

Parcae: Proactive, Liveput-Optimized DNN Training on Preemptible Instances

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 **ByteDance**

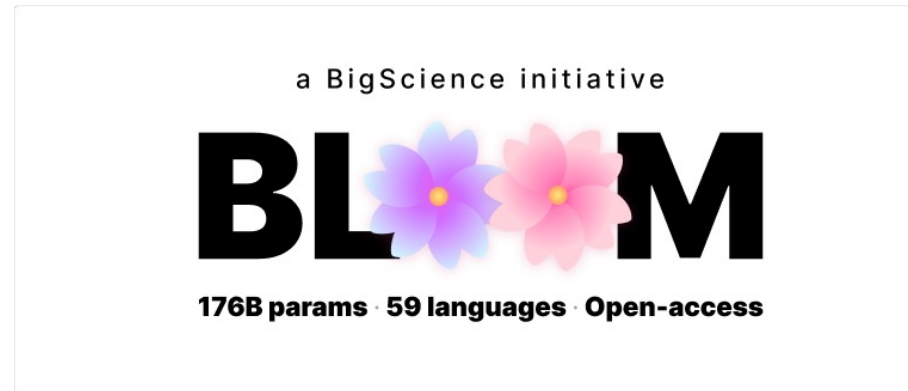
UCLA

 **Microsoft**

LLM Training Is Expensive



\$3.9 Million



\$2-5 Million



\$4.6 Million

Spot Instances Reduce Training Cost



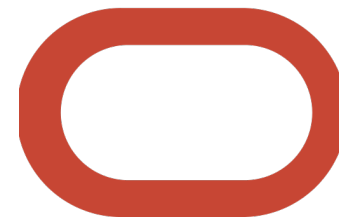
up to 10× Savings



up to 9× Savings

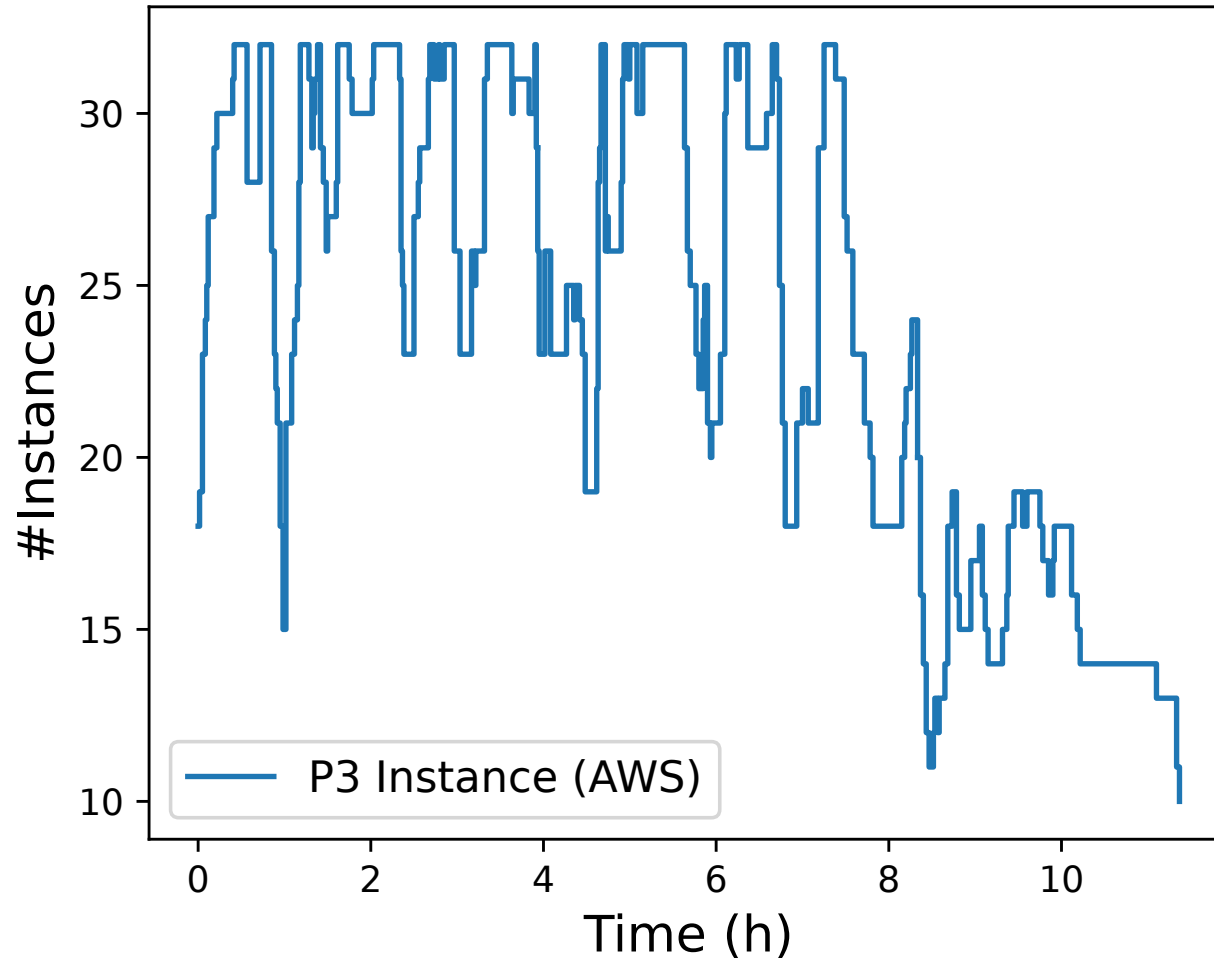


up to 3× Savings



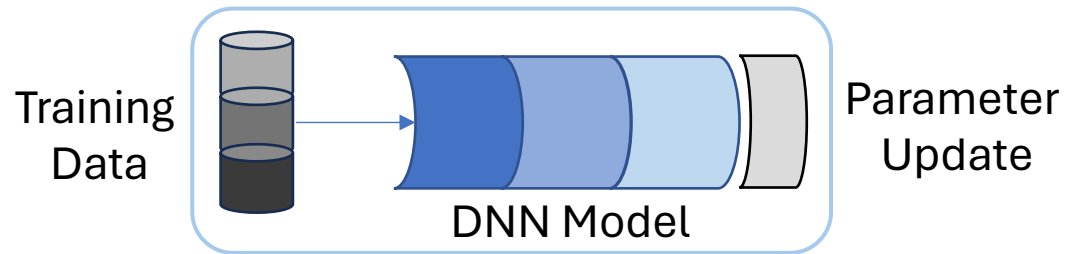
2× Savings

Spot Instances Reduce Training Cost, But Are Preemptible

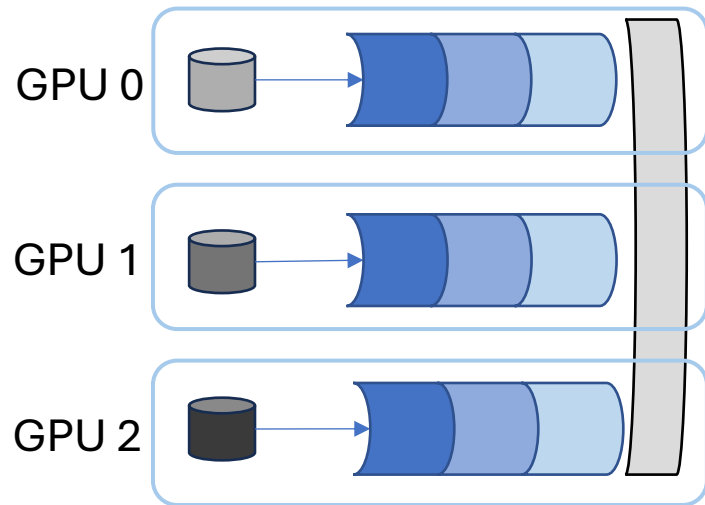


Preemption once every
5 minutes,
worst case **1** minute

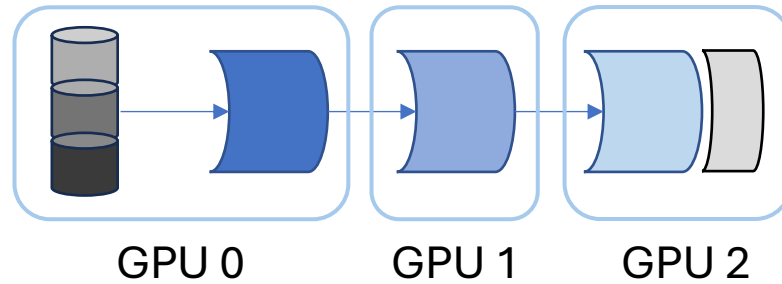
Background: Distributed DNN Training



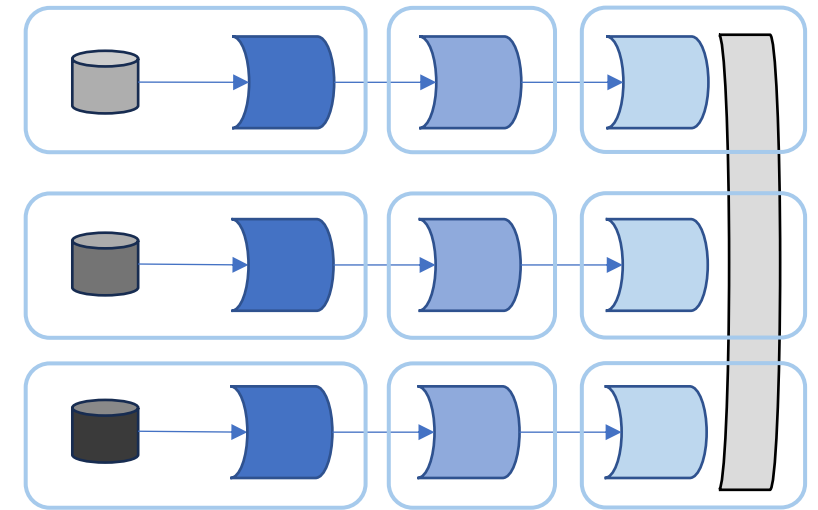
Single GPU Training



Data Parallelism

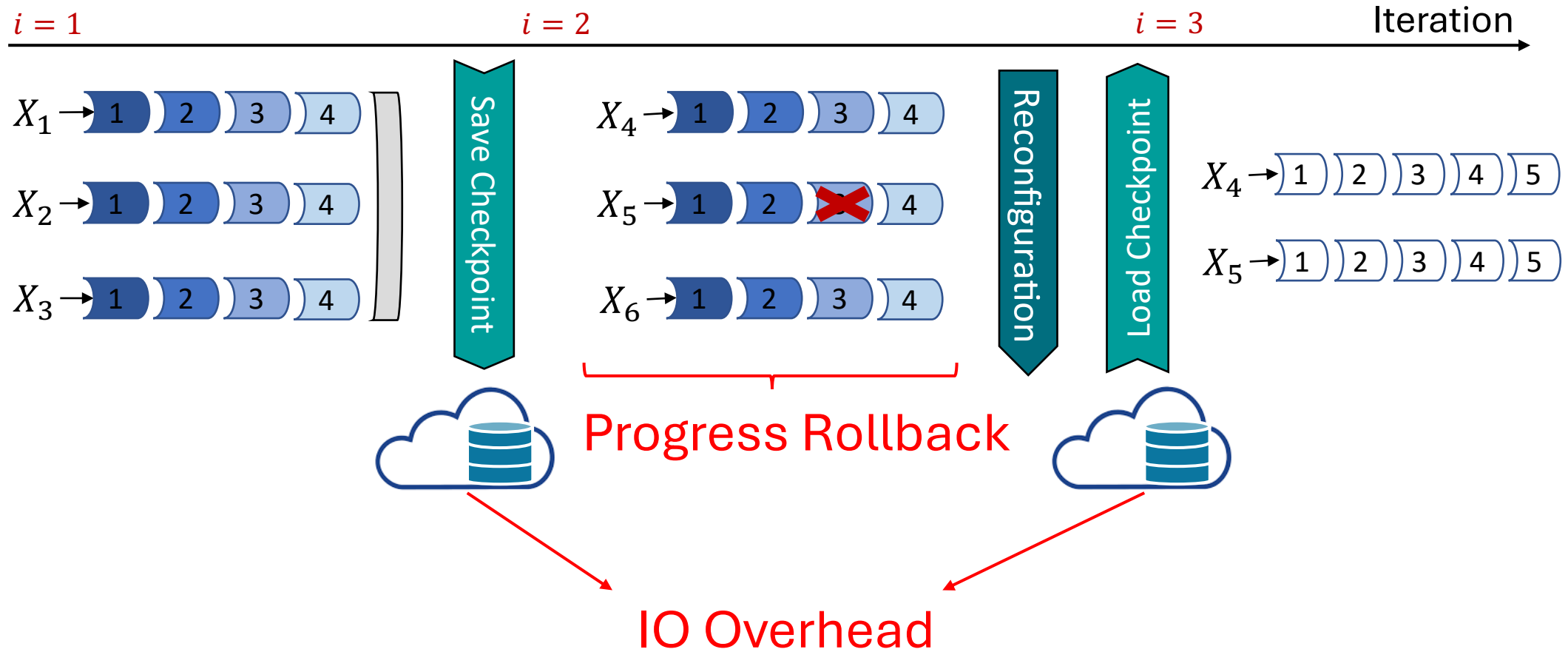


Pipeline Parallelism

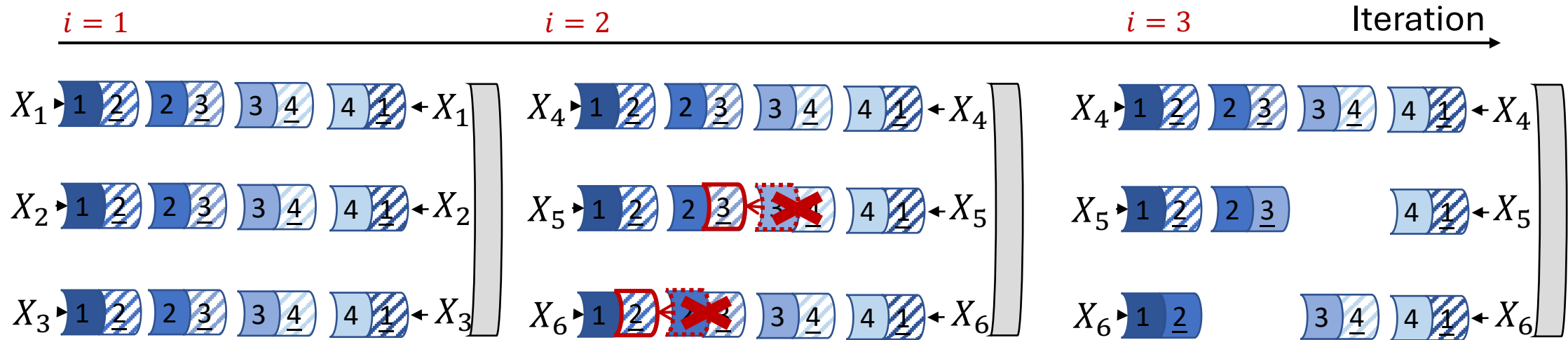


Hybrid Parallelism

Checkpoint-based: Varuna (EuroSys '22)

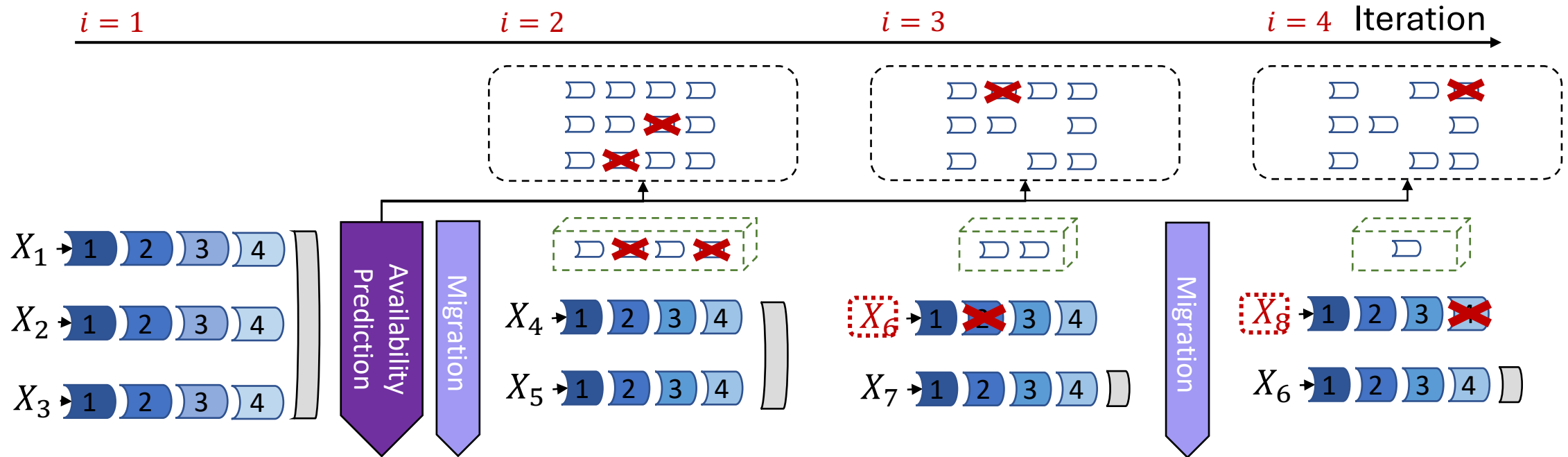


Redundancy-based: Bamboo (NSDI '23)



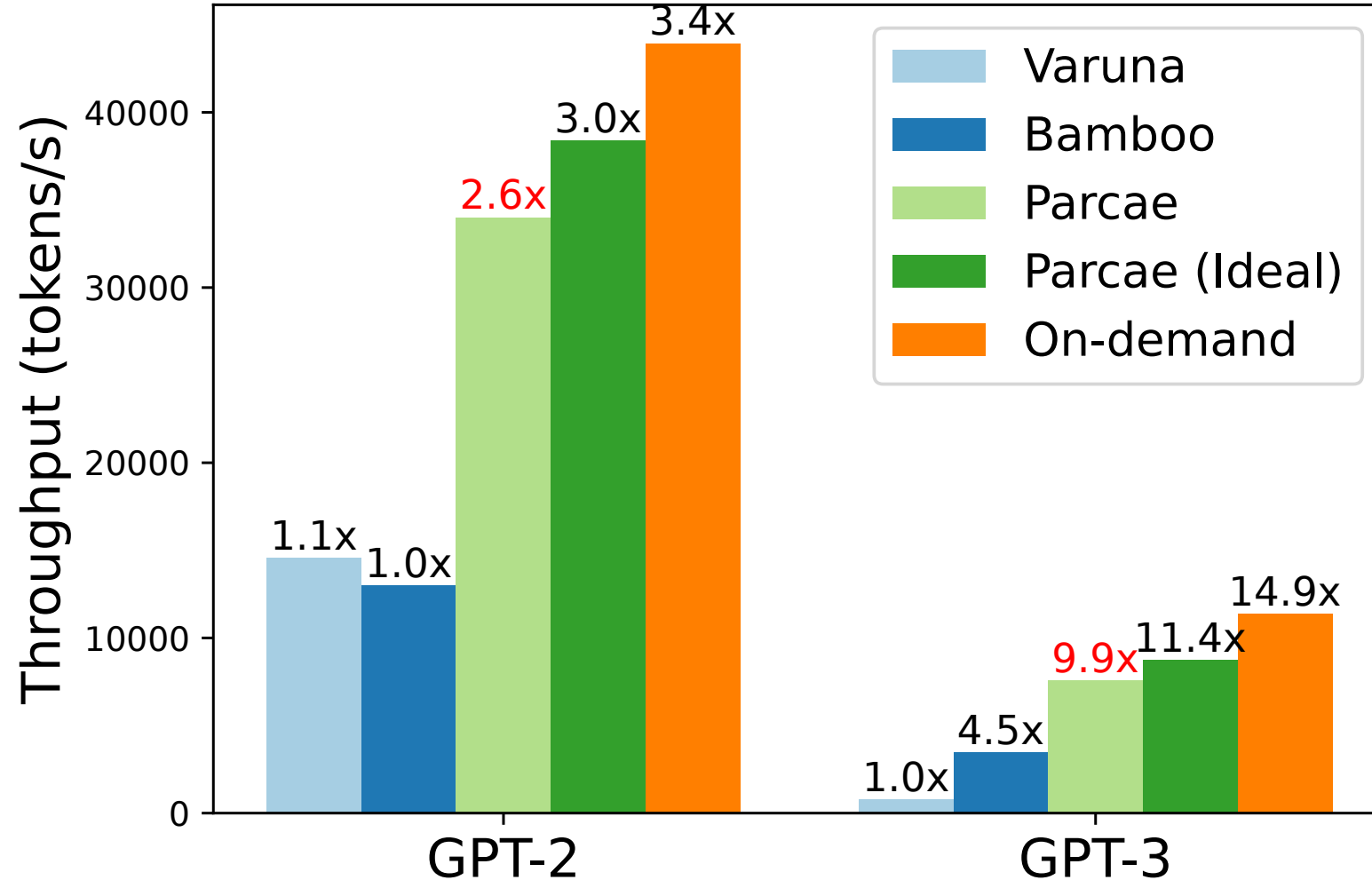
Redundant Computation and Memory Overhead

Parcae: Proactive, Liveput-Optimized



Proactively plan for future preemptions to maximize *preemption-aware* throughput (i.e., **Liveput**)

Up To **10x** Improvement, Near Optimal



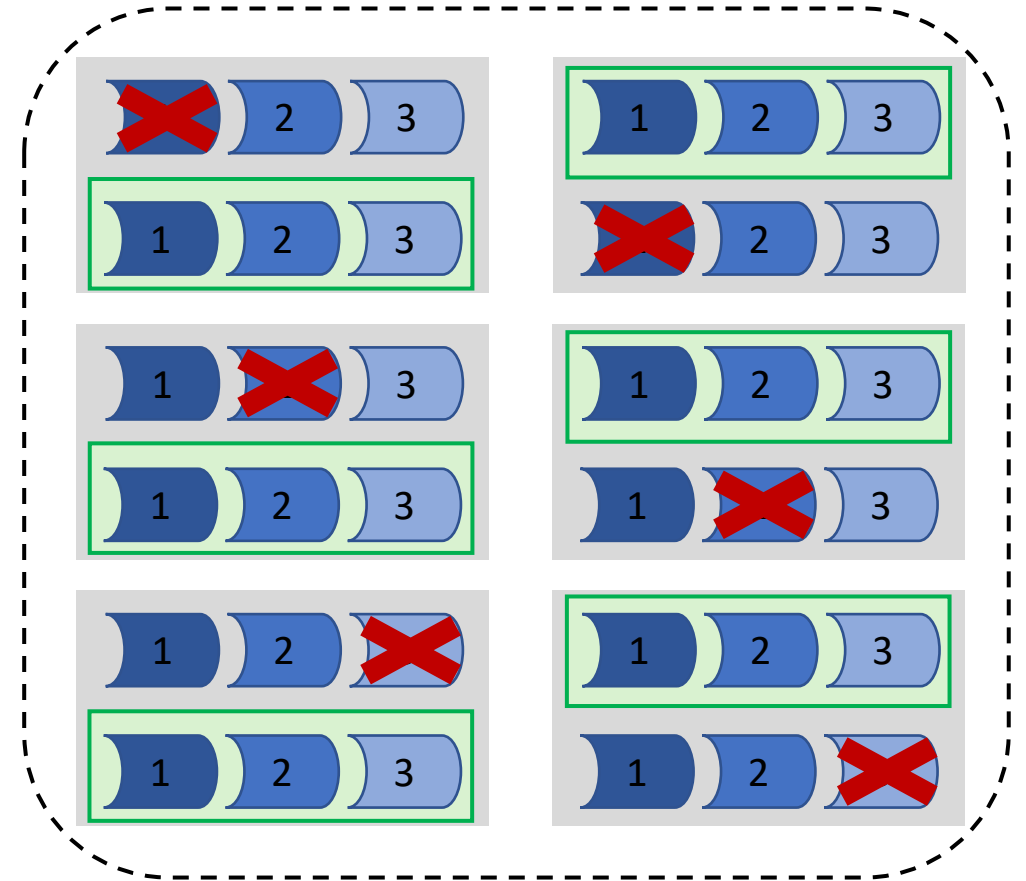
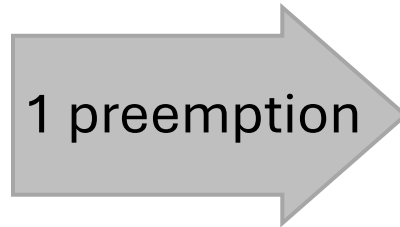
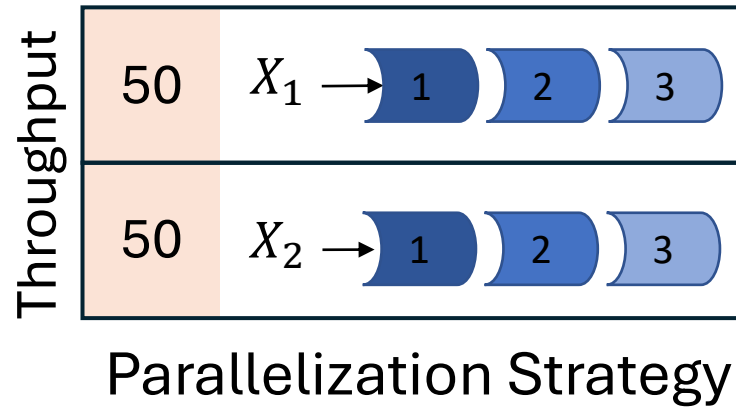
Liveput

$$\text{LIVEPUT}(D, P, \mathcal{V}) = \mathbb{E}_{\vec{v} \sim \mathcal{V}} [\text{THROUGHPUT}(D_{\vec{v}}, P_{\vec{v}})]$$

Preemption Scenario

*Parallelization Strategy
after Preemption*


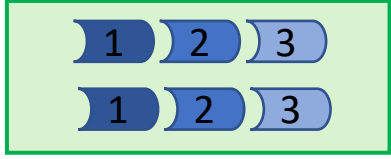

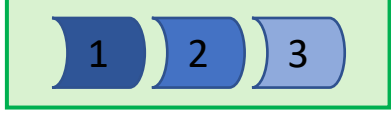

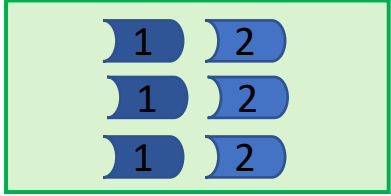

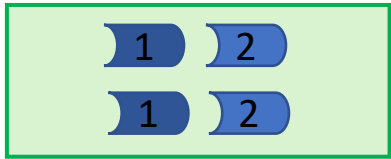

Liveput Example



LIVEPUT (1 preemption) = 100% × 50 = 50

Preemption Scenarios
100% remain 1 pipeline

Maximizing Throughput Is Sub-Optimal

Configurations	#preemptions	Distribution After Preemption	Throughput	LIVEPUT
50 $X_1 \rightarrow$ 	0	100% 	100 👍	$100\% \times 100 = 100$ 👍
50 $X_2 \rightarrow$ 	1	100% 		$100\% \times 50 = 50$
30 $X_1 \rightarrow$ 	0	100% 	90	$100\% \times 90 = 90$
30 $X_2 \rightarrow$ 	1	100% 		$100\% \times 60 = 60$ 👍
30 $X_3 \rightarrow$ 				

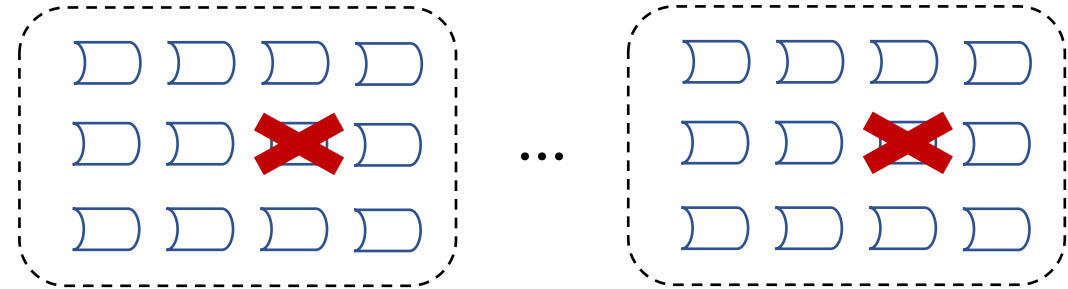
Challenges

1. Predict Liveput
2. Maximize Liveput
3. Handle Preemption

Our Contributions

1. Availability Predictor
2. Liveput Optimizer
3. Live Migration

Instance-wise Preemption Is Unpredictable



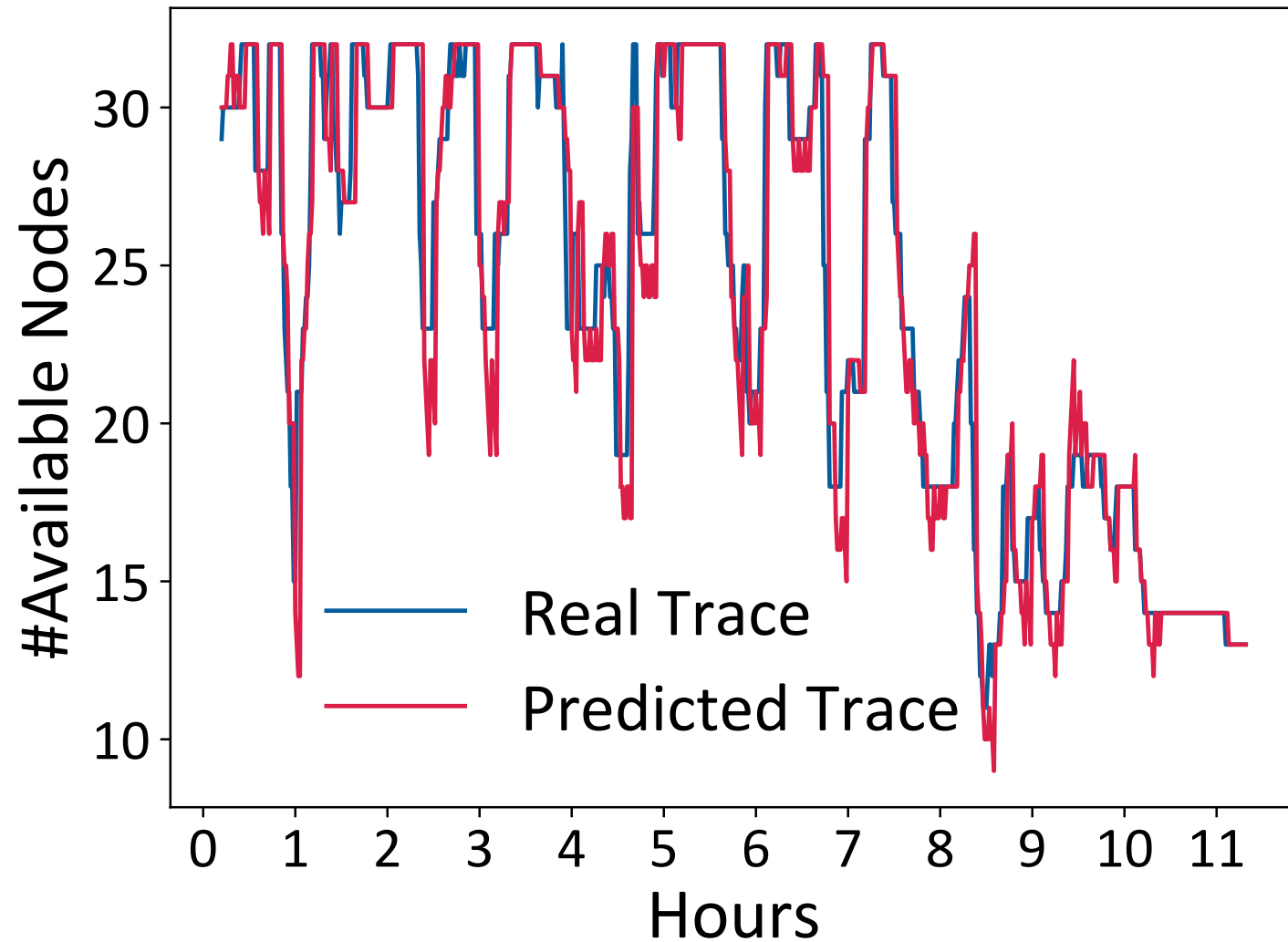
Parallelization Strategy

Preemption Mapping 😞

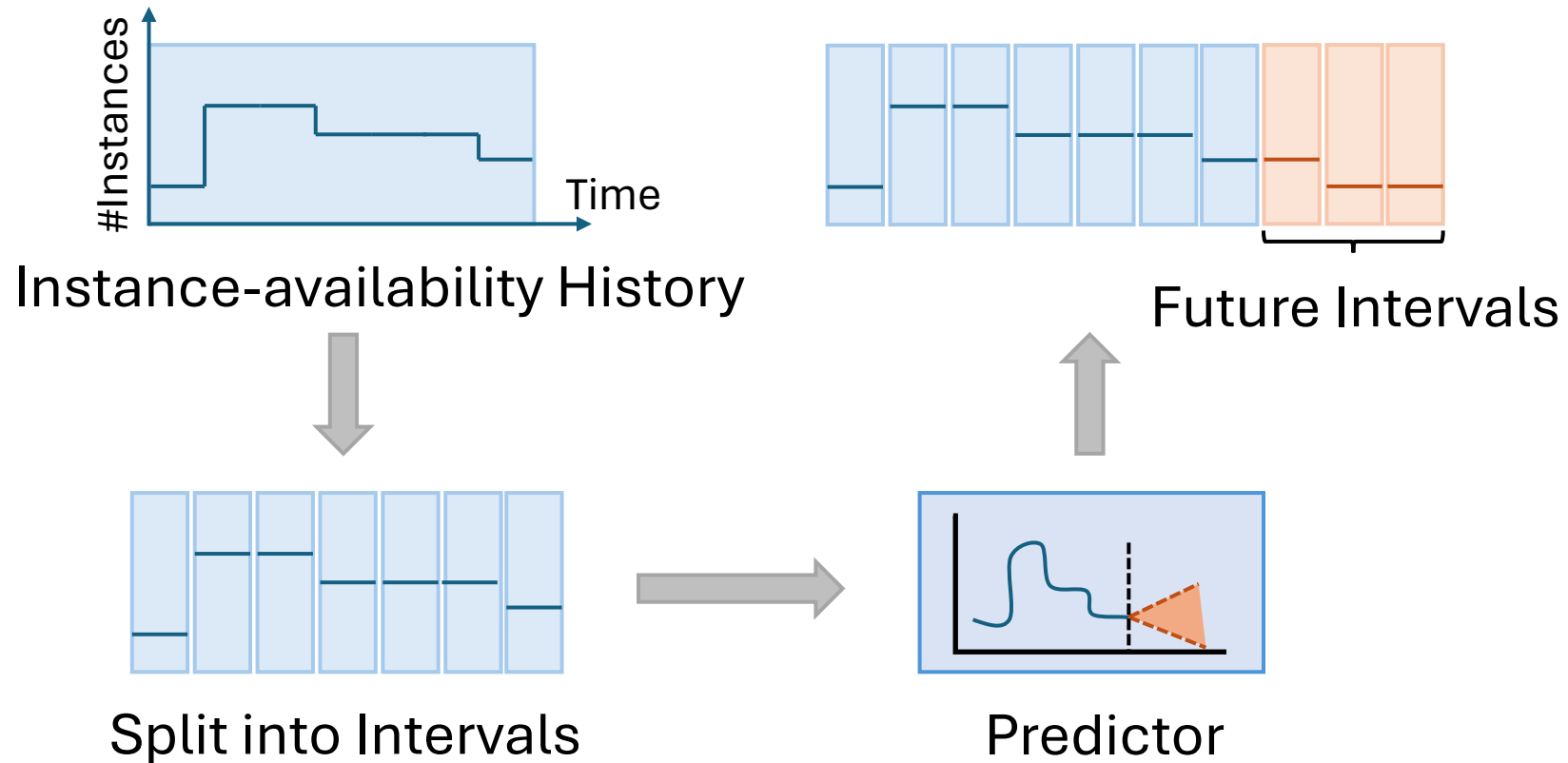


LIVEPUT

Short-Term Instance Availability Is Predictable

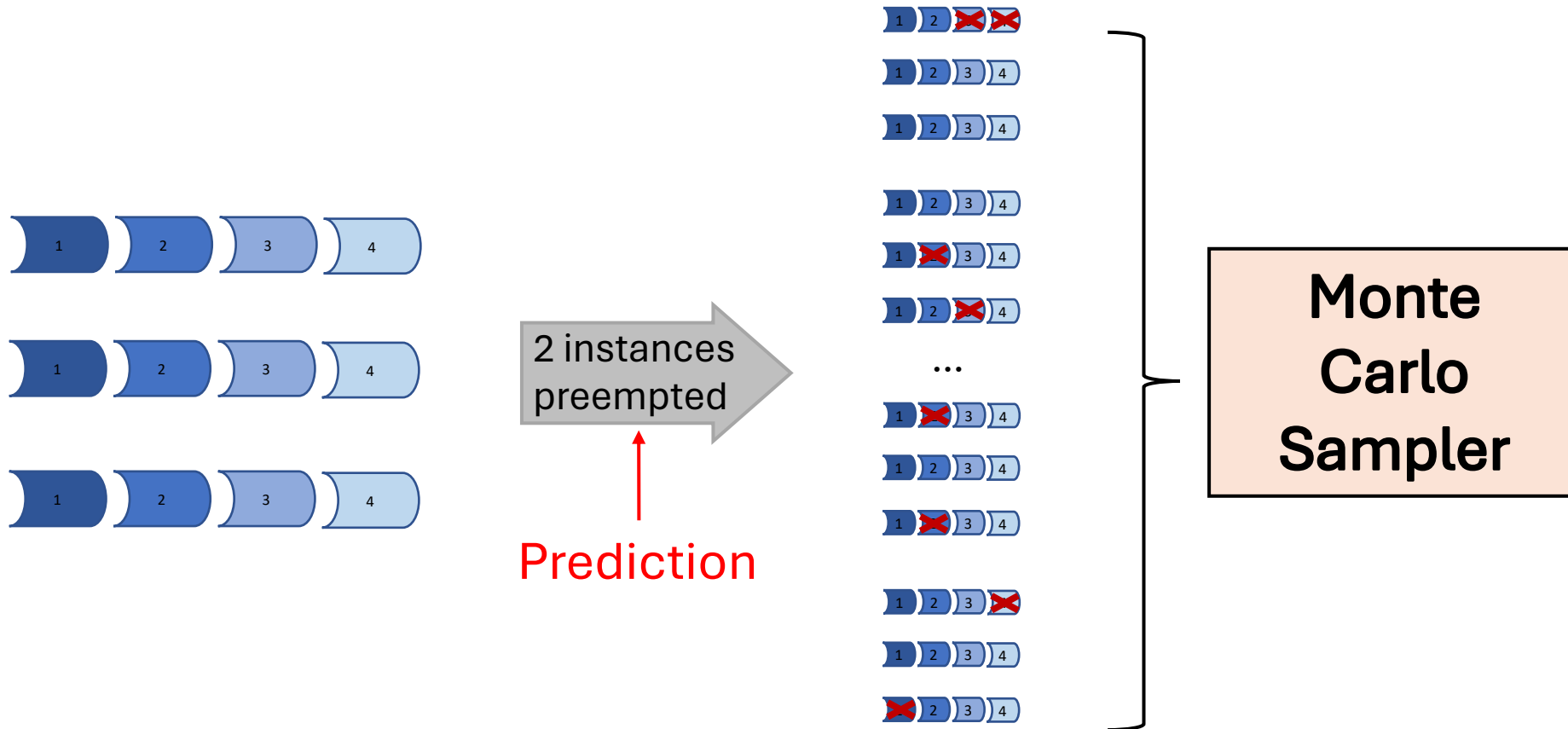


Availability Predictor



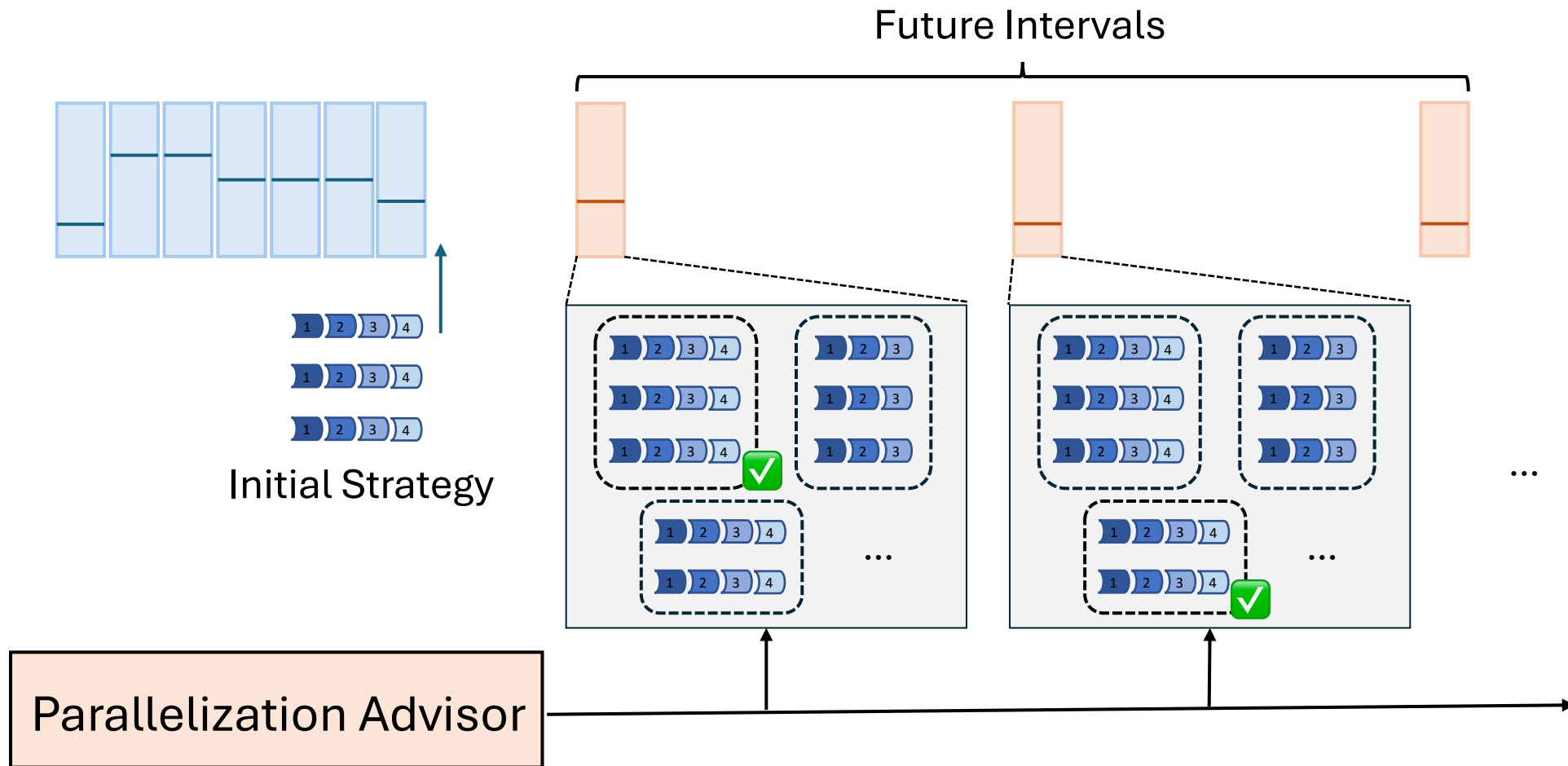
Time-series forecasting problem!

Preemption Mapping Sampling



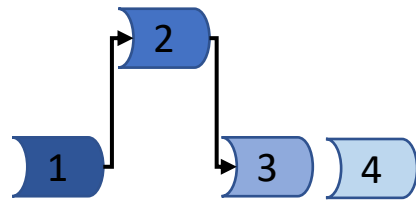
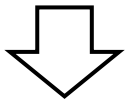
Large preemption mapping space

Liveput Optimizer

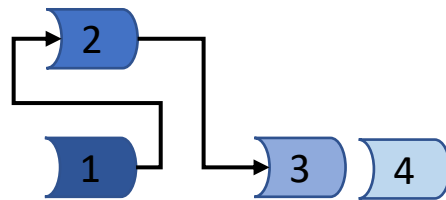
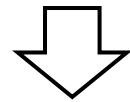


Combinatorial optimization space!

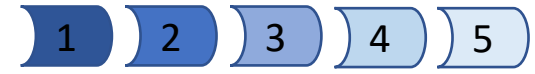
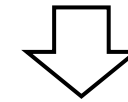
Fine-grained Live Migration



Intra-stage Migration

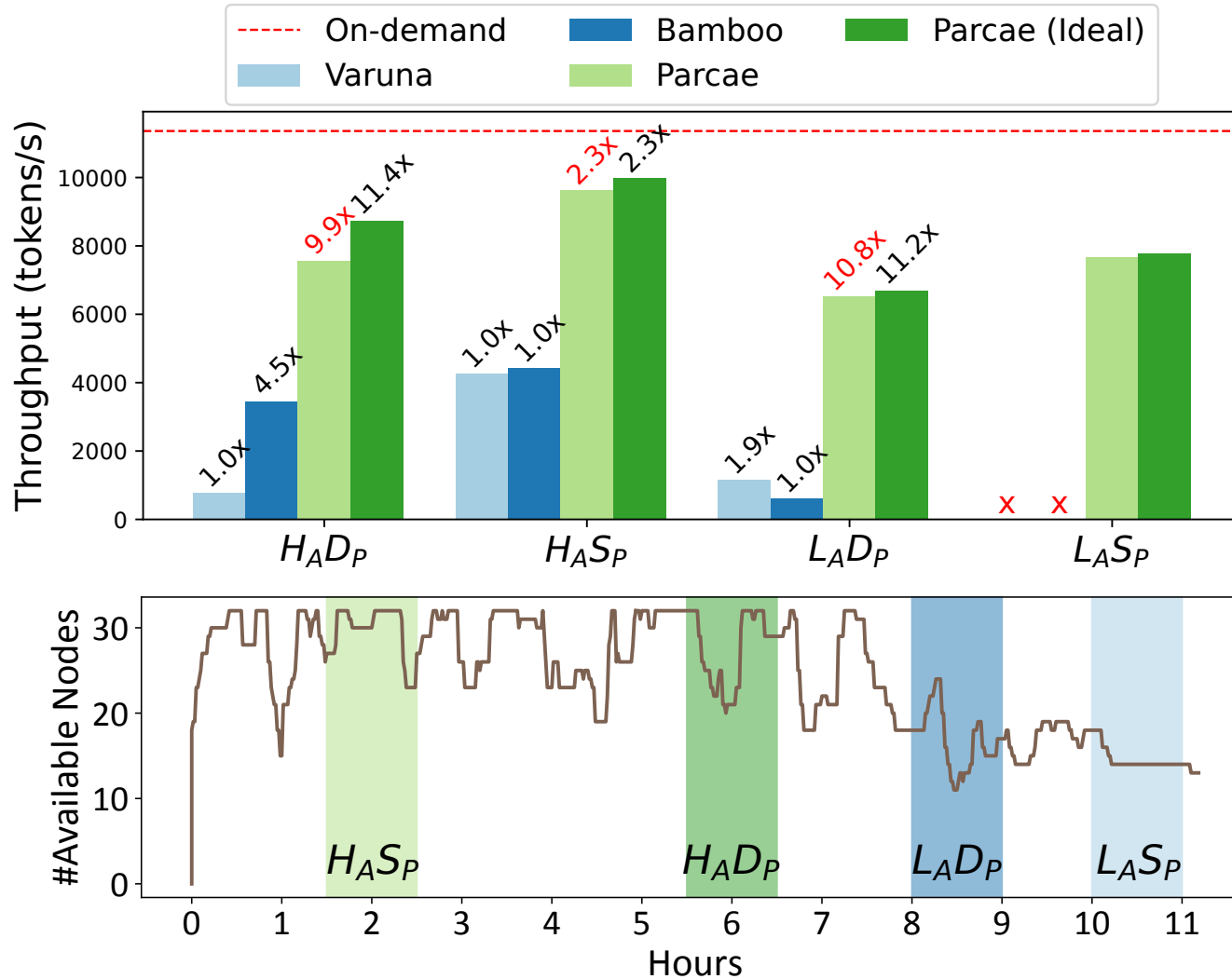


Inter-stage Migration



Pipeline Migration

GPT-3 (6B) Training on 32 Instances with Deterministic Replay



- Up to 10× higher throughput
- Better scalability
- Near optimal performance

Parcae



- Proactive, Liveput-Optimized DNN training on spot instances
- Up to 10× higher throughput than checkpoint- and redundancy-based systems
- Code: <https://github.com/JF-D/Parcae>