

Driving the Generative AI Powered Enterprise with NIMs



Enterprise are on the Generative Al Journey



Explosion

ChatGPT gets announced late in 2022, gaining over 100 million users in just two months. Users of all levels can experience AI and feel the benefits firsthand.



Experimentation

Enterprise application developers kick off POCs for generative AI applications with API services and open models including Llama 2, Mistral, NVIDIA, and others.



Production

Organizations have set aside budget and are ramping up efforts to build accelerated infrastructure to support generative AI in production.



Generative AI Can Learn and Understand Everything



MULTI-MODAL

BRAINWAVES









Enterprises Face Challenges Experimenting with Generative Al Organizations must choose between ease of use and control

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Easy to use APIs for development

Fast path to getting started with Al

Infrastructure limited to managed environment

Data and prompts are shared externally

> **Limited control** for overall generative AI strategy



Run anywhere across data center and cloud

Securely manage data in self hosted environment

Tuning required for different infrastructure

Custom code for APIs and fine-tuned models

Ongoing maintenance and updates



NVIDIA NIM Optimized Inference Microservices

Prebuilt container and Helm chart

Industry standard APIs

Support for custom models

Domain specific code

Optimized inference engines







Accelerated runtime for generative AI

NVIDIA NIM



d	ORACLE	DGX &	D&LL
		DGX Cloud	

Deploy anywhere and maintain control of generative AI applications and data

Simplified development of AI application that can run in enterprise environments

Day 0 support for all generative AI models providing choice across the ecosystem

Improved TCO with best latency and throughput running on accelerated infrastructure

Best accuracy for enterprise by enabling tuning with proprietary data sources

Enterprise software with feature branches, validation and support

Technologies









NVIC Reduces eng

	NVIDIA NIM	
Deployment Time	5 minutes	
API Standardization	Industry standard protocol OpenAI for LLMs, Google Translate Speech	
Pre-Built Engine	Pre-built TRT-LLM engines for NV and community models	User co swee
Triton Ensemble/ BLS Backend	Pre-built with TRT-LLM to handle pre/post processing (tokenization)	
Triton Deployment	Automated	
Customization	Supported – P-tuning and LORA, more planned	
Container Validation	Pre-validated with QA testing	
Support	NVIDIA AI Enterprise - Security and CVE scanning/patching and tech support	

NVIDIA NIM is the Fastest Path to Al Inference

Reduces engineering resources required to deploy optimized, accelerated models

Triton + TRT-LLM Opensource

~1 week

User creates a shim layer (reducing performance) or modify Triton to generate custom endpoints

converts checkpoint to TRT-LLM format and creates and runs eps through different parameters to find the optimal config

User manually sets up + configures

User manually sets up + configures

User needs to create custom logic

No pre-validation

No enterprise support



Inference Microservices for Generative Al NVIDIA NIM is the fastest way to deploy AI models on accelerated infrastructure across cloud, data center, and PC





Google



ADEPT









NVIDIA API Catalog









gettyimages

Microsoft

shutterstsck[.]







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NVIDIA NIM for Every Domain







NVIDIA AI Enterprise

High Performance and Efficient Runtime for Generative AI



Experience and Run Enterprise Generative AI Models Anywhere Seamlessly integrate AI in business applications with NVIDIA AI APIs



Experience Models

Prototype with APIs

Deploy with NIMs







Additional slides to be used as needed



Industry Standard APIs

Text, Speech, Image, Video, 3D, Biology

Triton Inference Server

cuDF, CV-CUDA, DALI, NCCL, Post Processing Decoder

Cloud Native Stack

GPU Operator, Network Operator

Enterprise Management

Health Check, Identity, Metrics, Monitoring, Secrets Management

Kubernetes

NVIDIA NIM



100's of Millions of CUDA GPUs Installed Base

TensorRT and TensorRT-LLM cuBLAS, cuDNN, In-Flight Batching, Memory **Optimization**, FP8 Quantization

Optimized Model Single GPU, Multi-GPU, Multi-Node

Customization Cache

P-Tuning, LORA, Model Weights

NVIDIA CUDA









Triton has 417 packages/libraries across OSS, 3rd party and NVIDIA

Anatomy of a NIM



TensorRT has 333 packages/libraries across OSS, 3rd party and NVIDIA

