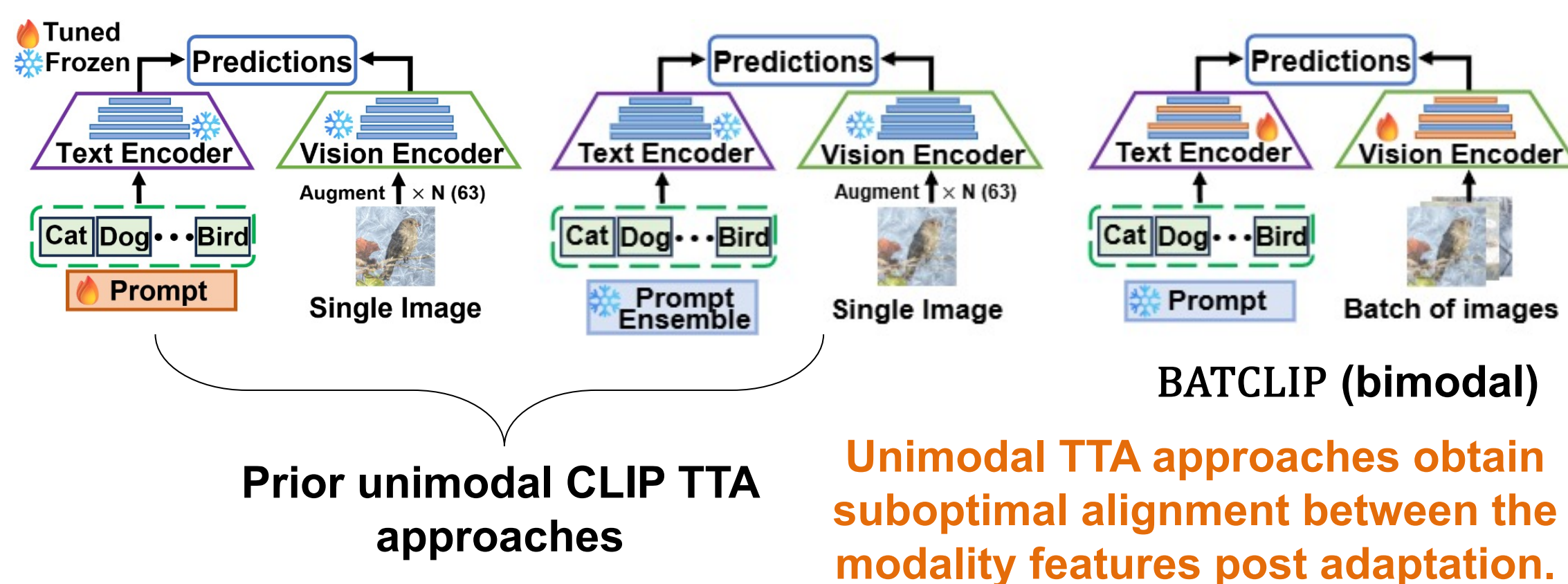


Across backbones, CLIP is very sensitive to severe visual corruptions (up, ViT-L/14); at test-time, “relevant” prompting doesn’t help (bottom). So, there’s a need for minimal adaptation as text and visual features are independent.

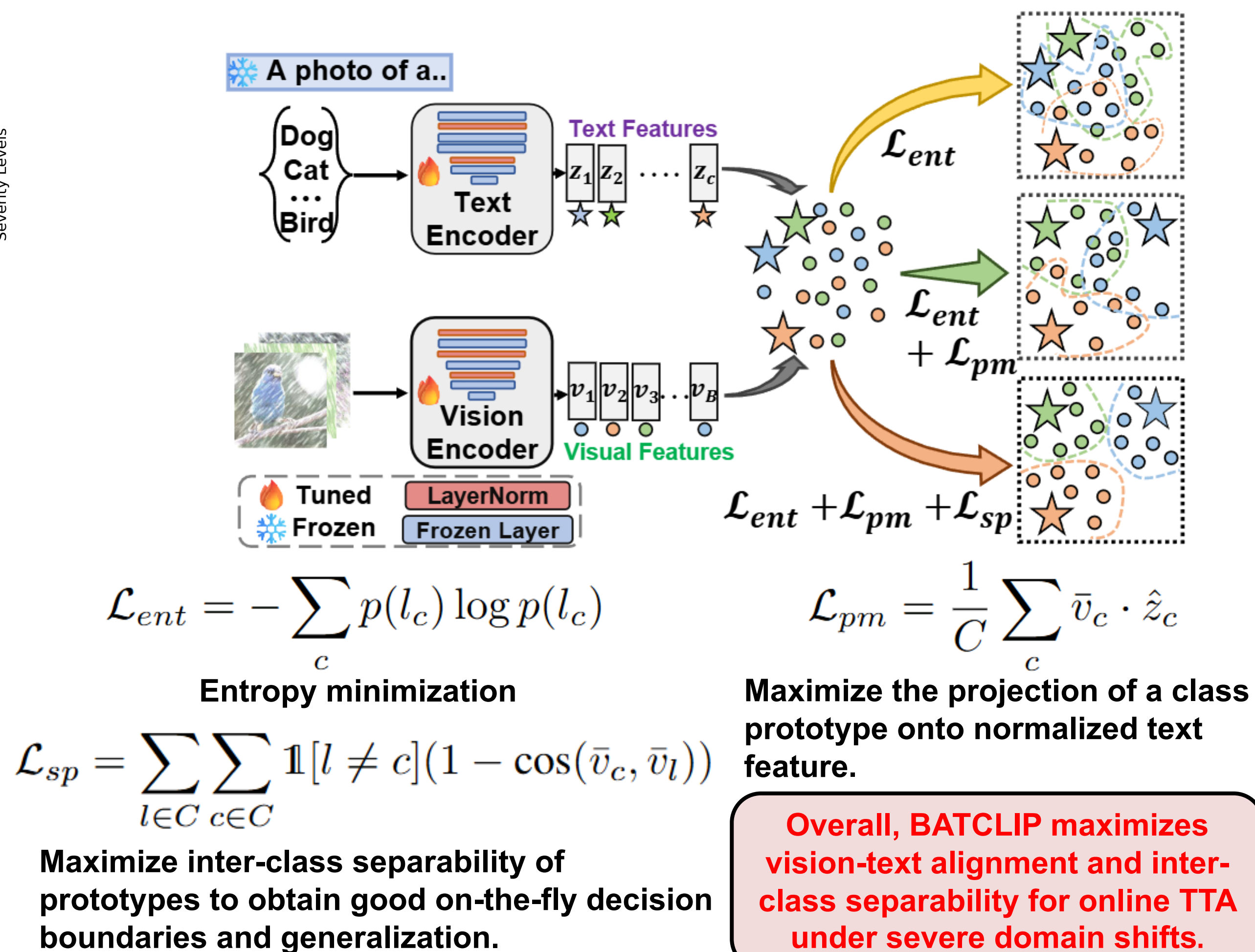


Proposed Methodology

- Existing online TTA approaches using CLIP vs ours.

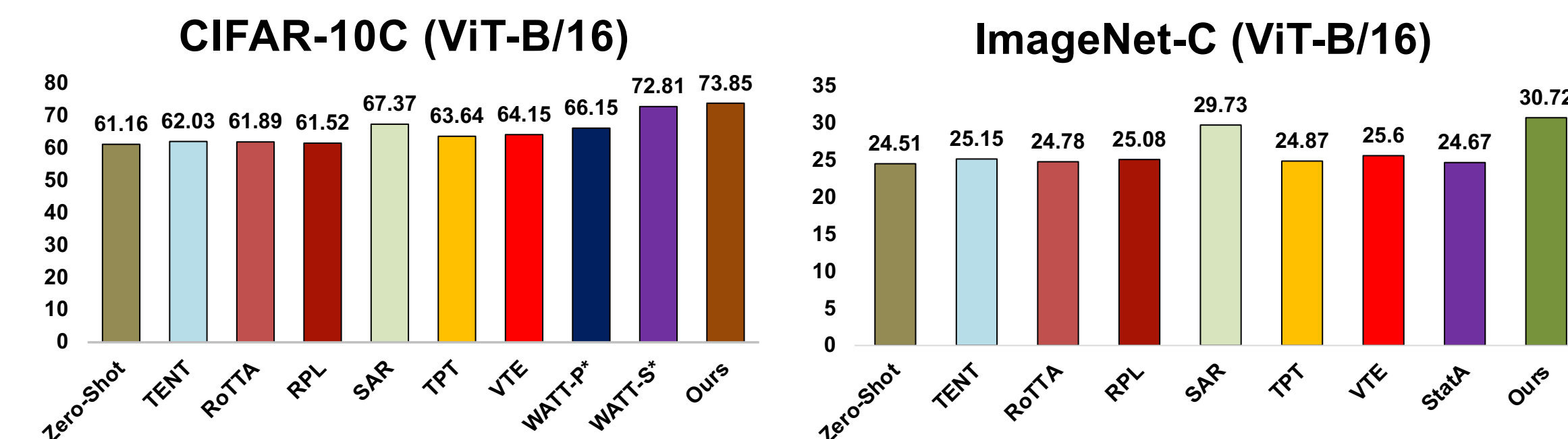


- Ours for **online** CLIP (or any contrastively pre-trained VLM) TTA

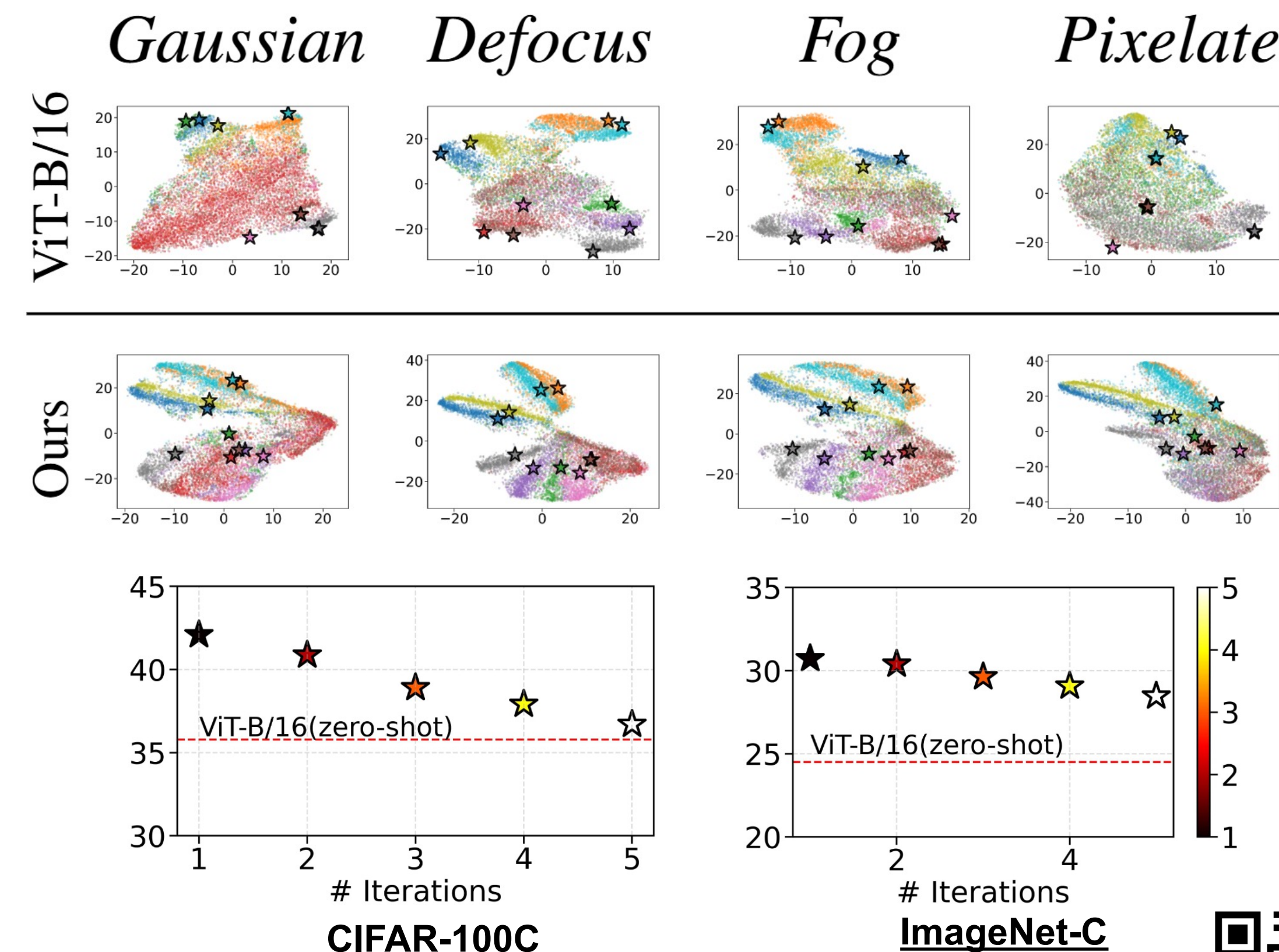


Results and Discussions

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



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



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

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

