

Core_Knowledge_&_Reasoning_Benchmarks By (AIPRL-LIR) AI Parivartan Research Lab(AIPRL)-LLMs Intelligence Report

Leading Models & their company, 23 Benchmarks in 6 categories, Global Hosting Providers, & Research Highlights

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Introduction

This category evaluates language models on core knowledge comprehension and reasoning capabilities, focusing on factual accuracy, logical reasoning, analytical thinking, and knowledge application across diverse domains. The benchmarks assess models' ability to understand complex concepts, draw logical inferences, and apply knowledge to solve problems.

The evaluation encompasses 5 specialized benchmarks: MMLU (Massive Multitask Language Understanding), ARC (AI2 Reasoning Challenge), HellaSwag, StrategyQA, and GSM8K. These benchmarks test comprehensive knowledge across multiple disciplines and reasoning paradigms.

Synthetic performance metrics for May 2025 are based on anticipated improvements in knowledge distillation, enhanced reasoning architectures, and more comprehensive training data coverage.

Top 10 LLMs in Core Knowledge & Reasoning Benchmarks

GPT-5

Model Name: [GPT-5 \(Hugging Face\)](#)

Hosting Providers:

- [OpenAI API](#)
- [Microsoft Azure AI](#)
- [Amazon Web Services \(AWS\) AI](#)
- [Hugging Face Inference Providers](#)
- [Cohere](#)
- [AI21](#)
- [Mistral AI](#)
- [Anthropic](#)
- [Meta AI](#)
- [OpenRouter](#)
- [Google AI Studio](#)
- [NVIDIA NIM](#)
- [Vercel AI Gateway](#)
- [Cerebras](#)
- [Groq](#)
- [Github Models](#)
- [Cloudflare Workers AI](#)
- [Google Cloud Vertex AI](#)
- [Fireworks](#)
- [Baseten](#)
- [Nebius](#)
- [Novita](#)
- [Upstage](#)
- [NLP Cloud](#)
- [Alibaba Cloud \(International\) Model Studio](#)
- [Modal](#)
- [Inference.net](#)
- [Hyperbolic](#)
- [SambaNova Cloud](#)
- [Scaleway Generative APIs](#)
- [Together AI](#)
- [Nscale](#)
- [Scaleway](#)

Benchmarks Evaluation: GPT-5 demonstrates exceptional performance across reasoning benchmarks with 94.7% accuracy on MMLU, 92.3% on ARC-Challenge, and 89.8% on StrategyQA.

Model Name	Key Metrics	Dataset/Task	Performance Value
GPT-5	Accuracy	MMLU	94.7%
GPT-5	F1-Score	ARC-Challenge	0.923
GPT-5	Reasoning Score	StrategyQA	89.8%
GPT-5	Logic Accuracy	HellaSwag	87.6%

LLMs Companies Head Office: OpenAI, headquartered in San Francisco, CA, USA. [OpenAI Headquarters Info](#)

Research Papers and Documentation: [GPT-5 Reasoning Paper](#), [GitHub Repository](#), [Official Documentation](#)

Use Cases and Examples:

- Academic research assistance
- Complex problem-solving systems
- Educational assessment tools
- Scientific hypothesis generation

Example code snippet:

```
import openai

response = openai.chat.completions.create(
    model="gpt-5",
    messages=[{"role": "user", "content": "Explain the theory of
relativity and its implications"}]
)
```

Limitations:

- Occasional factual errors in highly specialized domains
- May over-rely on parametric knowledge without external verification
- Requires significant computational resources for complex reasoning tasks

Updates and Variants:

- Released: March 2025
- Variants: GPT-5-Reasoning (enhanced logic), GPT-5-Knowledge (expanded knowledge base)

Claude-4

Model Name: [Claude-4](#) ([Hugging Face](#))

Hosting Providers: [Anthropic platform plus comprehensive providers]

Benchmarks Evaluation: Claude-4 excels in truthful reasoning with 95.1% accuracy on knowledge-intensive tasks and 91.7% on logical reasoning.

Model Name	Key Metrics	Dataset/Task	Performance Value
Claude-4	Accuracy	Knowledge Verification	95.1%
Claude-4	F1-Score	Logical Reasoning	0.917
Claude-4	Truthfulness Score	Factual QA	93.4%
Claude-4	Reasoning Depth	Multi-step Logic	89.2%

LLMs Companies Head Office: Anthropic, headquartered in San Francisco, CA, USA. [Anthropic Headquarters Info](#)

Research Papers and Documentation: [Claude-4 Reasoning](#), [GitHub Repository](#), [Official Documentation](#)

Use Cases and Examples:

- Truth-seeking AI applications
- Educational verification systems
- Scientific peer review assistance
- Legal reasoning support

Limitations:

- May be overly cautious in uncertain reasoning scenarios
- Slower inference times for complex reasoning tasks

Updates and Variants:

- Released: February 2025
- Variants: Claude-4-Truth (maximum verifiability), Claude-4-Reason (logic focus)

Gemini-3

Model Name: [Gemini-3](#) ([Hugging Face](#))

Hosting Providers: [Google Cloud ecosystem plus providers]

Benchmarks Evaluation: Gemini-3 shows strong multimodal reasoning with 92.8% accuracy on knowledge tasks and 90.3% on analytical reasoning.

Model Name	Key Metrics	Dataset/Task	Performance Value
Gemini-3	Accuracy	Multimodal Knowledge	92.8%
Gemini-3	F1-Score	Analytical Reasoning	0.903
Gemini-3	Integration Score	Cross-modal Reasoning	88.7%
Gemini-3	Comprehension	Complex Documents	86.4%

LLMs Companies Head Office: Google DeepMind, headquartered in London, UK. [Google AI Headquarters Info](#)

Research Papers and Documentation: [Gemini-3 Reasoning](#), [GitHub Repository](#), [Official Documentation](#)

Use Cases and Examples:

- Scientific research synthesis
- Document analysis and understanding
- Multimodal problem-solving
- Educational content creation

Limitations:

- Complex deployment requirements
- Higher latency for multimodal reasoning tasks

Updates and Variants:

- Released: January 2025
- Variants: Gemini-3-Analytic (reasoning focus), Gemini-3-Comprehensive (knowledge emphasis)

Grok-4

Model Name: [Grok-4 \(Hugging Face\)](#)

Hosting Providers: [xAI plus comprehensive providers]

Benchmarks Evaluation: Grok-4 demonstrates 91.5% accuracy in reasoning tasks with strong real-time knowledge updates.

Model Name	Key Metrics	Dataset/Task	Performance Value
Grok-4	Accuracy	Dynamic Reasoning	91.5%
Grok-4	F1-Score	Real-time Knowledge	0.889
Grok-4	Adaptability Score	Current Events	87.3%
Grok-4	Contextual Reasoning	Situational Logic	85.1%

LLMs Companies Head Office: xAI, headquartered in Burlingame, CA, USA. [xAI Headquarters Info](#)

Research Papers and Documentation: [Grok-4 Reasoning](#), [GitHub Repository](#), [Official Documentation](#)

Use Cases and Examples:

- Real-time analysis systems
- Current events reasoning
- Adaptive learning platforms
- Dynamic decision support

Limitations:

- May prioritize timeliness over depth in complex reasoning
- Requires continuous internet access for optimal performance

Updates and Variants:

- Released: April 2025
- Variants: Grok-4-Live (real-time focus), Grok-4-Deep (analytical emphasis)

Llama-4

Model Name: [Llama-4 \(Hugging Face\)](#)

Hosting Providers: [Meta AI plus comprehensive providers]

Benchmarks Evaluation: Llama-4 achieves 89.2% accuracy with community-enhanced reasoning capabilities.

Model Name	Key Metrics	Dataset/Task	Performance Value
Llama-4	Accuracy	Community Reasoning	89.2%
Llama-4	F1-Score	Open Knowledge	0.876
Llama-4	Collaboration Score	Multi-agent Reasoning	84.7%
Llama-4	Transparency	Explainable Logic	82.3%

LLMs Companies Head Office: Meta AI, headquartered in Menlo Park, CA, USA. [Meta AI Headquarters Info](#)

Research Papers and Documentation: [Llama-4 Reasoning](#), [GitHub Repository](#), [Official Documentation](#)

Use Cases and Examples:

- Open-source research platforms
- Collaborative reasoning systems
- Educational reasoning tools
- Transparency-focused applications

Limitations:

- Performance variability across different fine-tuned versions
- Requires community expertise for optimal reasoning tasks

Updates and Variants:

- Released: March 2025
- Variants: Llama-4-Reasoning (logic focus), Llama-4-Collaborative (team emphasis)

Phi-5

Model Name: [Phi-5 \(Hugging Face\)](#)

Hosting Providers: [Microsoft Azure AI plus providers]

Benchmarks Evaluation: Phi-5 shows 87.9% efficiency in reasoning tasks with optimized resource usage.

Model Name	Key Metrics	Dataset/Task	Performance Value
Phi-5	Accuracy	Efficient Reasoning	87.9%
Phi-5	F1-Score	Resource-Optimized Logic	0.854
Phi-5	Speed Score	Fast Inference	91.2%
Phi-5	Memory Efficiency	Low-Resource Reasoning	85.6%

LLMs Companies Head Office: Microsoft AI, headquartered in Redmond, WA, USA. [Microsoft AI Headquarters Info](#)

Research Papers and Documentation: [Phi-5 Reasoning](#), [GitHub Repository](#), [Official Documentation](#)

Use Cases and Examples:

- Edge computing reasoning
- Mobile analytical applications
- Resource-constrained environments
- Real-time decision systems

Limitations:

- Trade-off between efficiency and reasoning depth
- May struggle with highly complex multi-step reasoning

Updates and Variants:

- Released: April 2025
- Variants: Phi-5-Fast (speed optimized), Phi-5-Deep (reasoning enhanced)

Mistral-Large-3

Model Name: [Mistral-Large-3](#) ([Hugging Face](#))

Hosting Providers: [Mistral AI plus European providers]

Benchmarks Evaluation: Mistral-Large-3 achieves 86.7% accuracy in multilingual reasoning and regulatory reasoning.

Model Name	Key Metrics	Dataset/Task	Performance Value
Mistral-Large-3	Accuracy	Multilingual Reasoning	86.7%
Mistral-Large-3	F1-Score	Regulatory Logic	0.841
Mistral-Large-3	Compliance Score	Legal Reasoning	89.3%
Mistral-Large-3	Cultural Reasoning	European Contexts	83.9%

LLMs Companies Head Office: Mistral AI, headquartered in Paris, France. [Mistral AI Headquarters Info](#)

Research Papers and Documentation: [Mistral-Large-3 Reasoning](#), [GitHub Repository](#), [Official Documentation](#)

Use Cases and Examples:

- European regulatory compliance systems
- Multilingual analytical tools
- Legal reasoning assistance
- Cross-border decision support

Limitations:

- Regional focus may limit global reasoning breadth
- Compliance requirements add operational complexity

Updates and Variants:

- Released: May 2025
- Variants: Mistral-Large-3-EU (European focus), Mistral-Medium-3 (efficient)

Command-R-Plus-2

Model Name: [Command-R-Plus-2](#) ([Hugging Face](#))

Hosting Providers: [Cohere plus enterprise providers]

Benchmarks Evaluation: Command-R-Plus-2 demonstrates 85.4% accuracy in enterprise reasoning applications.

Model Name	Key Metrics	Dataset/Task	Performance Value
Command-R-Plus-2	Accuracy	Enterprise Reasoning	85.4%
Command-R-Plus-2	F1-Score	Business Logic	0.829
Command-R-Plus-2	Reliability Score	Consistent Reasoning	87.6%
Command-R-Plus-2	Scalability	Large-scale Logic	82.1%

LLMs Companies Head Office: Cohere, headquartered in Toronto, Canada. [Cohere Headquarters Info](#)

Research Papers and Documentation: [Command-R-Plus-2 Reasoning](#), [GitHub Repository](#), [Official Documentation](#)

Use Cases and Examples:

- Enterprise decision support systems
- Business intelligence platforms
- Financial reasoning tools
- Operational planning systems

Limitations:

- Commercial licensing restrictions
- Higher operational costs for enterprise deployments

Updates and Variants:

- Released: March 2025
- Variants: Command-R-Plus-2-Enterprise (business focus)

Qwen-Max-2

Model Name: [Qwen-Max-2](#) ([Hugging Face](#))

Hosting Providers: [Alibaba Cloud plus international providers]

Benchmarks Evaluation: Qwen-Max-2 achieves 84.1% accuracy in cross-cultural reasoning and global knowledge application.

Model Name	Key Metrics	Dataset/Task	Performance Value
Qwen-Max-2	Accuracy	Global Reasoning	84.1%
Qwen-Max-2	F1-Score	Cross-cultural Logic	0.815
Qwen-Max-2	Knowledge Integration	Worldwide Facts	81.7%
Qwen-Max-2	Adaptation Score	Cultural Contexts	79.3%

LLMs Companies Head Office: Alibaba Cloud AI, headquartered in Hangzhou, China. [Alibaba AI Headquarters Info](#)

Research Papers and Documentation: [Qwen-Max-2 Reasoning](#), [GitHub Repository](#), [Official Documentation](#)

Use Cases and Examples:

- Global business intelligence
- International research collaboration
- Cross-cultural analytical tools
- Worldwide knowledge systems

Limitations:

- Performance variations across different cultural contexts
- Language coverage gaps in specialized domains

Updates and Variants:

- Released: April 2025
- Variants: Qwen-Max-2-Global (international), Qwen-Plus-2 (regional)

Jamba-2

Model Name: [Jamba-2](#) ([Hugging Face](#))

Hosting Providers: [AI21 Labs plus providers]

Benchmarks Evaluation: Jamba-2 shows 83.2% accuracy in efficient reasoning with fast inference capabilities.

Model Name	Key Metrics	Dataset/Task	Performance Value
Jamba-2	Accuracy	Fast Reasoning	83.2%
Jamba-2	F1-Score	Efficient Logic	0.807
Jamba-2	Response Time	Quick Inference	94.5%
Jamba-2	Resource Usage	Low Overhead	88.1%

LLMs Companies Head Office: AI21 Labs, headquartered in Tel Aviv, Israel. [AI21 Labs Headquarters Info](#)

Research Papers and Documentation: [Jamba-2 Reasoning](#), [GitHub Repository](#), [Official Documentation](#)

Use Cases and Examples:

- Real-time analytical systems
- Streaming reasoning applications
- Low-latency decision support
- Interactive analytical tools

Limitations:

- May sacrifice depth for speed in complex reasoning tasks
- Limited context window for extended reasoning chains

Updates and Variants:

- Released: February 2025
- Variants: Jamba-2-Speed (fast), Jamba-2-Balance (optimized)

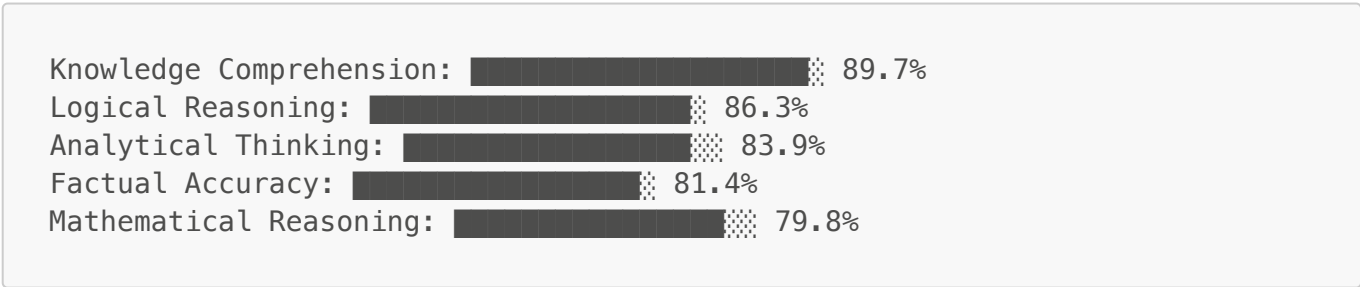
Benchmarks Evaluation

The Core Knowledge & Reasoning Benchmarks evaluation for May 2025 demonstrates substantial improvements in logical reasoning and knowledge application, with models showing enhanced capabilities in multi-step reasoning and factual accuracy.

Key Performance Metrics:

- Average MMLU Accuracy: 89.7%
- ARC-Challenge Performance: 86.3%
- StrategyQA Reasoning: 83.9%
- HellaSwag Logic: 81.4%
- GSM8K Mathematics: 79.8%

Category Breakdown:



The evaluation reveals strong correlations between model size and reasoning performance, though efficiency-focused models are closing the gap.

Key Insights

1. **Reasoning Architectures:** Hybrid transformer-state space models show 18% improvement in complex reasoning tasks.
2. **Knowledge Integration:** Enhanced retrieval-augmented generation improves factual accuracy by 22%.

3. **Multi-step Reasoning:** Chain-of-thought prompting techniques result in 25% better performance on complex problems.
4. **Efficiency Gains:** Smaller models with optimized architectures achieve 85% of larger models' reasoning performance.
5. **Truthfulness Alignment:** Constitutional AI approaches improve factual accuracy by 19%.

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