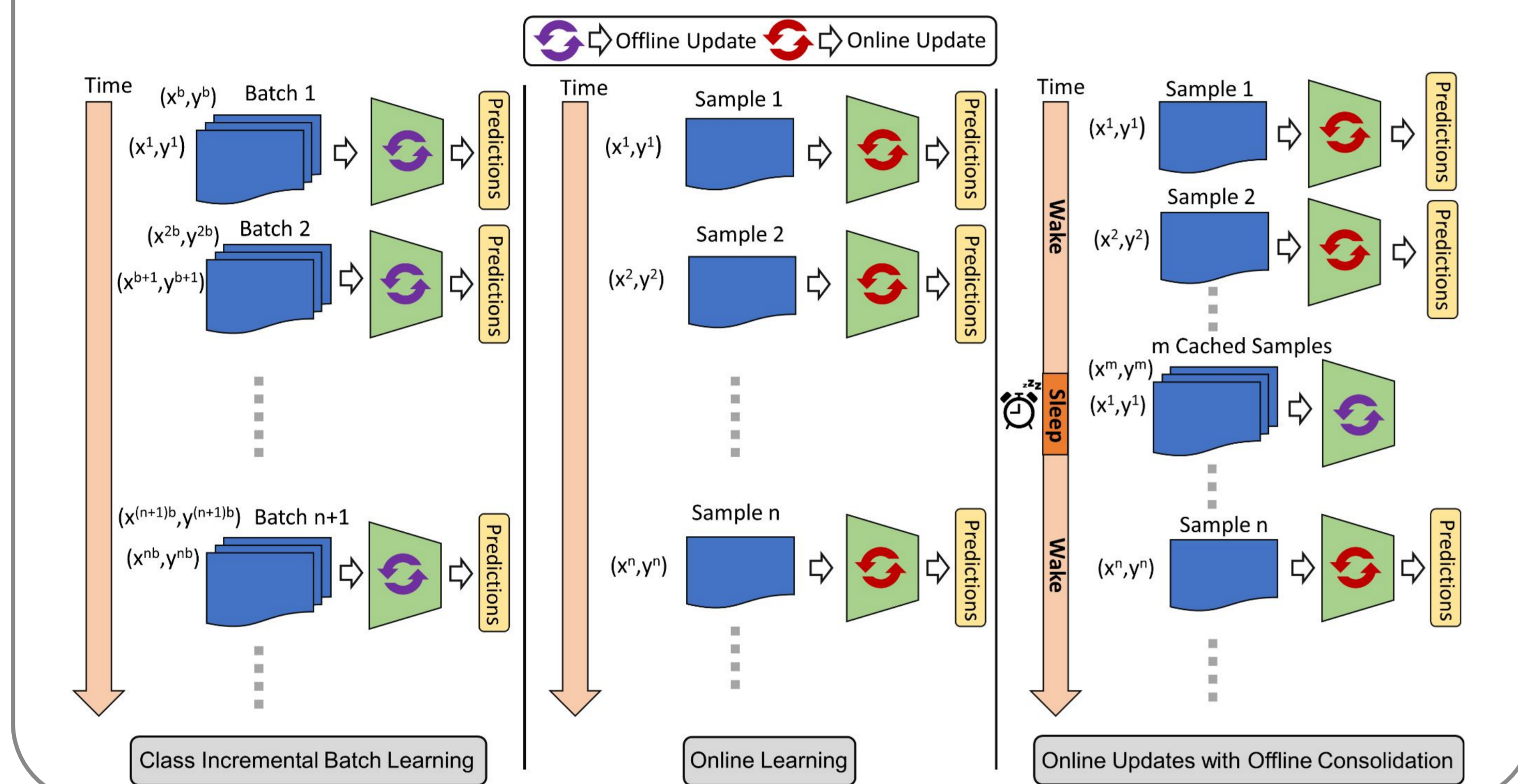


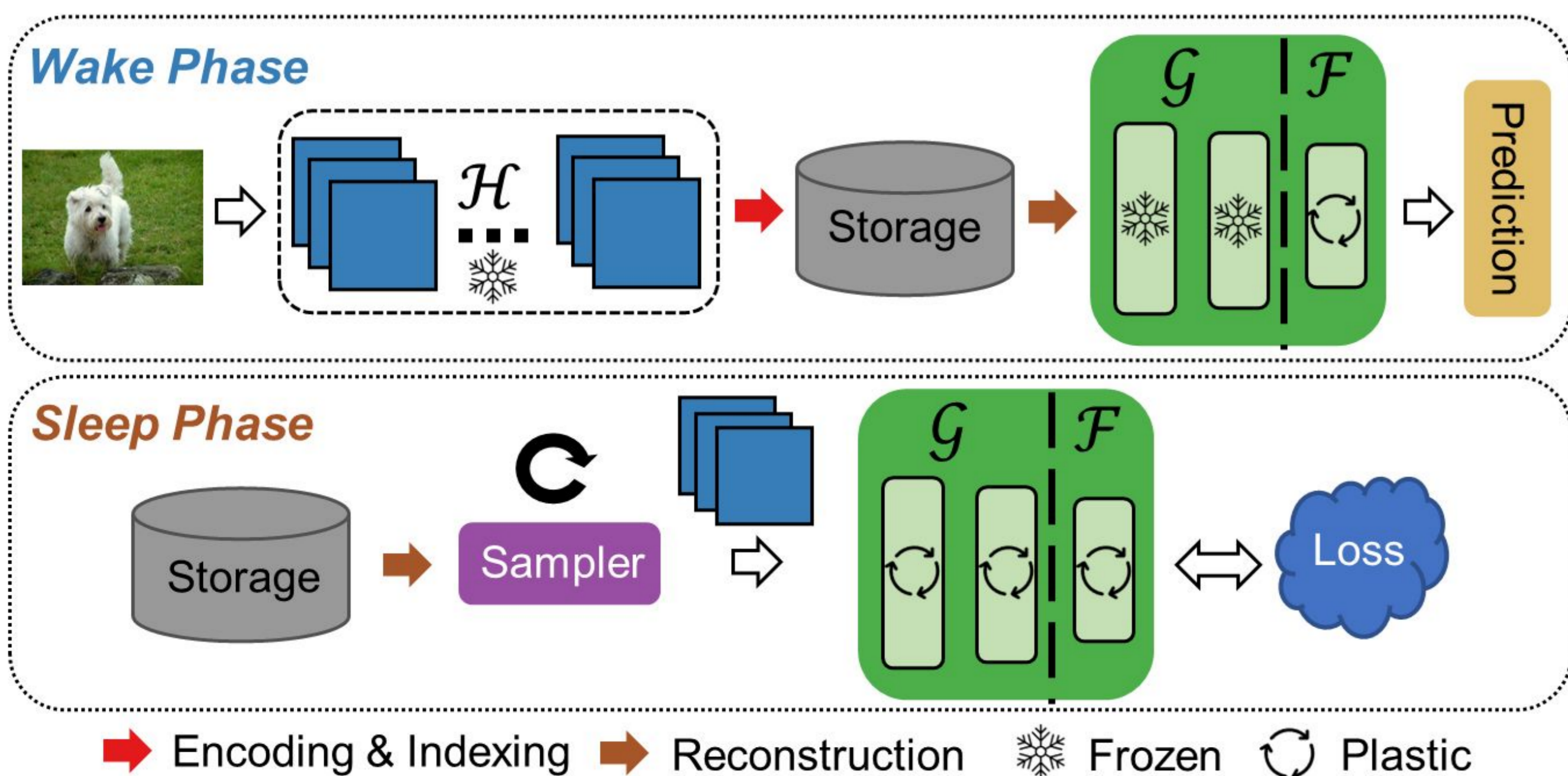
Overview

- Continual learning (CL) research solely focused on **catastrophic forgetting** overlooking **compute efficiency**.
- To make a real-world impact, CL must provide compute efficiency and **rival batch learners** trained from scratch.
- We propose a novel method, **SIESTA** based on a **wake/sleep** framework for rapid online CL. **SIESTA performs CL on ImageNet-1K in under 2 hours**.
- We propose a **new paradigm** combining offline and online CL, facilitating many applications e.g., on-device CL.

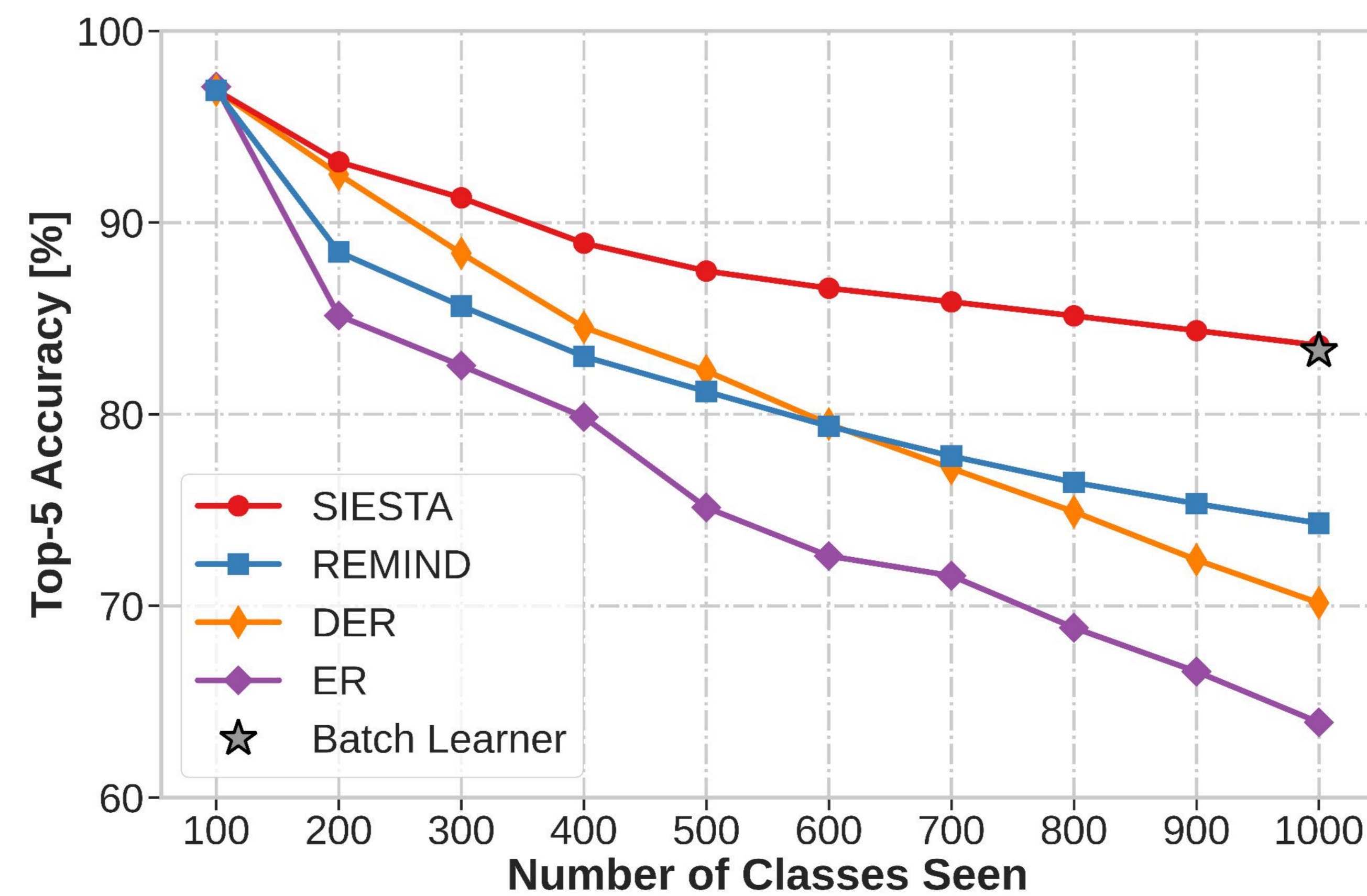


How Does SIESTA Work?

During the **wake phase**, SIESTA transforms raw inputs into feature representations using H and stores them in a buffer in compressed form. F is updated in an online manner with running class means. During the **sleep phase**, a sampler retrieves examples from the buffer to reconstruct and update G & F with backpropagation.

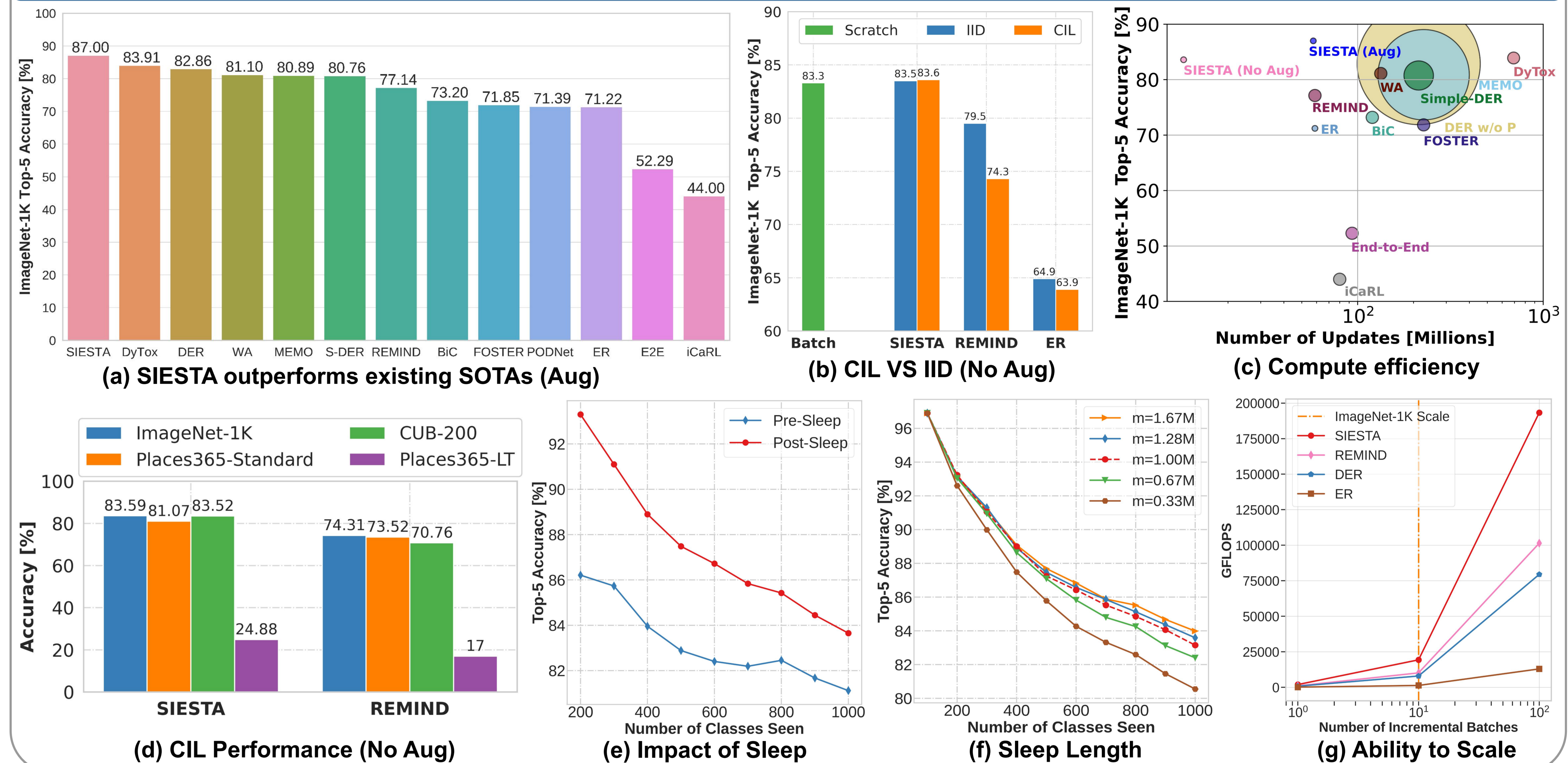


SIESTA Achieves Zero Forgetting



SIESTA matches offline batch learner and outperforms ER, DER, and REMIND in CIL on ImageNet-1K.

Results



Summary

- SIESTA outperforms SOTA CL methods while requiring **7x-60x fewer updates**, **10x less memory**, and **2x-20x fewer parameters**.
- Compared to REMIND, SIESTA is **3.4x** (ImageNet-1K) and **4.4x** (Places365-Standard) faster.
- Unlike others, SIESTA maintains similar performance across data distributions e.g., CIL and IID CL.
- SIESTA demonstrates the benefits of sleep for memory consolidation and learning efficiency.
- SIESTA rivals offline batch learners and achieves **“Zero Forgetting”** in large-scale CL.
- SIESTA unlocks opportunities for many real-world applications and has potential to advance Green AI.

Acknowledgements

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