Inference trillion parameters everywhere

Redesign LLM for everyday device



LLM is better with more parameters, but it's memory bound







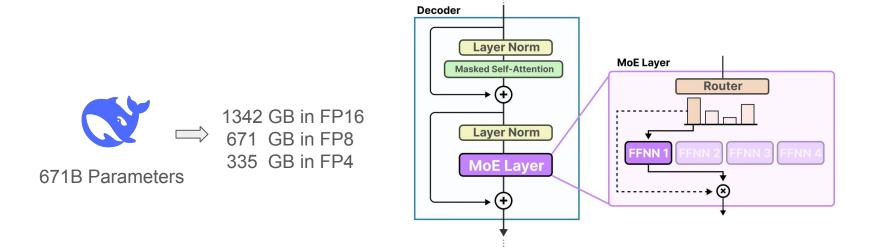


1999\$ + PC cost	2999\$	3600\$	1999\$
RTX5090 GPU	DGX Spark	Mac Studio M4	AMD AI Max+ 395
32 GB VRAM	128 GB UM	128 GB UM	128 GB UM
1792 GB/sec	273 GB/sec	273 GB/sec	256 GB/sec
30B model max	100B model max	100B model max	100B model max
	32B model ~20 t/s	32B model ~20 t/s	32B model ~20 t/s

Personal AI PC: up to 32B model usable



MoE: reduce computing, but not memory



A **router** (also called a **gate network**) is added which is trained to choose which expert to choose for a given **token**.



Problem:

- MoE architecture doesn't fit general PC's architecture
- ☐ High bandwidth memory is expensive and limited
- Distilled models lose task specific ability dramatically
- Every vendor has their own DSL & runtime

Solution:

- Break LLM into smaller pieces
- ☐ Load them into HBM by tasks
- "Vulkanised"



Training Process



task specific corpora hundreds of categories







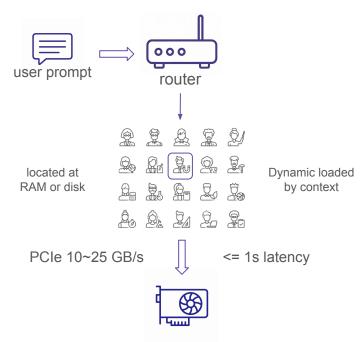
source large model 100B ~ 1T parameters

agent based router small encoder classifier



distilled small agent models 100+ of 1~7B models

Inference Process



Stay in HBM for the same context

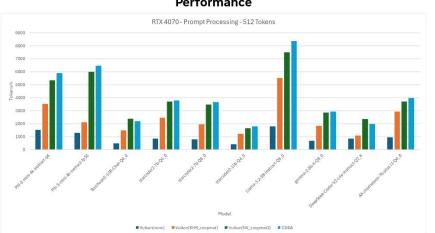


Potential of Vulkan

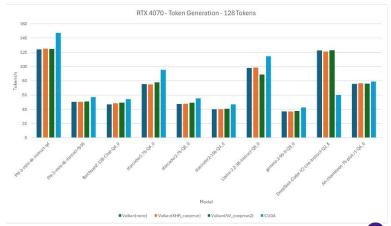
- Cross platform
- High performance
- Example: ggml



Performance



Performance







Outcomes:

- Unbound inference at edge side
 - Much cheaper inference
 - ☐ Makes private large scale inference possible
 - Cross-platform code base
- Modularity brings scalability
 - ☐ Fine tuned agents instead of prompt-based
 - Much lower cost for continual knowledge update
- Sustainability
 - ☐ Build ecosystem for difference models
 - provide long term service for deployment system



Business Model

- APP based platform for models. On PC and mobiles.
- Target all customers:
 - To Business: Privacy
 - To Individual: Flexibility
- We provide:
 - Thousands of small proprietary models varied by sizes and tasks
 - On demand subscription and download
 - Automatic router to load model by context
 - ☐ Ecosystem for small models



Business case: to individual

Aim for high performance, easy access, lower cost, privacy and flexibility.

- ☐ A high school student is preparing for exam, parabase has models trained with latest exam questions.
- ☐ A gaming PC owner want to use his discrete GPU as a personal AI host.
- A French learner wants lower latency and a more stable connection to an LLM, so they can run a French language model directly on their phone.
- An author or innovator wants to use an LLM for support but doesn't want to risk their ideas being exposed to the service provider.
- **□** ...



Business case: to enterprise

Aim for local inference and fine-tuning for full control, and low cost and risk for continual knowledge updating.

- ☐ A university department wants to train proprietar models tailored to their subject matter for student use.
- A law firm wants to train a model using financial case data from the past five years.
- ☐ A car company wants its AI assistant to remain accessible even in tunnels.
- An AI company wants to host its own LLM server to avoid relying on OpenAI

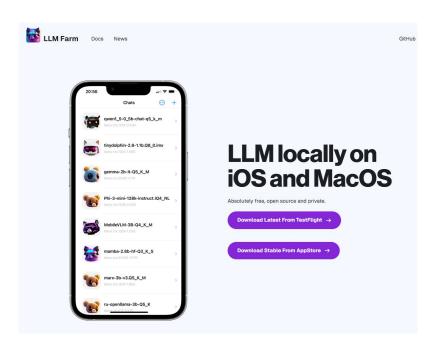


Why now?

- Private Al environment on its way to launch
 - > AI PCs
 - ➤ Vendor's SDK
 - Efficient hardware
- Dense model shows a good performance
 - ➤ Llama 1B, 3B, 8B
 - > Qwen 0.5B, 1.5B, 3B, 7B
 - > 0.5~16 GB: fits in most HBM.



Other players: LLM farm



Absolutely free

LLM Farm provides all features absolutely free of charge! No hidden fees, subscriptions or feature limitations - all features are available for use at no additional cost.



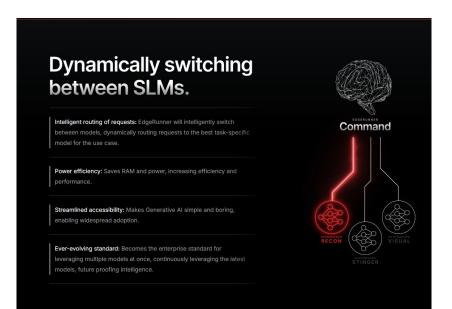


Open Source

The core is a Swift library based on llama.cpp, ggml and other open source projects that allows you to perform various inferences. A class hierarchy has been developed that allows you to add your own inference.



Other players: Edgerunner





Our founding team has deep experience in developing SOTA open source foundation models, optimizing models for edge use cases, and building technology that solves real-world challenges.

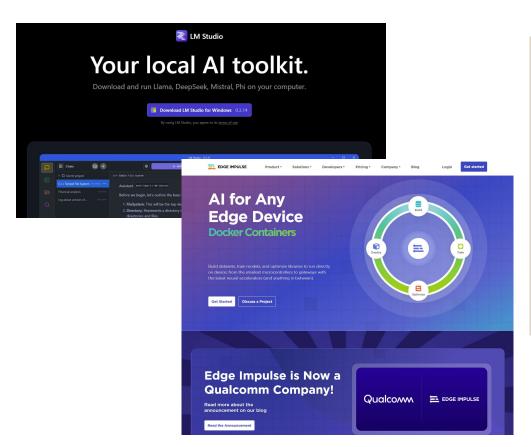
We advocate for the deployment of Generative Al locally and privately, responding to data gravity and privacy concerns. Our strategy involves developing a suite of domain-specific Small Language Models (SLMs) that operate synergistically. This catalyzes the start of collective intelligence, where multiple SLMs interact and collaborate to solve problems.

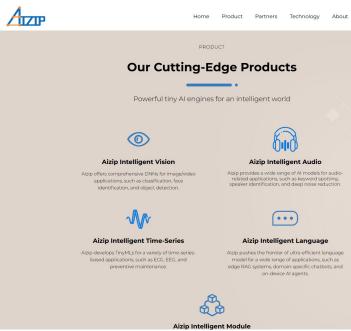
Generative AI will transform all legacy unstructured data into relevant intelligence that you can interact with through natural language like a human conversation. In the future, you will be able to speak to your data and it will speak back to you.

We believe technology this transformative should be under your control, living on your device, and isolated from external networks to ensure a private and hyper-personalized experience.



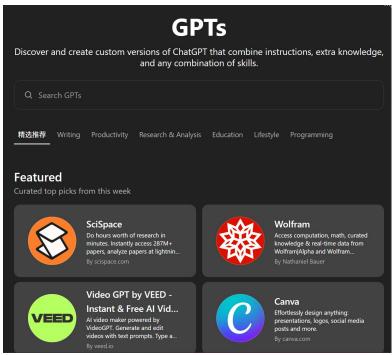
Other players: to customer local inference tool chain





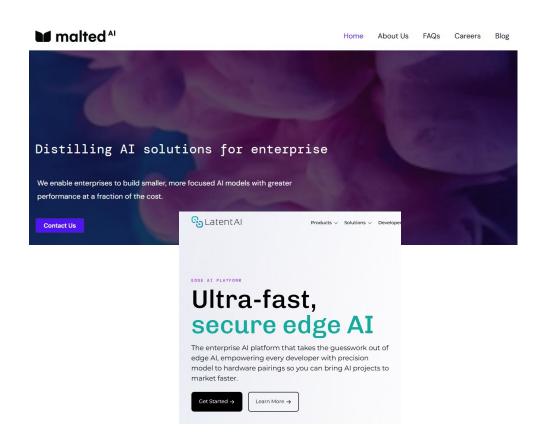


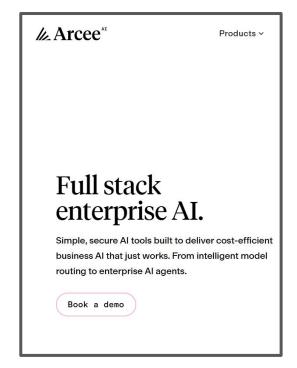
Other players: Prompt based agents





Other players: Enterprise demand for private inference







Related Research

- → "Beyond distillation: Task-level mixture-of-experts for efficient inference", Google Al. 2021
 - * "Determines routing based on task IDs, ensuring that different tasks are routed to distinct experts, thereby effectively minimizing interference between tasks. Compared to token-level routing strategies, this approach only requires loading a subset of experts relevant to the current task during inference, rather than loading all experts of the entire model. This results in reduced communication costs between devices and lower memory usage. "
 - "A comprehensive survey of mixture-of-experts: Algorithms, theory, and applications". Siyuan Mu and Sen Lin, 2025
- → "Modular deep learning", Google Deepmind. 2023
- → "CoServe: Efficient Collaboration-of-Experts (CoE) Model Inference with Limited Memory", Beihang University, ASPLOS '25



Expense:

? for software development.

Fine tuning each small model < ?\$. ?\$ GPU cost in total if rent.

Timeline:

Model training: 6~? months.

App Development: 6~? months.

parabase.ai Active

Renewal Date: 4/22/2027 Renew Now

domain name is secured:)

