

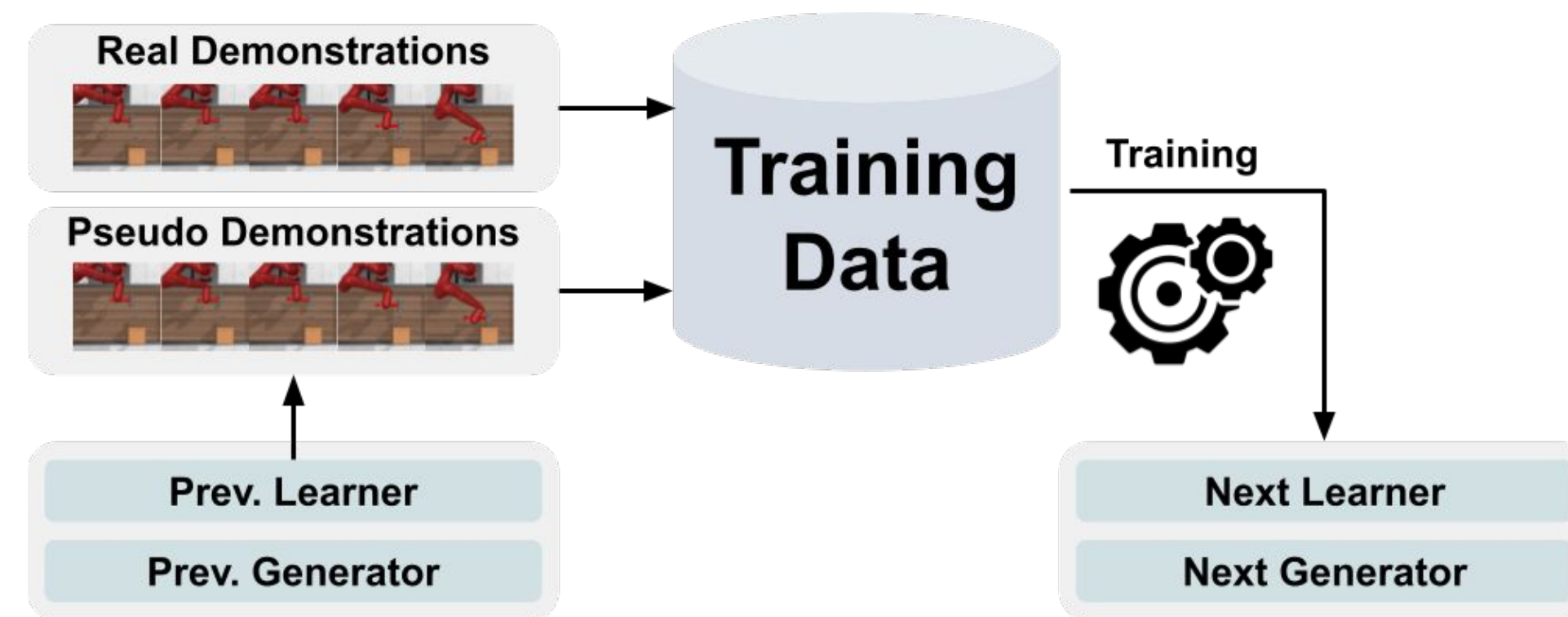
# t-DGR: A Trajectory-Based Deep Generative Replay Method for Continual Learning in Decision Making

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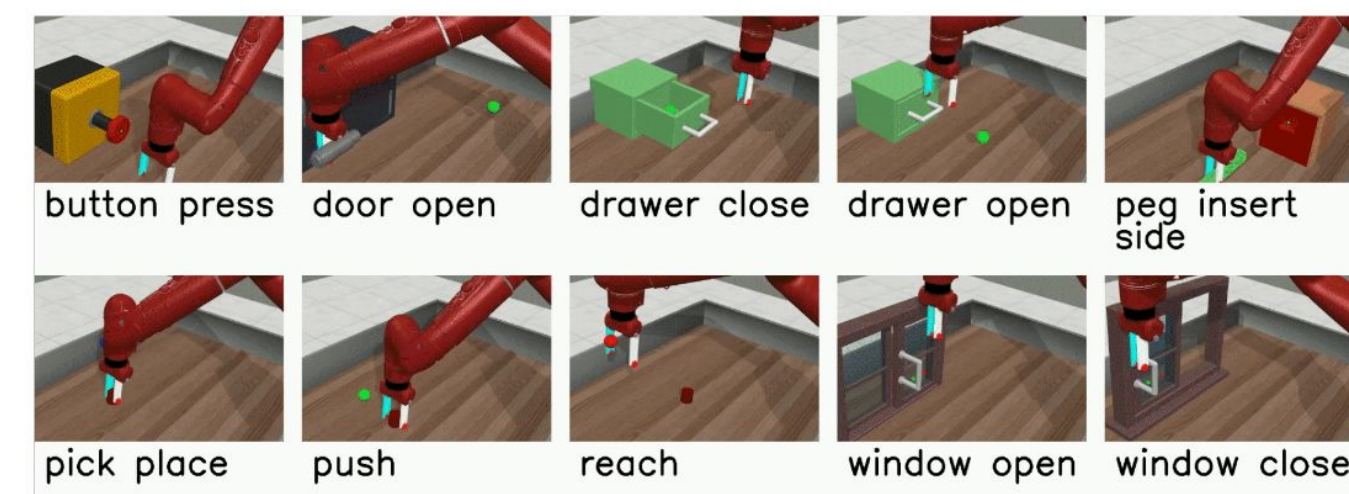


## Continual Learning and Deep Generative Replay



Deep generative replay in machine learning involves training a neural network to generate samples from past experiences, mitigating catastrophic forgetting in continual learning.

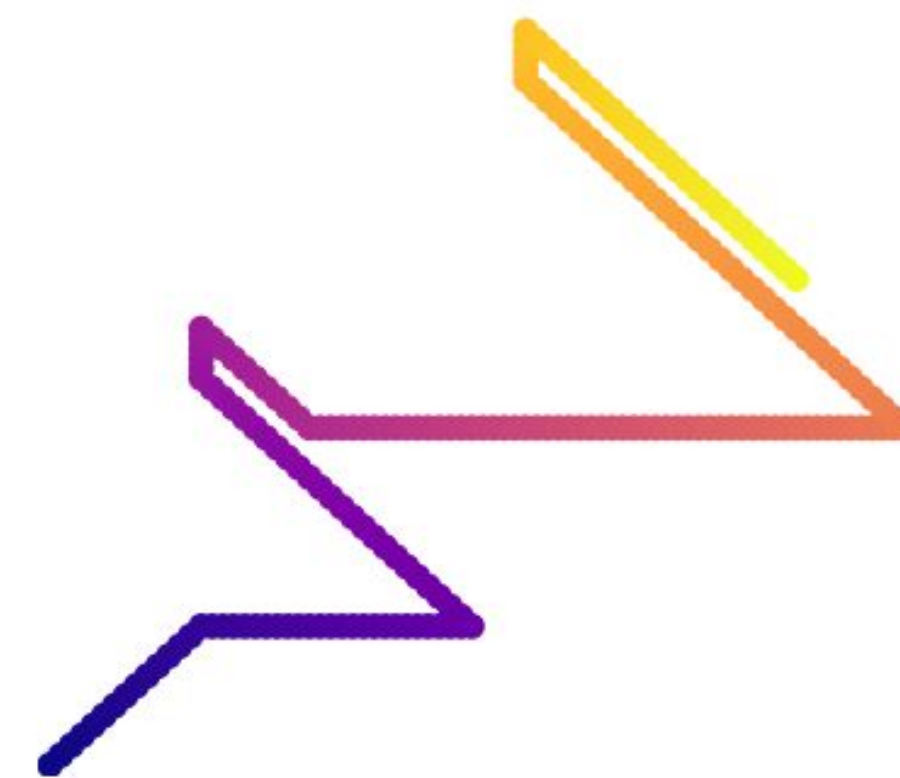
## Continual World Benchmark Results



The Continual World CW20 benchmark consists of 10 tasks, each repeated twice, where a Sawyer arm manipulates one or two objects in the MuJoCo physics simulator.

Method	Success Rate $\uparrow$	FT $\uparrow$	Forgetting $\downarrow$
Finetune	14.2 $\pm$ 4.0	-0.5 $\pm$ 3.0	82.2 $\pm$ 5.6
Multitask	97.0 $\pm$ 1.0	N/A	N/A
oEWC	19.4 $\pm$ 5.3	-2.8 $\pm$ 4.1	75.2 $\pm$ 7.5
PackNet	74.1 $\pm$ 4.1	-20.4 $\pm$ 3.4	<b>-0.2</b> $\pm$ 0.9
DGR	74.1 $\pm$ 4.1	18.9 $\pm$ 2.9	23.3 $\pm$ 3.3
CRIL	50.8 $\pm$ 4.4	4.4 $\pm$ 4.9	46.1 $\pm$ 5.4
t-DGR	<b>83.9</b> $\pm$ 3.0	<b>30.6</b> $\pm$ 4.5	14.6 $\pm$ 2.9

## Ground Truth



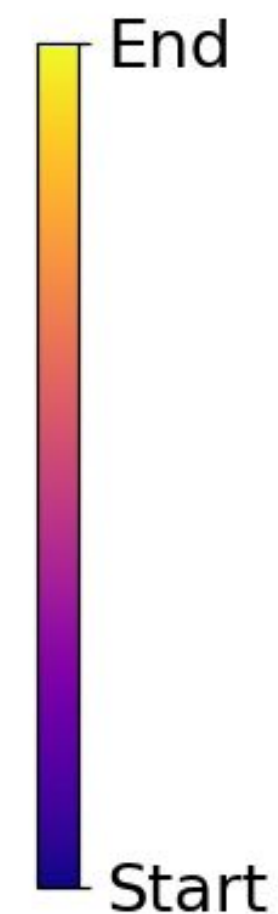
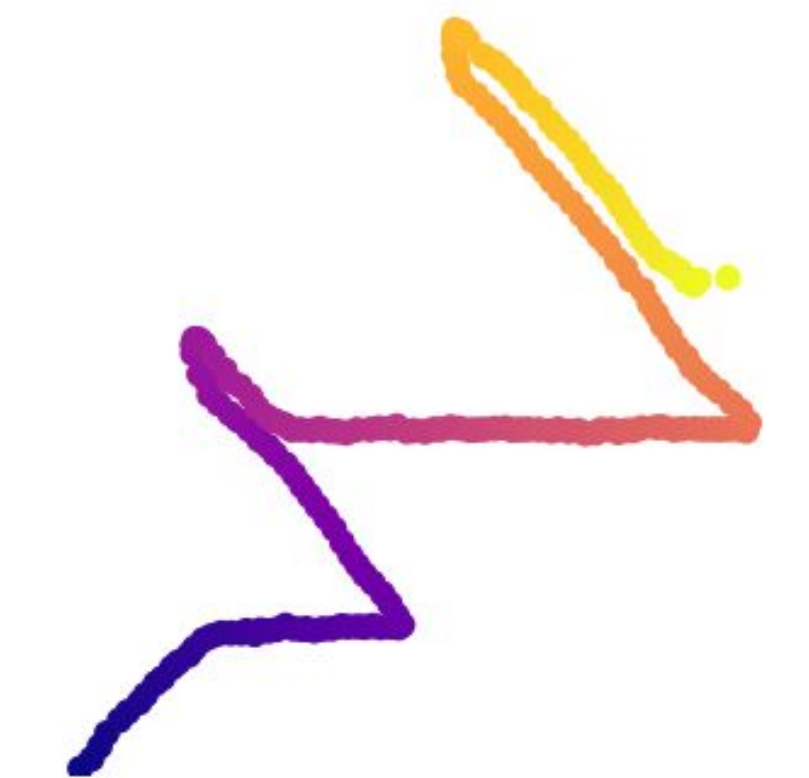
## CRIL



## DGR

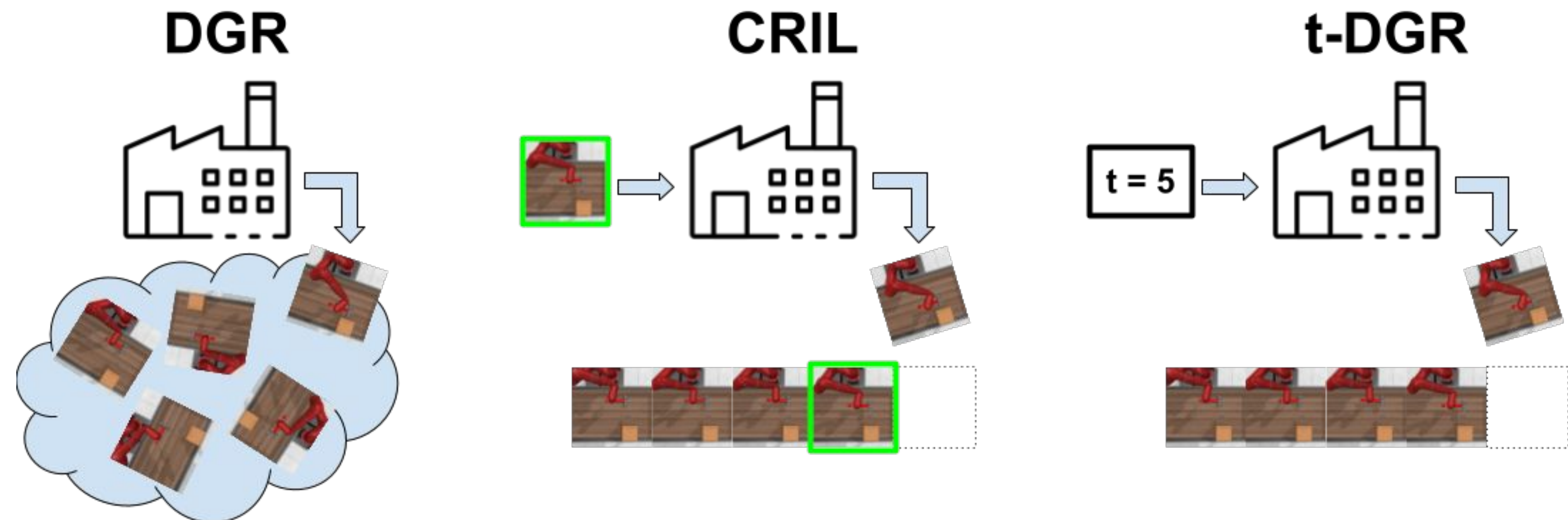


## t-DGR (ours)



We evaluate three methods for replicating an agent's movement in a 2D plane. CRIL deviates at a sharp turn and **never recovers**, while DGR produces a **fragmented path**. Our proposed method, t-DGR, overcomes these issues by sampling state observations based on the **trajectory timestep**.

## Current Pseudo-Rehearsal Methods



Generates video frames from the distribution observed during training, resulting in a **bag of unordered video frames**.

Generates the next video frame by **conditioning on the preceding video frame**.

Generates a video frame by **conditioning on the trajectory timestep**.