

# Agentic AI for Nuclear Materials Design: From Tool to Co-Investigator

*"When does a workflow become a collaborator?"*

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# Motivation: Why Nuclear Materials?

## Why it's hard



**Extreme Complexity Across Scales**  
From atomic defects (picoseconds) to fuel rods (m, years)  
Couple physics: radiation, chemistry, mechanics

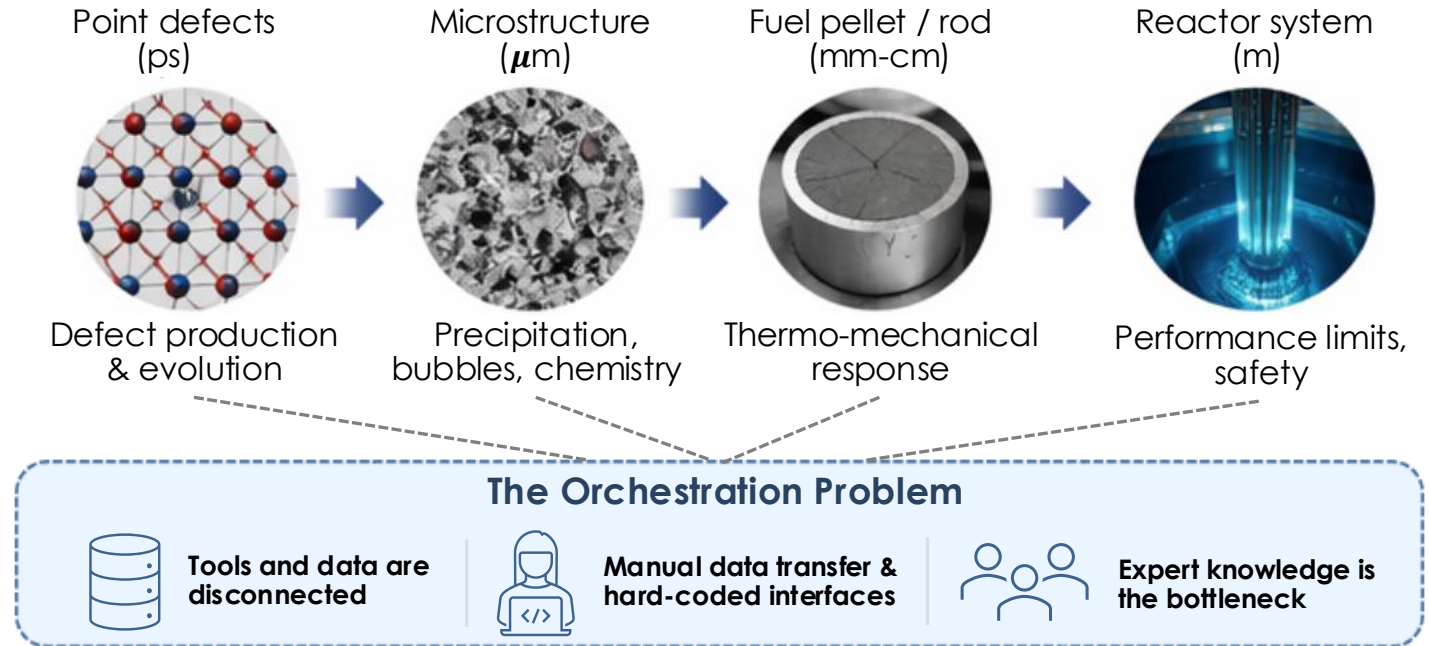


**Fragmented Knowledge**  
Experiments, simulations, ML models operate in isolation  
Integration relies on manual workflows and expert judgement



**Slow Qualification Cycle**  
Conventional sequential testing takes decades to qualify new fuels and architectures

## Multiscale Nuclear Materials Behavior



*We don't lack models or data—we lack a way to connect them into a coherent scientific workflow*



Agentic systems can **dynamically connect models, data and simulations**, and automate hypothesis  $\rightarrow$  simulation  $\rightarrow$  interpretation loops



From months of analysis to minutes

When agents start making scientific decisions, **are they still tools or co-investigators?**

# What We Mean by “Agentic System”

*From fragmented tools to an integrated, human-centered workflow*

## What Changes with Agentic AI



**Brings models, data and experiments** together as interoperable modules



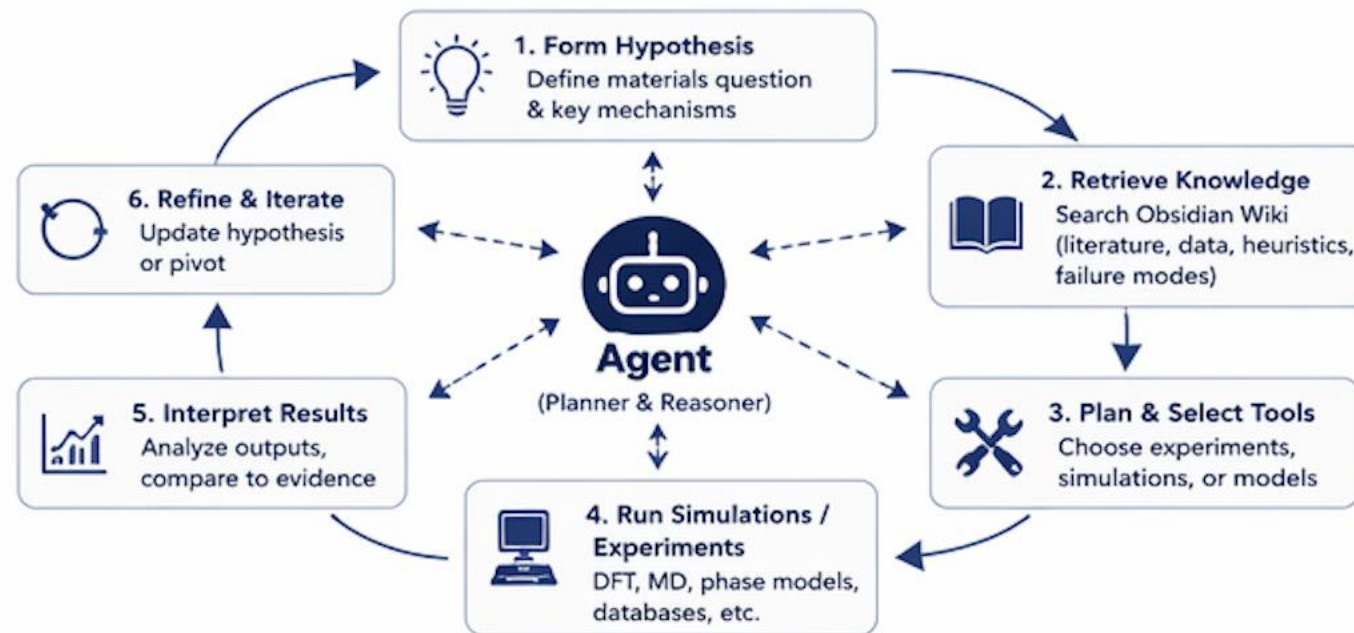
**Automates the scientific workflow** hypothesize – plan – run – analyze – loop



**Learns and remembers** from prior results, assumptions and failure modes



**Improves speed and reproducibility** by enabling traceable reasoning



## Agentic Framework

### LLM-based planner

**Knowledge base:** Wiki of prior experiments, assumptions, failure modes

**Tools:** DFT/MD simulation interfaces, materials database, surrogate models

## From tools to co-investigation

**Agentic AI does not replace scientists**, it augments them. The goal is faster, more informed decisions while **preserving human judgment** as the ultimate arbiter of knowledge.

## Degrees of freedom

**Accuracy/cost tuning** through task granularity: precise vs general tasks

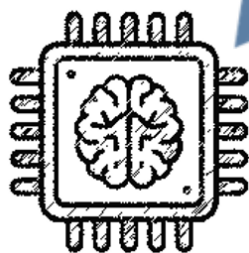
**Handling the knowledge base:** provenance tracking, GNN, recommendation system.

# Example Workflow

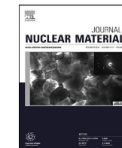
1.2 GB based on over 500 DFT simulations and 2 million downloaded entries



<https://github.com/ORNL-Inria/AERIS>



Reproduce the findings of a fission paper



## Stability of $U_5Si_4$ phase in U-Si system: Crystal structure prediction and phonon properties using first-principles calculations



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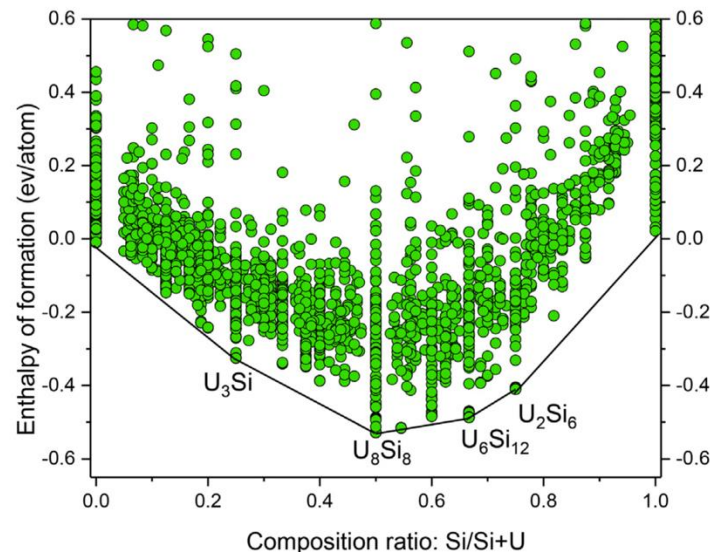
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### ABSTRACT

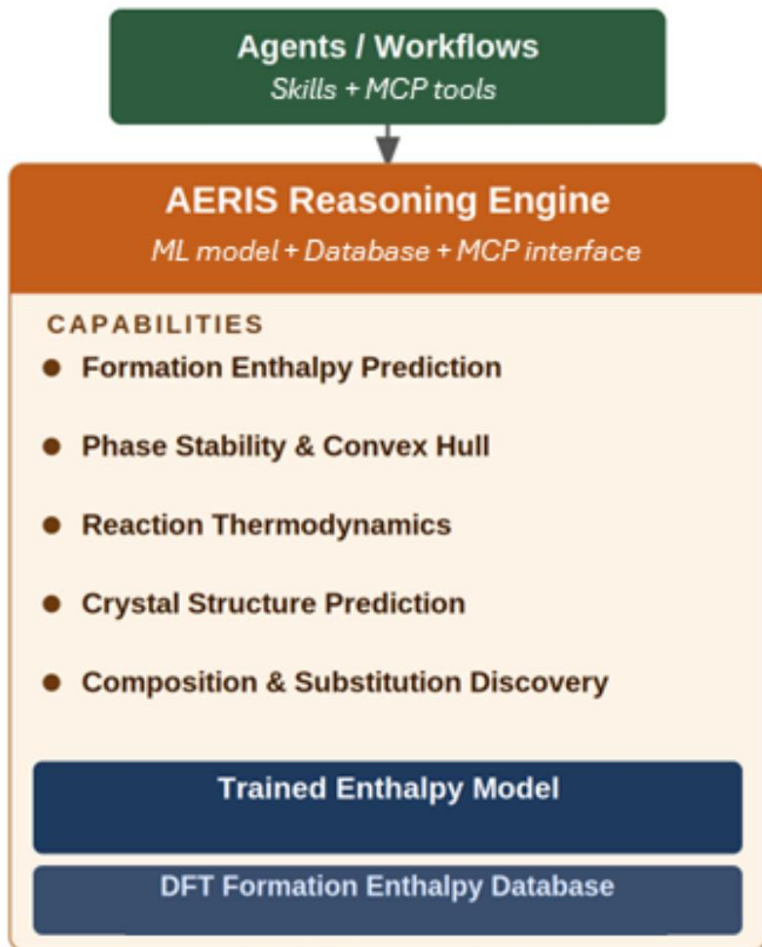
U-Si systems have recently received considerable attention due to the potential application of  $U_2Si_2$  as a high-density fuel under an accident tolerant fuel initiative. However, the thermodynamic stability of the more recently reported adjacent  $U_5Si_4$  phase is uncertain and could play a significant role in fuel performance. In this work, the enthalpy of formation of the phase predicted by density functional theory (DFT) using the DFT +  $U$  formalism is used with an evolutionary algorithm (USPEX) to evaluate stability and possible atomic structures for  $U_5Si_4$ . The structure of U-Si convex hull phases and the confirmed  $U_3Si_2$  structure were predicted providing confidence in the reliability of the evolutionary algorithm, as well as the stability of  $U_5Si_4$  with reasonable enthalpies of formation. Subsequently, the code was used to predict the structure of  $U_5Si_4$  with 18 and 36 atoms, predicting a 36-atom hexagonal structure as the lowest-energy configuration, agreeing with that calculated using the density functional perturbation method. The structure of  $U_5Si_4$  is dynamically unstable, exhibiting negative phonon frequencies that generated shears are directed toward the structure analogous to those occupied by carbon atoms in  $U_{20}Si_{16}C_3$ . The stability of  $U_5Si_4$  phase equilibria is currently under investigation.

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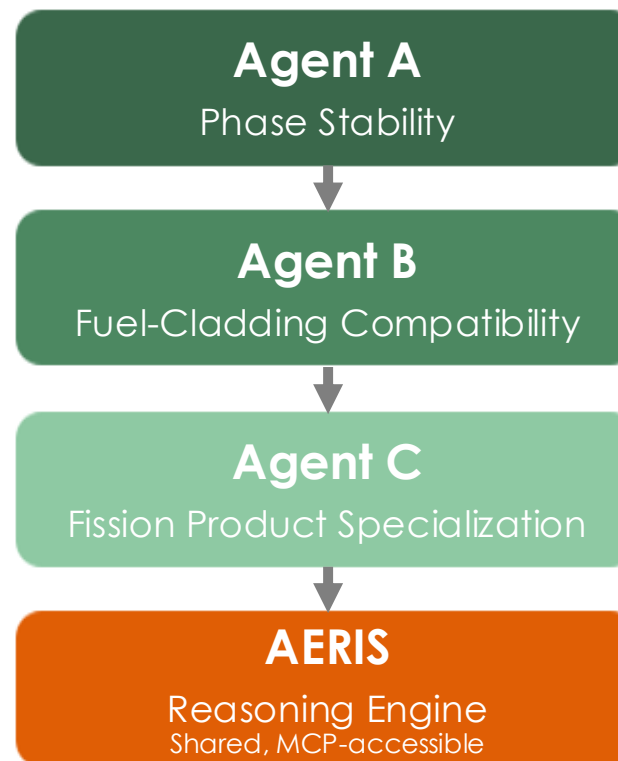


# Agentic Framework (AERIS)

AERIS reasoning engine

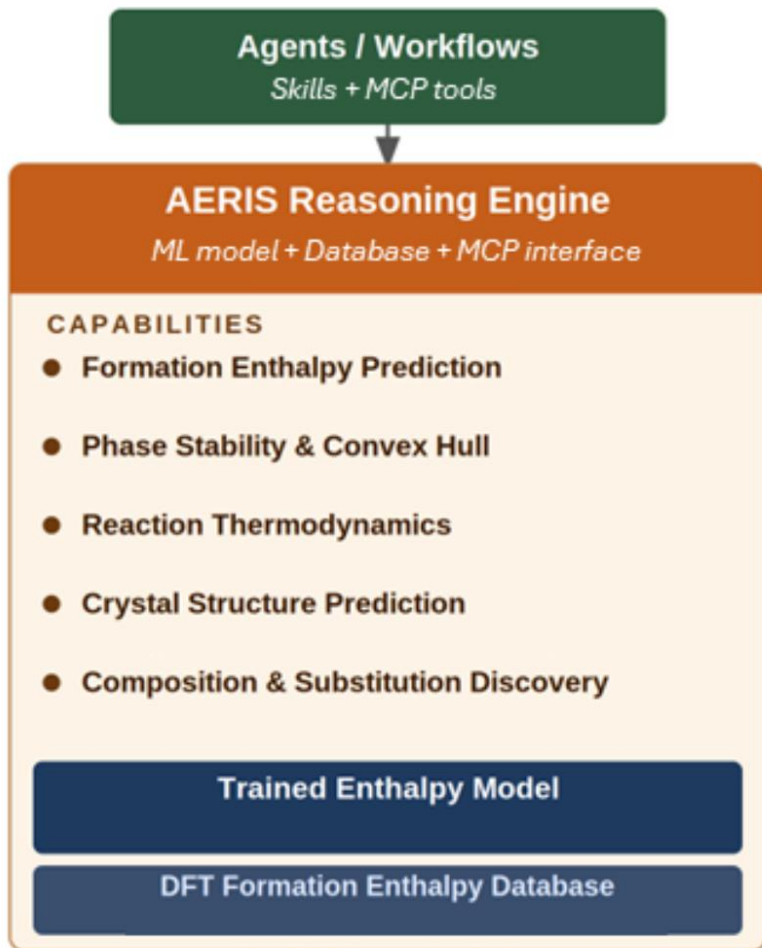


Single model multiple agents

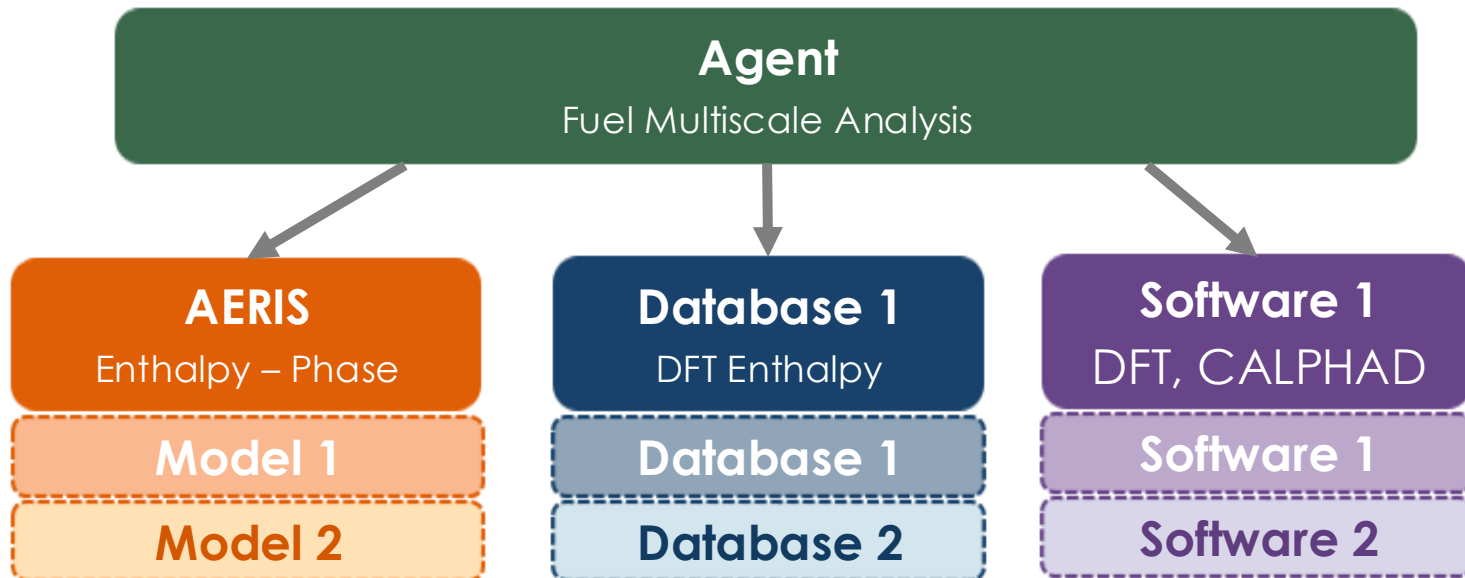


# Agentic Framework (AERIS)

AERIS reasoning engine



Single agent multiple resources



# Where It Starts Acting Like a Co-PI

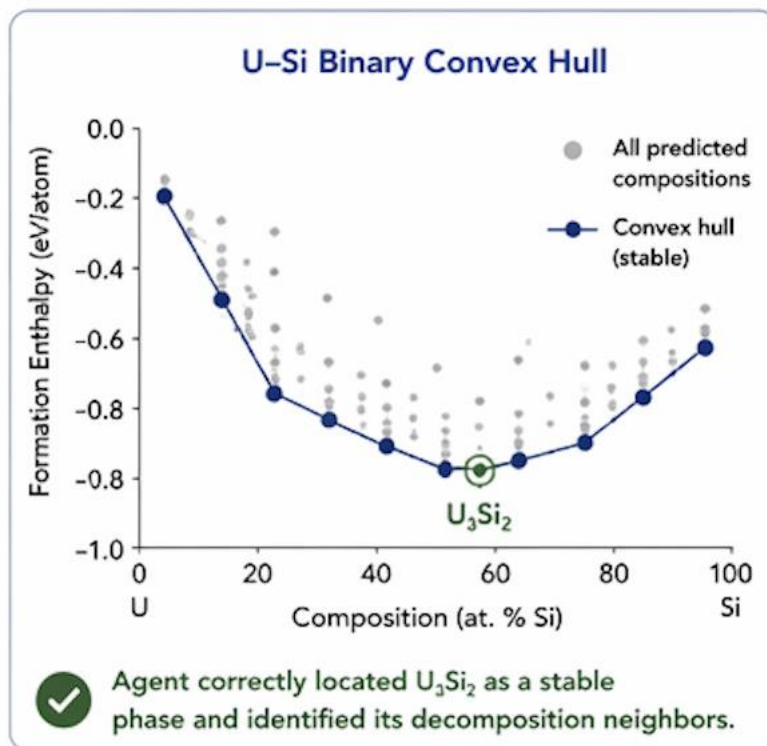
## Case Study: Reproducing the U-Si Binary Convex Hull



**Task given to agent:**  
"Characterize the U-Si system with formation enthalpies and determine stable phases."

### What the agent did autonomously

- Enumerated candidate compositions across U-Si
- Selected relevant structural templates
- Predicted formation enthalpies using AERIS Enthalpy (DFT-based)
- Constructed the convex hull
- Identified stable phases and decomposition neighbors



Why this is scientific reasoning—  
not just execution

**Chooses which components to evaluate:** Defines the search space (hypothesis framing)

**Decides which structures are relevant:** Encodes physical assumptions (structure selection)

**Selects which models to query:** Makes methodological choices (model, database)

**Determines lowest-energy phases:** Evaluates multiple structure prediction and filters based on relevance to the hypothesis

**Infers stability & decomposition pathways:** Produces scientific conclusions (knowledge generation)

The agent is not just running calculations- it is deciding what counts as evidence and what conclusions to draw



Runtime: minutes

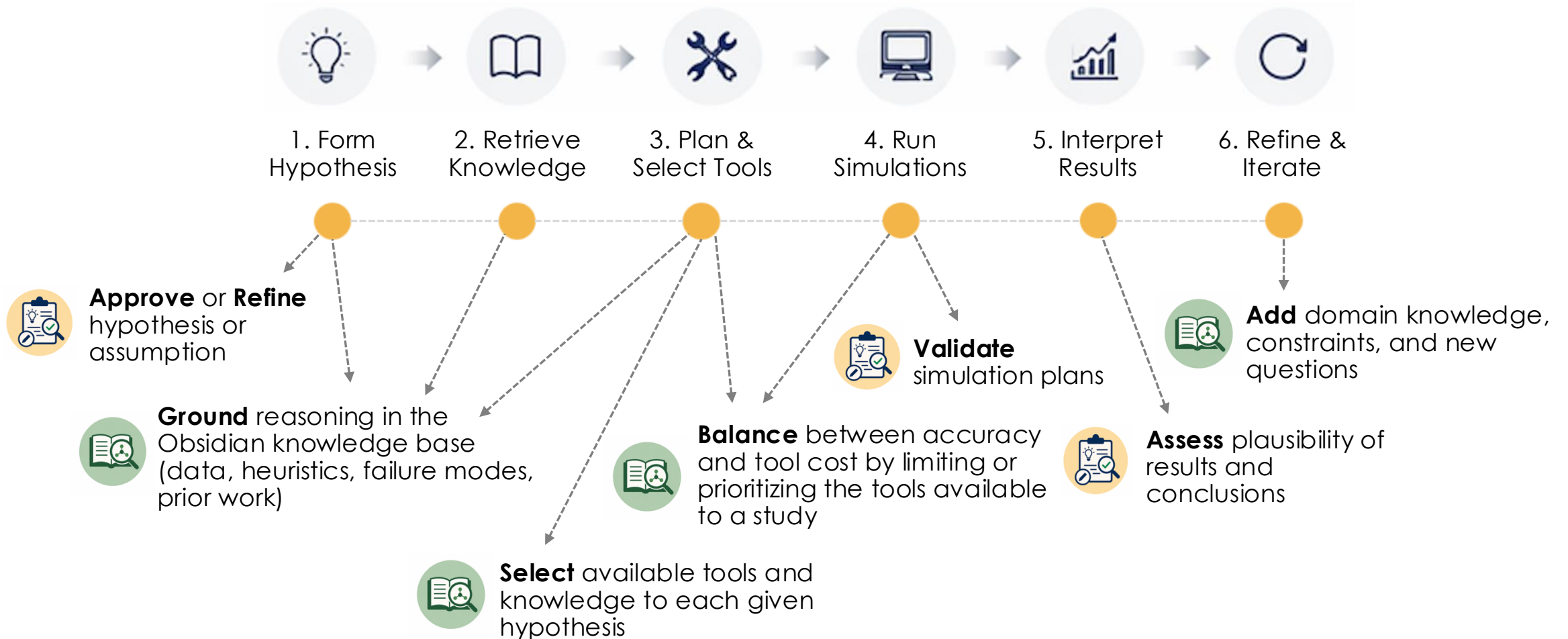


No predefined workflow – the agent determines the sequence from the task

(Adorno Lopes et al., 2025)

# Human as Arbiter: Keeping Scientists in the Driver's Seat

*Ensuring human judgment remains the ultimate authority on plausibility and knowledge*

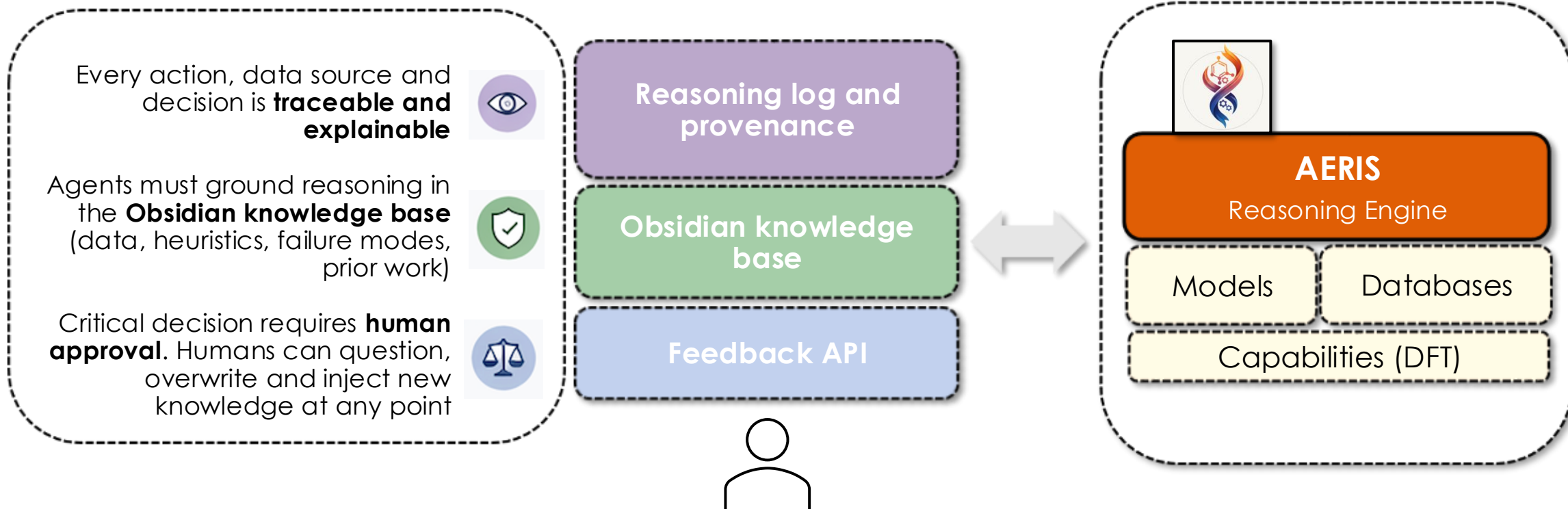


● = human checkpoint is required or recommended

# Human as Arbiter: Keeping Scientists in the Driver's Seat

*Ensuring human judgment remains the ultimate authority on plausibility and knowledge*

## Human Arbitration Principles



*“ The goal is not to slow the agent down-it is to ensure the science moves forward with transparency, rigor, and human judgment as the final authority “*



# Open Questions: When Agents Become Co-Investigators?

## What counts as a "scientific decision"?

- Is selecting a model a decision?
- Is defining the search space a hypothesis?
- Is interpreting results reasoning or post-processing?



## How do we evaluate agent-generated science?

- Accuracy vs scientific insight
- Agreement with known results vs novel discovery
- Benchmarks vs real scientific workflows



## What does "co-PI" actually mean?

- Autonomy in decisions?
- Contribution to knowledge?
- Attribution / credit?
- At what point does assistance become collaboration?



## Where should human oversight occur?

- At hypothesis generation?
- At tool selection?
- At result interpretation?
- How do we ensure agents amplify, not replace, human understanding.