



Connect

Why you want your AI 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 AI?

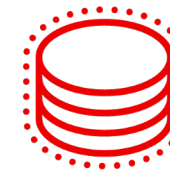
Enterprises are taking the AI leap



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



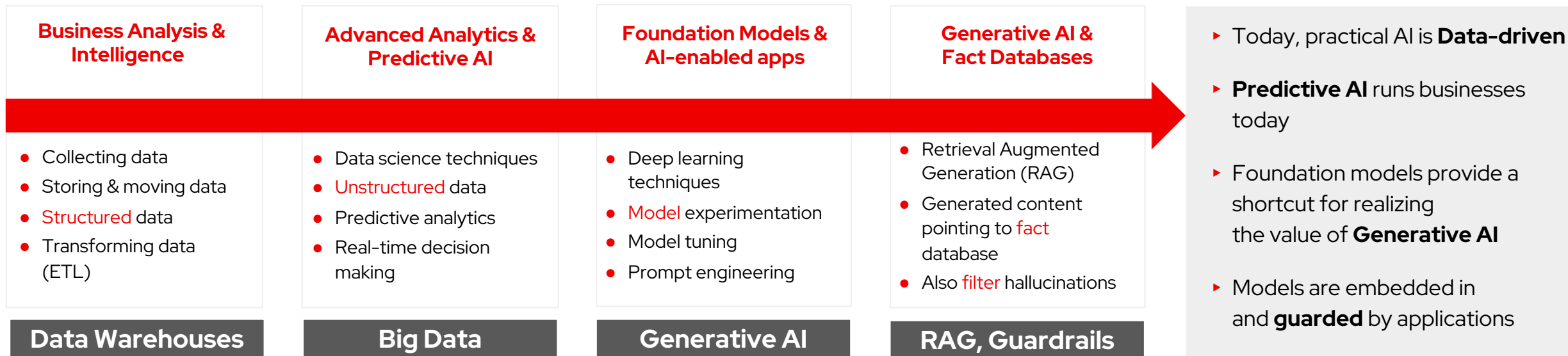
Companies can **realize the value** from AI-enabled applications and AI-support



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

AI has undergone significant evolution

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



Predictive AI vs. Generative AI

Most common types of AI for business applications

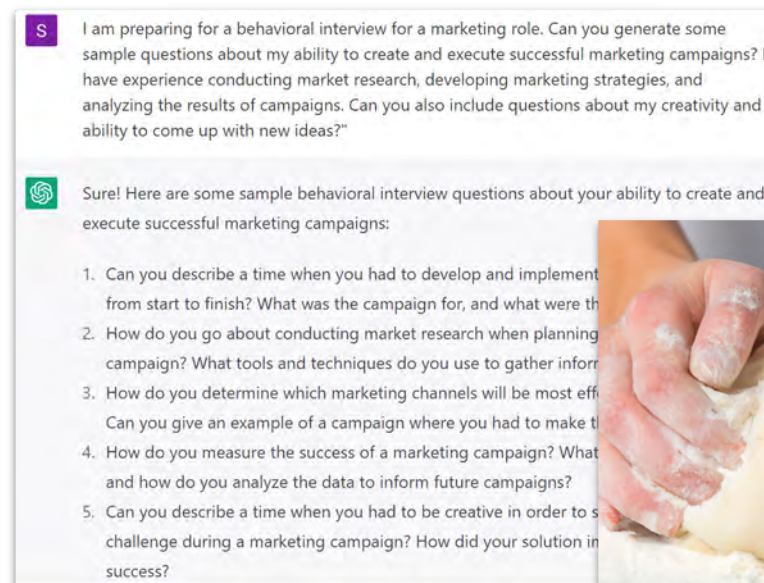
Predictive AI

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



Generative AI

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



<https://www.letsdive.io/blog/generative-ai-vs-predictive-ai-all-you-need-to-know#:~:text=Generative%20AI%20is%20primarily%20focused,events%20based%20on%20historical%20data.>

Predictive AI vs. Generative AI

Most common types of AI for business applications

Predictive AI

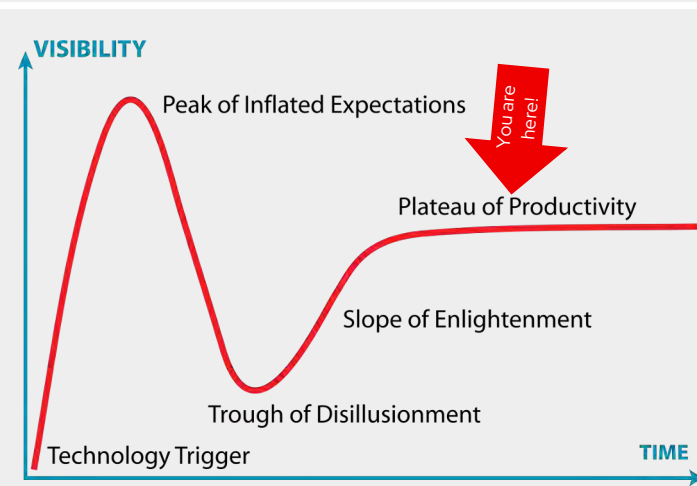
What is it for?

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

Penetration

90%

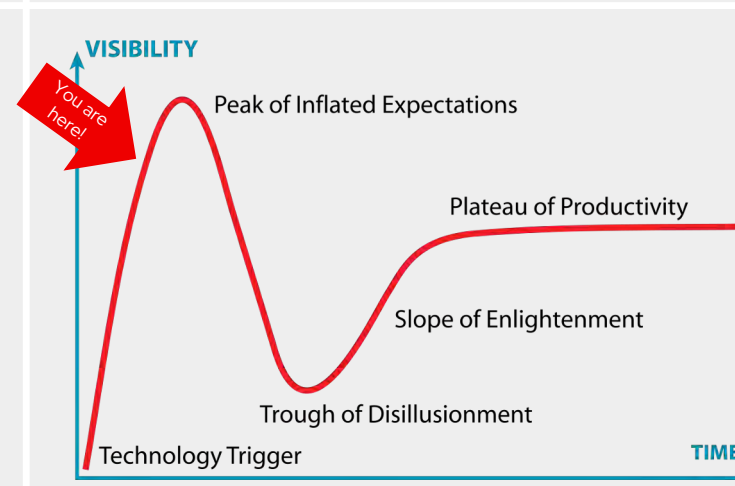
Maturity



Generative AI

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

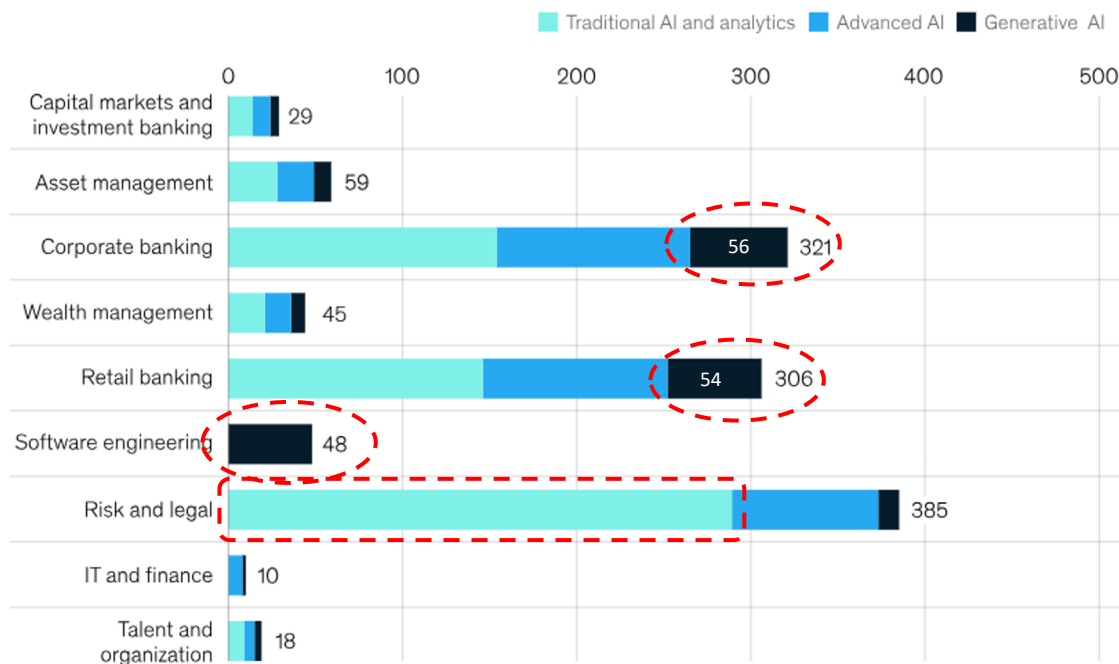
10%



Predictive AI and Generative AI in banking

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

Value created by AI at stake by segment and function,¹ \$ billion



¹Assumes 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 AI: The next productivity frontier, McKinsey Global Institute, June 2023; QuantumBlack, AI by McKinsey traditional advanced analytics and AI 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 AI 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 AI Use-Cases in Financial Services



Fraud Management

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



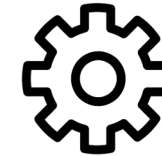
Hyper-Personalization

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



Operational Efficiency

Branch Location & Staff
Planning, ATM Cash on
Hand, Call Routing,
Workflow Automation



Risk Analytics

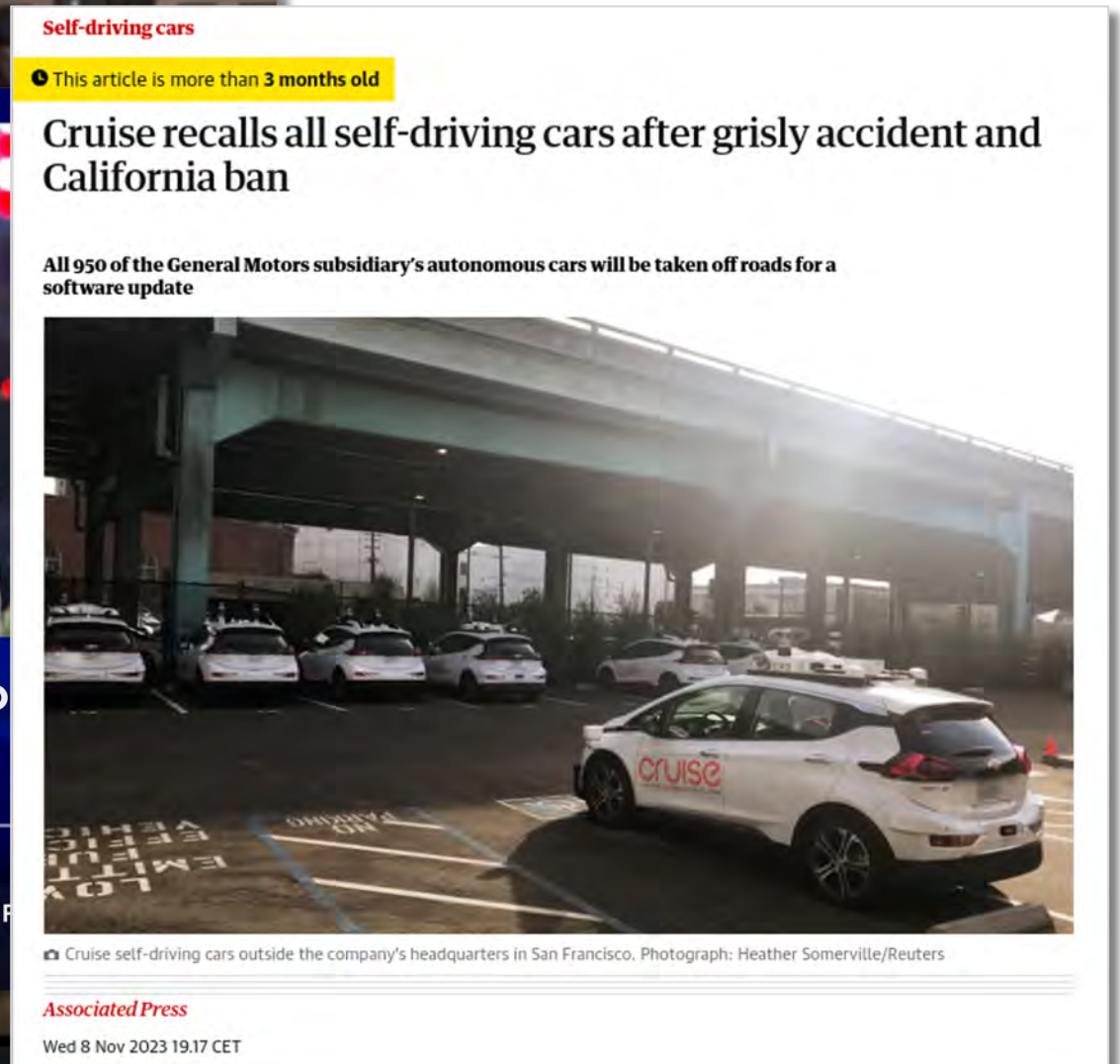
Automated Underwriting
Decisions, Mortgage
Prepayment Analytics,
Credit Scoring



These are great AI use-cases.

But there are also challenges.

Rage against the machine?



Driverless cars wake residents with nighttime honking

https://www.youtube.com/watch?v=9kn9yu_Cqfc

<https://www.bbc.com/news/videos/c17qqverq99o>

<https://www.theguardian.com/technology/2023/nov/08/cruise-recall-self-driving-cars-gm>

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 AI

- Lawful
 - Ethical
 - Robust
- accurateness
 - transparency
 - fairness
 - no (unintended) bias
 - 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.

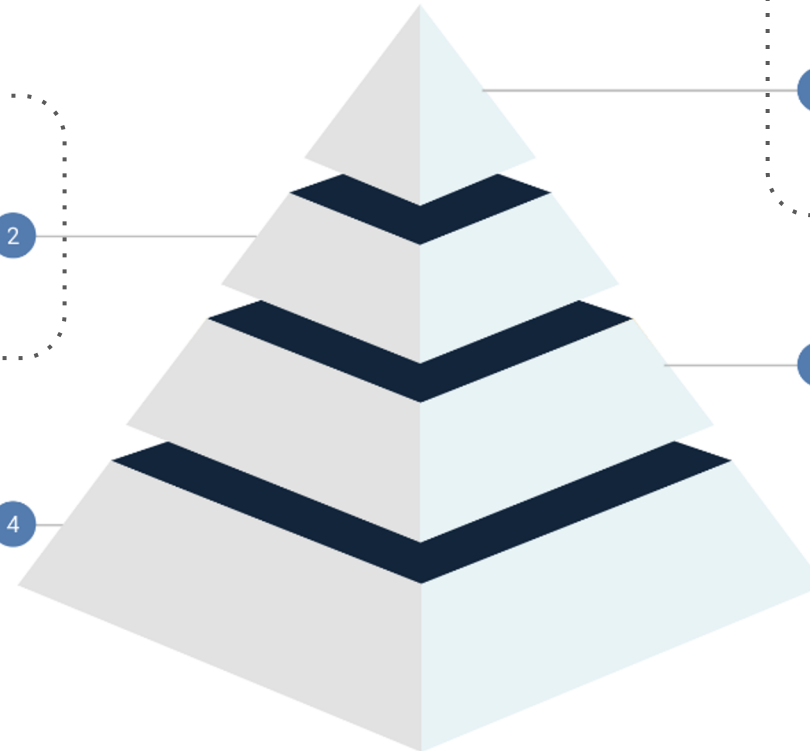
banned:

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?



Openwashing

1 language

Article Talk Tools

From Wikipedia, the free encyclopedia

Openwashing or open washing (a **compound word** modeled on “**whitewash**” and derived from “**greenwashing**”) is a term to describe presenting something as **open**, when it is not actually open. In the context of openwashing, ‘open’ refers to transparency, access to information, participation, and knowledge sharing.^[1]

Usage [edit]

The term was coined by Michelle Thorne, an Internet and climate policy scholar,

Rethinking open source generative AI: 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

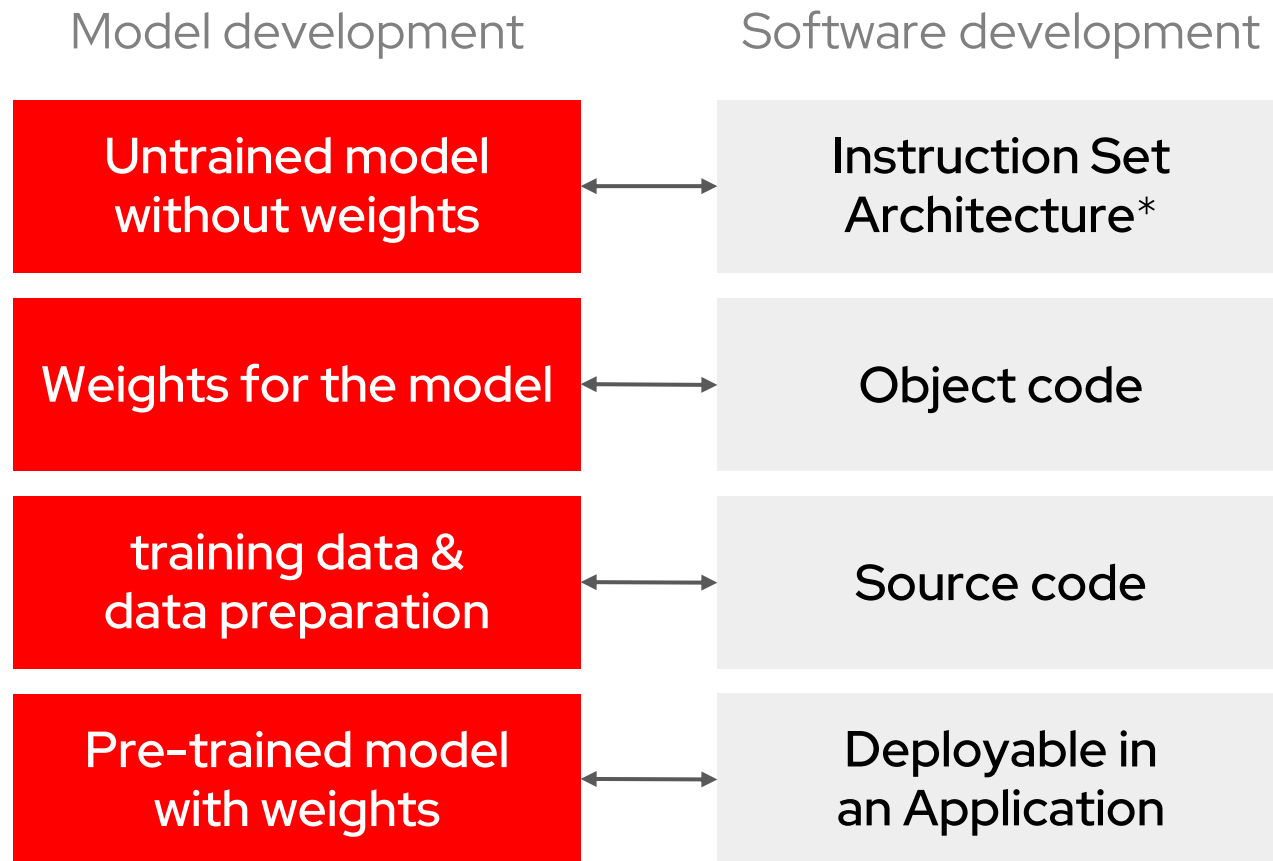
Janeiro, Brazil. ACM, New York, NY, USA, 14 pages. <https://doi.org/10.1145/3630106.3659005>

1 INTRODUCTION

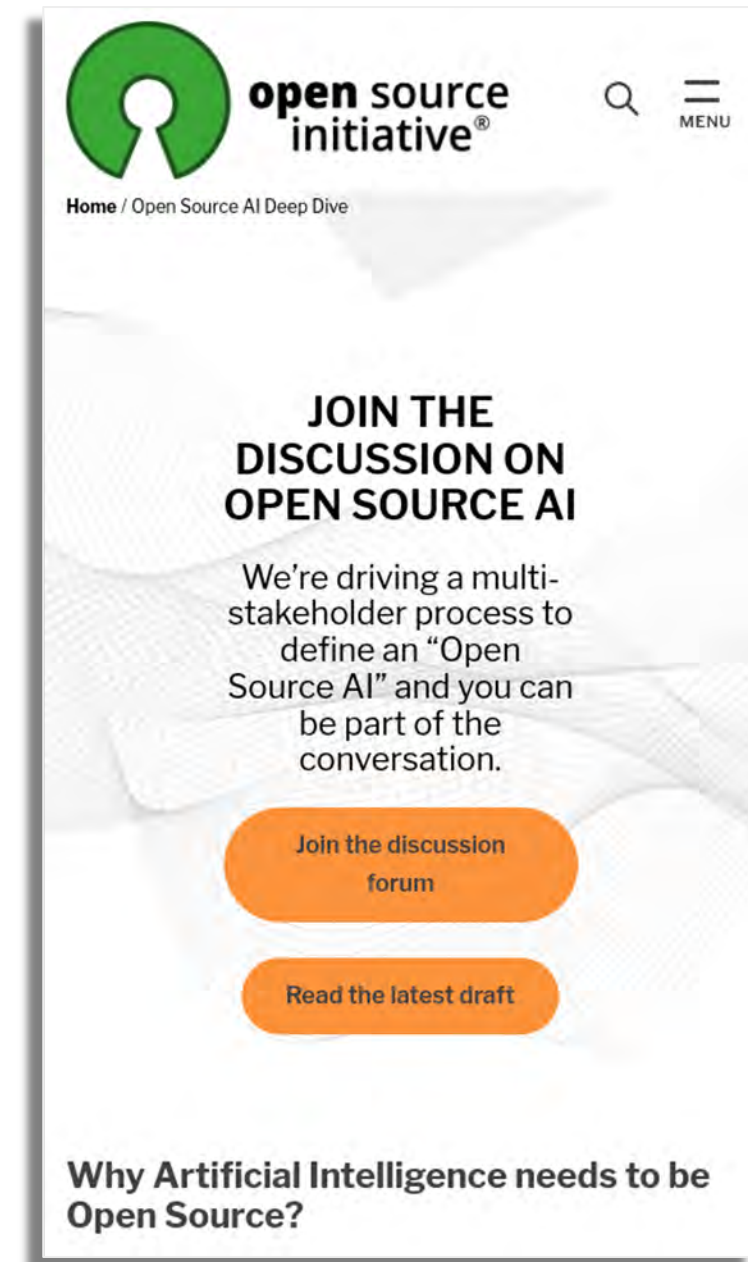
Open generative AI systems are on the rise, with small players and academic initiatives leading the way in open innovation and scientific documentation [20, 32, 61] and several larger corporations joining the fray by releasing models billed as ‘open’. But there are three critical challenges to openness in the domain of generative AI systems. The first is that openness is not a binary feature: today’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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59	511	538	565	592	619	21	48	7	10	13	2060				790	25	237	264																				


„Weights are code.“



* Instruction Set Architecture (ISA), such as RISC-V, P-Code Machine, Java Virtual Machine, WebAssembly



An open source **community** project for GenAI model development

instructlab

Overview


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InstructLab

Unfollow

README.md

Welcome to the 🐶 InstructLab Project



Instruct Lab

InstructLab 🐶 uses a novel synthetic data-based alignment tuning method for Large Language Models (LLMs.) The "lab" in InstructLab 🐶 stands for [Large-Scale Alignment for ChatBots](#) [1].

[1] Shivchander Sudalairaj*, Abhishek Bhandwaldar*, Aldo Pareja*, Kai Xu, David D. Cox, Akash Srivastava*. "LAB: Large-Scale Alignment for ChatBots", arXiv preprint arXiv: 2403.01081, 2024. (* denotes equal contributions)

Why InstructLab

There are many projects rapidly embracing and extending permissively licensed AI models, but they are faced with three main challenges:


- Contribution to the models themselves is not possible directly. They show up as forks, which forces consumers to choose a "best-fit" model that isn't easily extensible, and the forks are expensive for model creators to maintain.
- The ability to contribute ideas is limited by a lack of AI/ML expertise. One has to learn how to fork, train, and refine models in order to see their idea move forward. This is a high barrier to entry.
- There is no direct community governance or best practice around review, curation, and distribution of forked models.

Top discussions this past month

Discussions are for sharing announcements, creating conversation in your community, answering questions, and more.

[Start a new discussion](#)

People



[View all](#)

Top languages

Python

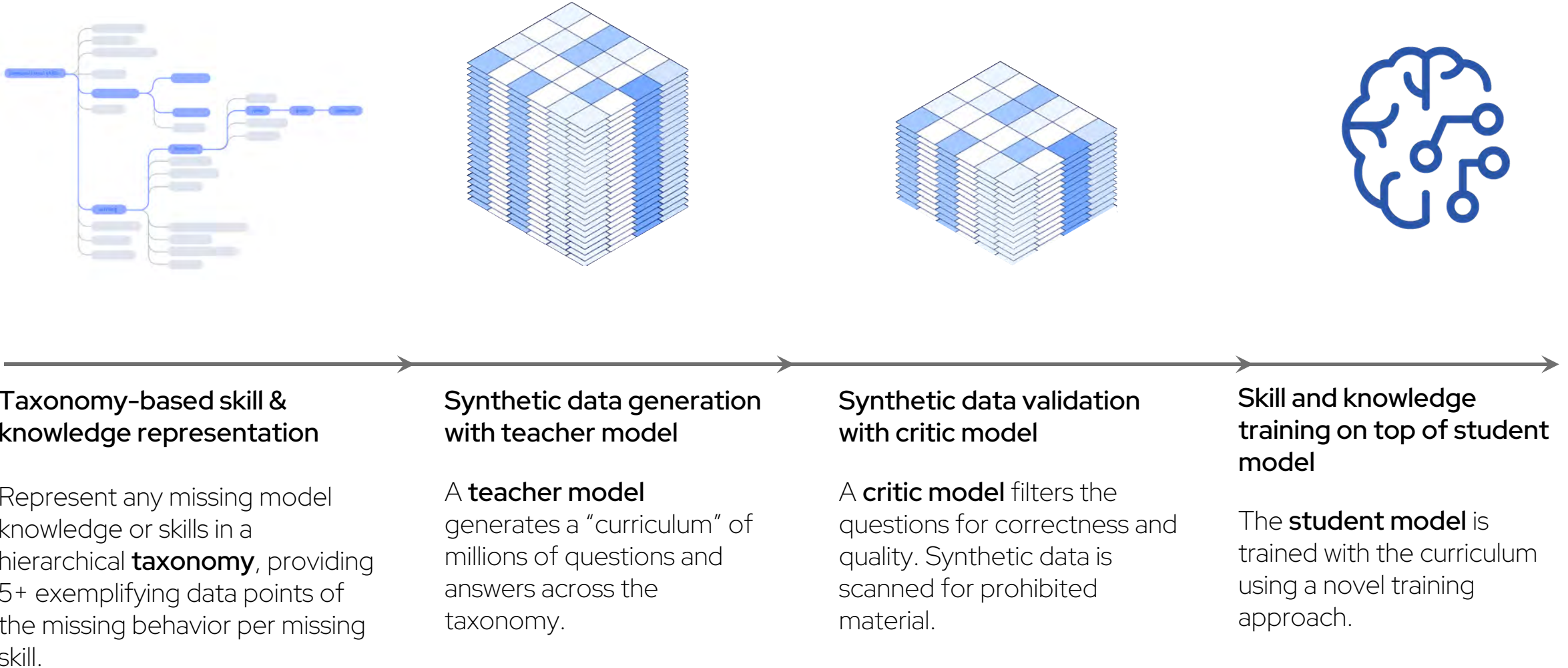
Shell

TypeScript

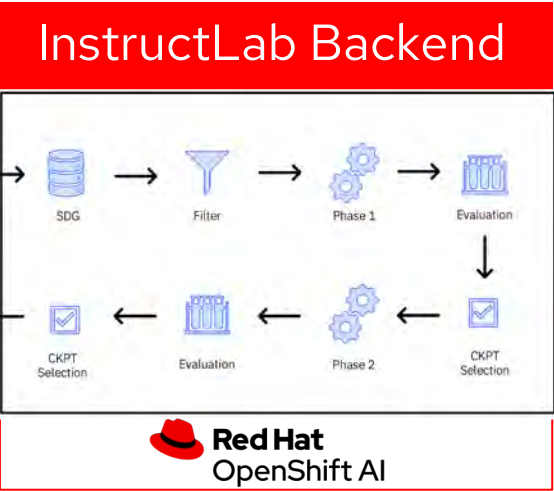
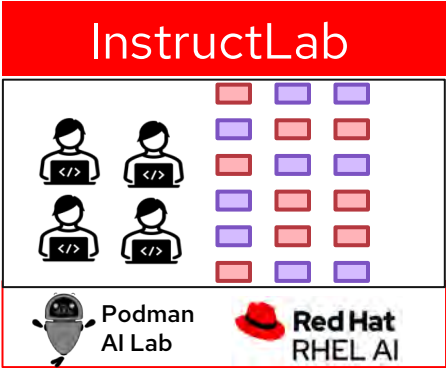
Jupyter Notebook

[Report abuse](#)

LAB (Large-scale Alignment for ChatBots) Method



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



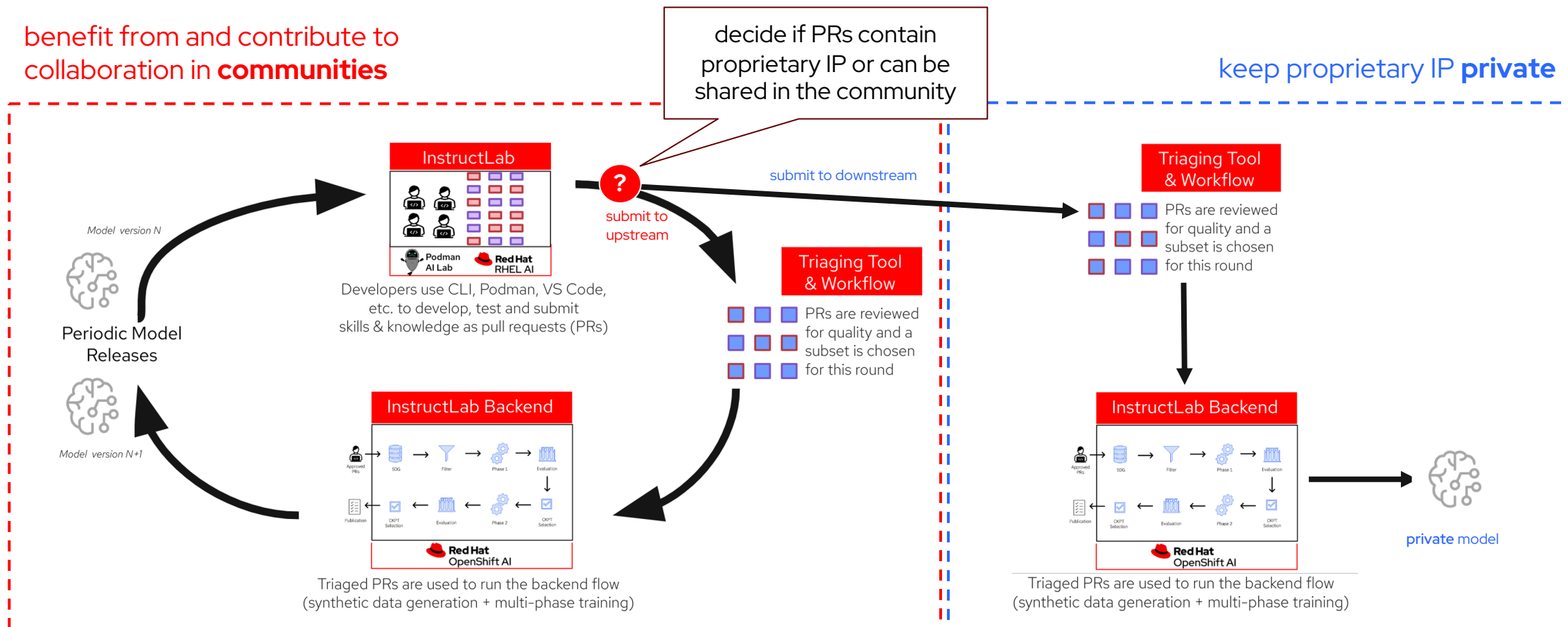
Model version N

Periodic Model Releases

Model version N+1

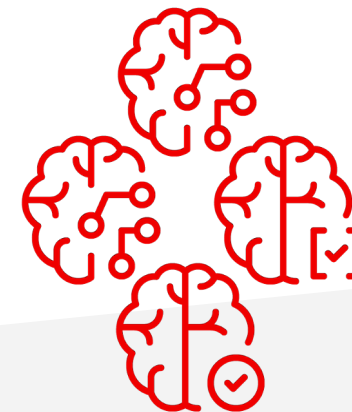
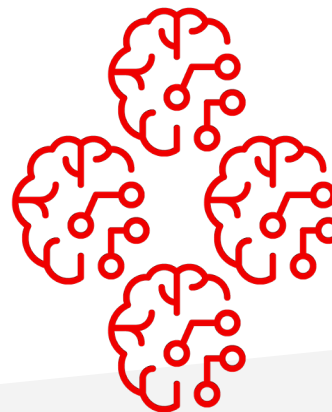
Triaged PRs are used to run the backend flow (synthetic data generation + multi-phase training)

benefit from and contribute to collaboration in **communities**



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.



InstructLab

STEP 1

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

Laptop / desktop



Red Hat Enterprise Linux AI

STEP 2

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

Server / VM



Red Hat OpenShift AI

STEP 3

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

Cluster

watsonx

STEP 4

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

Cluster

Why you want your AI to be Open Source



Innovate with Open Source

- Proven **Predictive-AI** 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 AI-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 AI 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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youtube.com/user/RedHatVideos



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awarda@redhat.com