

# Hybrid Agentic AI-FSM Framework for Instruction-based Industrial Manipulation Tasks

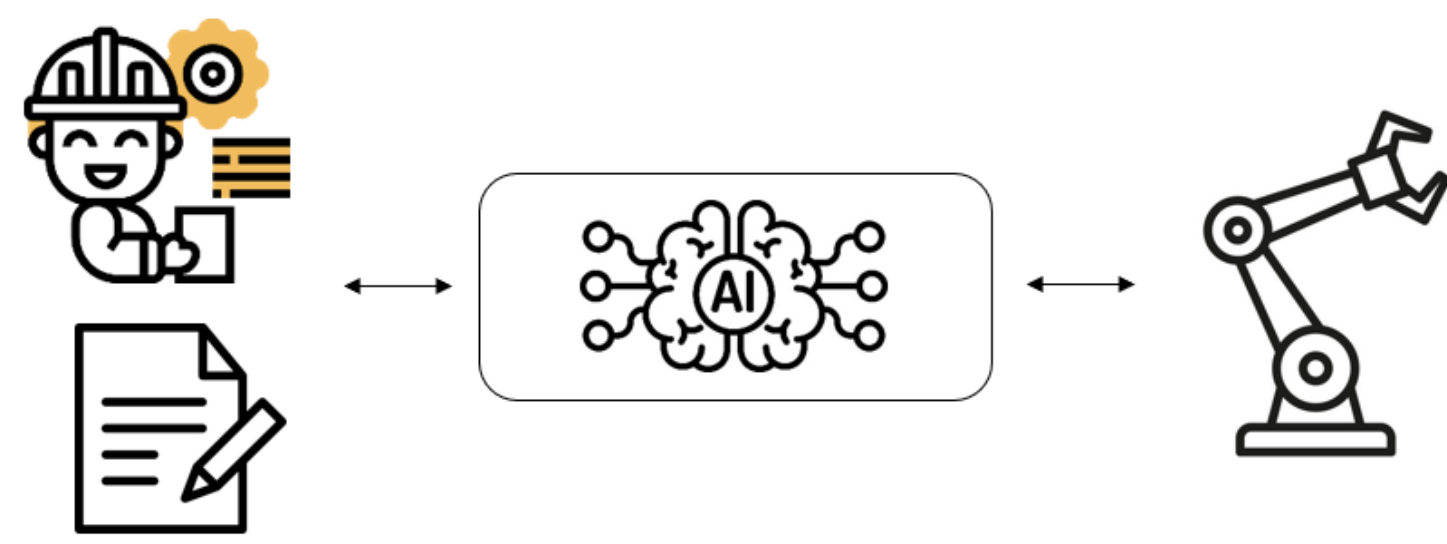
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## Motivation & Challenges

- In small- to medium-sized factories with flexible production lines, robotic systems must handle frequently changing assembly tasks without requiring time-consuming reprogramming.



- Industrial Needs:** High-reliability, high-risk, and safety-critical environments—such as manufacturing—demand robot stability that directly impacts productivity.
- Gap in Automation:** While industrial tasks are typically documented in natural language (SOPs), converting these instructions into executable robot programs currently requires intensive manual effort.
- Limitations of Current AI:** State-of-the-art Vision-Language-Action (VLA) models are "black-box" systems prone to hallucinations and non-determinism, making them unverifiable by industrial safety standards.
- The Dilemma:** A fundamental conflict exists between the flexible reasoning of LLMs and the deterministic reliability required for industrial control.



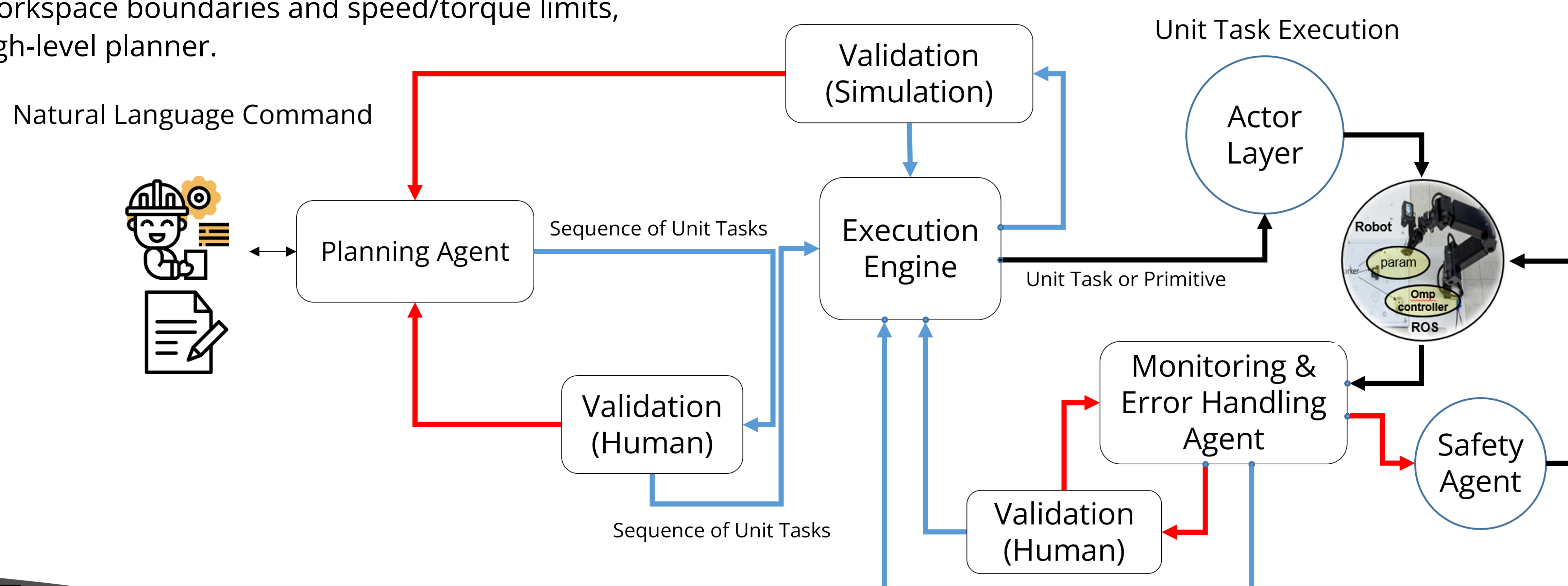
## Proposed Solutions: Hybrid Agentic AI-FSM Framework

### Key Idea

- A complex task is decomposed into simple unit tasks.
- A language model, restricted to offline planning, generates a unit-task-level task plan.
- A verified FSM-style execution engine manages the deterministic execution of the task plan.

### Key Elements

- Hierarchical Task Planning:** Instructions are decomposed into a three-level hierarchy—Composite Tasks, Unit Tasks, and Primitive Actions—ensuring semantic grounding and valid execution.
- Multi-stage Validation Pipeline:** Logical and physical validity is verified through schema checks, rule-based constraints, and simulation-based dry-runs before any real-world deployment.
- RAG-Enhanced Exception Handling:** Upon detecting runtime anomalies, a RAG-based agent retrieves relevant manuals and proposes recovery actions. The proposed recovery plan must be approved by a human operator (Human-in-the-Loop) before task execution resumes.
- Independent Safety Agent:** A dedicated physical filter enforces real-time constraints, such as workspace boundaries and speed/torque limits, independent of the high-level planner.

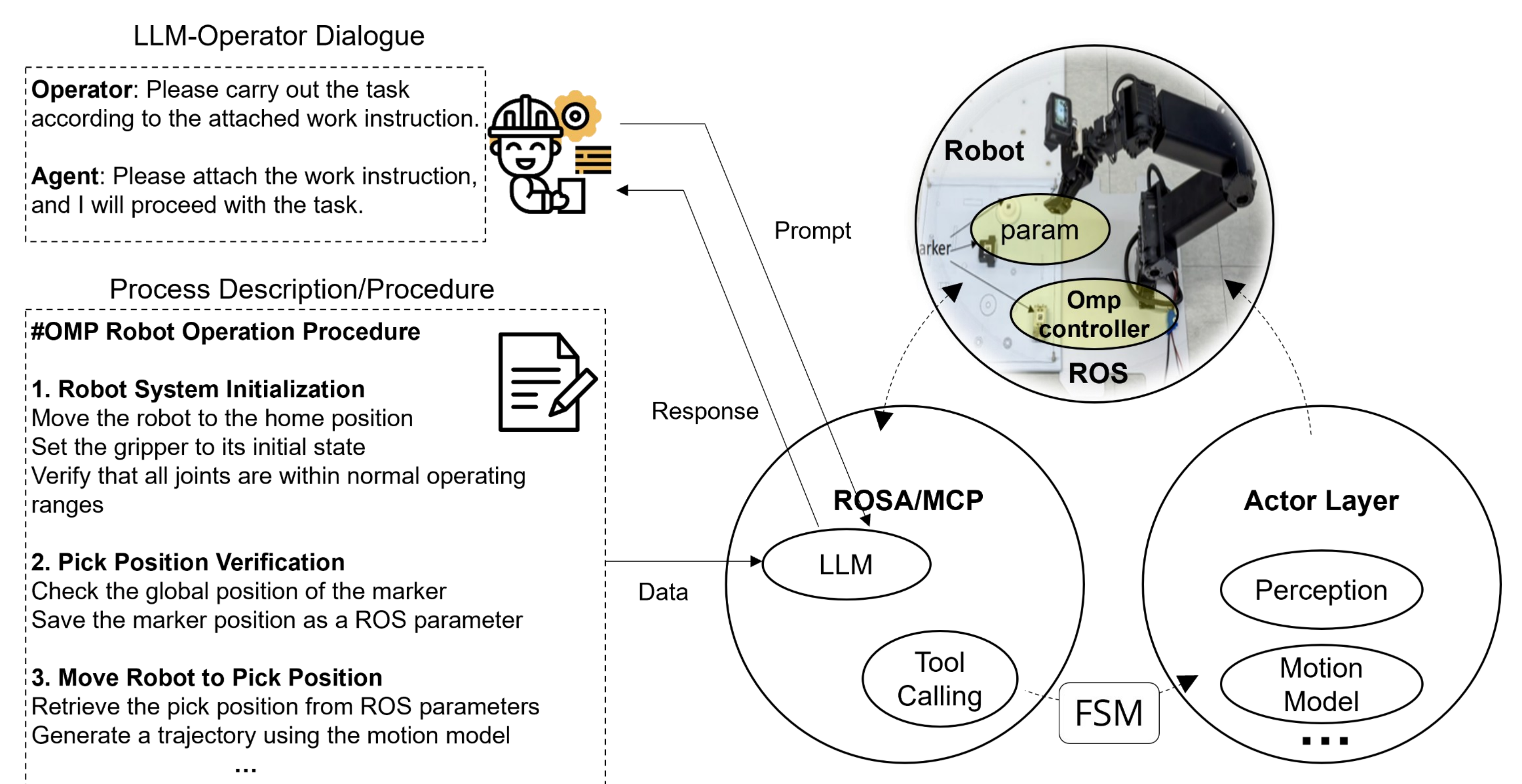


## Implementation

### Agentic AI Framework for Robot Control

- ROSA:** ROSA is an AI-powered assistant for ROS1 and ROS2 systems. Built on the Langchain framework, ROSA enables operators to interact with robots using natural language (<https://github.com/nasa-jpl/rosa>)
- ROS-MCP-SERVER:** ROS-MCP-SERVER connects large language models (such as Claude, GPT, and Gemini) to robots, enabling bidirectional communication with no changes to existing robot source code (<https://github.com/robotmcp/ros-mcp-server>)

Component	Implementation Tool	Role & Responsibility
Planner Agent	ROSA (LangChain-based)	High-level reasoning and Unit Task composition.
Monitoring & EH Agent	ROSA	Real-time log/event auditing and autonomous recovery.
High-level Safety Agent	ROSA	Strategic safety validation (Logical Fail-safe).
Context Gateway	ROS-MCP-SERVER	Exposes ROS context to external LLM clients (Claude, Cursor).
Execution Engine (Outside Agent)	BehaviorTree or State Machine	Deterministic execution of the Unit Task sequence.
Low-level Safety Agent (Outside Agent)	Native C++ / ROS Node	Real-time collision avoidance and hardware protection.
Actor Layer (Outside Agent)	PyTorch (ROS Inference Node)	Execution of Imitation-Learned (IL) policies.



## Summary & Conclusion

- The proposed framework implements a multi-layered protection strategy (aka "Defense-in-Depth" in nuclear industry), integrating logical guardrails with physical safety controllers to mitigate both reasoning errors and physical risks.
- This hybrid approach overcomes the uncertainty of end-to-end models, offering a practical, explainable, and safe solution for natural-language-driven industrial automation.

