Time-Efficient Sparse and Lightweight Adaptation for Real-Time Mobile Applications

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Background

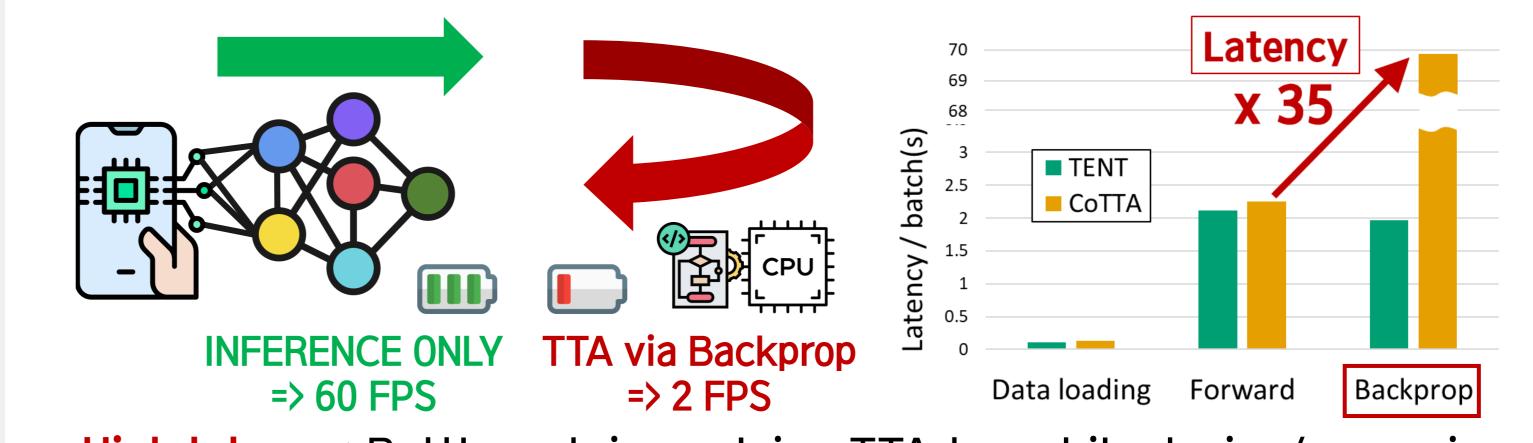
- Deep learning models on mobile applications often suffer from domain shifts
 - Lighting changes / Sensor noises resulting
 - from different weather conditions or time

Original Gaussian Noise Shot Noise





Time-Efficient TTA Suitable for Mobile App is Needed

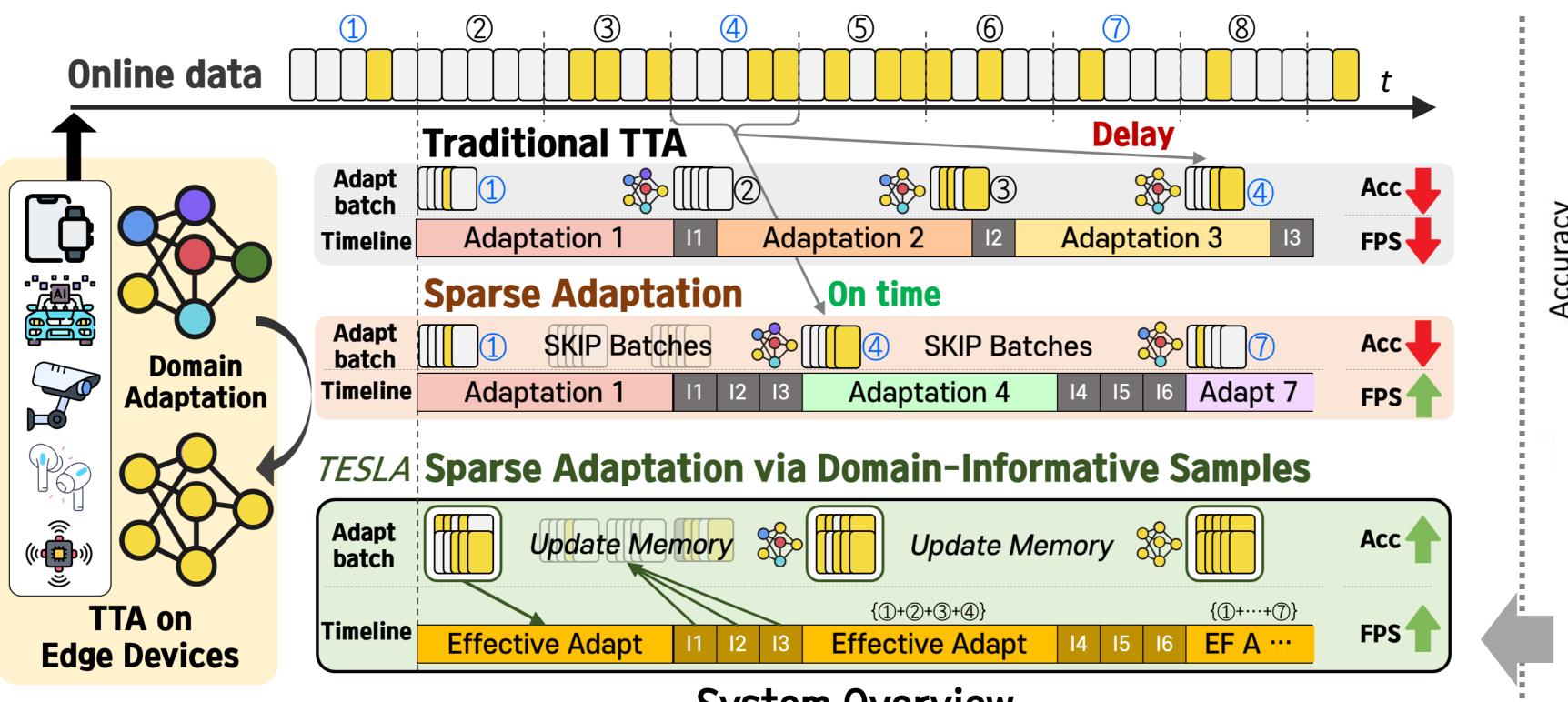


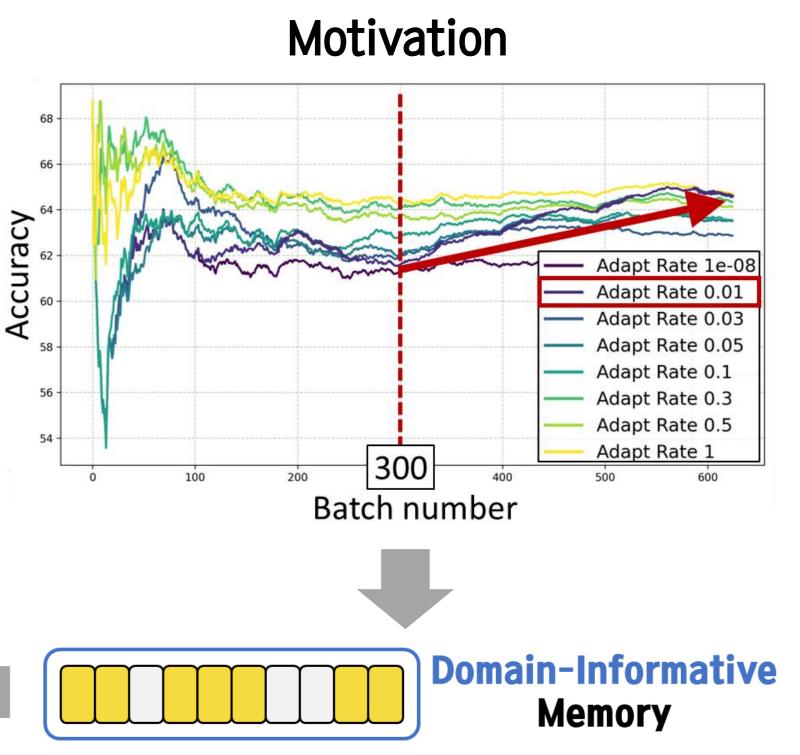


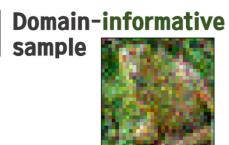


- Test-Time Adaptation (TTA) rapidly adapts models without any source or labeled data
- High latency: Bottleneck in applying TTA to mobile device/scenario
 - Backpropagation, Augmentations, Teacher-Student Models …
- State-of-the-art TTA algorithms have been designed and evaluated mainly on GPU servers, focusing on improving accuracy

Sparse Adaptation Framework : Strategically Skip Batches and Effectively Update Model







Domain-irrelevant

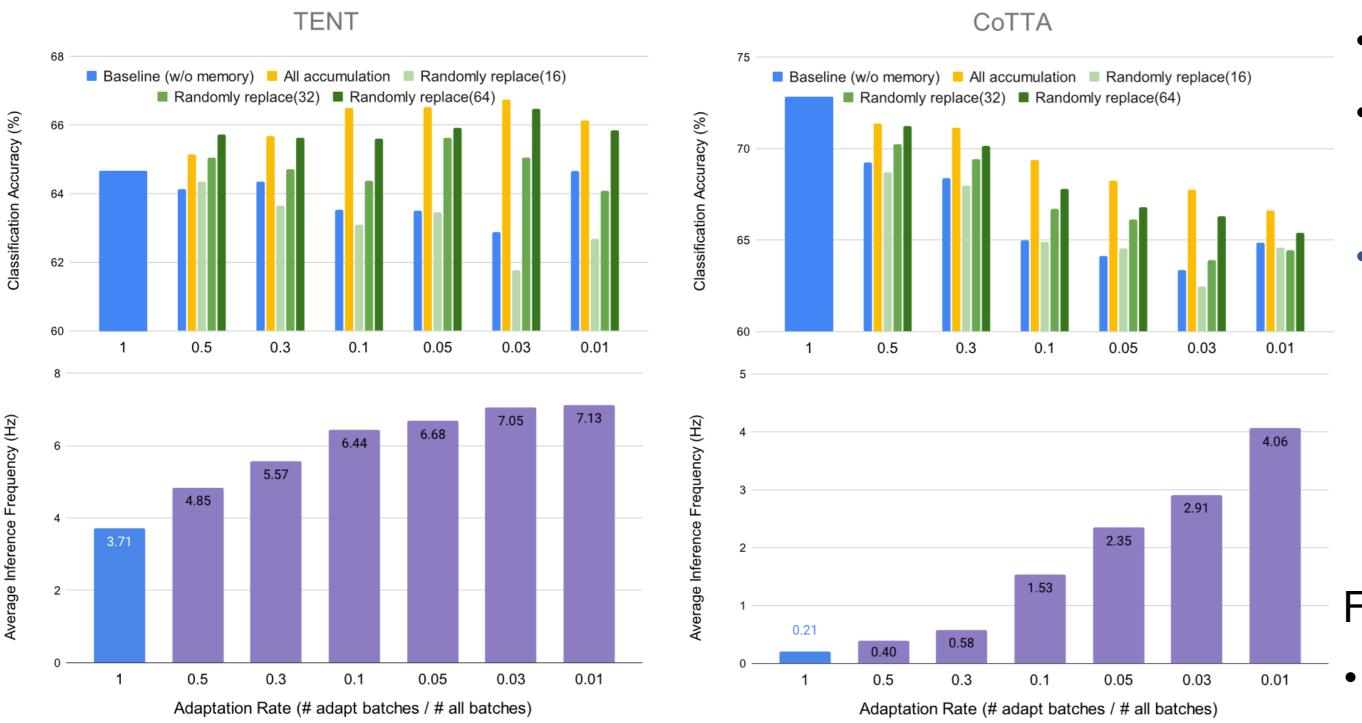
sample

System Overview

- Sparse adaptation: Skip adaptation boldly and compensate for it
- by strong update via domain-informative samples
- By strategically controlling the ADaptation Rate (ADR), our system balances inference fps and model accuracy
- Extremely sparse ADR of 0.01 can achieve
- competitive accuracy even against 1.
- *⇒ Some samples can greatly contribute*

to domain adaptation loss.

Experimental Results & Discussions



- [FPS] Average inference speed improves up to 20x
- [Acc] Sparse adaptation with memory can achieve even higher accuracy than adapting every batch (baseline)
- Seamlessly integrate with existing lightweight adaptation and optimization algorithms, further accelerating inference across diverse mobile systems

TESLA: enabling efficient and effective TTA for

resource-constrained real-time mobile applications Future works

• Memory optimization: maintaining a large buffer is impractical







