

**CLOSED MODELS**  
**OPEN WEIGHT MODELS**  
**OPEN SOURCE MODELS**

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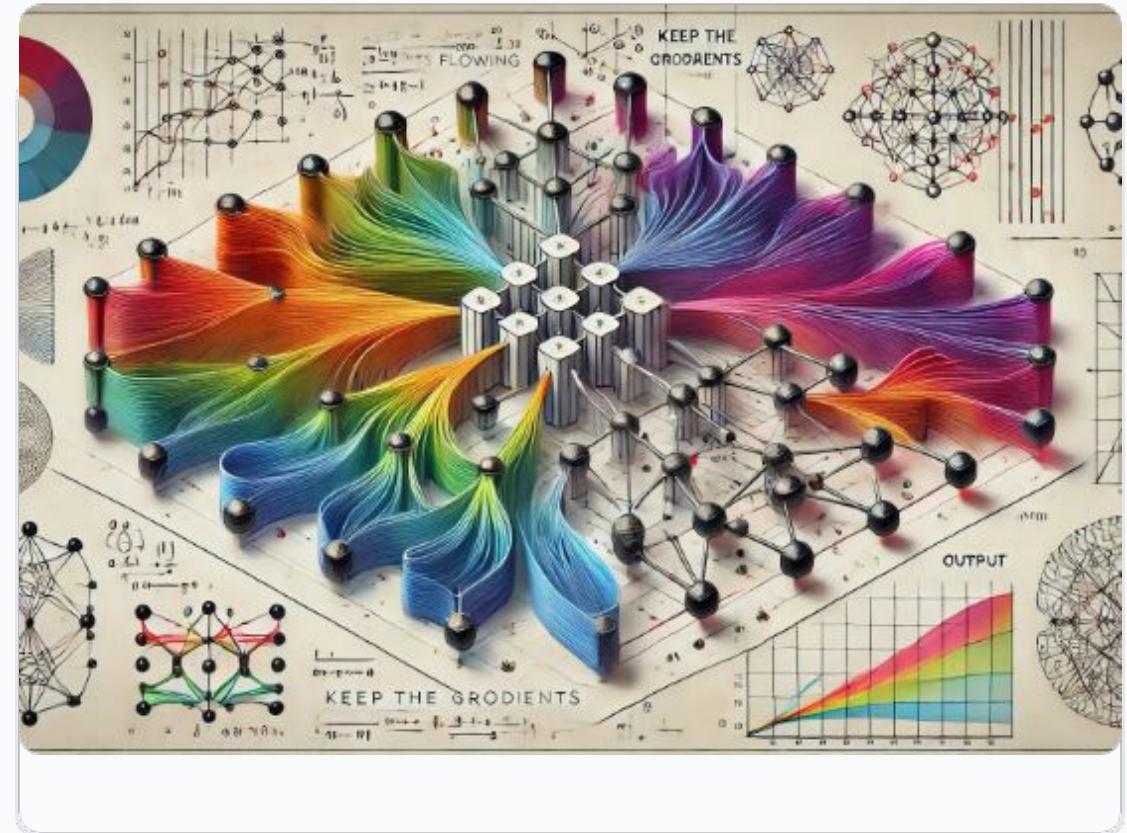
# WHAT IS AN AI "MODEL"?

For today's topic, a model is a **Large Language**

**Model (LLM)**: a neural network trained to predict the next token in a sequence.

We will discuss and compare three characteristics of models when it comes to how "open" they are:

- **The Recipe:** The training code, architecture and datasets used to calculate the weights.
- **Parameters (Weights):** Numerical values defining how data is processed. These are the "learned intelligence."
- **Inference:** Executing the weights on a GPU to generate a result based on user input.

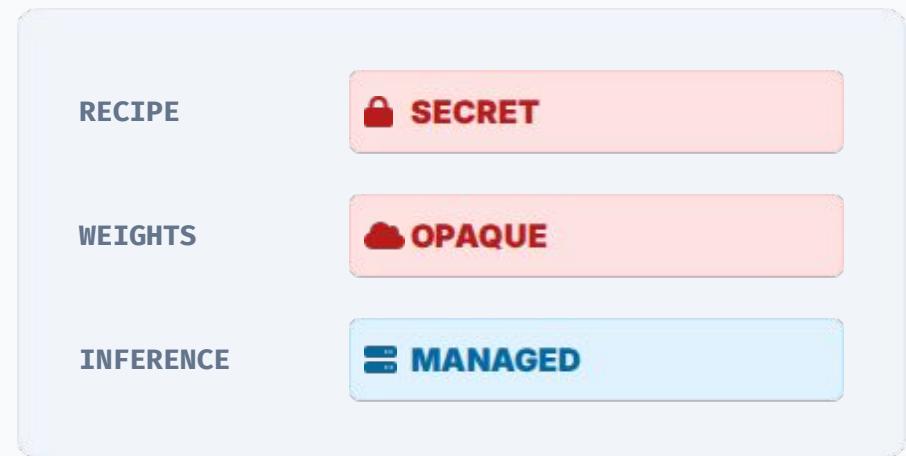


# 01. CLOSED (PROPRIETARY) MODELS

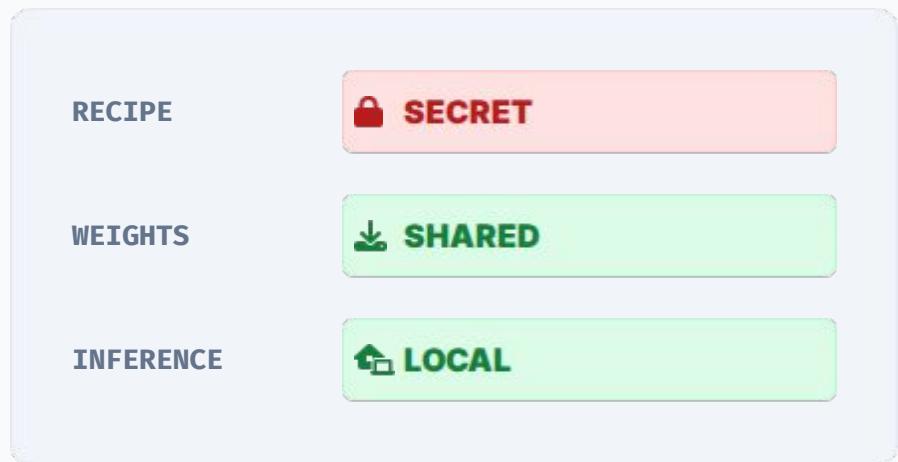
## The Vendor Lock-In Reality

Proprietary models are managed entirely by providers. You are fully dependent on their infra, alignment, and internal updates.

- ❖ **Total Dependency:** Every feature relies on the provider's uptime, pricing, and API accessibility.
- ❖ **Silent "Nerfing":** Models are updated "behind-the-scenes." A prompt that works today might fail tomorrow due to mid-cycle weight changes.
- ❖ **Opaque Guardrails:** Mandatory safety filters can change without warning, breaking specific business or logic use-cases.



## 02. OPEN WEIGHT MODELS



### Immutable Sovereignty

Models where the final weights are shared. Once you download the file, it is yours to control and run indefinitely.

- **Version Immutability:** The model never changes unless YOU update it. No silent nerfing or mid-project prompt drift.
- **Local Execution:** Deploy air-gapped on private hardware for maximum data sovereignty and privacy.
- **Full Control:** Ability to apply custom fine-tuning (LoRA), quantization (GGUF/EXL2), and in some cases even remove guardrails.

# 03. OPEN SOURCE MODELS

## The Glass Box Paradigm

Total transparency. Providers release the weights, the full training datasets, and the code used to train them.

- 💡 **Scientific Rigor:** Public datasets allow for auditable research on how models actually learn and generalize.
- 🔍 **Auditability:** Scrutinize pre-training data mixtures for bias, copyright, and safety training effectiveness.
- 📅 **Total Lineage:** Absolute control over the model's history. No proprietary "secret sauce" in the recipe.



# Performance comparison



**Frontier Closed Models** still hold the absolute "bleeding edge" in multimodal integration (native video/audio reasoning) and complex agentic reliability.

**Open-Weight Models** often reach parity with these closed models in coding and mathematics within one quarter of their release.

**Open Source Models** are generally about **9–12 months behind**, as the massive compute and data cleaning required for frontier-level training are still largely gated by corporate resources.

## // BENCHMARK SOURCES & METHODOLOGY

**LMSYS Arena:** Human-preference Elo ratings for GPT-5.2 and Claude 4.5 Opus.

**MMLU-Pro / GPQA:** Graduate-level reasoning and STEM task understanding scores.

**SWE-bench Verified:** Real-world software engineering issue resolution benchmarks.

OPENAI

## GPT-5.2

- > Unified Reasoning: Default CoT logic
- > 400,000 Token Context Window
- > Agentic Orchestration: 50+ sub-tasks

ANTHROPIC

## Claude 4.5 Opus

- > Effort Parameter: Reflection toggle
- > 64,000 Token single-pass output limit
- > Vision-based UI interaction ("Zoom")

# CLOSED FRONTIER

GOOGLE

## Gemini 3 Pro

- > Native Temporal Video/Audio Processing
- > 2M - 10M Context Window (Ultra)
- > Deep Workspace & Disco integration

OPENAI

## o1-pro

- > Reasoning-time compute scaling
- > RL-trained internal thinking cycles
- > Hidden CoT for security & logic

META

## Llama 4 Behemoth

- > 405B+ Params; 15T token pre-training
- > Native Multi-Token Prediction (MTP)
- > Scout variant: 10M context capability

OPENAI

## gpt-oss-120b

- > MXFP4 Quantization: Single 80GB GPU
- > 117B MoE (5.1B active per token)
- > Optimized for agentic tool-use calls

# OPEN WEIGHT LEADERS

DEEPSEEK

## DeepSeek-R1

- > MLA Architecture (671B MoE)
- > RL-based logic thinking steps (CoT)
- > MIT License (Unrestricted)

MINIMAX

## Minimax M2.1

- > Interleaved Thinking Verification
- > iOS/Android native UI optimization
- > 204,000 Token local context window

MOONSHOT AI

## KIMI 2.5

- > Multimodal (image and video)
- > 15T tokens training
- > Performance rivalling closed models

AI2

## OLMo 2

- > Full Dolma v3 Dataset Disclosure
- > 500+ intermediate checkpoints public
- > Scientific glass-box architecture

# TRULY OPEN SOURCE

DATACOMP-LM

## DCLM-7B v2

- > Dataset: 240T token curated pool
- > Public filtering & quality-scoring scripts
- > Maximum parameter efficiency (SOTA)

ZHIPU AI / THUDM

## GLM-4.7-9B

- > Agentic Terminal: Native self-correction
- > 200K long-context code window
- > SFT & RLHF strategies disclosed

```
$ cat architecture_summary.csv
```

METRIC	CLOSED	OPEN WEIGHT	OPEN SOURCE
Weights Access	Locked (API)	Public (Local)	Public (Local)
Training Recipe	Secret	Secret	Public & Auditable
Model Stability	Variable (Nerfing)	Fixed (Immutable)	Fixed (Immutable)
Strategic Edge	Frontier Scaling	Sovereignty	Scientific Rigor

# QUESTIONS