Connect 2025 Supercharging Generative AI: KleidiAI [™], PyTorch, and Arm[®] Neoverse

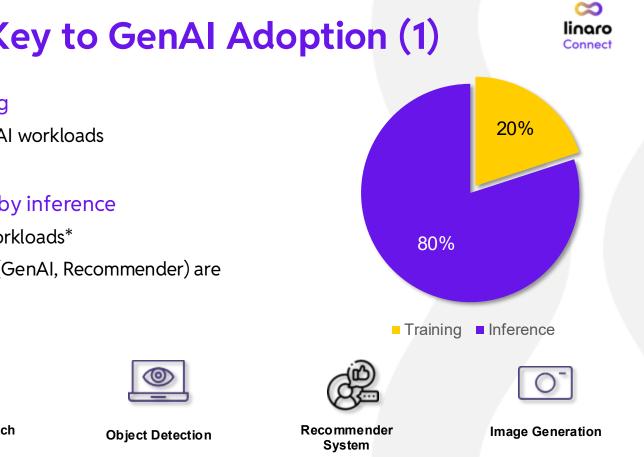
Nikhil Gupta Arm Ltd.

Arm Neoverse: Powering GenAl at Scale



- Arm Neoverse and Armv9: Foundation for Cloud, HPC and GenAI
- Scalable Vector Extension 2 (SVE2): Unlocking Data-Level Parallelism
- Massive Compute and Hardware Acceleration for AI and HPC
- Widespread Cloud Adoption: Neoverse in Major Hyperscalers





Inferencing is Key to GenAl Adoption (1)

Training is the tip of iceberg

Training = Only 15-20% of AI workloads ٠

The Growth of AI is driven by inference

- Inference = ~80-85% AI workloads* ٠
- Real-world AI applications (GenAI, Recommender) are ٠ inference heavy





Natural Language Processing

Automatic Speech Recognition

Inferencing is Key to GenAl Adoption (2)



Efficient Inference needs high throughput, low latency and scalable performance

Challenges with Efficient Inferencing on CPUs

- Memory bandwidth bottlenecks: Large models require high memory access rates
- **Limited parallelism**: CPUs must maximize vectorization and threading efficiency
- Handling large models: Scaling inference efficiently across CPU cores

Solution Requirements

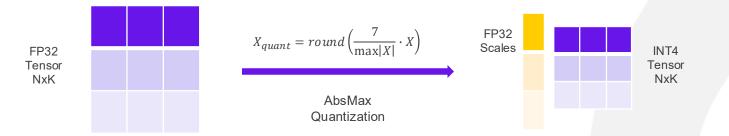
- Fast and Memory-Efficient: Optimize vectorization and bandwidth usage
- **CPU-optimized**: Leverage Arm Neoverse features (low precision instructions)
- **Scalable** across model types, sizes and newer platforms

Supercharging GenAl at scale

Low-bit Quantization



- Converts high-precision numbers (e.g., FP32) to lower-precision formats (e.g., INT4).
- Performs computation directly in the quantized form.
- 8× higher data density: INT4 model can fit eight times more parameters into the same cache space as FP32.
- Higher effective memory bandwidth: more data moved per cycle, less traffic to main memory.
- Enables Arm **low-bit intrinsics** (**i8mm, dotprod**) to achieve much **higher compute throughput** than FP32 matmuls.
- Possible to preserve near-original model accuracy with LLMs





Scalability Through PyTorch Integration

- **PyTorch-native integration** ensures **seamless adoption** and **ease of use**.
- Leverages **PyTorch's massive ecosystem** tools, libraries, and active community support.
- **Out-of-the-box compatibility** for newly trained GenAI models no custom rework needed.
- PyTorch is **the Leading training framework** for GenAI (e.g., Llama, Gemma, Stable Diffusion).
- **Scalable** across model sizes and deployment targets from raspberry-pies to arm neoverse.

orm + Orch

Leveraging KleidiAI & PyTorch on Arm Neoverse

Introduction to KleidiAI





arm Kleidi

- Optimized Al Micro Kernels: Provide low-level, highly optimized microkernels designed for ARM CPUs
- Quantized Matmul Kernels: Designed for GenAl use-cases
- Fast Packing Routines: Offers fast weight & input packing kernels for memory efficient computations
- Independent & Stateless: No memory allocation & dependencies

Target Users

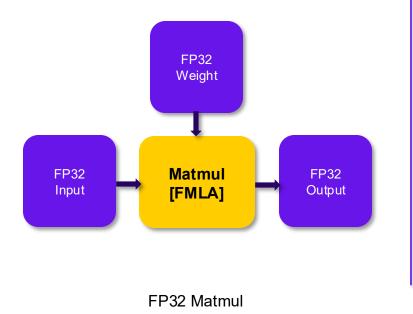
- AI Framework Developers
- AI SDK Developers
- Al pipeline Developers

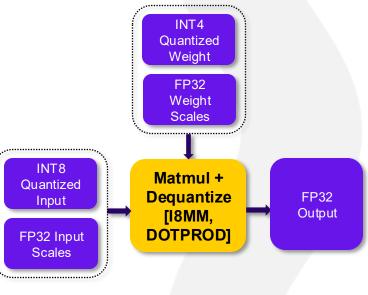
KleidiAI Domain

- Classic ML
- GenAl
- Agentic Al

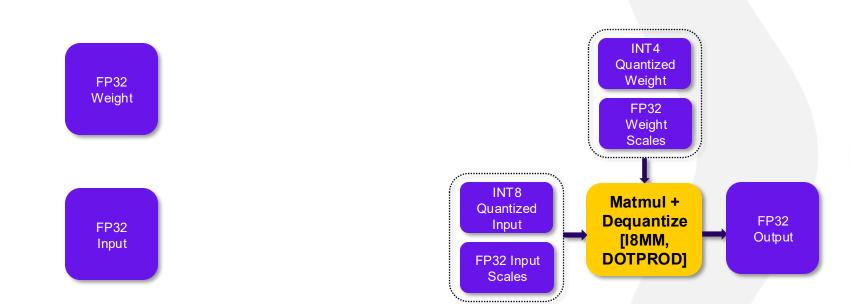


KleidiAI 4-bit Matmul





INT4 KleidiAl Matmul

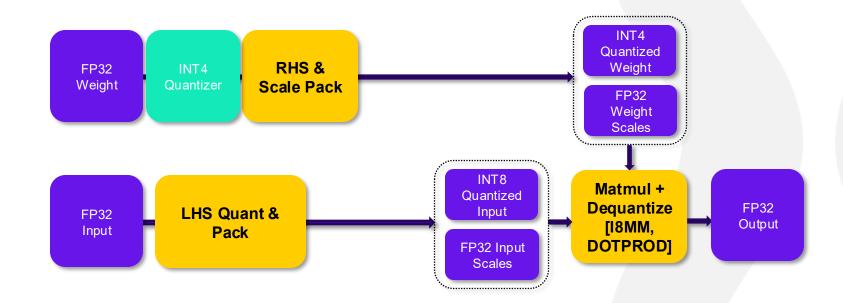


Input and Weight Packing



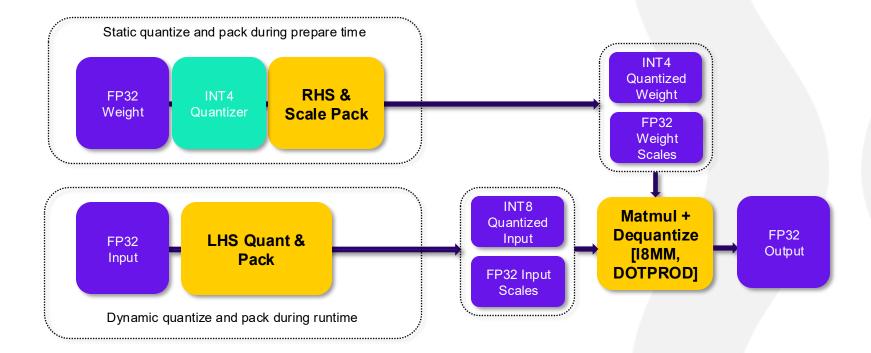


Input and Weight Packing





Input and Weight Packing



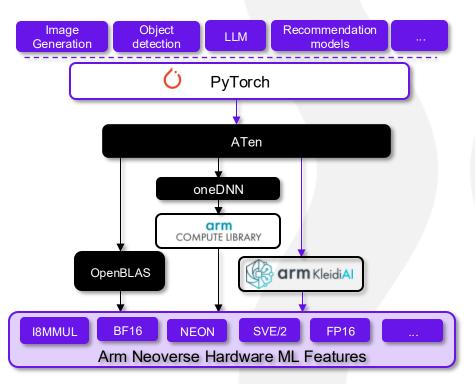
KleidiAl Integration into PyTorch



- Integrated directly into PyTorch backend via custom ATen ops.
- Introduces two new 4-bit quantized ops:

torch.ops.aten._dyn_quant_pack_4bit_weight

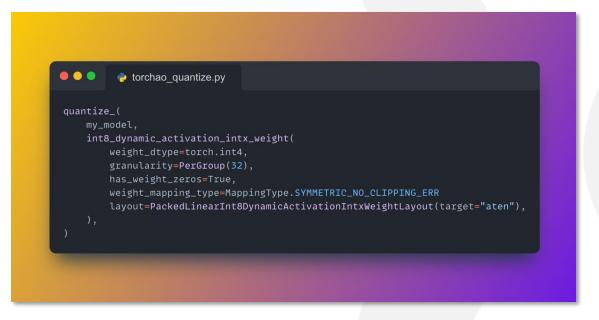
torch.ops.aten._dyn_quant_matmul_4bit





TorchAO - Simplifying Quantization

- One API call to quantize any PyTorch model
- Seamless Integration with PyTorch + KleidiAI
- Precise control over model accuracy
- Layer-wise quantization support



Text Generation Speedups



User: The lemon tree produces a pointed oval yellow fruit. Botanically this is a hesperidium, a modified berry with a tough, leathery rind. The rind is divided into an outer colored layer or zest, which is aromatic with essential oils, and an inner layer of white spongy pith.

Assistant: [thinking]

13 tok/s | Llama 3.1 8B FP32 PyTorch User: The lemon tree produces a pointed oval yellow fruit. Botanically this is a hesperidium, a modified berry with a tough, leathery rind. The rind is divided into an outer colored layer or zest, which is aromatic with essential oils, and an inner layer of white spongy pith.

Lemons need a minimum temperature of around 7 °C, so they are not hardy year-round in temperate climates, but become hardier as they mature. Citrus require minimal pruning by trimming overcrowded branches, with the tallest branch cut back to encourage bushy growth. Throughout summer, pinching back tips of the most vigorous growth assures more abundant canopy development. As mature plants may produce unwanted, fast-growing shoots called water shoots, these are removed from the main branches at the bottom or middle of the plant. There is reputed merit in the tradition of urinating near a lemon tree. Lemons need a minimum temperature of around 7 °C, so they are not hardy year-round in temperate climates, but become hardier as they mature. Citrus require minimal pruning by trimming overcrowded branches, with the tallest branch cut back to encourage bushy

> 68 tok/s | Llama 3.1 8B INT4 PyTorch + KleidiAl

Performance Impact with KleidiAI





Arm Neoverse 2 | 32 Threads | PyTorch Compile

Key Takeaways

- Arm Neoverse: Designed for AI at Scale
- Inferencing: The Driver of GenAl
- Low-Bit Quantization: Core to Performance Gains for GenAl
- **PyTorch Integration:** Mature Ecosystem with ease of Use at Scale



Resources & References



- Arm Neoverse 2 https://www.arm.com/products/silicon-ip-cpu/neoverse/neoverse-v2
- KleidiAl <u>https://gitlab.arm.com/kleidi/kleidiai</u>
- PyTorch https://github.com/pytorch/pytorch
- TorchAO https://github.com/pytorch/ao
- Arm Learning Path https://learn.arm.com/learning-paths/servers-and-cloud-computing/pytorch-llama/

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Thank You!