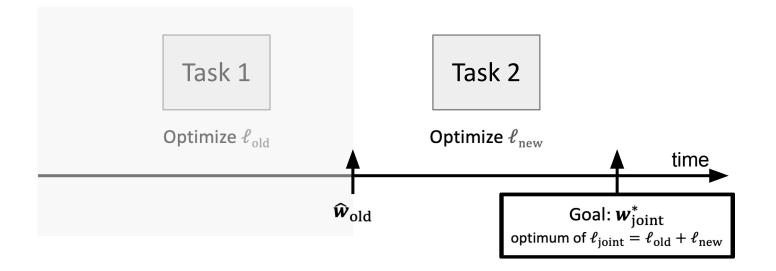
#### Research Project 2024/2025 Q4

# The Stability Gap in Continual Learning with Deep Neural Networks

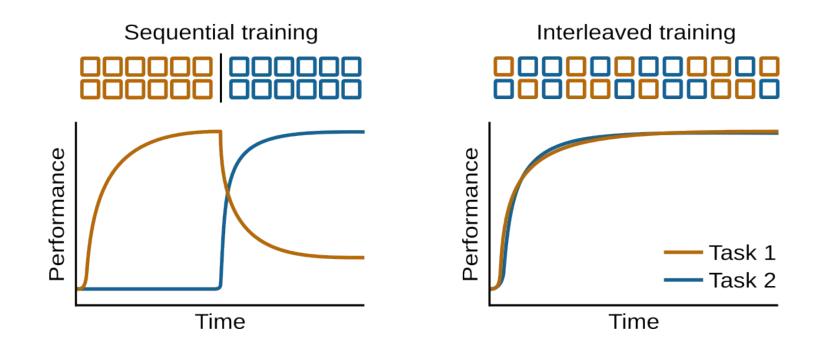
Gido van de Ven (Supervisor), Tom Viering (Responsible Professor)

# The continual learning problem

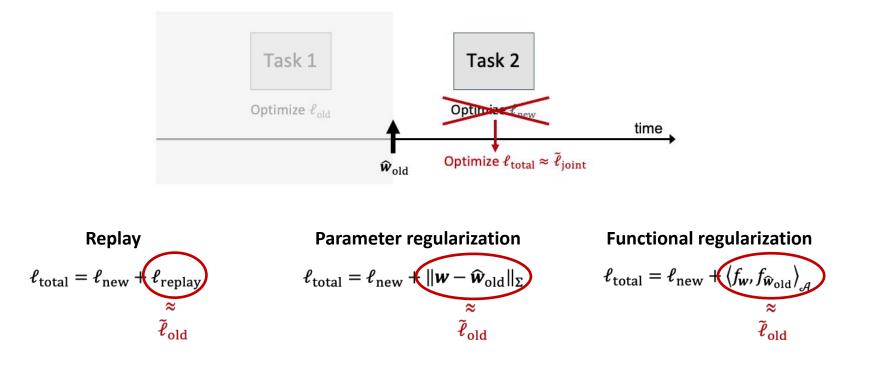
→ Optimize the parameters w of a neural network  $f_w$  for two tasks that are observed one after the other



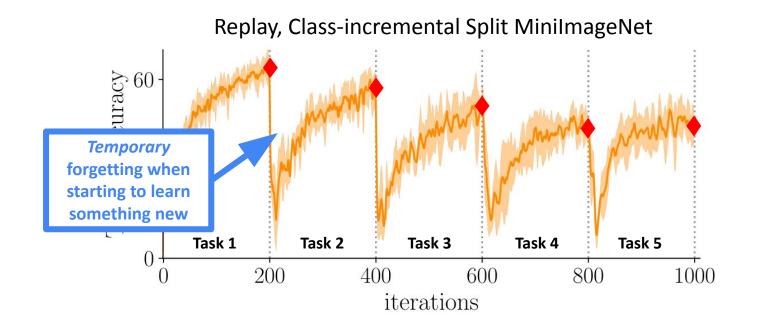
## Catastrophic forgetting



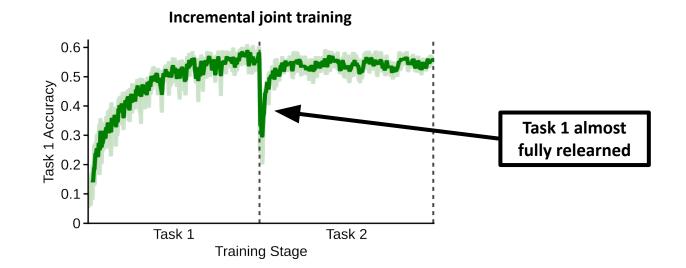
# Current approach to continual learning: make changes to the loss



### Does replay prevent forgetting?



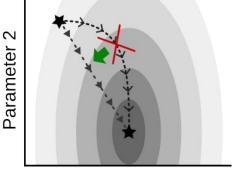
## The stability gap occurs even with "perfect" replay!



- Problematic for safety-critical applications
- Seems highly inefficient

## Continual learning needs a new perspective: **optimization**

- To overcome the stability gap, changes must be made not only to *which* loss function is optimized, but also to *how* it is optimized
- Standard optimization routines for deep learning have been developed for the stationary setting
- No guarantees in continual setting, yet widely used
- Fundamental difference between both settings:
  - Stationary  $\rightarrow$  start from random initialization
  - $\circ$  Continual  $\rightarrow$  start from partial solution



Parameter 1

#### **Research Questions**

- (1) How does learning rate affect the stability gap?
- (2) How does momentum and/or optimizer type affect the stability gap?
- (3) How does mini-batch size affect the stability gap?
- (4) To what extent can the method Layerwise Proximal Replay [Yoo et al., 2024]
  ICML] reduce the stability gap? (code available)
- (5) To what extent can using second-order optimization methods reduce the stability gap? (hint available)
- (6) **[BONUS]** Can you propose and test your own method to reduce the stability gap?

#### **Practicalities**

- You will learn to implement continual learning experiments with deep neural networks in PyTorch. (Prior experience with PyTorch not needed.)
- Code that can be used as starting point is available: <u>https://github.com/GMvandeVen/continual-learning</u>
- Relatively simple, publicly available datasets (MNIST, CIFAR) will be used