

Connect

Why you want your Al to be Open Source

Business Track FSI - Financial Services and Insurances





Armin Warda

EMEA FSI Chief Technologist Red Hat





Why is NOW a good time for companies to invest in Al?

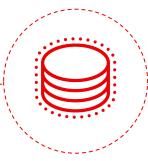
Enterprises are taking the Al leap



Al technologies are becoming more accessible and affordable for businesses of all sizes



Companies can realize
the value from
Al-enabled applications
and Al-support



Organizations are better prepared to manage, transform and use their ever-increasing data



Al has undergone significant evolution

The evolution of **Data-driven** Al: from Business Intelligence to Generative Al

Business Analysis & Foundation Models & **Generative AI & Advanced Analytics &** ► Today, practical Al is **Data-driven** Intelligence **Fact Databases Predictive Al** Al-enabled apps Predictive AI runs businesses today • Retrieval Augmented • Data science techniques Deep learning Collecting data Foundation models provide a Generation (RAG) techniques Unstructured data Storing & moving data shortcut for realizing Generated content Model experimentation • Predictive analytics Structured data pointing to fact the value of **Generative AI** Model tuning Real-time decision database Transforming data (ETL) making Prompt engineering Also filter hallucinations Models are embedded in and **guarded** by applications **Data Warehouses Big Data Generative Al** RAG, Guardrails



Predictive Al vs. Generative Al

Most common types of AI for business applications

Predictive Al

Predicts or classifies outcomes with models trained on use-case specific data sources



Generative Al

Generates new content with models trained on vast amounts of data from many sources

- I am preparing for a behavioral interview for a marketing role. Can you generate some sample questions about my ability to create and execute successful marketing campaigns? I have experience conducting market research, developing marketing strategies, and analyzing the results of campaigns. Can you also include questions about my creativity and ability to come up with new ideas?"
 - Sure! Here are some sample behavioral interview questions about your ability to create and execute successful marketing campaigns:
 - Can you describe a time when you had to develop and implement from start to finish? What was the campaign for, and what were th
 - How do you go about conducting market research when planning campaign? What tools and techniques do you use to gather inform
 - How do you determine which marketing channels will be most efform the control of the contro
 - 4. How do you measure the success of a marketing campaign? What and how do you analyze the data to inform future campaigns?
 - 5. Can you describe a time when you had to be creative in order to s challenge during a marketing campaign? How did your solution in success?



Predictive Al vs. Generative Al

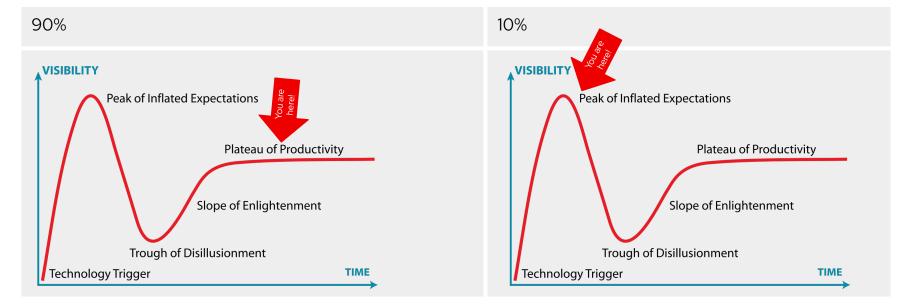
Most common types of AI for business applications

Predictive Al

Generative Al

Penetration

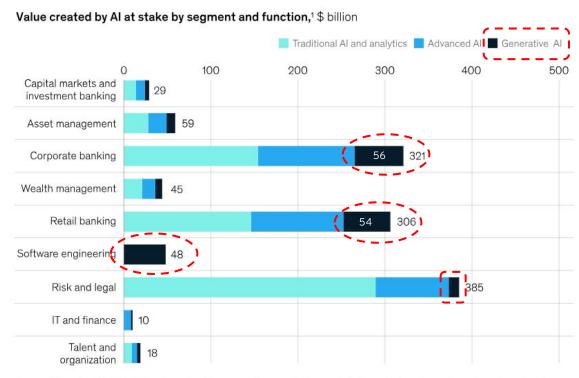
Maturity





Predictive Al and Generative Al in banking

McKinsey Insights: Capturing the full value of generative AI in banking



'Assumes 0% overlap of traditional AI and generative AI (generative AI assumes the lower end of value at stake), top-down estimation based on projected growth and value pools.

Source: The economic potential of generative Al: The next productivity frontier, McKinsey Global Institute, June 2023; QuantumBlack, Al by McKinsey traditional advanced analytics and Al analysis

Among industry sectors, banking is expected to have one of the largest opportunities, largely from increased productivity

- The economic impact will likely benefit all banking segments and functions, with the greatest absolute gains through Generative Al in the corporate and retail sectors with \$56 billion and \$54 billion, respectively
- No surprise: software engineering LLMs
- Risk and legal get largest value from AI, but that's mostly traditional AI

Proven Al Use-Cases in Financial Services



Fraud Management

Anomaly detection,
Countering financial crime
such as money laundering,
terror financing, tax evasion



Hyper-Personalization

Improve customer and employee experience,
Customer Next Best Offer,
Chatbots, Onboarding



Operational Efficiency

Planning, ATM Cash on Hand, Call Routing, Workflow Automation



Risk Analytics

Automated Underwriting

Decisions, Mortgage

Prepayment Analytics,

Credit Scoring



İŞBANK



FriendliAl





JPMorganChase 🚺



These are great Al use-cases.

But there are also challenges.





Rage against the machine?

This article is more than 3 months old

Cruise recalls all self-driving cars after grisly accident and California ban

All 950 of the General Motors subsidiary's autonomous cars will be taken off roads for a software update



Cruise self-driving cars outside the company's headquarters in San Francisco. Photograph: Heather Somerville/Reuters

Associated Press

Wed 8 Nov 2023 19.17 CET

San Francisco



Crowd burns Waymo

A parking lot full

Crowd burns Waymo self-driving vehicle in San

a nuisance for a Television of 2.64M subscribers



Regulating AI: The EU-AI Act (March 13)



Unregulated, irresponsible or abusive use of AI could lead to negative consequences for individuals or the society, create public opposition and **hinder AI innovation in the EU**.

The EU is committed to strive for a balanced approach to Al

- Lawful
- → accurateness
- Ethical
- → fairness
- Robust
- → no (unintended) bias

transparency

ust → security

EU AI Act Requirements:

Explainability, Documentation, Process & Data Governance, Human Oversight, Risk Management, Auditability.

There are some exceptions for AI systems released under **Open Source** licenses.

High Risk

Most regulated AI systems, as these have the potential to cause significant harm if they fail or are misused, e.g. if used in law enforcement or recruiting.

Minimal Risk

All other AI systems, e.g. a spam filter, which can be deployed without additional restrictions.

banned:

Unacceptable Risk

Highest level of risk prohibited in the EU. Includes AI systems using e.g. subliminal manipulation or general social scoring.

Limited Risk

Includes AI systems with a risk of manipulation or deceit, e.g. chatbots or emotion recognition systems. Humans must be informed about their interaction with the AI.

Open Source **Software**



Open Source Hardware



Open Source AI/ML Models ?



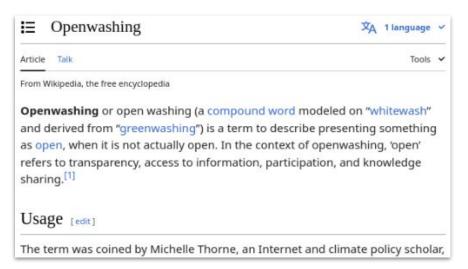
How open are today's "Open Source" Models?

Open Source Software	Today's "Open Source" LLMs
Frequent releases (sometimes nightly)	Irregular releases (e.g. 1y between LLaMA versions)
Incremental contributions	Monolithic development
Feature roadmaps	"Emergent behaviour", no one knows what's coming
Community contributions (pull requests)	Largely single-party development (expensive collection of training data)
Contributions from many contributors can be merged and reconciled	Contributions to model, in the form of fine-tuning, are mutually incompatible between contributors, leading to fragmentation in model families (forks)
Almost any developer can, in principle, contribute	High barrier to contribution (clusters, GPUs for fine-tuning)



Openwashing?





Rethinking open source generative Al: open-washing and the EU AI Act

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ABSTRACT

The past year has seen a steep rise in generative AI systems that claim to be open. But how open are they really? The question of what counts as open source in generative AI is poised to take on particular importance in light of the upcoming EU AI Act that regulates open source systems differently, creating an urgent need for practical openness assessment. Here we use an evidence-based framework that distinguishes 14 dimensions of openness, from training datasets to scientific and technical documentation and from licensing to access methods. Surveying over 45 generative AI systems (both text and text-to-image), we find that while the term open source is widely used, many models are 'open weight' at best and many providers seek to evade scientific, legal and regulatory scrutiny by withholding information on training and fine-tuning

Janeiro, Brazil. ACM, New York, NY, USA, 14 pages. https://doi.org/10.1145/ 3630106.3659005

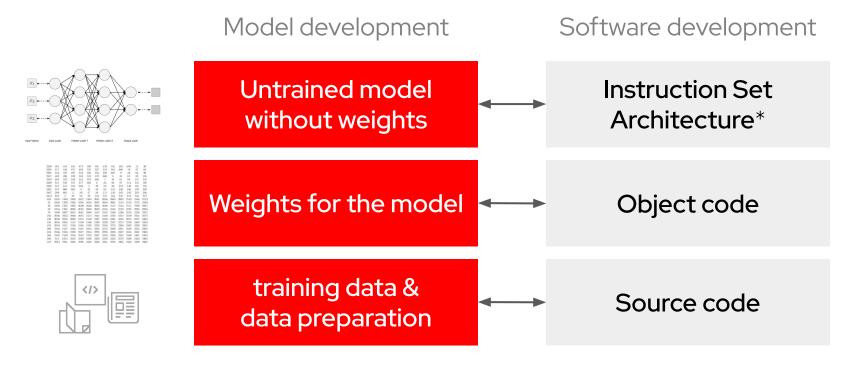
1 INTRODUCTION

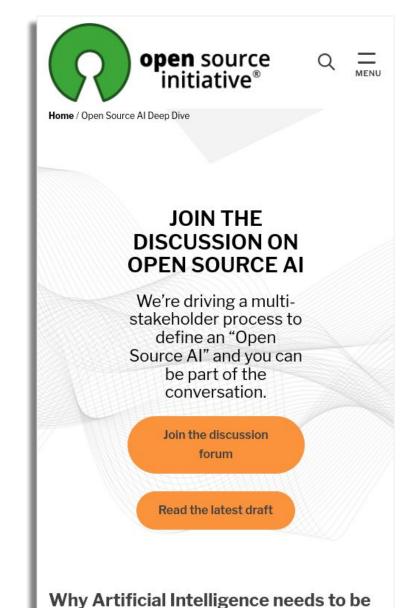
Open generative AI systems are on the rise, with small players and academic initiatives leading the way in open innovation and scientific documentation [20, 32, 61] and several larger corporations joining the fray by releasing models billed as 'open'. But there are three critical challenges to openness in the domain of generative AI systems. The first is that openness is not a binary feature: to-day's transformer-based system architectures and their training procedures are complex, and they can only be classified into open or closed at the price of severe information loss. Secondly, some



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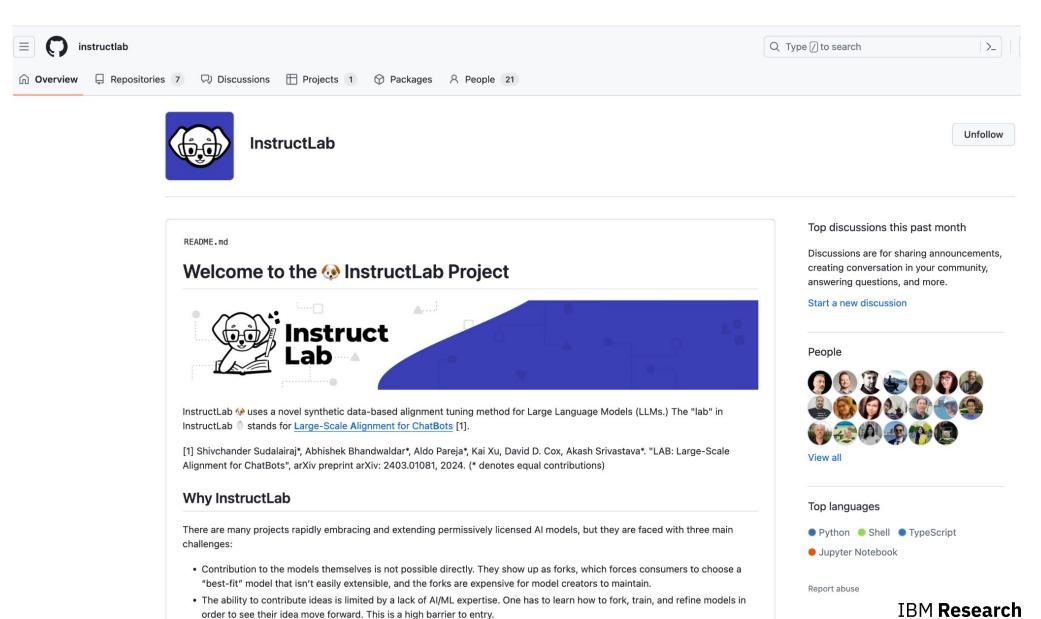
"Weights are code."



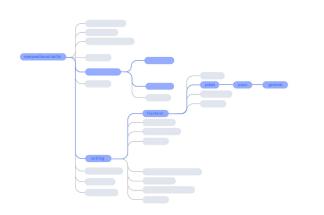


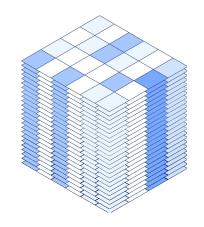
Open Source?

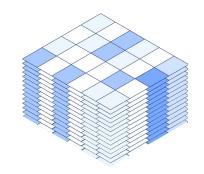
An open source community project for GenAl model development



LAB (Large-scale Alignment for ChatBots) Method









Taxonomy-based skill & knowledge representation

Represent any missing model knowledge or skills in a hierarchical **taxonomy**, providing 5+ exemplifying data points of the missing behavior per missing skill.

Synthetic data generation with teacher model

A **teacher model** generates a "curriculum" of millions of questions and answers across the taxonomy.

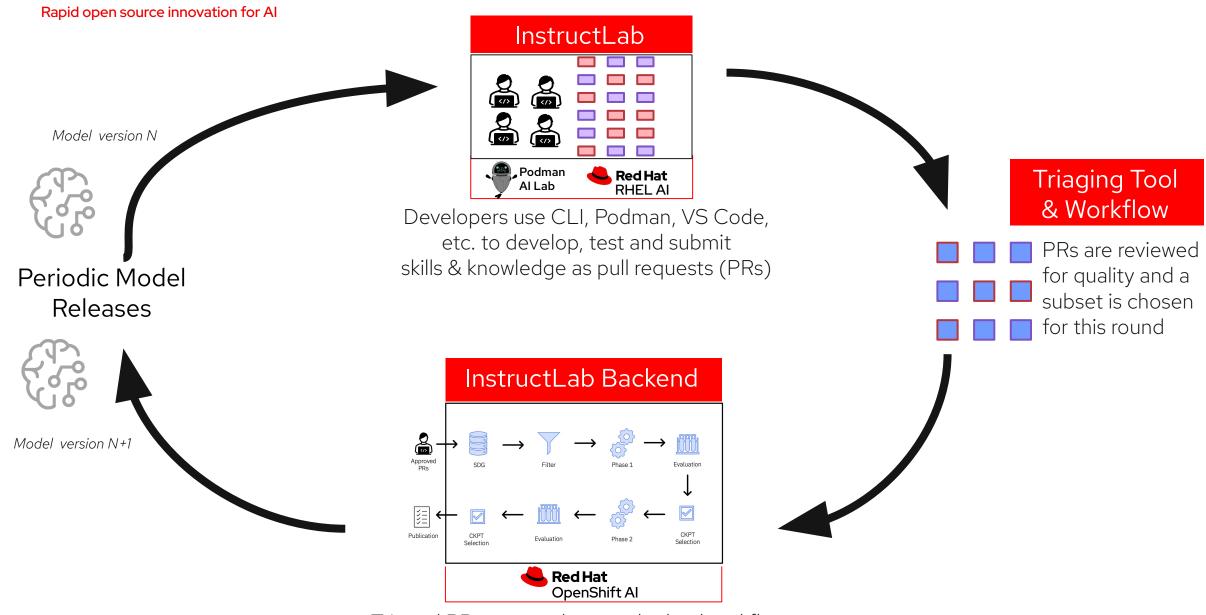
Synthetic data validation with critic model

A **critic model** filters the questions for correctness and quality. Synthetic data is scanned for prohibited material.

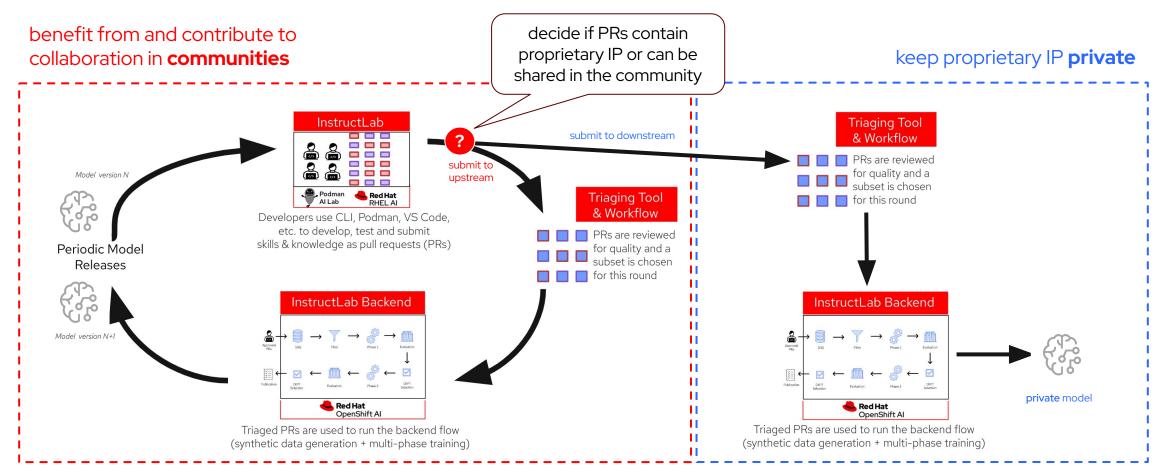
Skill and knowledge training on top of student model

The **student model** is trained with the curriculum using a novel training approach.

IBM Research publication: https://arxiv.org/html/2403.01081v1
IBM Think keynote: https://www.youtube.com/watch?v=SuGedexBudQ







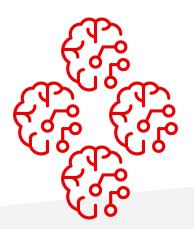
Skills and knowledge that can be shared with the community are contributed upstream. These come back for free with the next version of the model, thus reducing the resources required for in-house fine-tuning of the private model, and potentially improved by other collaborators.

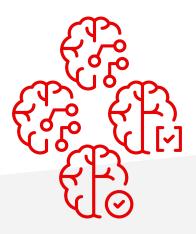
Proprietary skills and knowledge, that shall not be shared, are not submitted upstream but retained in-house. These have to be re-added to each new version of the upstream base model.













STEP 1

Learn & experiment via limited desktop-scale training method (qlora) on small datasets. Future potential Podman Desktop integration.







STEP 2

Production-grade model training using full synthetic data generation, teacher and critic models. Tooling focused on scriptable primitives.





STEP 3

Production-grade model training as in RHEL AI, using full power of Kubernetes scaling, automation and MLOps services.



watsonx

STEP 4

Comprehensive Al solution including Al optimized infrastructure, runtimes, middleware, data services, governance and applications.





Why you want your AI to be Open Source







Innovate with Open Source

- Proven Predictive-Al use-cases with Open Source can provide faster time-to-business value,
- Open Source & Open Research is where Innovation in Generative-AI happens,
- Open Source avoids Lock-Ins to hyperscalers or HW vendors.

But there are challenges

- Open Source provides better
 Transparency and Auditability,
- the EU Al-Act regulation is a bit lighter on Open Source,
- Collaboration on AI model development can solve common challenges faster, while allowing to keep unique IP private.

Red Hat can help

- We have the capabilities and partnerships to help speed-up your Al initiatives,
- allowing faster delivery of intelligent software applications,
- providing faster time-to-business value
- and to control the risks.





Connect

Thank you



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