

Al-Driven Accelerated Discovery for NASA's Science Mission Directorate

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Science at NASA

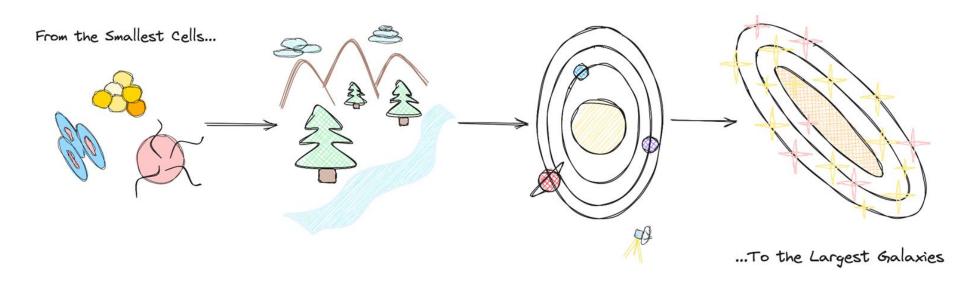


Image Credit: SDE team

SMD by the Numbers



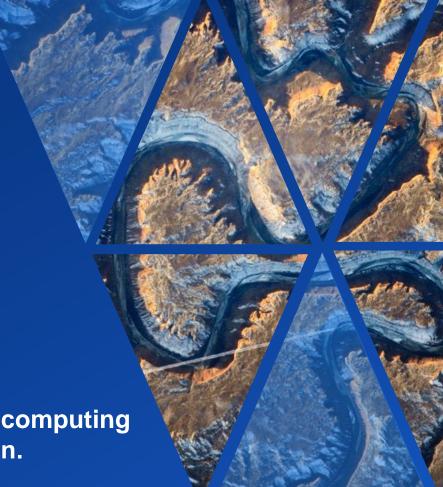








Efficient data management and computing are essential for NASA's mission.



Motivation and Foundational Principles



Motivation to Use LLMs for Science

- Large Language Models (LLMs) increase research efficiency, saving scientists significant time.
 - Streamline workflows: data discovery, access, literature review, and coding.
 - Reduce "data-wrangling" time from 80% to significantly less, accelerating scientific discovery.
- LLMs enhance data system value: improving data visibility, access, usability, and value.
 - Facilitate new discovery pathways and applications, increasing data system use.
 - Enable easier and more contextual information retrieval, aligning with user expectations.

LLMs offer *transformative benefits* for NASA science, streamlining both *science and data system operations*, thereby accelerating the pace of discovery and enhancing data utility.



Foundational Principle: AI Ethics for Science

- Prioritize open models, data, workflows, and code for transparency and collaboration.
- Devote significant effort to formulating clear, concise, and correct questions.
 - Embrace Albert Einstein's approach: Spend most of the time defining the problem accurately to enable rapid solutions.
 - Ensure questions are precisely worded to retrieve relevant answers.
- Apply Carl Sagan's "baloney detection kit" for critical analysis and fact verification.
 - Adhere to the principle of "trust but verify" to maintain rigorous scrutiny.
 - Avoid unquestioning acceptance of Al-generated outputs.
- Approach Al as a collaborative partner, not a sole decision-maker.
 - Employ the co-pilot analogy, emphasizing shared responsibility in AI interaction.
 - Remain accountable for the integrity and accuracy of Al-assisted outcomes.

Ethical AI use demands *openness, critical questioning, collaborative partnership, and vigilant verification* to ensure responsible and effective outcomes

Large Language Models for Science



Large Language Models for Science

- At NASA's Science Mission Directorate (SMD), we aim to understand the appropriate use
 of LLMs within the scientific enterprise by both researchers and developers of scientific
 applications.
- This involves investigating:
 - Whether NASA should build its own LLM for science;
 - Determining the type of model needed (encoder vs. decoder/generative);
 - Deciding between building from scratch or fine-tuning an existing open model like Meta's Llama;
 - Curating pre-training materials such as journal papers and technical reports.

Ramachandran, R., & Bugbee, K. (2025). Balancing practical uses and ethical concerns: The role of large language models in scientific research. *Perspectives of Earth and Space Scientists*, 6, e2024CN000258. https://doi.org/10.1029/2024CN000258



Large Language Models Study

- Leveraged our partnership with IBM Research to develop a LLM framework for NASA Science Mission Directorate (SMD).
- Curated sources from the five SMD topical areas for model pretraining.
 - Collaboration with subject matter experts (SMEs) from across the NASA science topical areas was essential
 - Resulted in a diverse data set totaling 66.2 billion tokens for model pre-training
- Developed an encoder-only transformer model, named INDUS, tailored for SMD applications.
 - Useful for various tasks such as named entity recognition, extractive question answering, text classification, semantic equivalence for document retrieval, and knowledge extraction for relationships
 - Also developed a distilled version of the model, five times smaller (30M parameters)
 than the original 125M model, with only marginally reduced performance.



Large Language Models Study

- Teams from the different science areas created a targeted question and answer (Q/A) suite to support the encoder model's training, focusing on precision in scientific inquiry and retrieval.
- Also created a sentence transformer:
 - Assists in information retrieval by efficiently understanding text semantics.
 - Generates embeddings for queries and sentences, enhancing information retrieval by converting them into high-dimensional vectors that capture deep semantic meanings.

Learn more about the technical details of INDUS here:

Bhattacharjee, B., Trivedi, A., Muraoka, M., Ramasubramanian, M., Udagawa, T., Gurung, I., et al. (2024). INDUS: Effective and efficient language models for scientific applications. arXiv, 98–112. https://doi.org/10.18653/v1/2024.emnlp-industry.9

Find INDUS on Hugging Face here:

Maraoka, M., Bhattacharjee, B., Ramasubramanian, M., Gurung, I., Ramachandran, R., Maskey, M., et al. (2023). Nasa-impact/nasa-smd-ibm-v0.1 (Version 0.1) [Software]. Hugging Face. https://doi.org/10.57967/hf/1429



Key Takeaways from the Study

- Collaboration and partnerships were critical to this study.
 - IBM Research for LLM expertise and needed compute resources.
 - SMD subject matter experts for domain expertise.
- A single type of model is not the right answer for all solutions.
 - A suite of models, including both encoders and decoders, is essential for supporting diverse downstream applications.
- Retrieval augmented generation (RAG) is a cost-effective, low-risk strategy for more responsibly using LLMs for science.
 - RAG involves using domain-specific encoder models, such as INDUS, within a RAG framework for information retrieval, which can be integrated with existing off-theshelf generative models like Llama or GPT.
 - RAG combines document retrieval (via an encoder) and generative modeling to enhance answer accuracy and relevance by providing contextual grounding.



LLM Applications: Stewardship Workflows

Image Credit: SDE Team

- Automated content curation.
 - Using LLMs to automate document classification for curated search and discovery experiences.
 - TDAMM, division classifier, GCMD keyword recommender, division classifier.



- Example: Time Domain and Multi-Messenger Astronomy Search Interface.
 - Relatively new field in astrophysics observations cover a wide range of time-varying and types of phenomena/messengers where messengers can be cosmic rays, electromagnetic radiation, gravitational waves and neutrinos.
 - Relevant data and information is dispersed across a number of archives and repositories—
 there is a need to make search and discovery easier for the TDAMM community.
 - Used astroBERT to build a classifier to streamline content curation.



LLM Applications: Stewardship Workflows

- Example: Earth Science Keyword Recommender.
 - Global Change Master Directory (GCMD)
 Keywords are a set of controlled
 vocabularies for the Earth sciences. Used to label metadata and other documentation.
 - GCMD Keyword recommender helps human curators accurately tag their data with relevant GCMD keywords.
 - We enhanced the keyword recommender by fine-tuning INDUS for the classification task.
 - New fine-tuned model outperforms other models including RoBERTa and the existing model.

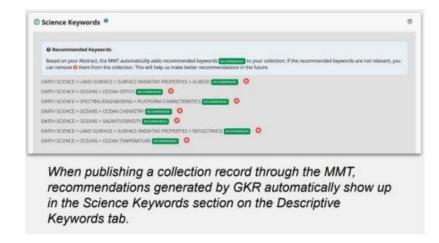


Image Credit: Bhattacharjee et al. https://ntrs.nasa.gov/api/citations/20240015188/downloads/AGU%202024%20INDUS%20Applications.pdf



LLM Applications: Enhanced Search

- Integrated the sentence transformer and passage re-ranker adapted from the base INDUS encoder model with a general model ChatGPT within a RAG framework into a prototype Science Discovery Engine (SDE) environment.
- Qualitative and quantitative assessments indicate improved relevancy of search results.

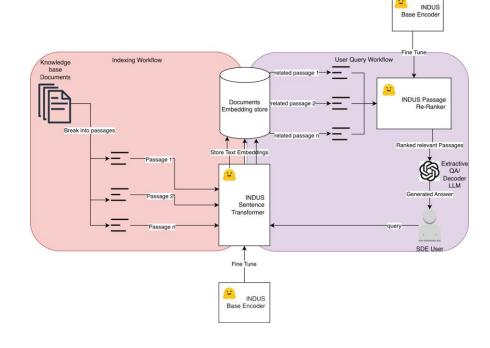


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LLM Applications: Enhanced Search

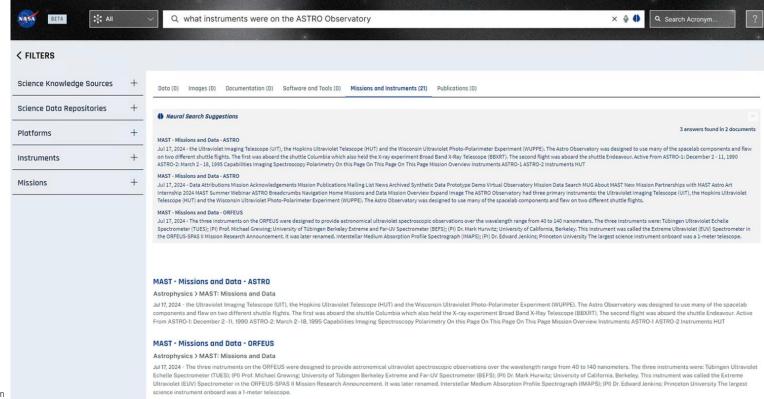
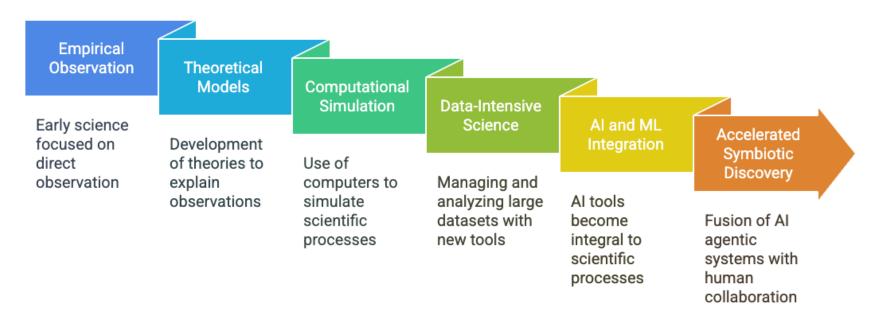


Image Credit: SDE Team



Evolution of Scientific Paradigms

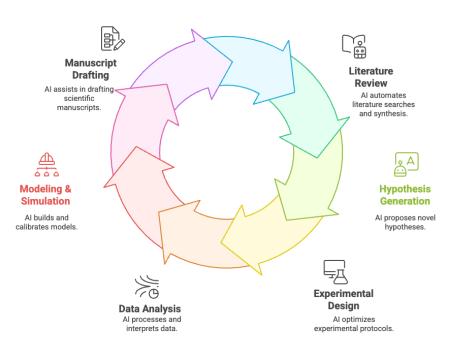


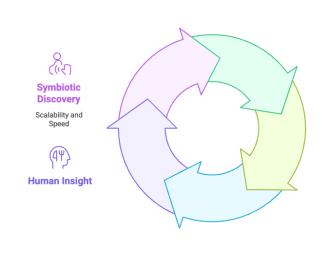
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Repositions AI from a tool to collaborator

Symbiotic Scientific Discovery Cycle

Closed-Loop Research Cycle







Phase
Initial findings
inform subsequent



Hypothesis Refinement

Insights refine research questions

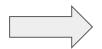


Task Automation

Routine tasks handled by machines

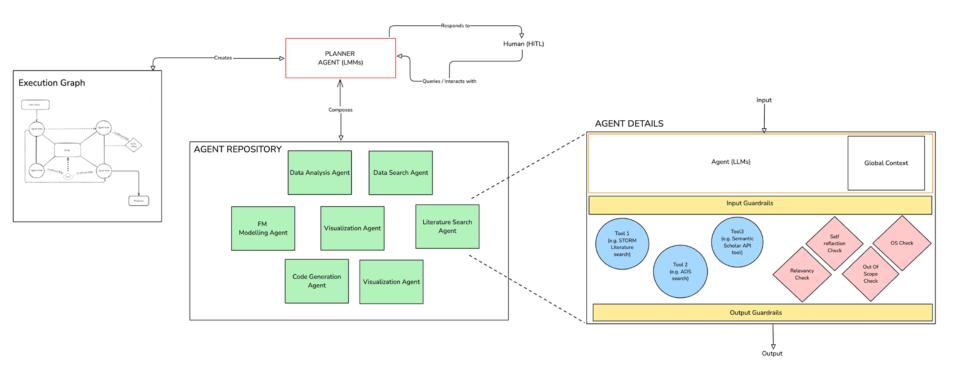
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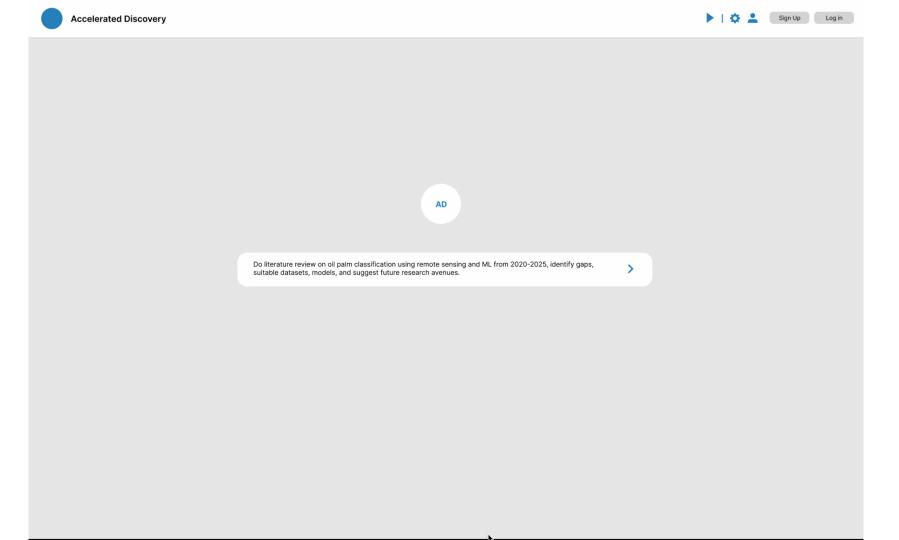
The scientific research process augmented by Al.

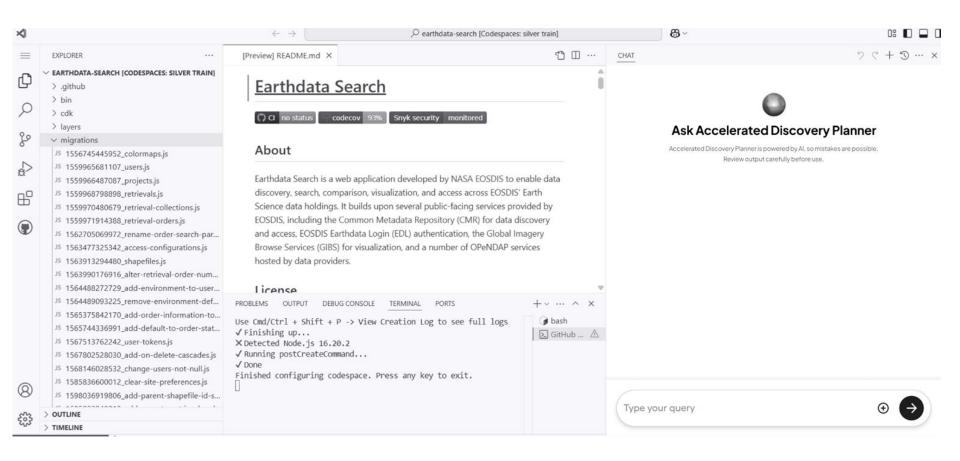


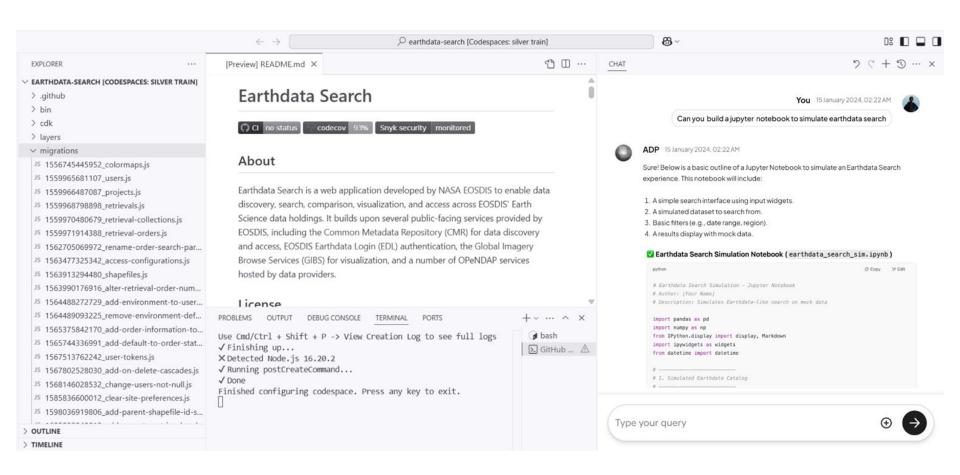
Al can help a scientist iterate and refine each step of the research process.

Conceptual Multi-Agent Architecture











Conclusions

- NASA is positioned to enable a transformative shift in scientific discovery.
- Accelerated Symbiotic Discovery will
 - Integrate agentic AI, multimodal foundation models, scalable cloud and high-performance computing (HPC) infrastructure, and existing knowledge-representation frameworks
 - Accelerate scientific discoveries from the smallest of cells to the largest of galaxies.



Call to Action

- We envision this agentic infrastructure to be open, transparent, and reproducible
- We welcome collaboration, thoughts and feedback on this effort

THANK YOU!

Questions?

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