

The Challenges of Continuous Self-Supervised Learning

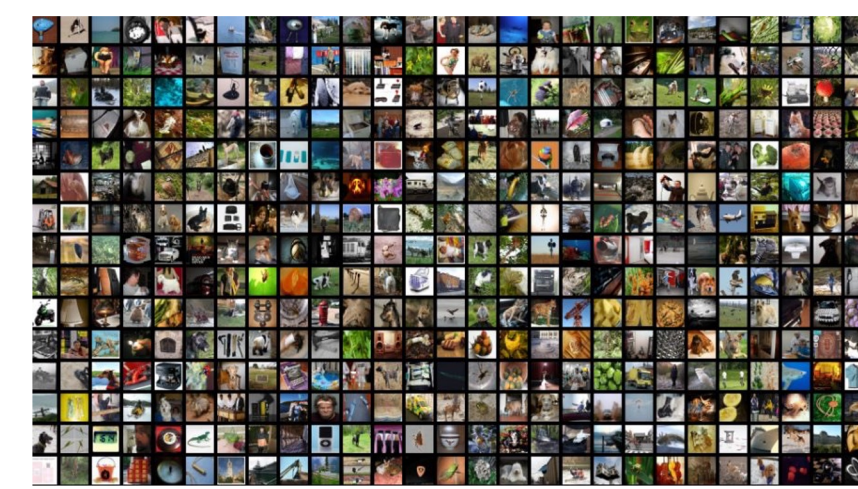
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<http://www.senthilpurushwalkam.com/publication/continuousssl/>

Carnegie Mellon University

The Robotics Institute

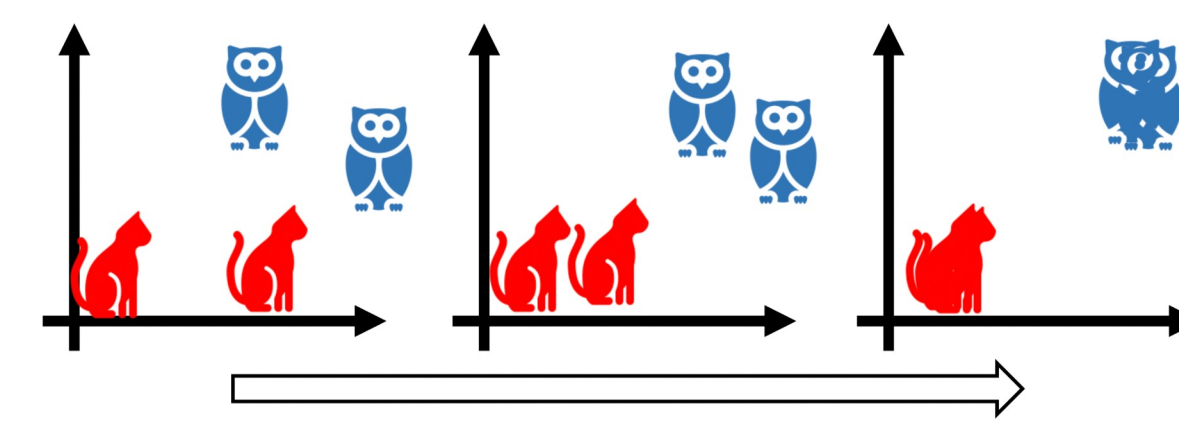
Conventional SSL



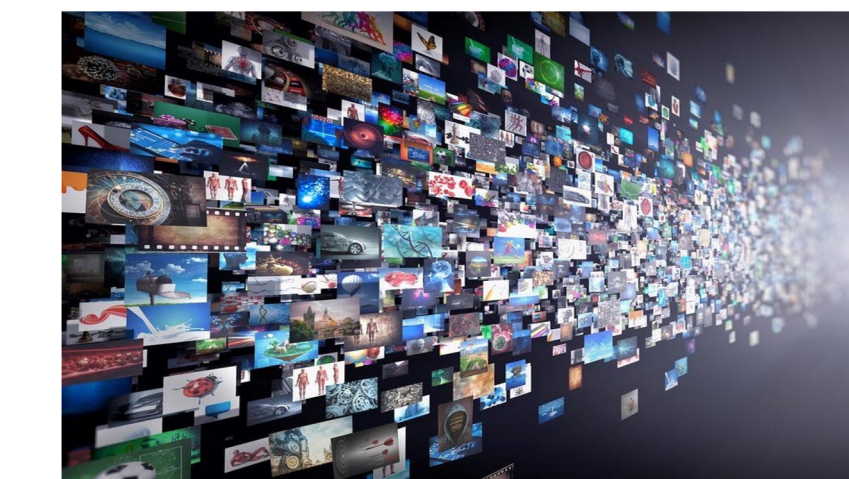
Finite Data



IID Data



Stationary Semantics

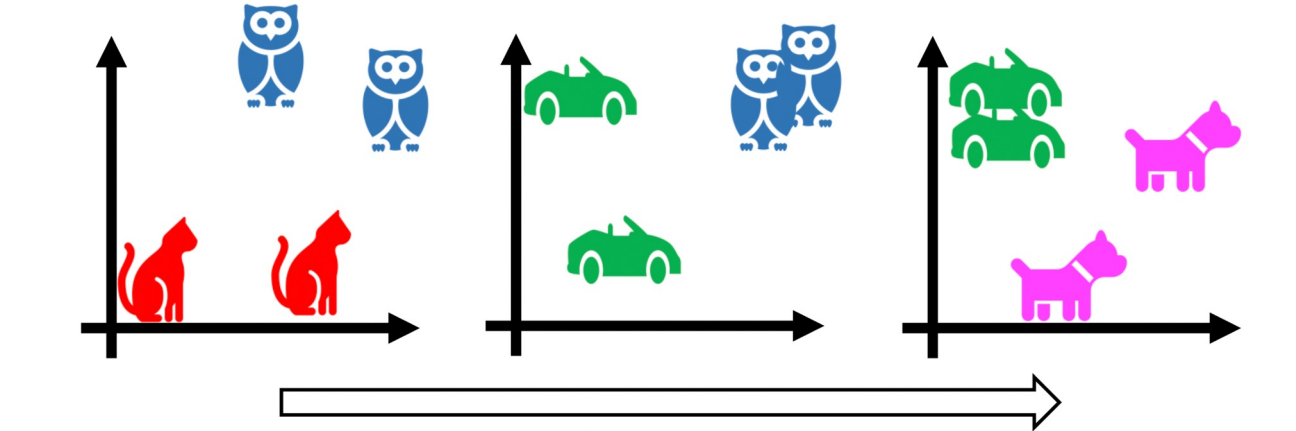


Infinite Data

Continuous SSL



Non-IID Data



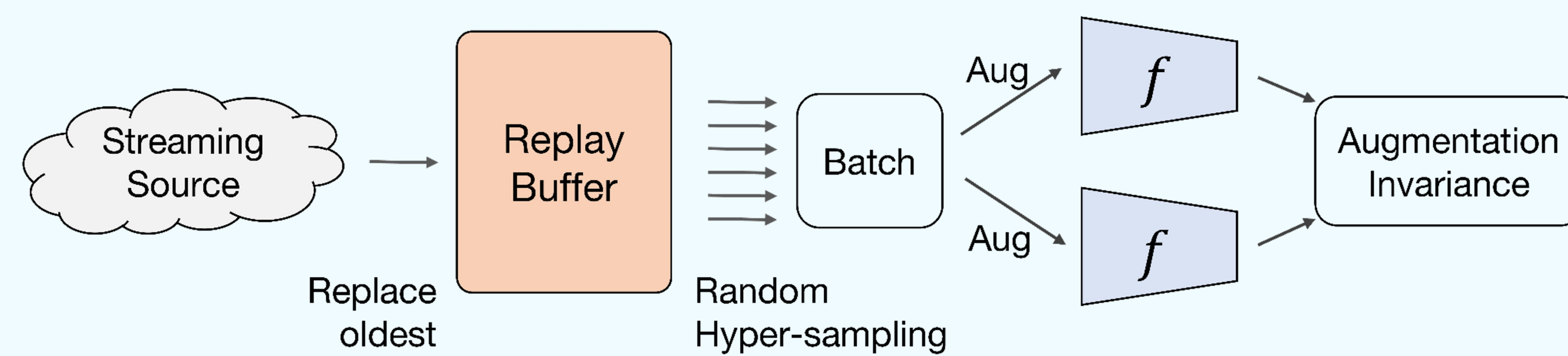
Non-stationary Semantics

Infinite Data

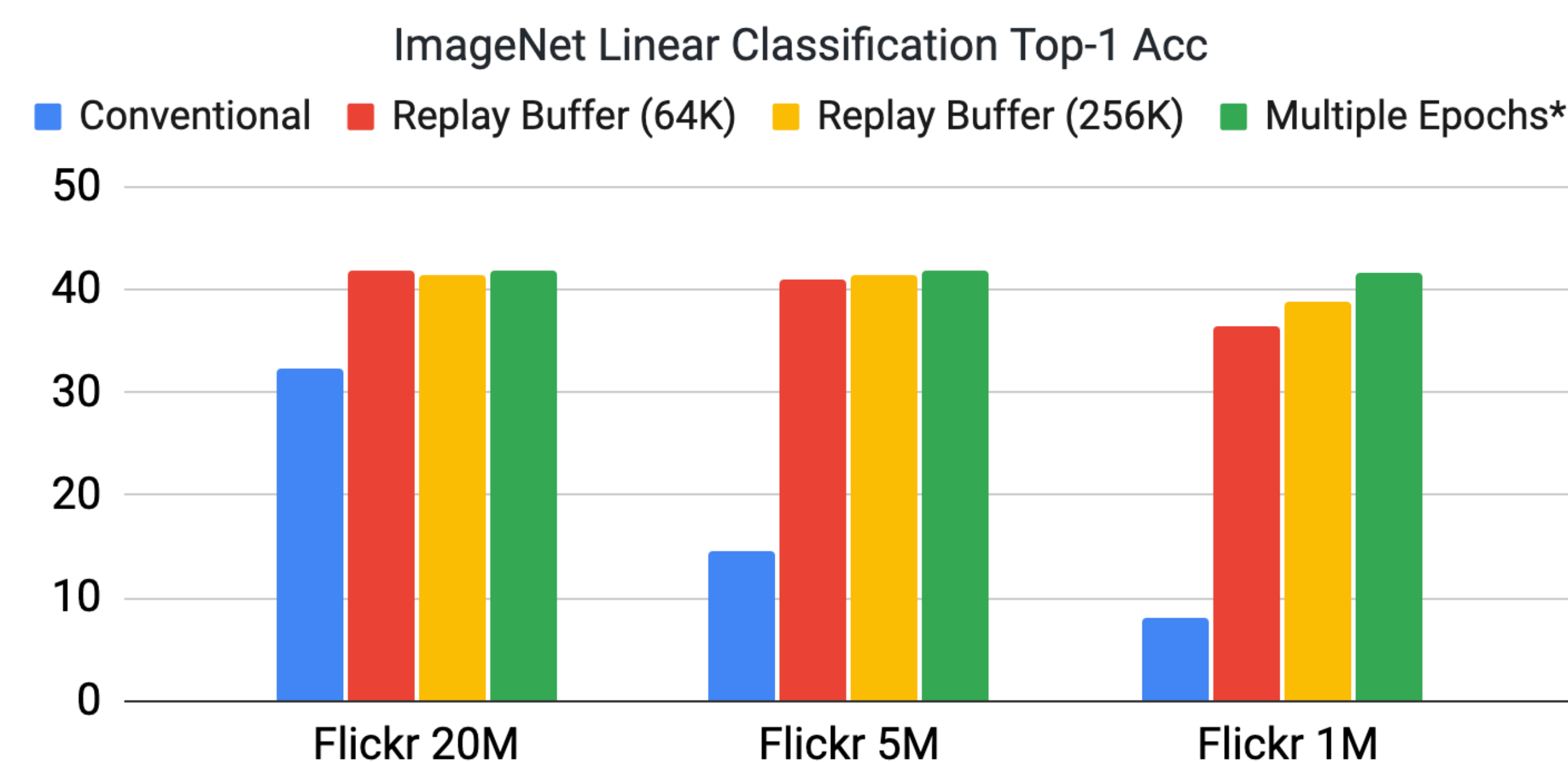
⇒ **Single Pass Training**

Data: Images sampled from 

Solution: Replay Buffers



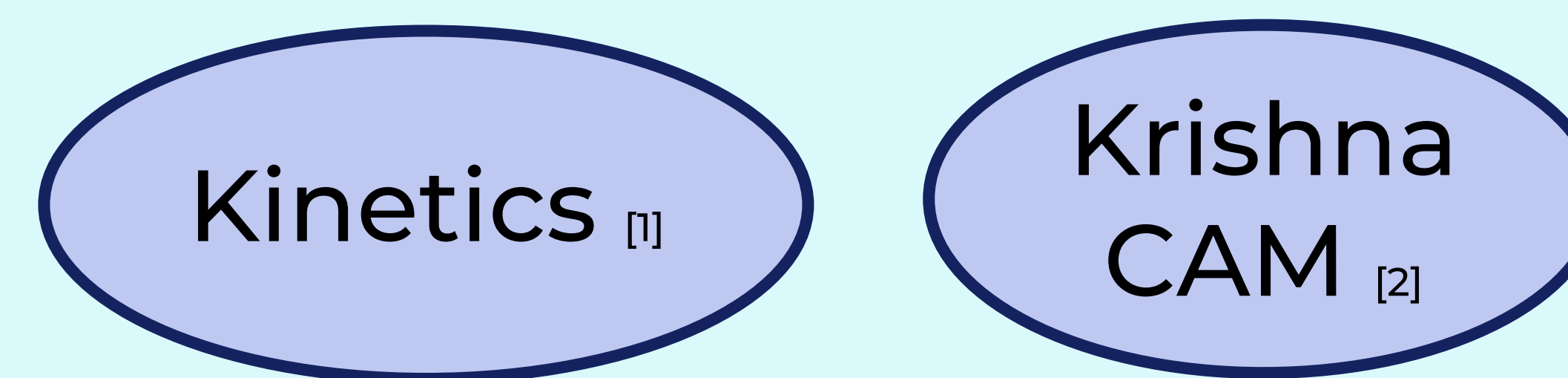
Allows re-using samples from streaming source



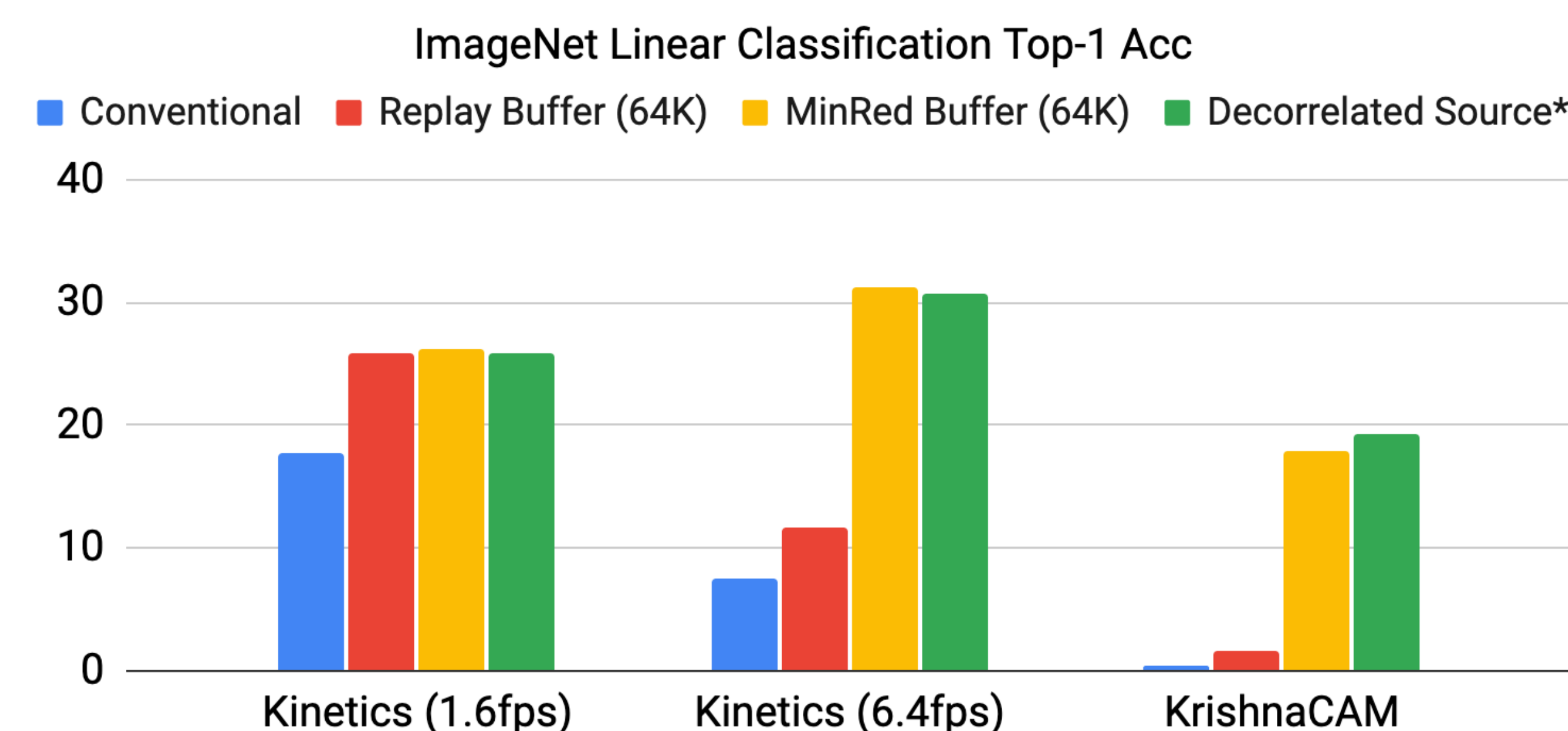
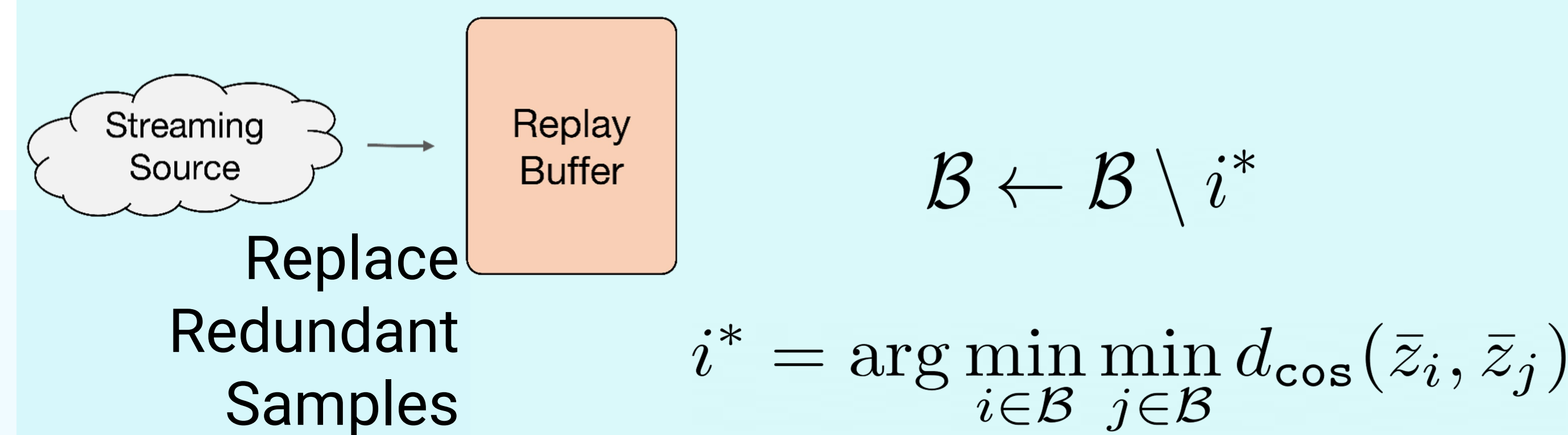
Non-IID Data

⇒ **Poor Optimization**

Data: Sequential frames from videos



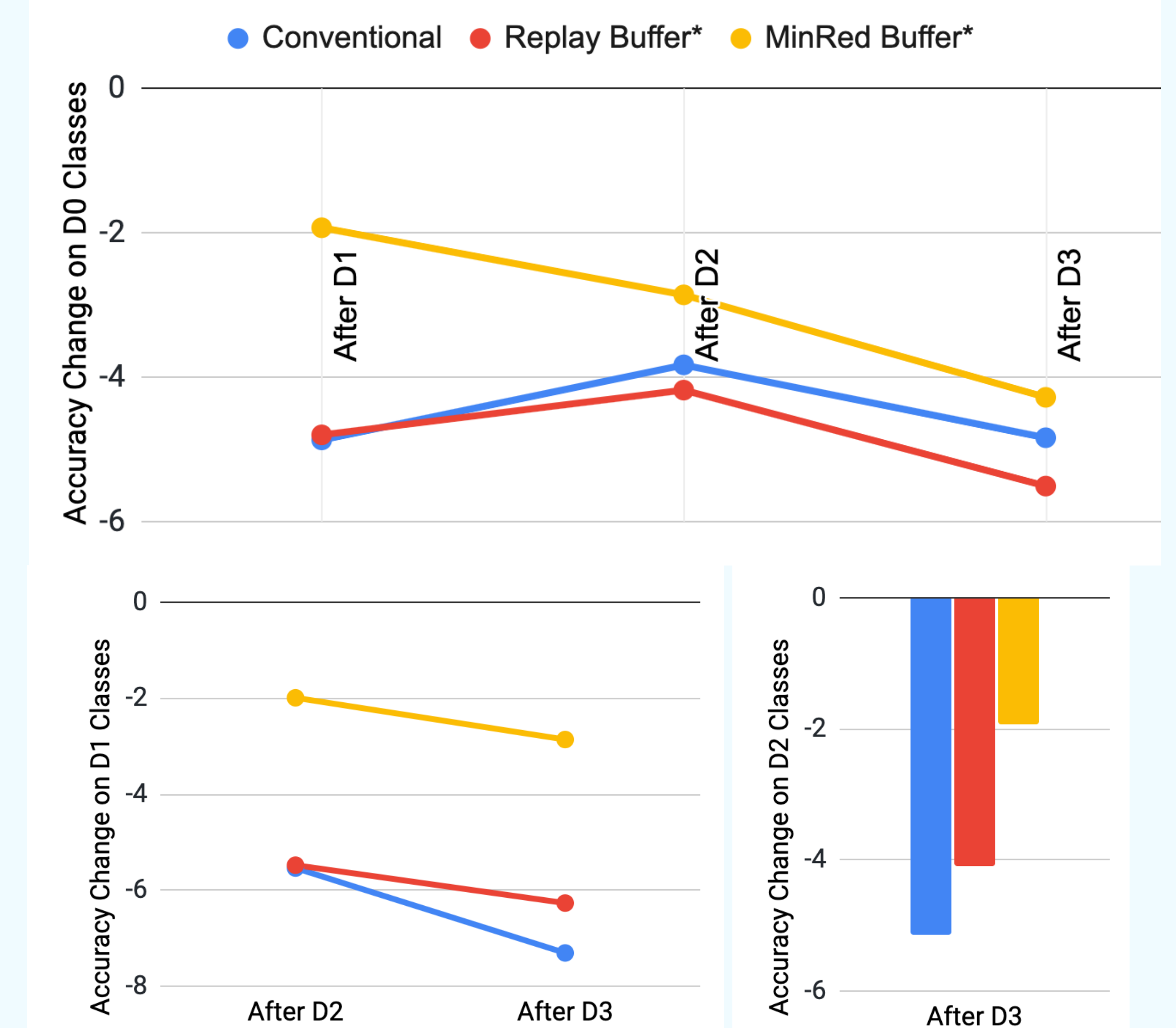
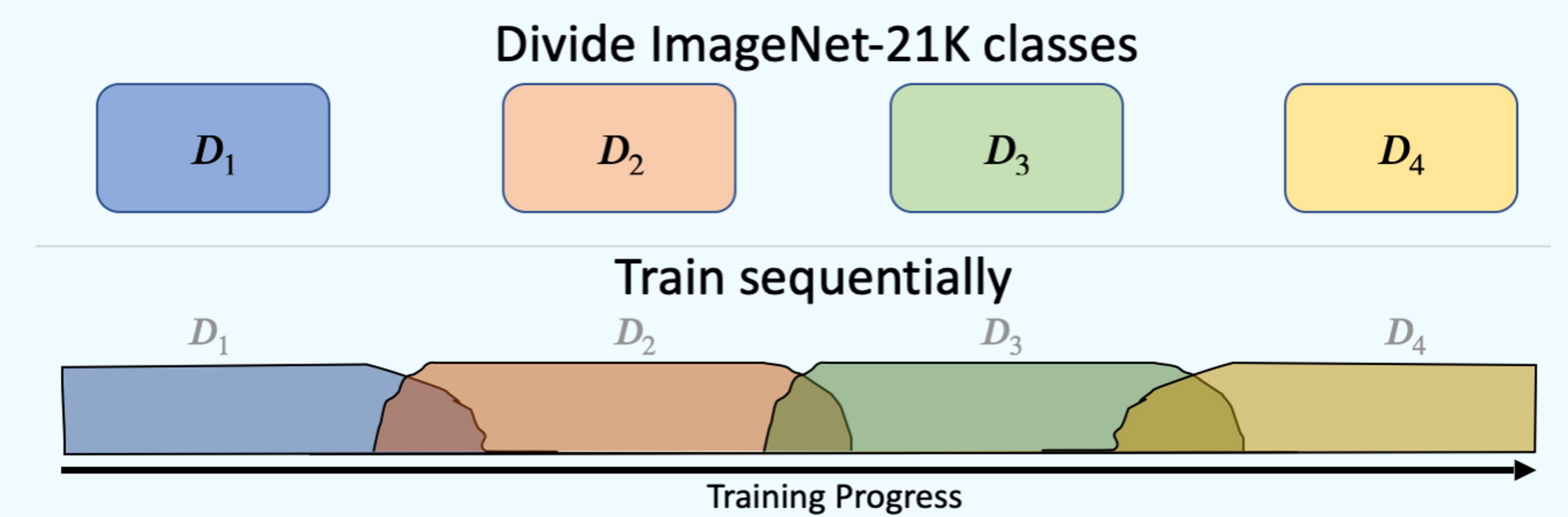
Solution: Minimum Redundancy Buffers



Non-stationary Semantics

⇒ **Catastrophic Forgetting**

Data:



1. Carreira, J., Zisserman, A.: Quo vadis, action recognition? a new model and the kinetics dataset. In: Proceedings of the IEEE/CVF Conference on Computer Vision and Pattern Recognition. pp. 6299–6308 (2017)
 2. Singh, K.K., Fatahalian, K., Efros, A.A.: Krishnacam: Using a longitudinal, single-person, egocentric dataset for scene understanding tasks. In: 2016 IEEE Winter Conference on Applications of Computer Vision (WACV). pp. 1–9. IEEE (2016)
 3. Chen, X., He, K.: Exploring simple siamese representation learning. In: Proceedings of the IEEE/CVF Conference on Computer Vision and Pattern Recognition. pp. 15750–15758 (2021)