

A 3D Attention U-Net GAN for Global Precipitation Nowcasting

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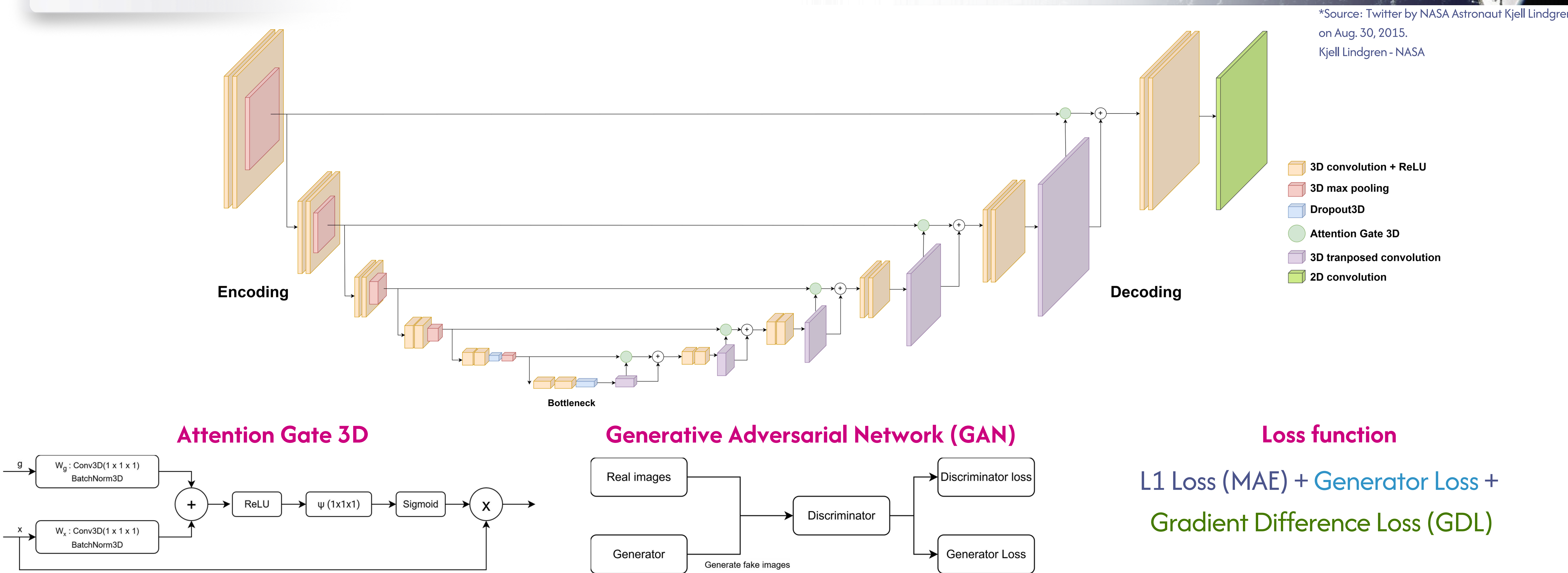
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INTRODUCTION

- **Goal:** Accurate, high-resolution precipitation nowcasting is critical for disaster management.
- **Current Challenges:**
 - **Blurry Outputs:** Traditional models with L2 (MSE) loss fail to capture high-intensity rain, resulting in blurry outputs.
 - **2D Limitations:** Previous models (e.g., SmaAt-UNet) treat time as a 2D channel, thereby missing complex 3D weather dynamics.
 - **Computational Cost:** Recurrent models (e.g., ConvLSTM) are computationally intensive.

METHODOLOGY



Experimental Setup:

- Dataset: Global Satellite Mapping of Precipitation (GSMaP).
- Resolution: Half of the original (512 x 1536).
- Metric: Critical Success Index (CSI).

Training Infrastructure:

- Compute Nodes: 8x NVIDIA A100 80GB GPUs.
- Host CPU: AMD EPYC™ 9684X (96 Cores).
- Memory: 2TB RAM.
- Strategy: Distributed Data Parallel (DDP) for multi-GPU scaling.

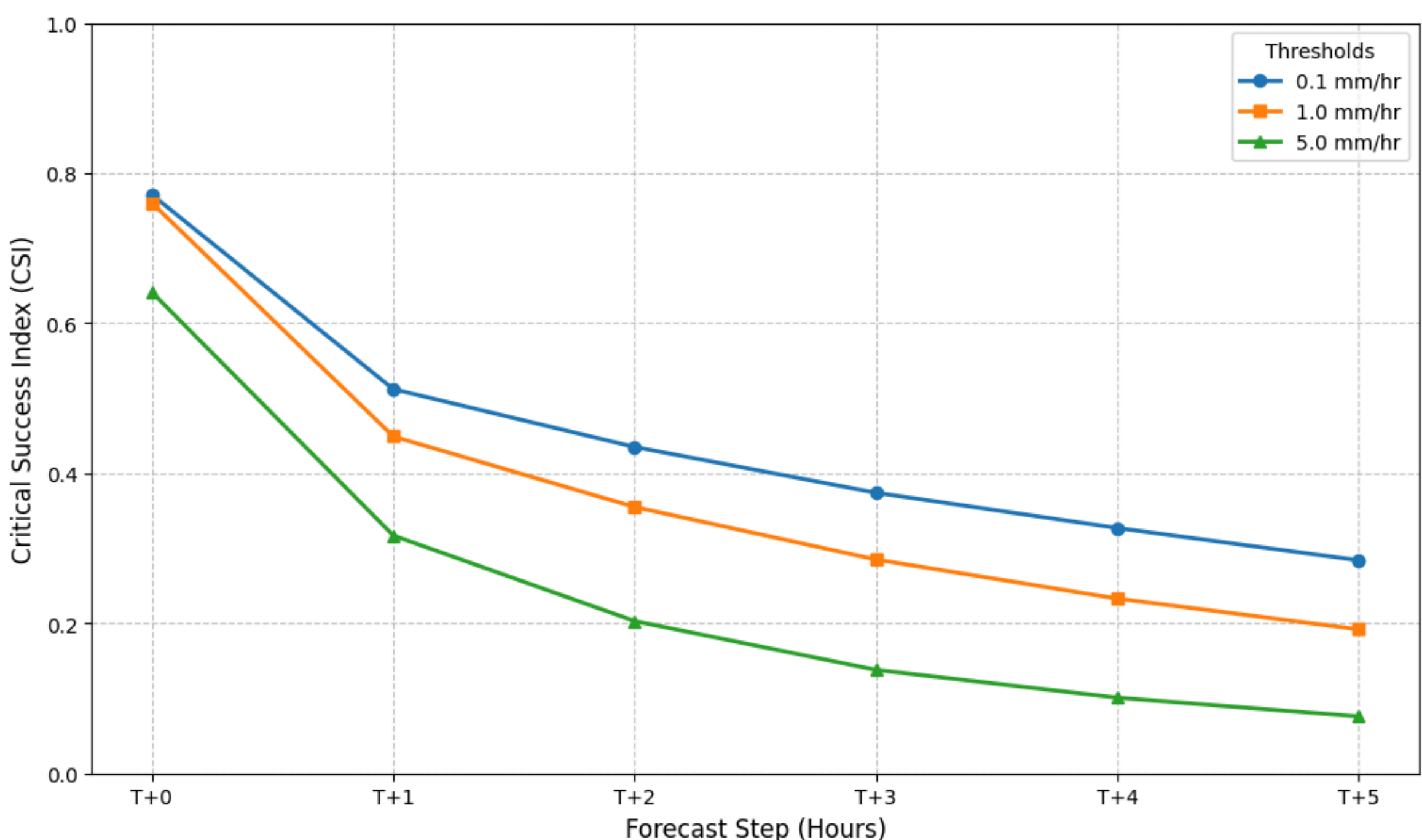
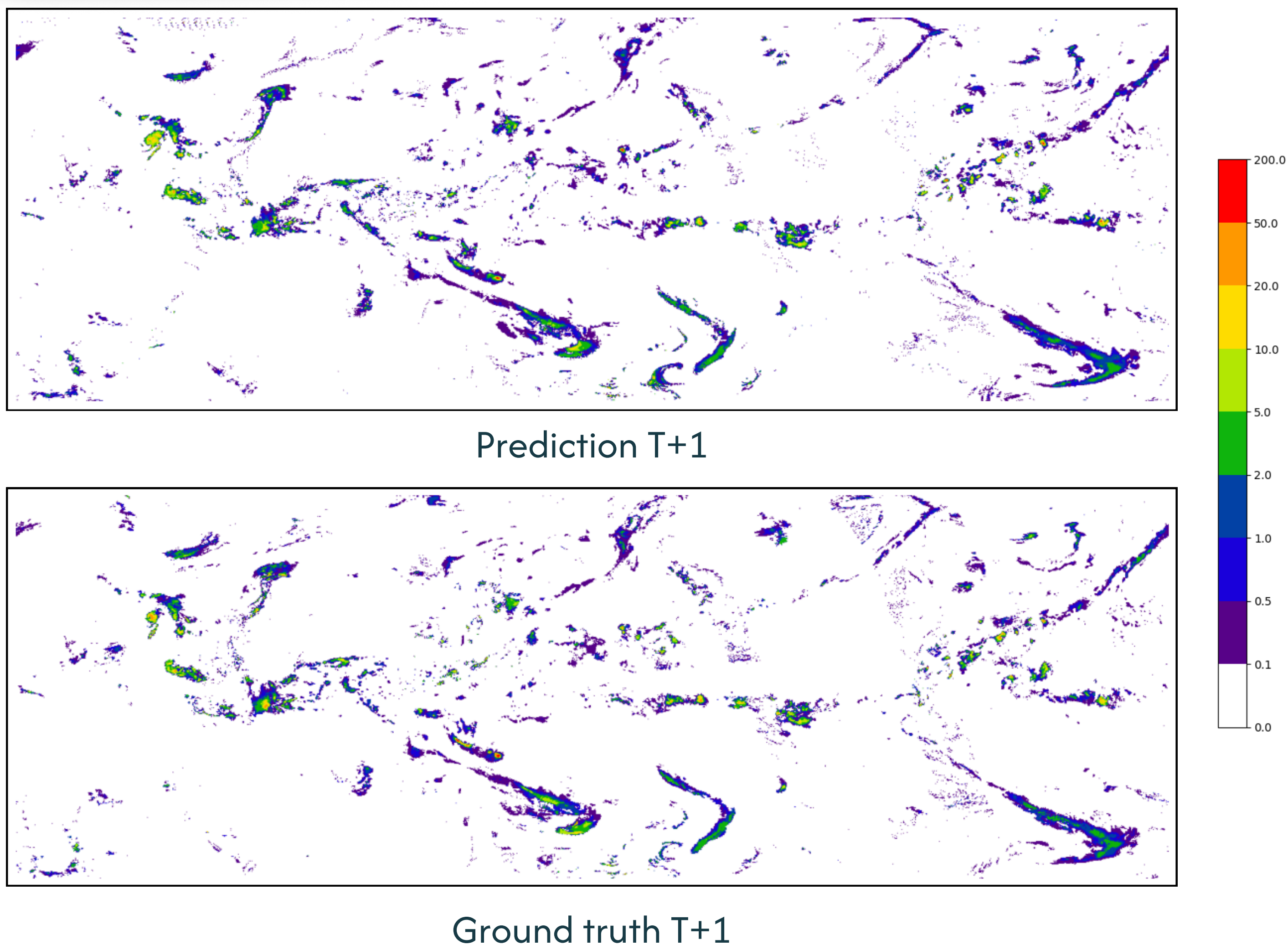


Table 1: 5-Step Validation CSI for January 2024 (higher is better)

RESULTS



CONCLUSION

The model delivers sharp, high-resolution forecasts for disaster response without the latency penalties of sequential processing

FUTURE WORK

Leverage Fugaku's massive node parallelism to train on full-resolution global data

HPC CHALLENGES

- I/O Bottleneck: We identified that the data loader could not keep up with the model's high throughput, leading to GPU starvation.
- Memory Optimization: Standard DDP replication consumed excessive VRAM. → Solution: Transitioning to Fully Sharded Data Parallel (FSDP2).