Akshay Umesh

🜙 +1-862-423-9778 💌 akshayumesh2000@gmail.com 📊 https://www.linkedin.com/in/akshayumesh/ᠺ https://github.com/Akshay879

Skills

- Programming Languages: Python, Java, Javascript
- Frameworks: PyTorch, Langchain, Expressis, Nodejs
- Core Skills: Machine Learning, Deep Learning, Generative AI, Cross-Modal/Multimodal Generative Modelling, Computer Vision, Natural Language Processing, AWS, RAG, Autoregressive Modelling, SQL

Education

New Jersey Institute of Technology

Sep 2023 - May 2025

M.S., Computer Science

JSS Academy Of Technical Education

Aug 2017 - Jul 2021

B.E., Computer Science Engineering

Research

Micro-Budget Latent Diffusion Transformer for High-Resolution Image Synthesis

Jan 2025 - Mar 2025

- Developed a 1.2B-param Latent Diffusion Transformer (PyTorch) from scratch, achieving a 12.7 FID on the COCO dataset at a \$1314 compute cost; trained for 54.96 8×H100 hours on 25M image-caption pairs (SA1B, JourneyDB, DiffusionDB).
- Engineered a progressive pretraining-finetuning pipeline, seamlessly scaling the model's denoising capabilities across resolutions by dynamically adjusting positional embeddings, enabling effective masked and unmasked denoising representation learning.
- Optimized pretraining by implementing a 75% patch masking ratio, minimizing input sequence length, and integrating a lightweight patchmixer to retain crucial global semantic information, leading to superior representation learning.
- Implemented a ResNet-style Diffusion Transformer, integrating a 6-layer patchmixer and 28-layer DiT backbone, which leverages cross-attention on OpenCLIP text embeddings, noise-modulated information flow, and Expert Choice Mixture of **Experts** within a VAE latent space.
- Modeled the diffusion process as a Stochastic Differential Equation (SDE), as explained in score-based generative models and denoising score matching.
- Scaled model training efficiently using Fully Sharded Data Parallel (FSDP) and the Composer Hydra framework.

Projects

Transformer based Autoregressive Image generation using VQGAN

Nov 2024 - Jan 2025

- Developed a generative model in PyTorch from scratch, combining a 108M-param VQGAN (vocab size: 8192) for discrete neural representation learning and a 91.5M-param GPT-2 (257-token context) for autoregressive image generation; implemented a VQ-VAE with adversarial training using a ResNet-based encoder-decoder, self-attention, and multi-objective loss (reconstruction, perceptual, adversarial, quantization, commitment) to learn structured latent codes.
- Designed quantization with gradient flow reparameterization for effective backprop through discrete latents; built a pre-layer norm GPT-2 with 4-headed causal self-attention to model chained categorical posteriors; trained on the Oxford Flowers dataset (8189 images) in BFloat16 mixed precision for high-fidelity synthesis.

Foundation Model for Text Completion

Sep 2024 - Nov 2024

- Implemented a GPT-2-based autoregressive language model with 124M trainable parameters and a 1024-token context length from scratch.
- Trained on 10B tokens (FineWebEdu dataset), achieving a validation loss of 3.1440 and HellaSwag score of 30.22%, outperforming OpenAI-GPT2 124M (trained on 100B tokens) in learning efficiency (~10x).
- Scaling to **50B** tokens matched OpenAI-GPT3 124M's HellaSwag score of **33.70%**.
- Optimized training on an 8x A100 SXM cluster (12 hours, 50B tokens) using PyTorch DDP, BFloat16 mixed precision, flash attention, and kernel fusion.

U-Net for Image Reconstruction

- Dec 2024 Jan 2025
- Implemented a U-Net architecture in PyTorch from scratch, incorporating a **4-stage convolutional encoder** (**64->512 channels**) and a **4-stage decoder** (**512->64 channels**) for high-fidelity image reconstruction.
- Designed an encoder with multi-stage downsampling, self-attention (1024 latent channels), and residual feed-forward layers, optimizing for 16x compression in latent space (256x256 ->16x16 feature maps).
- Developed a decoder using **transposed convolutions** (2x upsampling per stage) and skip connections, improving feature retention and generating realistic RGB images (256x256, 3 channels).
- Applied custom Kaiming initialization for weight scaling, reducing initialization variance and stabilizing training.
- Trained using BFloat16 mixed precision, achieving a 50% memory footprint reduction, enabling larger batch sizes on A100 GPUs.

Work Experience

Accenture Sep 2021 - Aug 2023

QA Team-Kaiser Permanente

- Worked in the QA team for KP-HIE using **Selenium**, **Jira**, and **Jenkins** framework, improving testing processes and ensuring project milestones were met on time
- Embraced Agile methodologies to expedite project delivery, slashing timelines by 30% and boosting team adaptability.
- Enhanced testing efficiency and accuracy for Kaiser's **Genomics** and **NIPT** projects, leading to more reliable test outcomes and improved project quality
- Ensured adherence to quality assurance and testing standards, resulting in the delivery of reliable and robust software solutions