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LEVERAGING AGENTIC DATA FLOWS WITH LARGE LANGUAGE MODELS

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Outlook

Introduction

• Day-to-**Day**ta tasks

Motivation

- Diving into Large language Models (LLMS)
- A shared Vision

Showcase

Practical Application

Conclusion

- Challenges
- Future Directions
- Questions





The (linked) data life cycle

ANALYSIS

Insights from linked data

Discover and analyze complex - and often hidden - relations in your data by means of customized data portals and state-of- the-art data visualization based on Semantic Data Lakes.

PROVISIONING

Data-driven personalization

Use semantic technologies to make your content and data management workflows smarter and more dynamic. Create personalized digital experiences. Feed your bots and machine learning algorithms with higher-quality data.

ENRICHMENT

Actionable data

Provide automatic tagging functions based on controlled vocabularies to ensure precise content classification. Semantic models help to enrich all content and data consistently and based on standards.

INGESTION

Avoid data migration

PoolParty is a semantic middleware that enriches and contextualizes your data with relevant information. Link and integrate different datasets by applying semantic knowledge models. Enterprise data gets smarter while remaining in the original storage system.

CLEANING

Data quality management

PoolParty allows organizations to ensure that their data is consistent, reliable and usable for analysis. Benefit from highly automated 'Data Repair' features.

AUTHORING

Knowledge graph capabilities

Enable collaboration and develop synergies between your departments by establishing defined knowledge graphs and controlled vocabularies

LINKING

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Data

Ingestion

Data

Cleaning

Data Analysis

LINKED DATA

Lifecycle

Data Linking

Data

Provisioning

Data

Enrichment

The power of connected data

Data Authoring

Link internal and external data resources and provide your organization with a customized and relevant knowledge platform.



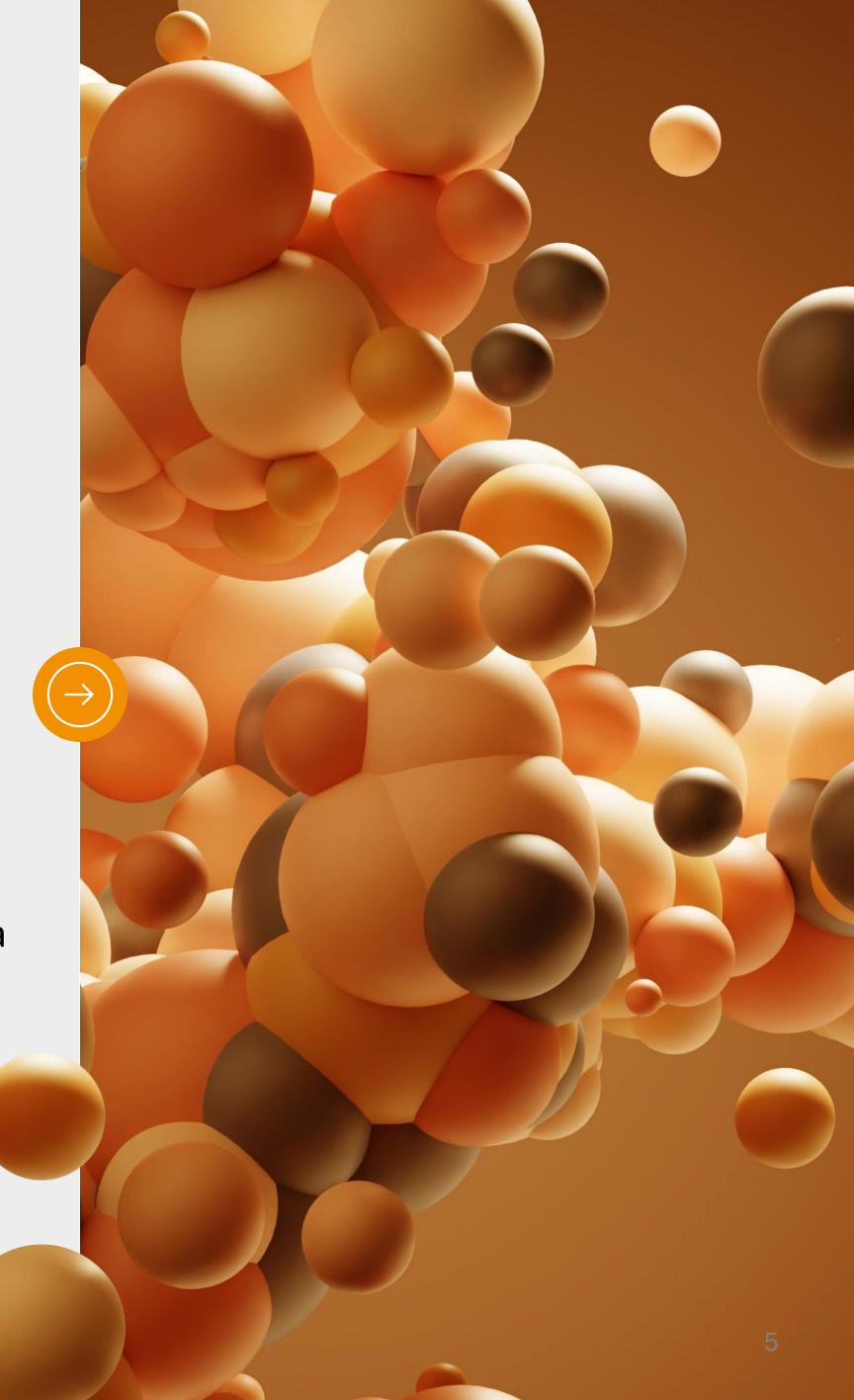
Day-to-Dayta tasks



"Data practitioners spend 80% of their valuable time finding, cleaning, and organizing the data"

Challenges in Data Tasks

- Handling Missing Data
- Removing Duplicate Entries
- Correcting Data Types
- Standardizing Data Formats
- Dealing with Outliers
- Data Cleaning Tools and Techniques
- Addressing Inconsistent Data



Clonflicting Visions

"In fact, we're not even close to matching the understanding of the physical world of any animal, cat or dog."

Lecun

Sam Altman: OpenAl's New Model Passes AGI Threshold



Meta Chief Al Scientist Slams Quest for Human-Level Intelligence



TECH

Elon Musk Says Tesla Vehicles Will Drive Themselves in Two Years



December 21, 2015 at 2:00 PM EST





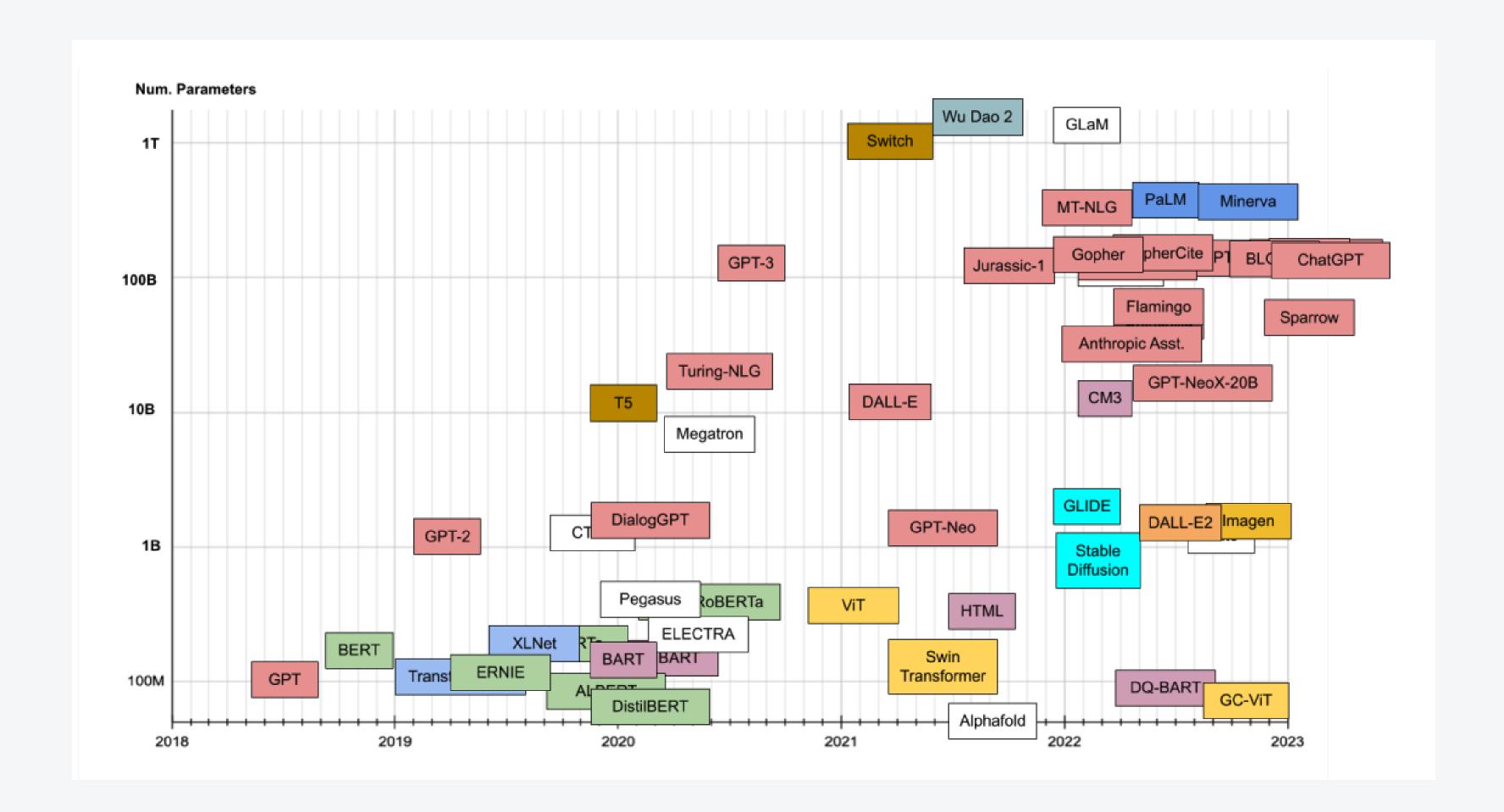
PHOTOGRAPH BY PATRICK FALLON — REUTERS



LLM Evolution

Open-Source LLMs

- MosaicML
- Falcon
- Open-Assistant
- Llama
- Qwen
- Mistral
- Phi
- DeepSeek





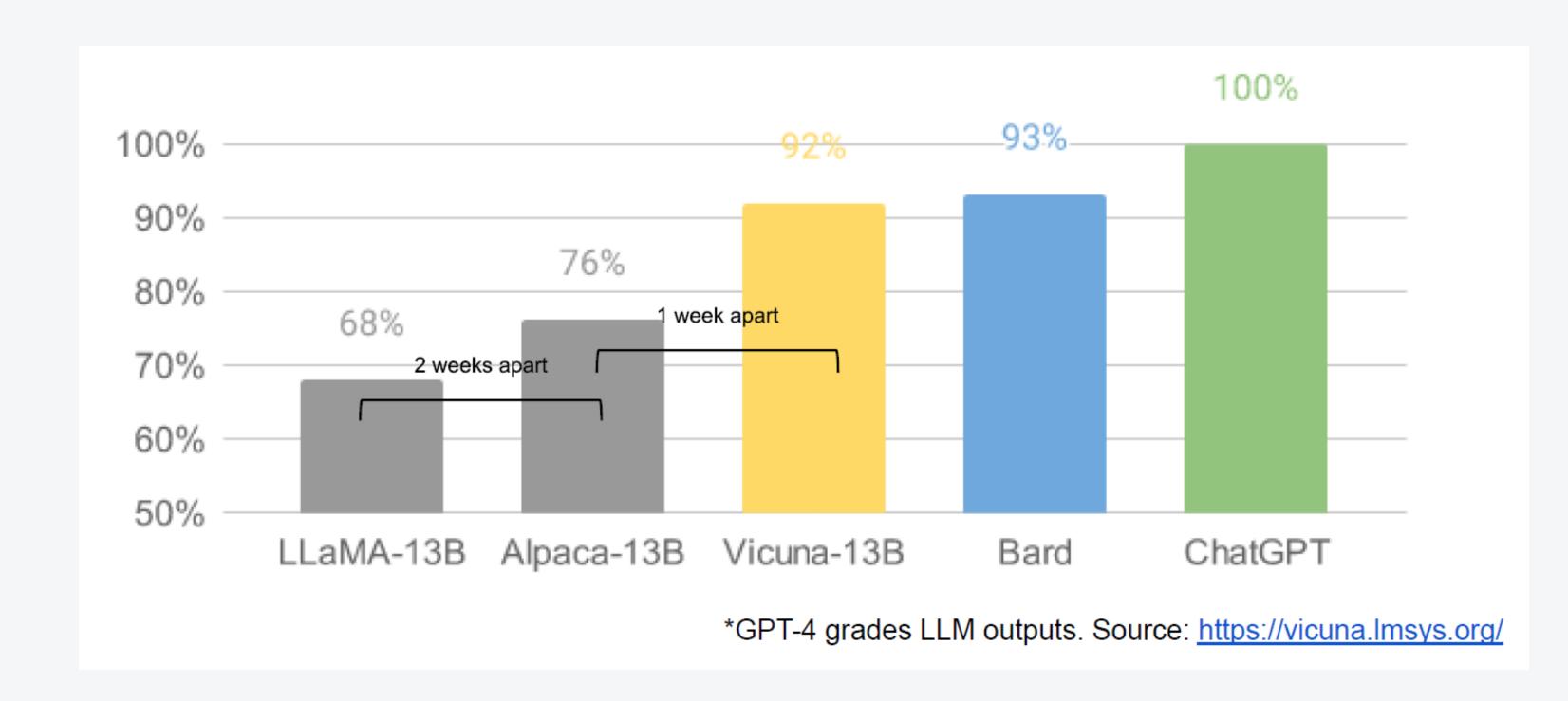
Fine-tuning

Deep Task Specialization

- Adapts LLMs to master niche domains (e.g., legal, medical, or brand-specific language).
- How? By training on small, high-quality datasets
 → reshapes model weights for domain expertise.
 - → Results in unmatched accuracy for targeted use cases.

Resource-Intensive but Powerful

- Requires:
 - Curated task-specific data (100s–1000s of examples).
 - Computational power (GPUs/TPUs) & technical expertise.
- Trade-off: Higher upfront effort than RAG, but delivers self-contained, optimized models needing no live retrieval.
- Key Takeaway: Fine-tuning creates purpose-built Al experts but demands significant investment.





Retrieval Augmented Generation (RAG)

Retrieval-First Architecture

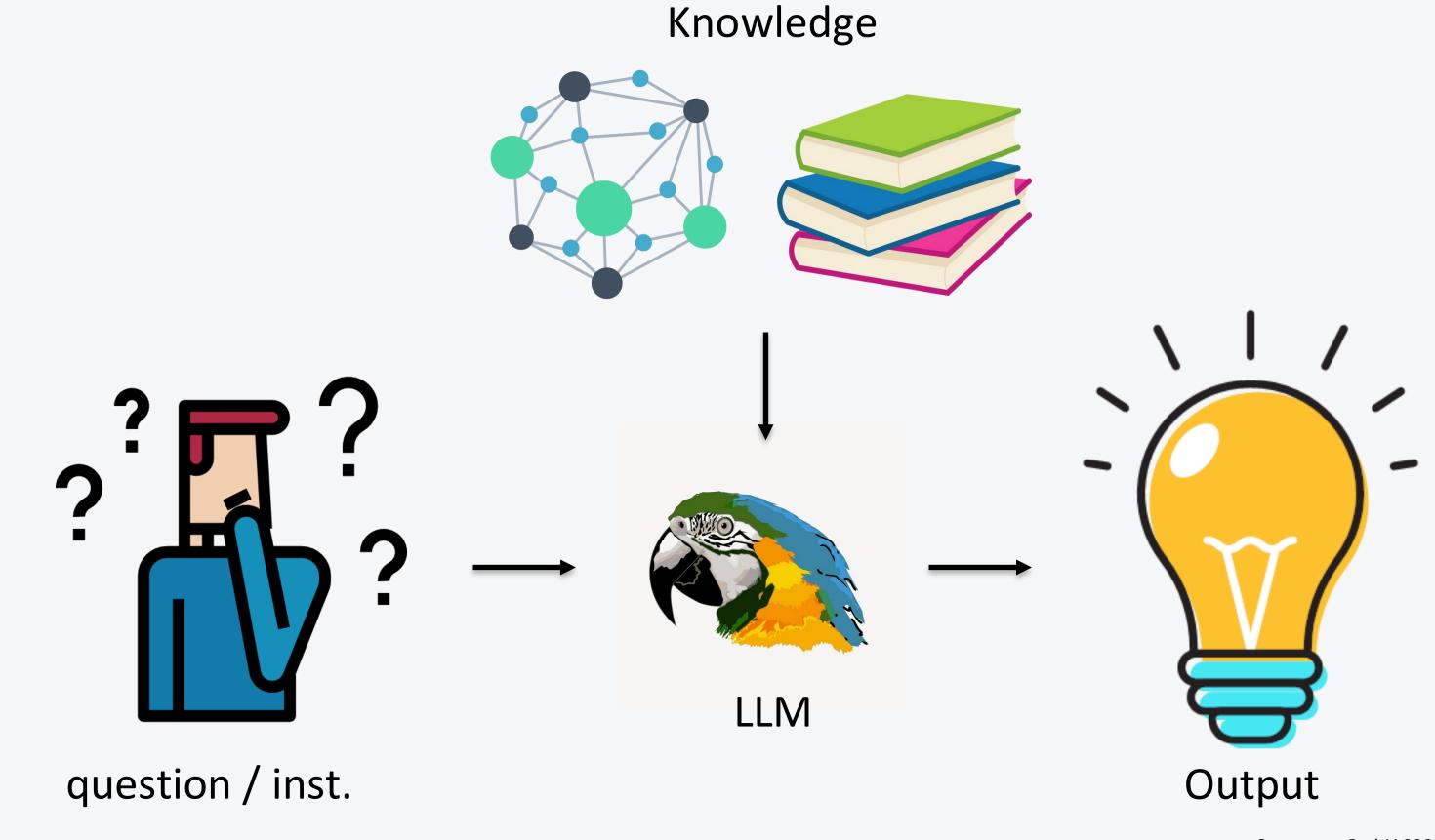
- Step 1: Retrieve relevant data from trusted sources (e.g., databases, documents).
- Step 2: Augment LLM prompts with this context.
 - → Grounds responses in real-time, domainspecific knowledge.

Mitigates Hallucinations & Outdated Info

- Uses live external data instead of relying solely on pre-trained model knowledge.
- Reduces factual errors and keeps outputs current/accurate.

Cost-Efficient & Adaptable

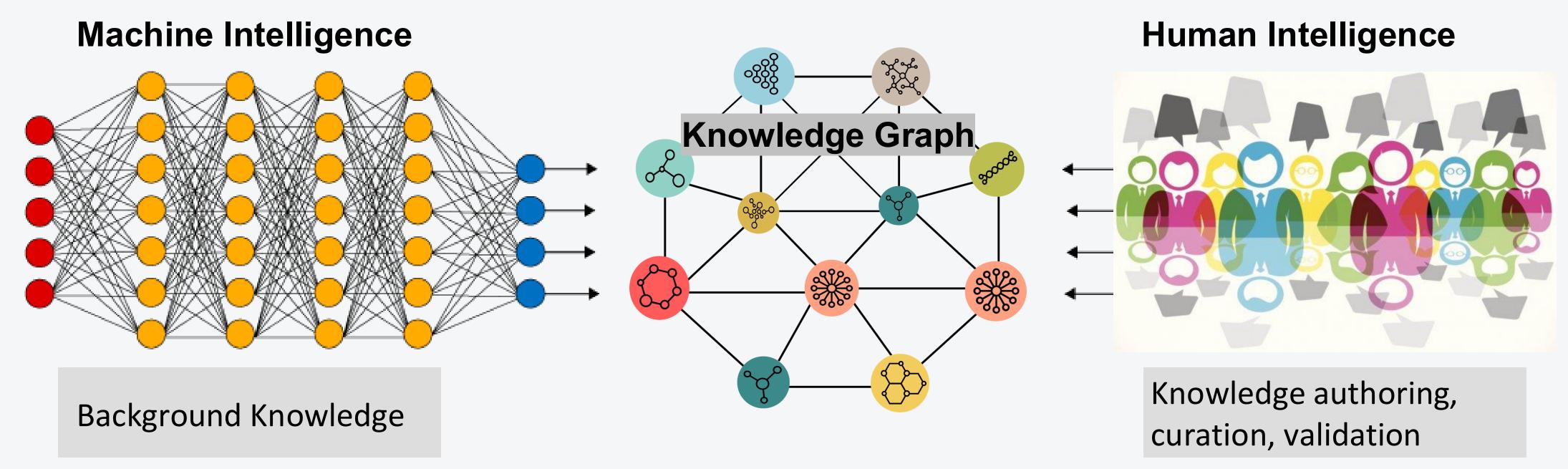
- No retraining needed for new data update the knowledge base instantly.
- Works with any LLM (e.g., GPT, LLaMA) for flexible deployment.





Best of both Worlds - Neuro-Symbolic Al

Integration of Knowledge Graphs with Machine Learning Models





Shared Vision

KG-LLM Integration Opportunities

Knowledge Graphs

- Represent a base of validated, trustworthy knowledge
- Help organize and integrate enterprise knowledge from various sources
- Provide input for enterprise and domain-specific training and fine-tuning of LLMs

Large Language Models

- Help curating knowledge in the KG by suggesting and recommending
- Create mappings, queries etc. for the KG
- Can become a frontend for human interaction with the KG







Challenges

Navigating Trade-offs in LLM Customization

Fine-tuning

Resource Intensity

- Demands heavy computation (GPU/TPU), time, and technical expertise
- Requires large curated datasets → costly/data-scarce domains struggle

Knowledge Rigidity

- Model "freezes" post-training → fails with new trends/updates
- Retraining needed for adaptation → operational bottlenecks

RAG

Retrieval Dependency

- Output quality = f(knowledge base quality + retrieval accuracy)
- Hallucinations persist if retrieval fails or sources are outdated

Latency & Complexity

- Real-time retrieval → slower responses than pure LLMs
- Synchronizing vector DBs/APIs adds architectural overhead
- Size of the Context



Any Questions? Get in Touch!





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