# **Knowledge Base System Improvements**

## **Technical Overview for Presentation**

### 1. Problem Statement

The Nova AI Coordinator system exhibited three critical issues in knowledge base functionality:

- **Intent Separation Issue**: When processing multi-intent prompts (e.g., "Calculate 10+20, what is the capital of Germany..."), the system stored the entire prompt as a single unit
- **Precision in History Recall**: When asked about previous tasks (e.g., "What was the last math question I asked?"), the system returned the entire multi-intent prompt rather than just the relevant part
- **Entity Extraction**: The system failed to extract specific entities (like "France" from energy model requests), limiting contextual awareness

### 2. Solution Architecture

We implemented a three-part solution architecture:

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### 1. Intent Separation Layer

- Uses Nova's existing intent detection capabilities
- Processes and separates multi-intent prompts
- Routes each intent to the appropriate agent

### 2. Categorized Storage Layer

- Stores each intent separately with proper metadata
- Maintains hierarchical session structure
- Implements entity extraction for enriched data

## 3. Contextual Retrieval Layer

- Enhances history query detection
- Implements intent-specific result extraction
- Provides contextually relevant responses

# 3. Key Code Improvements

### 3.1 Intent Separation and Storage

```
python
# NEW: For non-history queries, try to detect multiple intents using Nova
    # Try to use Nova's intent separation
    intents = await nova.identify_multiple_intents_async(prompt)
    if len(intents) > 1:
        print(f"DEBUG: Processing {len(intents)} separate intents for storage")
        # Process each intent separately
        for intent_info in intents:
            intent_text = intent_info["intent"]
            intent_lower = intent_text.lower()
            # Determine intent type
            if any(term in intent_lower for term in ['+', '-', '*', '/', 'calculate', 'math']):
                # Math intent
                history["math_questions"].append({
                    "session_id": session_count,
                    "prompt": intent_text,
                    "result": extract_math_result(result)
                })
                print(f"DEBUG: Stored math intent: {intent_text}")
```

# 3.2 Entity Extraction

```
python
```

```
def extract_country(text):
    """Extract country name from text."""
   # List of countries to check
    countries = [
        "Spain", "UK", "France", "Germany", "Italy",
        "Portugal", "Belgium", "Netherlands", "Greece",
        "Croatia", "Sweden", "Norway", "Denmark",
        "Finland", "Ireland", "Switzerland", "Austria"
    ]
   text_lower = text.lower()
    # Check for direct country mentions
    for country in countries:
        if country.lower() in text_lower:
            return country
    # Look for "for X" pattern
    import re
    match = re.search(r"for\s+([A-Za-z]+)(?:\.|\s|$)", text_lower)
    if match:
        country_candidate = match.group(1).capitalize()
        if country_candidate in countries:
            return country_candidate
    return None
```

### **3.3 Improved History Query Detection**

```
python
```

```
def answer_general_question(kb: KnowledgeBase, prompt: str, input2="-"):
    # Check if the question is about previous session content
    prompt_lower = prompt.lower()
    # Enhanced keywords for context detection
    past_indicators = ["did", "was", "asked", "last", "previous", "earlier", "before", "history
    contains past = any(word in prompt lower for word in past indicators)
    # Check if it's asking about math specifically
    math_keywords = ["math", "calculation", "+", "-", "*", "/", "plus", "minus", "multiply", "c
    about_math = any(kw in prompt_lower for kw in math_keywords)
    # If it's a history query about math
    if contains_past and about_math:
        # Get math questions history
        math_history = history.get("math_questions", [])
        if not math_history:
            return "I don't have any record of math questions."
        # Get the most recent math question
        latest_math = math_history[-1]
        response = (
            f"In a previous session, you asked the math question: '{latest_math['prompt']}'\n\r
            f"The result was: {latest_math['result']}"
        )
        return response
```

# 4. System Component Mapping

Component	File	Function	Purpose
Intent Separation	main.py	nova.identify_multiple_intents_async()	Splits multi-intent prompts
Intent Categorization	main.py	interactive_async_main()	Categorizes by intent type
Entity Extraction	main.py	extract_country()	Extracts entities from intents
Result Extraction	main.py	extract_math_result()	Extracts specific results from responses
History Query	general_knowledge.py	answer_general_question()	Identifies and processes
Detection			history queries
Report Generation	functions_registery.py	write_report()	Generates reports based on models
<b>▶</b>			

### 5. Test Results

## **Test Case 1: Multi-Intent Processing**

- **Input**: "Calculate 10+20, what is the capital of Germany, build an electricity energy model for solar for France and write a report"
- **Result**: Successfully separated and processed all 4 intents
- Storage: Each intent stored in appropriate category

DEBUG: Processing 4 separate intents for storage

DEBUG: Stored math intent: Calculate 10+20

DEBUG: Stored general intent: Determine the capital of Germany

DEBUG: Stored energy model intent: Build an electricity energy model for solar for France

(Country: France)

DEBUG: Stored general intent: Write a report

### **Test Case 2: History Query Precision**

- **Input**: "what was the last math question I asked?"
- **Result**: Correctly returned only the math portion
- **Response**: "In a previous session, you asked the math question: 'Calculate 10+20'"

DETECTED: Contains past tense indicators

DETECTED: History query about math

HISTORY CHECK: is\_history\_query=True, query\_type=math

Processing as a history query about: math

Found 3 math question entries
Returning math history response

### 6. Knowledge Base Structure

## 7. Summary and Benefits

## 1. Enhanced User Experience

- More precise recall of past interactions
- Better contextual awareness
- Improved response quality

### 2. Technical Improvements

- Proper separation of multi-intent prompts
- Structured knowledge base organization
- Entity extraction for contextual awareness

### 3. Future Capabilities

- Foundation for more complex entity extraction
- Improved cross-session memory
- Better support for conversational context

# 8. Simplified Overview

The improved knowledge base system:

1. **Separates** multiple requests in a single prompt

- 2. Categorizes each request appropriately
- 3. **Extracts** important entities (like countries)
- 4. **Stores** structured information for retrieval
- 5. **Recalls** precise information when asked about history

This creates a more natural, human-like interaction where the system remembers exactly what was discussed, not just that a conversation happened.