

Comparing Agentic AI Frameworks for Java

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April 2026

Who We Are



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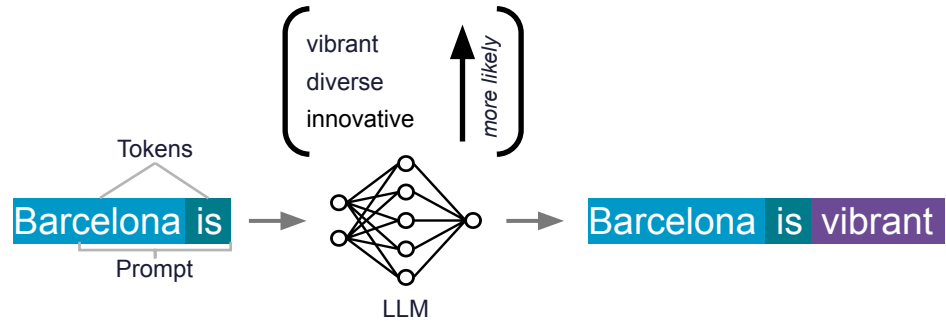
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Glossary

A **Prompt** is the inputs given to a model to guide its response and behavior

Tokens are the text units a model processes, often subwords chosen to balance efficiency and meaning. They are the currency of AI.



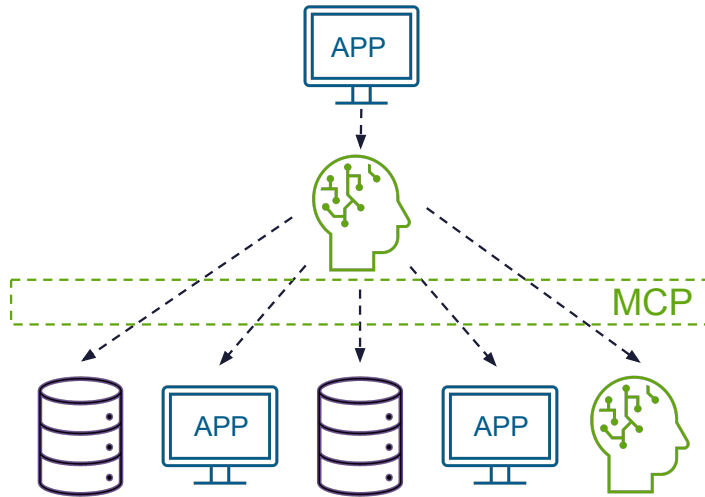
The **context length** is the maximum amount of information (tokens) a model can handle in one interaction

Retrieval-Augmented Generation (RAG) uses similarity search on a vector store to find the most relevant information and adds it to the context before the prompt is sent to the model, resulting in more accurate and up-to-date responses

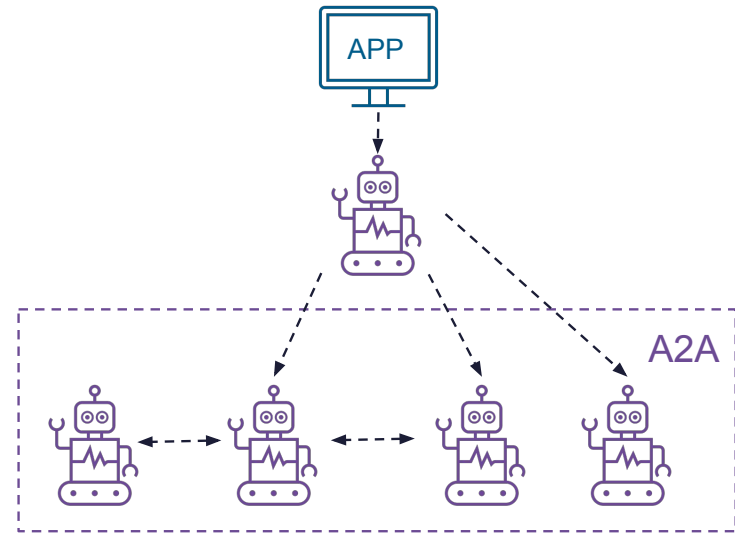
Tools are extra capabilities a model can invoke through **Tool Calling**, like searching the web, running code, or calling external services

Glossary

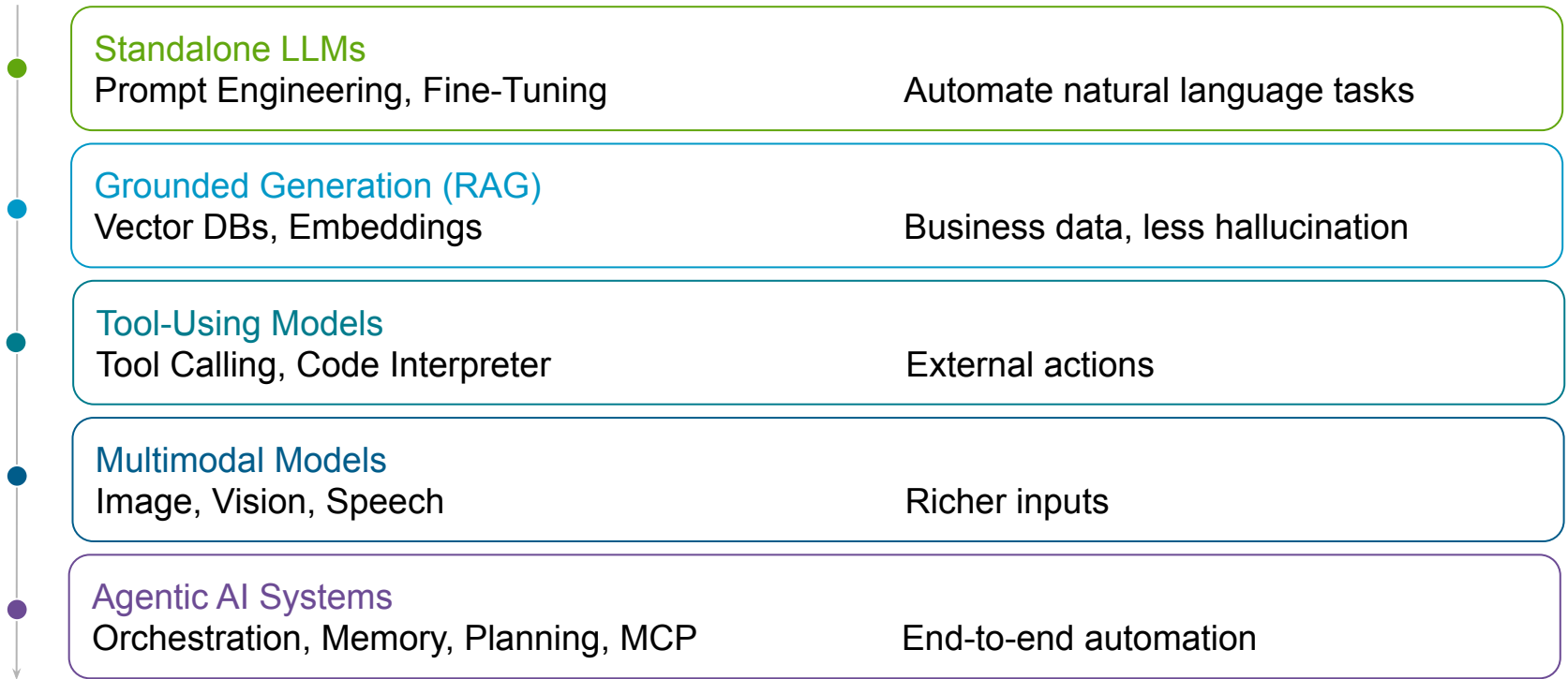
The **Model Context Protocol (MCP)** provides a standardized way to connect AI models to different data sources and tools



Agent-to-Agent (A2A) is a protocol that allows multiple AI agents to communicate, share work, and collaborate to complete complex tasks.



The Evolution of LLM Based Solutions



Agentic AI

*Agentic AI systems are designed to **pursue complex goals** with a high level of autonomy and predictability.*

*They are **productivity enablers** who can effectively incorporate humans in the loop via the use of multi-modality.*

Key Characteristics

Autonomy: takes goal-directed actions with minimal human oversight

Reasoning: contextual decision-making, judgment calls & tradeoffs

Adaptable Planning: dynamically adjusts plans based on changing conditions to complete processes efficiently

Context Understanding: comprehends and follows natural language and other modalities

Action Enabled: empowered to take action via access to web services delivering skills

Two Approaches to Orchestrate LLMs and Tools

Workflows

LLMs and tools are orchestrated through predefined code paths

- ✓ Predictable and consistent output
- ✓ Easier to debug and test
- ✓ Better for well-defined, repeatable tasks
- ✓ Lower cost and latency
- ✗ Inflexible — can't adapt to unforeseen subtasks
- ✗ Requires upfront task decomposition

Use workflows for predictability.

Use agents only when flexibility and model-driven decision-making are genuinely needed.

Pure Agents

The LLM dynamically directs its own processes and tool usage

- ✓ Flexible and adaptive for open-ended problems
- ✓ Handles tasks where steps can't be predicted in advance
- ✓ Scales well in trusted environments
- ✗ Higher cost and latency
- ✗ Risk of compounding errors
- ✗ Harder to debug and control
- ✗ Requires extensive testing and guardrails

Workflow Patterns

for Agentic Systems

Prompt Chaining

Decomposes tasks into sequential steps. Each LLM call processes the output of the previous one, with optional programmatic gates.

Parallelization

Runs LLM tasks simultaneously. Two variants: Sectioning (parallel subtasks) and Voting (multiple attempts for confidence).

Routing

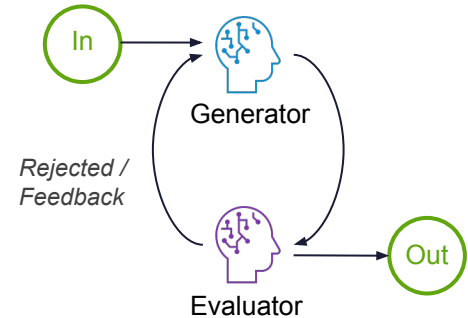
Classifies an input and directs it to a specialized follow-up task. Enables separation of concerns and optimized prompts per category.

Orchestrator-Workers

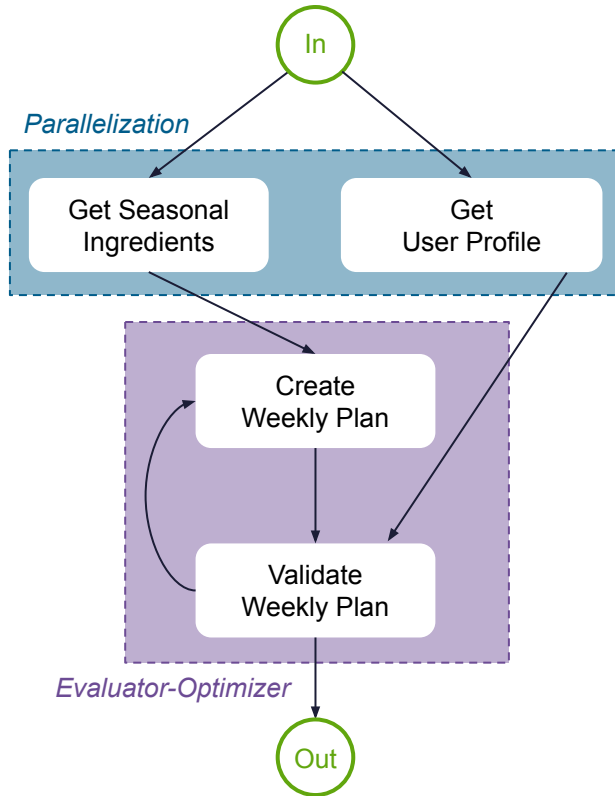
A central LLM dynamically breaks down tasks, delegates to worker LLMs, and synthesizes results. Subtasks aren't pre-defined.

Evaluator-Optimizer

One LLM generates a response while another evaluates and gives feedback in a loop. Continues until quality criteria are met.



Sample Application



Days & Meals

Monday	Tuesday	Wednesday	Thursday	Friday	Saturday	Sunday
<input checked="" type="checkbox"/> Breakfast <input checked="" type="checkbox"/> Lunch <input checked="" type="checkbox"/> Dinner	<input checked="" type="checkbox"/> Breakfast <input type="checkbox"/> Lunch <input type="checkbox"/> Dinner	<input checked="" type="checkbox"/> Breakfast <input checked="" type="checkbox"/> Lunch <input checked="" type="checkbox"/> Dinner	<input checked="" type="checkbox"/> Breakfast <input type="checkbox"/> Lunch <input type="checkbox"/> Dinner	<input type="checkbox"/> Breakfast <input checked="" type="checkbox"/> Lunch <input type="checkbox"/> Dinner	<input checked="" type="checkbox"/> Breakfast <input type="checkbox"/> Lunch <input type="checkbox"/> Dinner	<input checked="" type="checkbox"/> Breakfast <input checked="" type="checkbox"/> Lunch <input checked="" type="checkbox"/> Dinner

Country (ISO code)

Additional instructions

Your Weekly Plan

MONDAY

BREAKFAST

Wild Garlic Spinach Omelette 10 min

Ingredients

- 3 pieces Eggs
- 60 g Spinach
- 1/2 bunch Wild garlic
- 1 tbsp Chives
- 1 tsp Olive oil
- to taste Salt
- to taste Pepper

Instructions

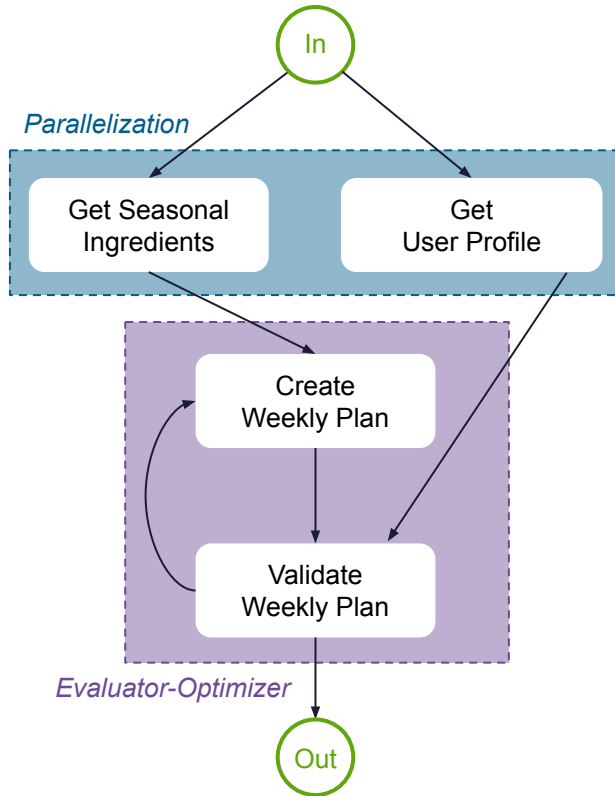
Beat eggs with salt and pepper. Sauté wild garlic in oil, add spinach until just wilted. Pour in eggs and cook gently, folding until set. Finish with chopped chives.

NUTRITION

Calories	270 kcal
Protein	18,0 g
Carbs	4,0 g
Fat	19,0 g
Sodium	280 mg

LUNCH

Sample Application



Days & Meals

Monday	Tuesday	Wednesday	Thursday	Friday	Saturday	Sunday
<input checked="" type="checkbox"/> Breakfast <input checked="" type="checkbox"/> Lunch <input checked="" type="checkbox"/> Dinner	<input checked="" type="checkbox"/> Breakfast <input type="checkbox"/> Lunch <input type="checkbox"/> Dinner	<input checked="" type="checkbox"/> Breakfast <input checked="" type="checkbox"/> Lunch <input checked="" type="checkbox"/> Dinner	<input checked="" type="checkbox"/> Breakfast <input type="checkbox"/> Lunch <input type="checkbox"/> Dinner	<input type="checkbox"/> Breakfast <input checked="" type="checkbox"/> Lunch <input type="checkbox"/> Dinner	<input checked="" type="checkbox"/> Breakfast <input type="checkbox"/> Lunch <input type="checkbox"/> Dinner	<input checked="" type="checkbox"/> Breakfast <input checked="" type="checkbox"/> Lunch <input checked="" type="checkbox"/> Dinner

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LUNCH

Meet the Frameworks



Official Spring project since
end of 2023

Aligns with Spring ecosystem
design principles

Supports Spring Boot \geq 3.4

Enterprise support available



LangChain4j

Initiated in early 2023 due to a
lack of Java counterparts to
Python libraries like LangChain,
LlamaIndex

Framework-agnostic with
integrations to Micronaut,
Quarkus, Spring Boot, ...



Embabel

Announced in 2025 to bring
Agentic AI to enterprise JVM
applications

Founded by Spring creator
Rod Johnson

Built on Spring AI

Spring AI

Advisors API

Advisors are a key concept to implement Agentic AI capabilities

They allow you to intercept, modify, and enhance AI interactions

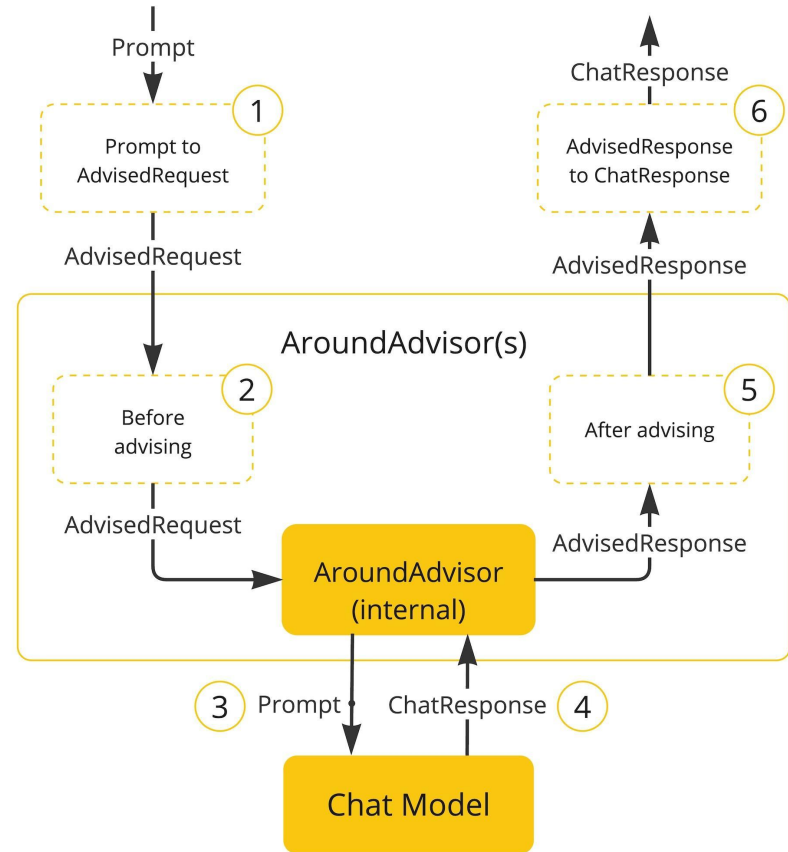
Examples: QuestionAnswerAdvisor, MessageChatMemoryAdvisor, SafeGuardAdvisor

A **Recursive Advisor** is a special type of Advisor

Allows to loop through downstream advisor chain multiple times

Enables controlled retries, validation, and multi-step tool execution

Examples: StructuredOutputValidationAdvisor, ToolCallAdvisor





Code Sample



LangChain4j

langchain4j-agentic module

The agentic module provides abstractions and utilities to define agentic workflows, manage tool usage, and maintain context across interactions with different LLM

Agents can be defined like AI services with the `@Agent` annotation. `outputKey` names the shared variable in the `AgenticScope` that stores the agent result

```
public interface NutritionPlanner {
    @UserMessage("...")
    @Agent(outputKey = "mealplan", description = "...")
    MealPlan createMealPlan(@V("ingredients") Ingredients ingred);
}
```

LangChain4j

langchain4j-agentic module

Instance and workflows can be build using the `AgenticServices.agentBuilder()` method

```
NutritionPlanner nutritionPlanner = AgenticServices
    .agentBuilder(NutritionPlanner.class)
    .chatModel(chatModel)
    .build();
```

```
NutritionPlanner nutritionPlanner = AgenticServices
    .sequenceBuilder(NutritionPlanner.class) // .loopBuilder .parallelBuilder .conditionalBuilder ...
    .subAgents(seasonalIngredientAgent, weeklyPlanCreator, validationLoop)
    .build();
```



LangChain4j

Code Sample



```
String representation,
    - String that represents location, as 2 double values split with coma. A
    framework.data.sokr.core.geo.Point instance

    Point parseLocation(String locationString) {
        Preconditions.checkNotNull(locationString, "Location String should not be null");
        Preconditions.checkNotNull(locationString, "Location must be split");
        locationString = locationString.trim();

        if (locationString.contains(", ")) {
            locationString = locationString.replaceAll(regex: ", ", replacement: ",");
        }

        if (locationString.contains(" ")) {
            locationString = locationString.replaceAll(regex: " ", replacement: ",");
        }

        String[] location = locationString.split(regex: ",");
        Preconditions.checkArgument(location.length >= 2, "Location should const");
        double lat = Double.parseDouble(location[0]);
        double lon = Double.parseDouble(location[1]);

        return new Point(lat, lon);
    }
}
```

Embabel

Agent Framework

Embabel goes beyond sequential execution with an AI planning step (A* Search)

Plans are dynamic, formulated at runtime, revised after every action (OODA* loop)

Embabel models agentic flows through **Actions**, **Goals**, **Conditions**, and a **Domain Model** and provides Spring-style annotations for it

```
@Agent(description = "...")
class NutritionPlanner {

    @Action(pre = {"spel:..."})
    Ingredients fetchSeasonalIngredients() {
        ...
    }

    @AchievesGoal(description = "Create meal plan")
    @Action
    MealPlan createMealPlan(Ingredients ingred) {
        ...
    }
}
```



Embabel

Code Sample

```
String representation,  
    - String that represents location, as 2 double values split with coma. A  
    framework.data.sotr.core.geo.Point instance  
  
    Point parseLocation(String locationString) {  
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        if (locationString.contains(", ")) {  
            locationString = locationString.replaceAll(regex: ", ", replacement: ",");  
        }  
        if (locationString.contains(" ")) {  
            locationString = locationString.replaceAll(regex: " ", replacement: ",");  
        }  
        String[] location = locationString.split(regex: ",");  
        Preconditions.checkArgument(location.length >= 2, "Location should const");  
        double lat = Double.parseDouble(location[0]);  
        double lon = Double.parseDouble(location[1]);  
        return new Point(lat, lon);  
    }  
}
```

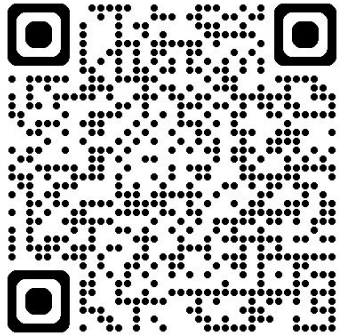


Comparing Agentic AI Frameworks' Capabilities for Spring Devs

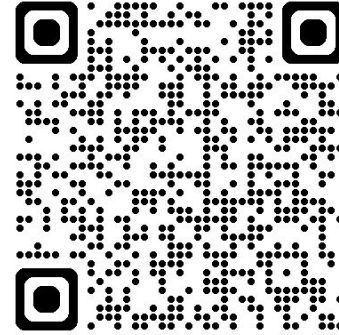
Aspect	Spring AI	LangChain4j	Embabel
Documentation	Modest effort to implement, or already available via a community project	Modest effort to implement, or already available via a community project	Good
Deterministic Workflow	Modest effort to implement, or already available via a community project	Good	Good
Pure Agents	Good	Good	Good
Subagents	Modest effort to implement, or already available via a community project	Good	Good
Advanced tool selection	Modest effort to implement, or already available via a community project	Modest effort to implement, or already available via a community project	Good
Human-in-the-loop	Modest effort to implement, or already available via a community project	Good	Good
AI skills	Modest effort to implement, or already available via a community project	Good	Good
Observability	Modest effort to implement, or already available via a community project	Good	Good
Testing	Modest effort to implement, or already available via a community project	Modest effort to implement, or already available via a community project	Good
MCP	Good	No HTTP Server	Good
A2A	Good	Good	Good

 *Modest effort to implement, or already available via a community project*

Related Previous Talks



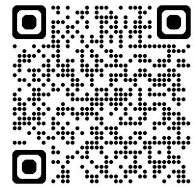
Spring AI Intro
[Slides](#) [Demo](#) [Video](#)



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Thank You



<https://github.com/SandraAhlgrimm/ai-nutrition-planner>