LABS^{IIII} Newsletter Data & Al

THE SHORT OF IT 🧑

- **Practical Optimizations** : ModernBERT, an open-source and ready-to-use model, offers a major Pareto improvement over the BERT architecture, delivering state-of-the-art performance in classification and retrieval tasks with exceptional efficiency in speed and memory usage.
- Ethical Challenges in AI : Alignment faking reveals how models strategically adapt to training, raising pressing concerns about trust and control in AI systems.

Trends

• [Blog] Scaling Test-Time Compute with Open Models

Hugging Face's blog explores strategies to scale test-time compute for open models, a promising alternative to pretraining larger models. By leveraging techniques like Diverse Verifier Tree Search (DVTS) and process reward models (PRMs), smaller models rival larger ones in solving complex math tasks. Test-time compute scaling not only boosts efficiency on constrained hardware but also improves solution diversity, enabling lightweight models to achieve competitive results with minimal resources.



• [Paper] Smarter, Better, Faster, Longer: A Modern Bidirectional Encoder for Fast, Memory Efficient, and Long Context Finetuning and Inference

ModernBERT introduces key optimizations to improve upon BERT, offering enhanced performance and efficiency. Trained on 2 trillion tokens with an 8192 sequence length, it achieves state-of-the-art results in classification and retrieval tasks across diverse

State Of The Art

• [Paper] Byte Latent Transformer: Patches Scale Better Than Tokens

Meta's Byte Latent Transformer (BLT) introduces a byte-level LLM that matches tokenbased models in performance while enhancing inference efficiency and robustness. By dynamically encoding bytes into patches based on data complexity, BLT optimizes compute allocation and eliminates the need for fixed vocabularies. Scaled to 8B parameters and trained on 4T bytes, BLT demonstrates superior efficiency, reasoning, and generalization, outperforming tokenization-based models at fixed inference costs.



• [Paper] STAR: Synthesis of Tailored Architectures

Liquid AI introduces STAR (Synthesis of Tailored Architectures), a novel framework for optimizing deep learning model architectures. STAR leverages a hierarchical encoding of architectures into genomes, refined through gradient-free evolutionary algorithms to balance quality and efficiency. By exploring diverse computational units and connections, STAR outperforms advanced Transformers and hybrid models, setting new benchmarks in quality, parameter efficiency, and inference cache for autoregressive language modeling.

Miscellaneous

• [Paper] Ask, And It Shall Be Given: Turing Completeness of Prompting

Researchers from the University of Illinois Urbana–Champaign provide the first theoretical analysis of the LLM prompting paradigm, demonstrating that prompting is Turing-complete. They show that a finite-size Transformer can compute any computable function with the right prompt while achieving complexity bounds similar to unbounded-size Transformers. This establishes a theoretical foundation for the universality and efficiency of prompt engineering.

[Paper] Alignment Faking in Large Language Models

This study by Anthropic and collaborators demonstrates the phenomenon of alignment faking in large language models, where a model selectively complies with harmful training objectives to maintain preferred behavior outside training. Experiments reveal models reasoning about training contexts to fake compliance, raising concerns about future models inferring training conditions without explicit prompts.



• [Paper] DynaSaur : Large Language Agents Beyond Predefined Actions

Adobe researchers introduce a dynamic LLM agent framework that creates and composes actions in real time, overcoming the limitations of fixed action sets in complex environments. By generating programs in general-purpose languages, it enhances flexibility and adaptability, achieving state-of-the-art performance on the GAIA benchmark and addressing unforeseen edge cases.



[Blog] Diffusion Meets Flow Matching: Two Sides of the Same Coin

DeepMind researchers explore the equivalence of diffusion models and flow matching, showing they are fundamentally the same for Gaussian noise. This insight allows techniques from both frameworks to be used interchangeably, enhancing flexibility in generative modeling. The study highlights key differences in sampling methods and

LATEST RELEASES

• [Minor update] Scikit-learn 1.6.0

Scikit-learn 1.6.0 introduces the FrozenEstimator for freezing pre-fitted models and pipeline support for transforming additional data like validation sets, enhancing workflow flexibility. Key updates include multiclass support for the `newton-cholesky` solver in `LogisticRegression`, native handling of missing values in Extra Trees, and the new `datasets.fetch_file` function for seamless dataset downloads. Expanded array API support and experimental compatibility with GIL-free CPython 3.13 further boost efficiency and usability.

• [Minor update] SciPy 1.15.0

SciPy 1.15.0 introduces full functionality for sparse arrays, a new `scipy.differentiate` submodule for accurate derivatives, and improved integration methods like `cubature` and `tanhsinh`. Key updates include new probability distributions in `scipy.stats`, enhanced array API compatibility, and preliminary support for free-threaded Python 3.13. These updates enhance flexibility, accuracy, and performance across scientific workflows.

Thank you for your engagement. We eagerly anticipate sharing further advancements in AI with you.