

# **NVIDIA NeMo Framework User Guide**

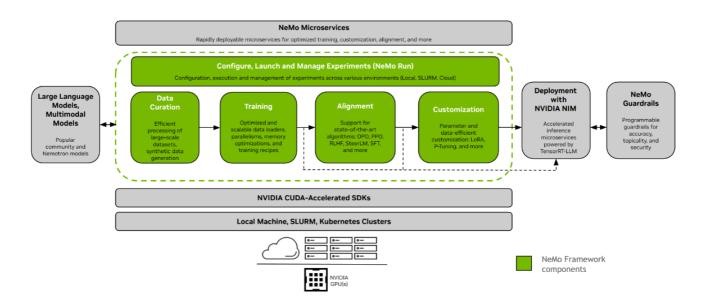
Home » Nvidia » NVIDIA NeMo Framework User Guide 🖺

#### **Contents**

- 1 NVIDIA NeMo Framework
- 2 Specifications
- **3 Product Usage**
- **Instructions**
- **4 Overview**
- 5 Training and
- Customization
- 6 Alignment
- 7 Multimodal Models
- **8 Supported Models**
- 9 Speech Al
- 10 FAQ
- 11 Documents / Resources
  - 11.1 References



**NVIDIA NeMo Framework** 



## **Specifications**

Product Name: NVIDIA NeMo Framework

• Affected Platforms: Windows, Linux, macOS

• Affected Versions: All versions prior to 24

• Security Vulnerability: CVE-2025-23360

• Risk Assessment Base Score: 7.1 (CVSS v3.1)

# **Product Usage Instructions**

#### **Security Update Installation:**

To protect your system, follow these steps:

- 1. Download the latest release from the NeMo-Framework-Launcher Releases page on GitHub.
- 2. Go to NVIDIA Product Security for further information.

#### **Security Update Details:**

The security update addresses a vulnerability in the NVIDIA NeMo Framework that could lead to code execution and data tampering.

#### **Software Upgrade:**

If you are using an earlier branch release, it is recommended to upgrade to the latest branch release to address the security issue.

#### Overview

NVIDIA NeMo Framework is a scalable and cloud-native generative AI framework built for researchers and developers working on Large Language Models, Multimodal, and Speech AI (e.g. Automatic Speech Recognition and Text-to-Speech). It enables users to efficiently create, customize, and deploy new generative AI models by leveraging existing code and pre-trained model checkpoints.

Setup Instructions: Install NeMo Framework

#### **Large Language Models and Multimodal Models**

NeMo Framework provides end-to-end support for developing Large Language Models (LLMs) and Multimodal

Models (MMs). It provides the flexibility to be used on-premises, in a data-center, or with your preferred cloud provider. It also supports execution on SLURM or Kubernetes enabled environments.

\_images/nemo-llm-mm-stack.png

#### **Data Curation**

NeMo Curator [1] is a Python library that includes a suite of modules for data mining and synthetic data generation. They are scalable and optimized for GPUs, making them ideal for curating natural language data to train or fine-tune LLMs. With NeMo Curator, you can efficiently extract high-quality text from extensive raw web data sources.

#### **Training and Customization**

NeMo Framework provides tools for efficient training and customization of LLMs and Multimodal models. It includes default configurations for compute cluster setup, data downloading, and model hyperparameters, which can be adjusted to train on new datasets and models. In addition to pre-training, NeMo supports both Supervised Fine-Tuning (SFT) and Parameter Efficient Fine-Tuning (PEFT) techniques like LoRA, Ptuning, and more.

Two options are available to launch training in NeMo – using the NeMo 2.0 API interface or with NeMo Run.

- With NeMo Run (Recommended): NeMo Run provides an interface to streamline configuration, execution and management of experiments across various compute environments. This includes launching jobs on your workstation locally or on big clusters both SLURM enabled or Kubernetes in a cloud environment.
  - Pre-training & PEFT Quickstart with NeMo Run
- Using the NeMo 2.0 API: This method works well with a simple setup involving small models, or if you are
  interested in writing your own custom dataloader, training loops, or change model layers. It gives you more
  flexibility and control over configurations, and makes it easy to extend and customize configurations
  programmatically.
  - Training Quickstart with NeMo 2.0 API
  - Migrating from NeMo 1.0 to NeMo 2.0 API

#### **Alignment**

- NeMo-Aligner [1] is a scalable toolkit for efficient model alignment. The toolkit has support for state-of-the-art
  model alignment algorithms such as SteerLM, DPO, Reinforcement Learning from Human Feedback (RLHF),
  and much more. These algorithms enable users to align language models to be more safe, harmless, and
  helpful.
- All the NeMo-Aligner checkpoints are cross-compatible with the NeMo ecosystem, allowing for further customization and inference deployment.

#### Step-by-step workflow of all three phases of RLHF on a small GPT-2B model:

- SFT training
- Reward model training
- PPO training

In addition, we demonstrate support for various other novel alignment methods:

- DPO: a lightweight alignment algorithm compared to RLHF with a simpler loss function.
- Self-Play Fine-Tuning (SPIN)
- SteerLM: a technique based on conditioned-SFT, with steerable output.

Check out the documentation for more information: Alignment Documentation

#### **Multimodal Models**

- NeMo Framework provides optimized software to train and deploy state-of-the-art multimodal models across several categories: Multimodal Language Models, Vision-Language Foundations, Text-to-Image models, and beyond 2D Generation using Neural Radiance Fields (NeRF).
- Each category is designed to cater to specific needs and advancements in the field, leveraging cutting-edge models to handle a wide range of data types, including text, images, and 3D models.

#### Note

We are migrating support for multimodal models from NeMo 1.0 to NeMo 2.0. If you want to explore this domain in the meantime, please refer to the documentation for the NeMo 24.07 (previous) release.

#### **Deployment and Inference**

NeMo Framework provides various paths for LLM inference, catering to different deployment scenarios and performance needs.

## **Deploy with NVIDIA NIM**

- NeMo Framework seamlessly integrates with enterprise-level model deployment tools through NVIDIA NIM.
   This integration is powered by NVIDIA TensorRT-LLM, ensuring optimized and scalable inference.
- For more information on NIM, visit the NVIDIA website.

#### **Deploy with TensorRT-LLM or vLLM**

- NeMo Framework offers scripts and APIs to export models to two inference optimized libraries, TensorRT-LLM and vLLM, and to deploy the exported model with the NVIDIA Triton Inference Server.
- For scenarios requiring optimized performance, NeMo models can leverage TensorRT-LLM, a specialized library for accelerating and optimizing LLM inference on NVIDIA GPUs. This process involves converting NeMo models into a format compatible with TensorRT-LLM using the nemo.export module.
  - LLM Deployment Overview
  - Deploy NeMo Large Language Models with NIM
  - Deploy NeMo Large Language Models with TensorRT-LLM
  - Deploy NeMo Large Language Models with vLLM

# **Supported Models**

# **Large Language Models**

Large Language Models

Large Language Models	Pretraini ng & SFT	P E F T	Alig nme nt	FP8 Training Convergence	TRT/T RTLL M	Convert To & Fro m Hugging Face	Eval uatio n
Llama3 8B/70B, Llama3.1 40 5B	Yes	Y es	х	Yes (partially v erified)	Yes	Both	Yes
Mixtral 8x7B/8x22B	Yes	Y es	х	Yes (unverified)	Yes	Both	Yes
Nemotron 3 8B	Yes	х	x	Yes (unverified)	x	Both	Yes
Nemotron 4 340B	Yes	х	x	Yes (unverified)	x	Both	Yes
Baichuan2 7B	Yes	Y es	х	Yes (unverified)	x	Both	Yes
ChatGLM3 6B	Yes	Y es	х	Yes (unverified)	x	Both	Yes
Gemma 2B/7B	Yes	Y es	х	Yes (unverified)	Yes	Both	Yes
Gemma2 2B/9B/27B	Yes	Y es	x	Yes (unverified)	x	Both	Yes
Mamba2 130M/370M/780M/1 .3B/2.7B/8B/ Hybrid-8B	Yes	Y es	х	Yes (unverified)	x	х	Yes
Phi3 mini 4k	х	Y es	x	Yes (unverified)	x	х	x
Qwen2 0.5B/1.5B/7B/72B	Yes	Y es	х	Yes (unverified)	Yes	Both	Yes
StarCoder 15B	Yes	Y es	х	Yes (unverified)	Yes	Both	Yes
StarCoder2 3B/7B/15B	Yes	Y es	х	Yes (unverified)	Yes	Both	Yes
BERT 110M/340M	Yes	Y es	х	Yes (unverified)	х	Both	х
T5 220M/3B/11B	Yes	Y es	х	х	x	х	х

# **Vision Language Models**

# Vision Language Models

Vision Languag e Models	Pretraining & SFT	PE FT	Align ment	FP8 Training Convergence	TRT/TR TLLM	Convert To & From H ugging Face	Evalu ation
NeVA (LLaVA 1. 5)	Yes	Ye s	x	Yes (unverified)	х	From	х
Llama 3.2 Vision 11B/90B	Yes	Ye s	x	Yes (unverified)	х	From	х
LLaVA Next (LLaVA 1.6)	Yes	Ye s	x	Yes (unverified)	х	From	х

# **Embedding Models**

# **Embedding Models**

Embedding Langu age Models	Pretrainin g & SFT	P EF T	Align ment	FP8 Training Co nvergence	TRT/TR TLLM	Convert To & From H ugging Face	Evalu ation
SBERT 340M	Yes	х	x	Yes (unverified)	х	Both	х
Llama 3.2 Embeddi ng 1B	Yes	х	х	Yes (unverified)	Х	Both	х

#### **World Foundation Models**

# World Foundation Models

World Foundation Models	Post-Training	Accelerated Inference
Cosmos-1.0-Diffusion-Text2World-7B	Yes	Yes
Cosmos-1.0-Diffusion-Text2World-14B	Yes	Yes
Cosmos-1.0-Diffusion-Video2World-7B	Coming Soon	Coming Soon
Cosmos-1.0-Diffusion-Video2World-14B	Coming Soon	Coming Soon
Cosmos-1.0-Autoregressive-4B	Yes	Yes
Cosmos-1.0-Autoregressive-Video2World-5B	Coming Soon	Coming Soon
Cosmos-1.0-Autoregressive-12B	Yes	Yes
Cosmos-1.0-Autoregressive-Video2World-13B	Coming Soon	Coming Soon

## Note

NeMo also supports pretraining for both diffusion and autoregressive architectures text2world foundation models.

# Speech Al

Developing conversational AI models is a complex process that involves defining, constructing, and training

models within particular domains. This process typically requires several iterations to reach a high level of accuracy. It often involves multiple iterations to achieve high accuracy, fine-tuning on various tasks and domain-specific data, ensuring training performance, and preparing models for inference deployment.

\_images/nemo-speech-ai.png

NeMo Framework provides support for the training and customization of Speech AI models. This includes tasks like Automatic Speech Recognition (ASR) and Text-To-Speech (TTS) synthesis. It offers a smooth transition to enterprise-level production deployment with NVIDIA Riva. To assist developers and researchers, NeMo Framework includes state-of-the-art pre-trained checkpoints, tools for reproducible speech data processing, and features for interactive exploration and analysis of speech datasets. The components of the NeMo Framework for Speech AI are as follows:

#### **Training and Customization**

NeMo Framework contains everything needed to train and customize speech models (ASR, Speech Classification, Speaker Recognition, Speaker Diarization, and TTS) in a reproducible manner.

#### **SOTA Pre-trained Models**

- NeMo Framework provides state-of-the-art recipes and pre-trained checkpoints of several ASR and TTS models, as well as instructions on how to load them.
- · Speech Tools
- NeMo Framework provides a set of tools useful for developing ASR and TTS models, including:
  - NeMo Forced Aligner (NFA) for generating token-, word- and segment-level timestamps of speech in audio using NeMo's CTC-based Automatic Speech Recognition models.
  - Speech Data Processor (SDP), a toolkit for simplifying speech data processing. It allows you to represent data processing operations in a config file, minimizing boilerplate code and allowing reproducibility and shareability.
  - Speech Data Explorer (SDE), a Dash-based web application for interactive exploration and analysis of speech datasets.
  - Dataset creation tool which provides functionality to align long audio files with the corresponding transcripts and split them into shorter fragments that are suitable for Automatic Speech Recognition (ASR) model training.
  - Comparison Tool for ASR Models to compare predictions of different ASR models at word accuracy and utterance level.
  - ASR Evaluator for evaluating the performance of ASR models and other features such as Voice Activity
     Detection.
  - Text Normalization Tool for converting text from the written form to the spoken form and vice versa (e.g. "31st" vs "thirty first").
- · Path to Deployment
- NeMo models that have been trained or customized using the NeMo Framework can be optimized and deployed with NVIDIA Riva. Riva provides containers and Helm charts specifically designed to automate the steps for push-button deployment.

#### **Other Resources**

#### **GitHub Repos**

- NeMo: The main repository for the NeMo Framework
- NeMo-Run: A tool to configure, launch and manage your machine learning experiments.
- NeMo-Aligner: Scalable toolkit for efficient model alignment
- NeMo-Curator: Scalable data pre-processing and curation toolkit for LLMs

#### **Getting Help**

Engage with the NeMo community, ask questions, get support, or report bugs.

- NeMo Discussions
- NeMo Issues

#### **Programming Languages and Frameworks**

- Python: The main interface to use NeMo Framework
- Pytorch: NeMo Framework is built on top of PyTorch

#### Licenses

- NeMo Github repo is licensed under the Apache 2.0 license
- NeMo Framework is licensed under the NVIDIA AI PRODUCT AGREEMENT. By pulling and using the container, you accept the terms and conditions of this license.
- The NeMo Framework container contains Llama materials governed by the Meta Llama3 Community License Agreement.

#### **Footnotes**

Currently, NeMo Curator and NeMo Aligner support for Multimodal models is a work in progress and will be available very soon.

# **FAQ**

#### Q: How can I check if my system is affected by the vulnerability?

A: You can check if your system is affected by verifying the version of the NVIDIA NeMo Framework installed. If it is below version 24, your system may be vulnerable.

#### Q: Who reported the security issue CVE-2025-23360?

A: The security issue was reported by Or Peles – JFrog Security. NVIDIA acknowledges their contribution.

# Q: How can I receive future security bulletin notifications?

A: Visit the NVIDIA Product Security page to subscribe to security bulletin notifications and stay informed about product security updates.

#### **Documents / Resources**



# NVIDIA NeMo Framework [pdf] User Guide NeMo Framework, NeMo, Framework

#### References

# • User Manual

#### Manuals+, Privacy Policy

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