



Deep Reinforcement Learning with Plasticity Injection

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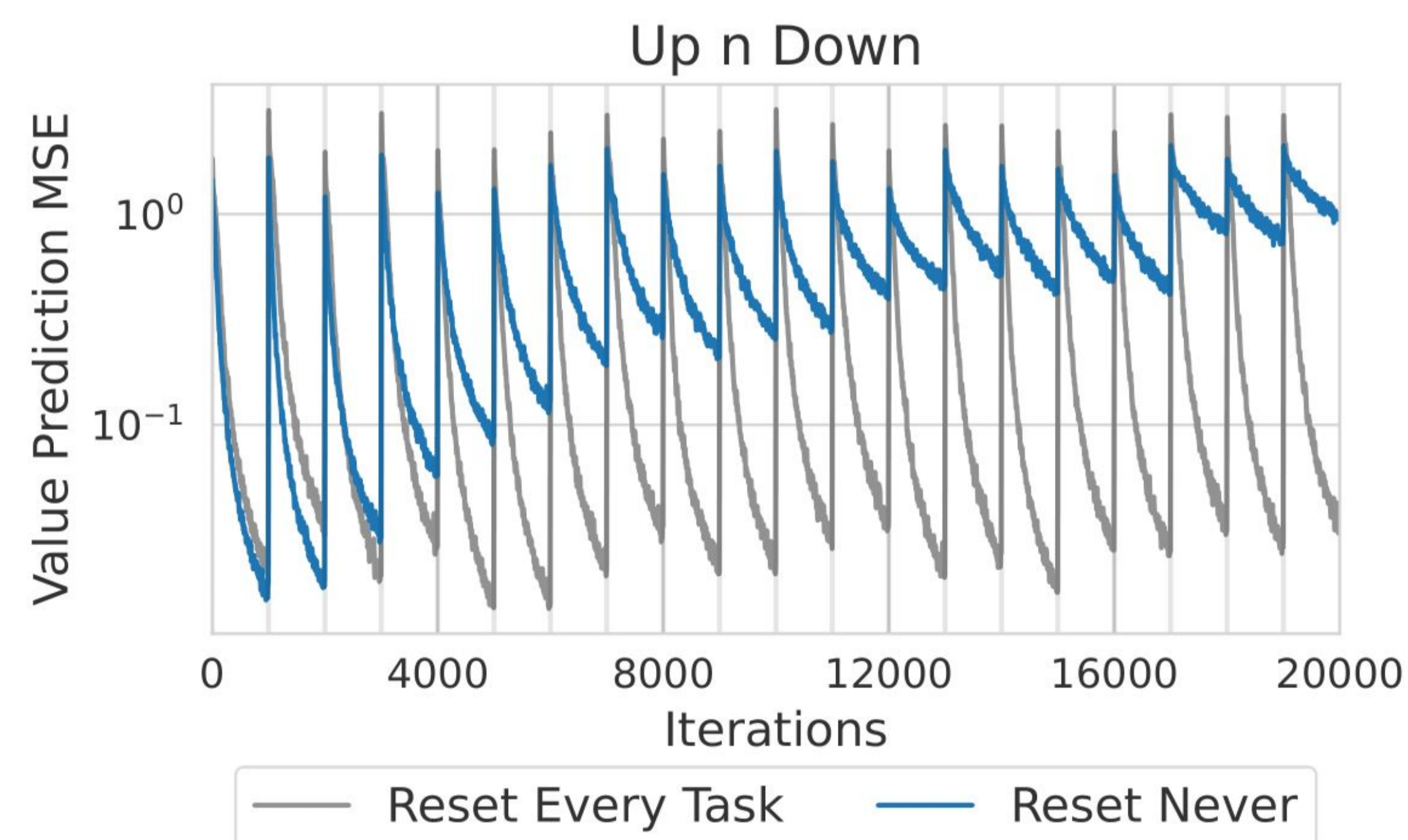


Overview

- Deep RL agents **lose plasticity** during training, but detecting it is challenging
- Plasticity Injection is a tool for careful **analysis** of the phenomenon
- Plasticity Injection **reincarnates** agents that lost plasticity without the need to re-train
- Agents with Plasticity Injection **save compute** by dynamically growing networks

Background on Plasticity Loss

On a sequence of policy evaluation tasks, a network becomes worse at predicting values



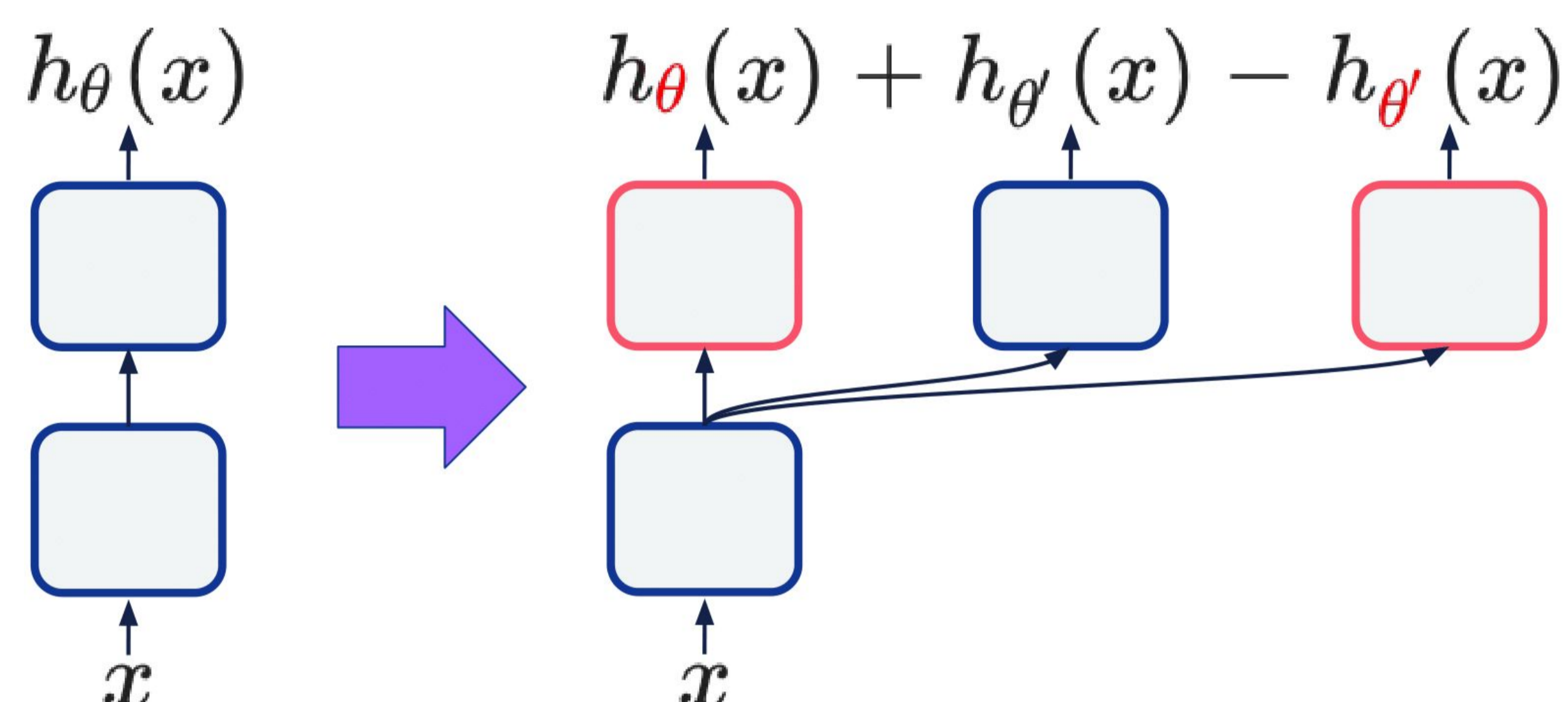
Challenges in Deep RL

- No clear measure of network plasticity
- Performance in deep RL can be bottlenecked by many factors including exploration

How to tell if an agent cannot improve due to loss of plasticity?

Plasticity Injection

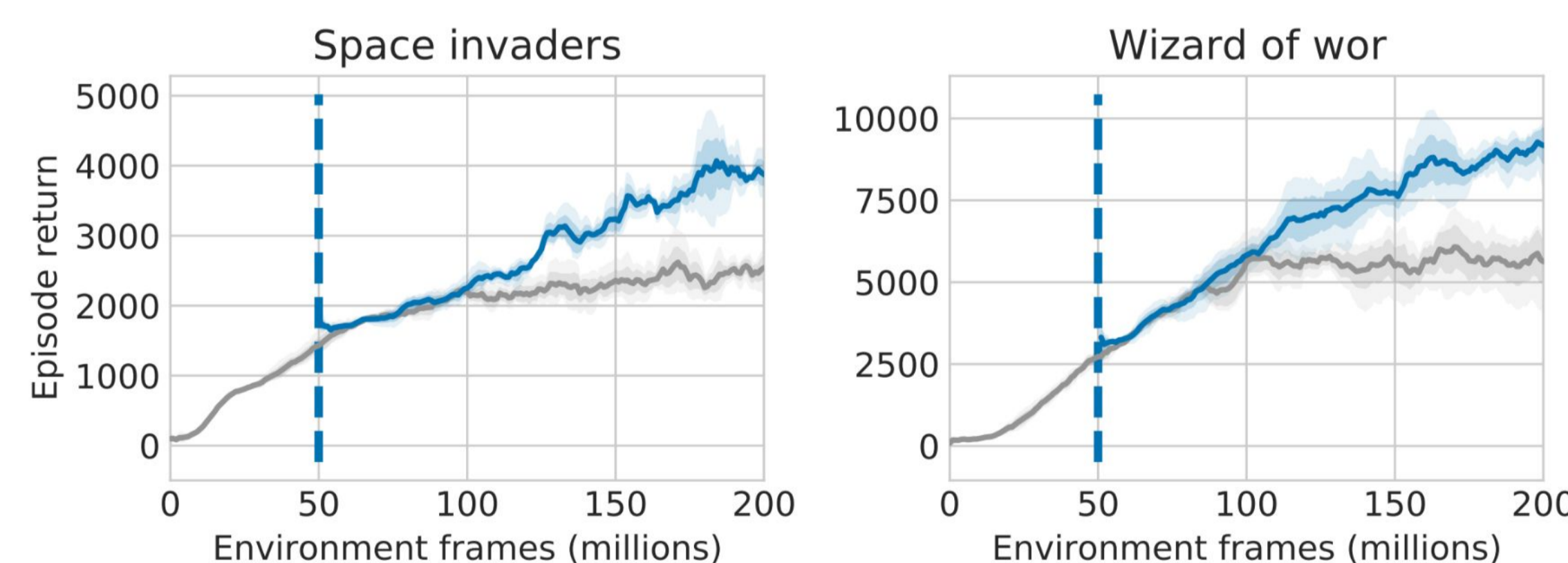
At some point, **freeze** the network and create a new, learning a residual to predictions



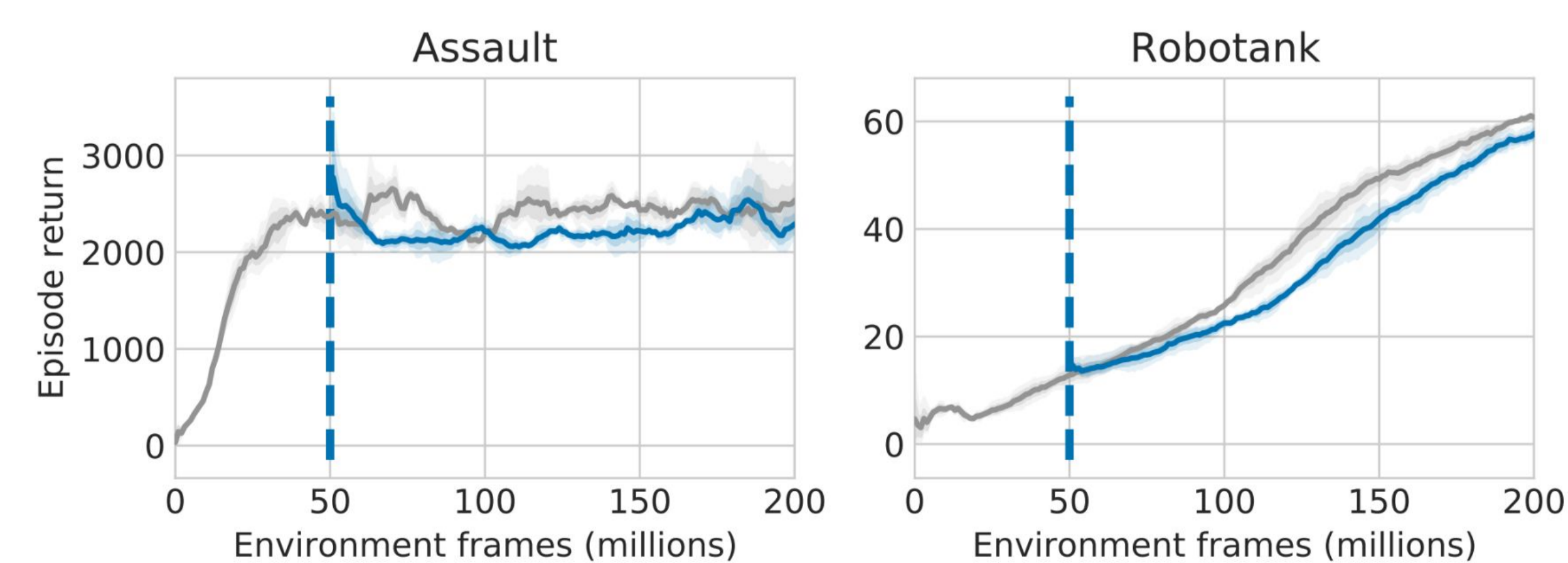
Key properties:

- Same number of trainable parameters
→ roughly **keeps representational power**
- Unaffected predictions after intervening
→ **excludes exploration** confounders

Diagnosing Loss of Plasticity

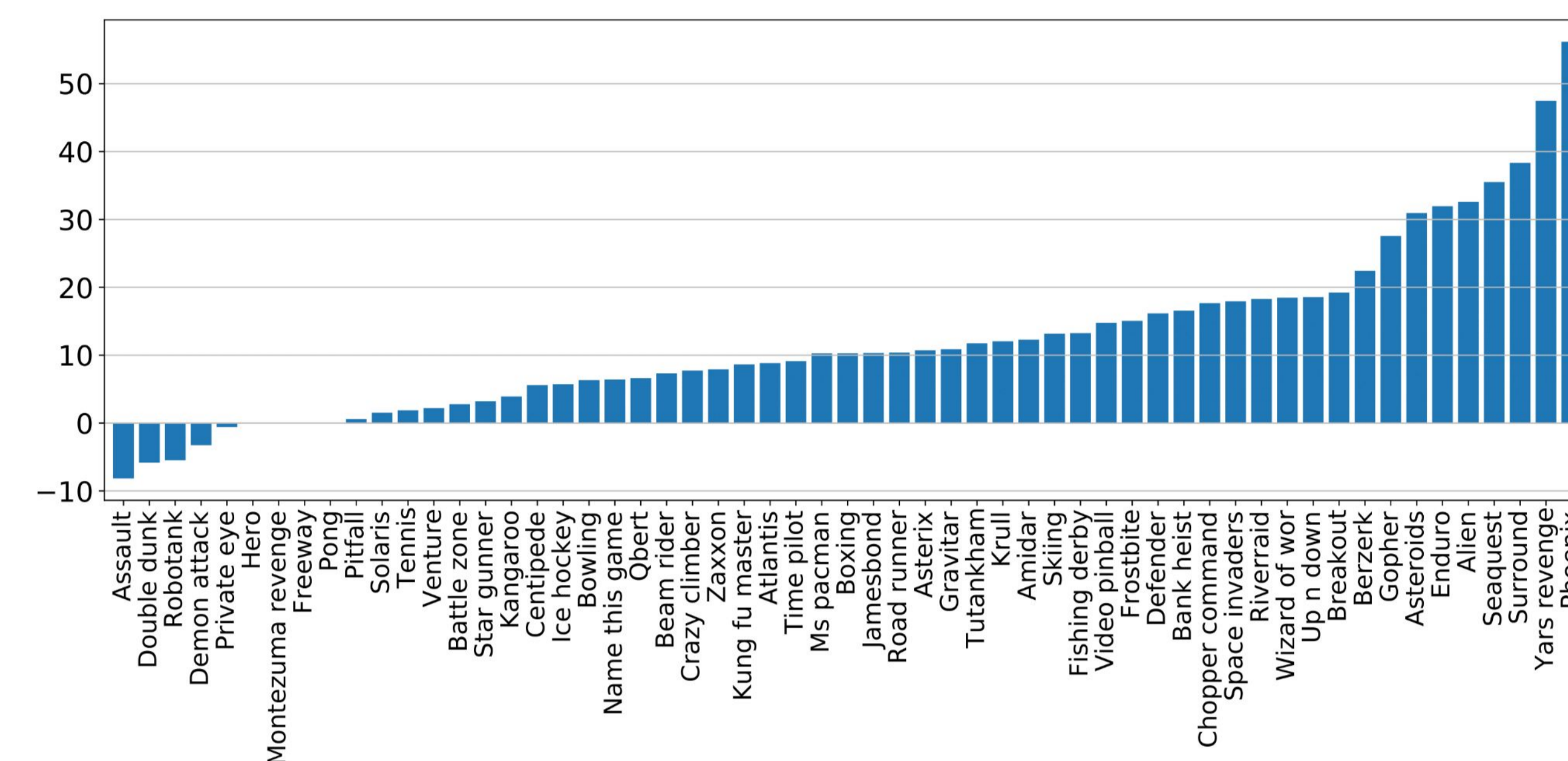


Improves performance → agent lost plasticity



Agent is stuck due to exploration / learns well → no effect from plasticity injection

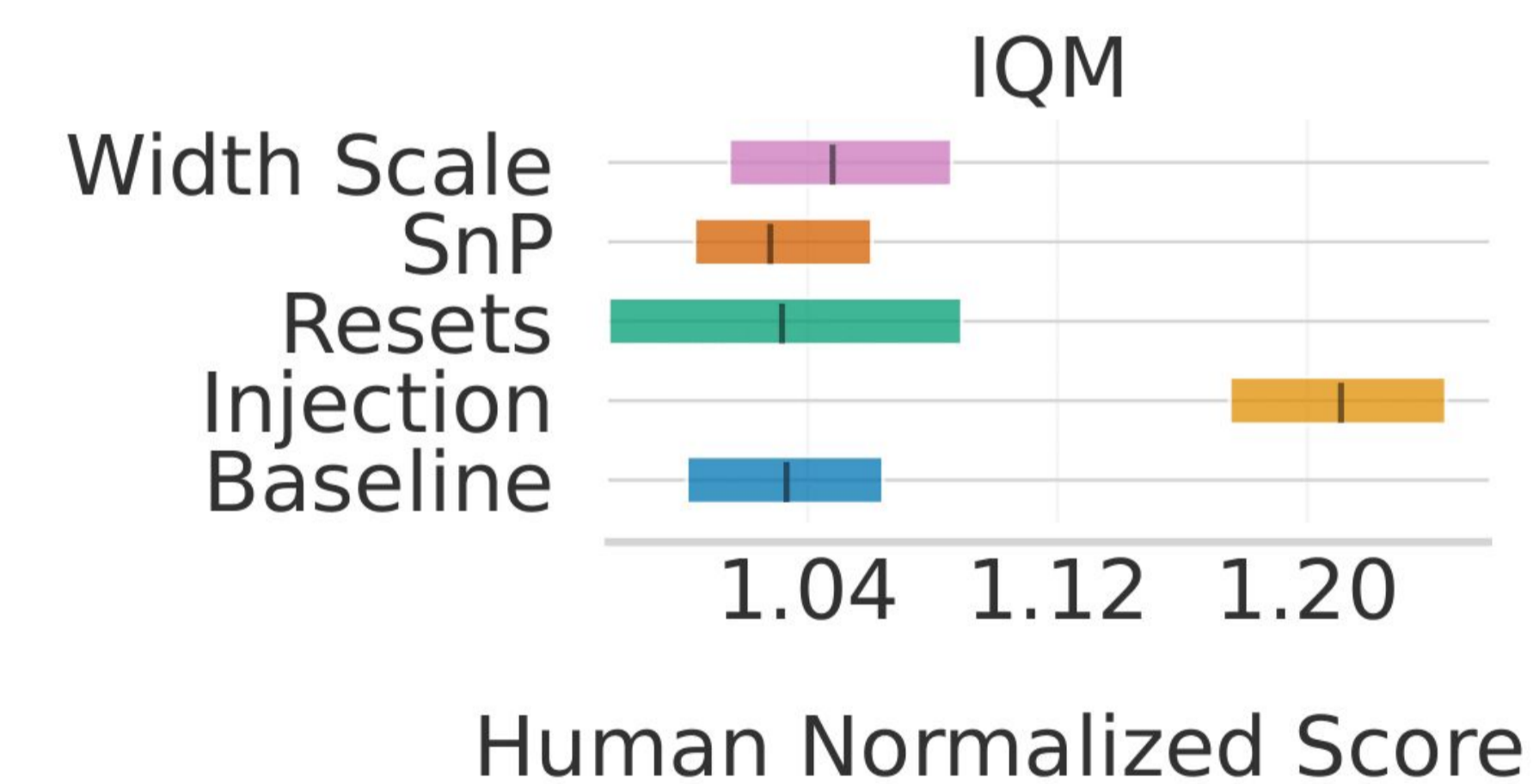
Results on 57 Atari Games



Agents lose plasticity within their lifetime

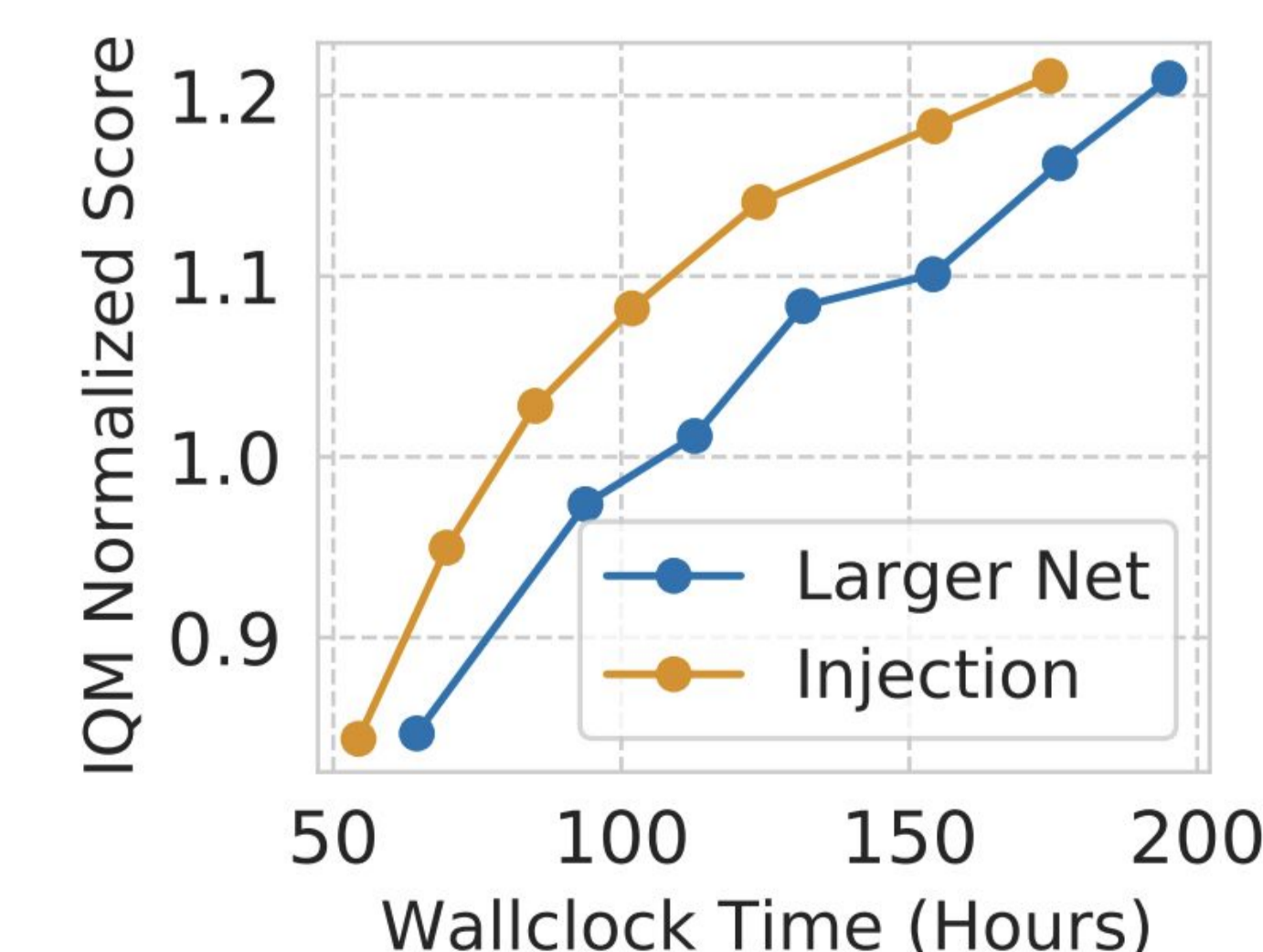
Reincarnating with Injection

Keep learning after plasticity loss efficiently



Better than re-training-free alternatives

Dynamic Network Growth



Same IQM as large nets from start but cheaper