Challenge Report

Preparation

1. Read the challenge carefully to match and achieve my mission goals

- 1. Understand the scenario.
- 2. Visit sandbox link and explore it.
- 3. Comprehend the expectation from the evaluation team.
- 4. Underline the deliverables to pass the challenge.
- 5. Considering time expectation to respect it.

2. Planning

- 1. Deep Dive through the 5G network environment sandbox.
- 2. Code NetworkSandbox class in engine.py and test_api.py to communicate with network endpoints, and verify if it is working well.
- 3. Build NetworkAgent in agent.py upon the NetworkSandbox class to use endpoints methods as Tools.
- 4. Program test agent.py file to test different scenarios.

Step-by-step walkthrough

1. NetworkSandbox class:

I decided to create an engine that would handle different API request through a class that would be modulable and flexible.

Core Request Handler

The heart of the NetworkSandbox class is the _make_request method, which provides a unified way to handle all API interactions:

API Endpoint Implementations

Based on the OpenAPI specification, I implemented the six required endpoints:

- 1. Network Control Operations: start network and stop network
- 2. Device Management Operations: connect device and disconnect device
- 3. Monitoring Operations: get network status and get system logs
- 4. (Optional) Utility Methods: is network running and wait for operation

Robust Error Handling and Retry Mechanism

A critical component I implemented was the <u>_retry_operation</u> method to ensure reliability in network operations

2. API Testing and Validation

Before building the intelligent agent, I created comprehensive testing utilities in test api.py:

3. Intelligent Agent Architecture

NetworkAgent Class Design

The core intelligent agent builds upon the NetworkSandbox foundation:

- 1. **Tool System Implementation:** The agent uses a dynamic tool system that maps natural language commands to API operations via initialize tools
- 2. **Intelligent Planning Engine:** The key innovation is the planning system that uses LLM reasoning to generate a plan generate plan
- 3. **Natural Language Processing:** Advanced NLP capabilities for parameter extraction in extract device id
- 4. **Execution Pipeline:** The agent follows a sophisticated four-step execution process in process command
- 4. Testing Framework Implementation

Comprehensive Scenario Testing

test agent.py provides a robust testing environment:

Interactive Mode

For real-time testing and demonstration:

End-to-End Pipeline

Stage 1:Input Phase

Engineer Natural Language Prompt → NetworkAgent.process_command()

- Input: Raw natural language query from network engineer
- Processing: Command received by the main entry point process command ()
- Validation: Basic input sanitization and initialization
- Example: "Why isn't John's phone connecting? His device ID is 90400"

Stage 2: Planning Phase

Plan Generation with Ollama \rightarrow JSON Plan Extraction \rightarrow Structured Action Plan

- LLM Analysis: generate plan() uses Ollama to analyze the request context
- JSON Schema: LLM produces structured plan following predefined format
- Plan Validation: Regex-based JSON extraction with fallback defaults

Action Sequencing: Logical ordering based on dependencies (check status before actions)

Stage 3: Parameter Extraction

Device ID Pattern Matching → Validated Parameters

- Pattern Recognition: extract device id() applies multiple regex patterns
- Priority Handling: Device ID from plan takes precedence over extracted ID
- Validation: Format verification before API usage

Stage 4: Execution Phase

Sequential Action Execution \rightarrow Tool Selection & Routing \rightarrow API Calls \rightarrow Response Processing

- Action Dispatch: _execute_actions() iterates through planned actions
- Tool Routing: Dynamic mapping to NetworkSandbox methods via tools registry
- API Integration: Each action triggers corresponding endpoint call:
 - get network status \rightarrow GET /network/status
 - connect device → POST /device/connect with device parameters
- Retry Logic: Automatic retry with exponential backoff for failed operations
- Result Collection: Structured storage of success/failure outcomes

Stage 5: Response Generation

Results Aggregation → LLM Response Synthesis → Final Natural Language Answer

- Result Compilation: All action outcomes aggregated into structured summary
- LLM Synthesis: _generate_response() uses Ollama to create human-readable response:
 - Context: Original query + executed plan + results
 - Format: Professional but friendly network operations style
 - Content: Actions taken, findings, issues, recommendations
- Quality Assurance: Temperature control (0.7) for natural yet consistent responses

Stage 6: Output Phase

Final Natural Language Answer → Engineer Response

- Delivery: Clean, formatted response returned to engineer
- Logging: Full execution trace for debugging and audit
- Performance Metrics: Execution time tracking and reporting

Future Enhancement Opportunities

Given more time, I would implement:

Immediate Improvements (1-2 days)

- 1. Conversation Memory: Maintain context across multiple interactions
- 2. Advanced Diagnostics: Root cause analysis for common network issues
- 3. Proactive Alerts: Automatic issue detection before user reports

Medium-term Enhancements (1-2 weeks)

- 1. Multi-modal Integration: Support for visual network topology
- 2. Performance Optimization: Caching and parallel execution
- 3. Learning System: Historical data analysis for improvement

Long-term Vision (1+ month)

- 1. Predictive Maintenance: Al-driven failure prediction
- 2. Natural Language Generation: More conversational responses
- 3. Integration Ecosystem: Plugins for other network management systems

Conclusion

This design provides a scalable foundation for an intelligent network agent. It demonstrates pragmatic use of AI techniques to solve real-world operations problems. Future enhancements could include integration with actual network APIs, role-based access control, and predictive analytics. The agent balances usability with technical feasibility, aligning with the challenge's goals.