

PART 2: DATA STRUCTURES & TECHNICAL IMPLEMENTATION

COMPLETE DATA STRUCTURE

Definitions (continued)

javascript

```
const definitions = [
    // ... (previous definitions)
    {
        term: "Vector Database",
        definition: "A database optimized for storing and retrieving vector embeddings for semantic search (e.g., Pinecone, Weaviate).",
        category: "tech-stack"
    },
    {
        term: "Embedding Model",
        definition: "Converts text/images into numerical vectors (embeddings) for semantic similarity (e.g., OpenAI, Voyage AI, Cohere).",
        category: "tech-stack"
    },
    {
        term: "Tool Use / Function Calling",
        definition: "Agent's ability to call external APIs, databases, calculators, or custom functions to perform actions.",
        category: "fundamentals"
    },
    {
        term: "MCP (Model Context Protocol)",
        definition: "Standardized protocol for connecting AI models to external tools and data sources with minimal code.",
        category: "tech-stack"
    },
    {
        term: "ReAct Pattern",
        definition: "Reasoning and Acting pattern where agents alternate between thinking (reason) and doing (action).",
        category: "patterns"
    },
    {
        term: "Chain-of-Thought Prompting",
        definition: "Technique that encourages LLMs to break down complex problems into step-by-step reasoning.",
        category: "patterns"
    },
    {
        term: "Self-Reflection",
        definition: "Pattern where agents evaluate their own outputs, identify errors, and iteratively improve.",
        category: "patterns"
    },
    {
        term: "Multi-Agent System",
        definition: "Multiple specialized agents working together, each with defined roles, to solve complex problems.",
        category: "patterns"
    },
    {

```

```
term: "Agent Orchestration",
definition: "The system that connects models, memory, tools, and steps in order (e.g., LangChain, CrewAI, AutoGen).",
category: "frameworks"
},
{
term: "Memory (Short-term)",
definition: "Current session or task history stored temporarily (e.g., episodic memory, working memory).",
category: "fundamentals"
},
{
term: "Memory (Long-term)",
definition: "Persistent user preferences, past decisions, and knowledge stored across sessions (e.g., Vector DB, SQL, file storage).",
category: "fundamentals"
},
{
term: "Guardrails",
definition: "Safety mechanisms that ensure agents behave responsibly, avoiding harmful or biased outputs.",
category: "best-practices"
},
{
term: "Multimodal AI",
definition: "AI systems that can process multiple types of data: text, images, audio, video (e.g., GPT-4V, Gemini).",
category: "advanced"
},
{
term: "HyDE (Hypothetical Document Embeddings)",
definition: "RAG technique where LLM generates hypothetical documents for better semantic search.",
category: "rag"
},
{
term: "Agentic Design Pattern",
definition: "Architectural approaches for building agents: ReAct, Planning, Multi-Agent, Self-Reflection, etc.",
category: "patterns"
},
{
term: "System Prompt",
definition: "Instructions that define agent's role, behavior, personality, and operational procedures.",
category: "fundamentals"
},
{
term: "Context Window",
definition: "Maximum amount of text (in tokens) an LLM can process at once (e.g., 128k tokens for GPT-4).",
category: "fundamentals"
},
```

```
{  
  term: "Token",  
  definition: "Basic unit of text processing in LLMs, roughly equivalent to 4 characters or 0.75 words.",  
  category: "fundamentals"  
},  
{  
  term: "Fine-tuning",  
  definition: "Training a pre-trained model on domain-specific data to improve performance for specialized tasks.",  
  category: "advanced"  
},  
{  
  term: "Prompt Engineering",  
  definition: "Craft of designing effective prompts to guide AI behavior and improve output quality.",  
  category: "best-practices"  
}  
];
```

Design Patterns (6 main patterns)

javascript

```
const designPatterns = [
  {
    id: "react",
    name: "ReAct Agent",
    category: "Single-Agent",
    description: "A reasoning and acting framework where an agent alternates between reasoning (using LLMs) and acting (using tools). It's designed for general-purpose tasks requiring flexible decision-making and tool use.",
    usedBy: "Most AI Agent Products",
    whenToUse: "General purpose tasks requiring flexible decision-making and tool use",
    diagram: "react-diagram.png",
    workflow: ["Query", "LLM (Reason)", "Tools (Action)", "Output"],
    pros: ["Simple to implement", "Flexible", "Works with most LLMs"],
    cons: ["Can be slow", "May loop unnecessarily"],
    codeExamples: {
      python: `from langchain.agents import create_react_agent
from langchain_openai import ChatOpenAI
from langchain.tools import WikipediaQueryRun

llm = ChatOpenAI(model="gpt-4o")
tools = [WikipediaQueryRun()]

agent = create_react_agent(
  llm=llm,
  tools=tools,
  prompt="You are a helpful research assistant"
)

result = agent.invoke({"input": "What is quantum computing?"}),
      javascript: `import { ReActAgent } from "@langchain/agents";
import { ChatOpenAI } from "@langchain/openai";

const model = new ChatOpenAI({ model: "gpt-4o" });
const agent = new ReActAgent({ model, tools: [] });

const result = await agent.invoke({
  input: "What is quantum computing?"
});`,
      go: `// Go example using Go-LangChain
package main

import (
  "github.com/tmc/langchaingo/agents"
  "github.com/tmc/langchaingo,llms/openai"
)`
```

```

func main() {
    llm := openai.New()
    agent := agents.NewReactAgent(llm, tools)
    result := agent.Run("What is quantum computing?")
}

java:// Java example using LangChain4j

import dev.langchain4j.agent.Agent;
import dev.langchain4j.model.openai.OpenAiChatModel;

OpenAiChatModel model = OpenAiChatModel.builder()
    .apiKey(System.getenv("OPENAI_API_KEY"))
    .modelName("gpt-4o")
    .build();

Agent agent = Agent.builder()
    .chatLanguageModel(model)
    .tools(tools)
    .build();

String result = agent.chat("What is quantum computing?");
},
links: [
    { title: "ReAct Paper", url: "https://arxiv.org/abs/2210.03629" },
    { title: "LangChain ReAct Docs", url: "https://python.langchain.com/docs/modules/agents/agent_types/react" },
    { title: "Tutorial Video", url: "https://youtube.com/watch?v=..." }
]
},
{
    id: "codeact",
    name: "CodeAct Agent",
    category: "Single-Agent",
    description: "The CodeAct architecture by Manus AI allows agents to autonomously execute Python code instead of using LLMs to generate text-based responses.",
    usedBy: "Manus AI",
    whenToUse: "Complex data processing, mathematical computations, file operations",
    diagram: "codeact-diagram.png",
    workflow: ["User", "Agent (Think)", "CodeAct", "Environment", "Observation", "Result"],
    pros: ["Can handle complex logic", "More precise than text", "Native code execution"],
    cons: ["Security risks", "Requires sandboxing", "Limited LLM code quality"],
    codeExamples: {
        python: `# CodeAct pattern with code execution
from langchain_experimental.utilities import PythonREPL

python_repl = PythonREPL()`
```

```

def code_agent(query):
    # Agent decides to write code
    code = llm.generate_code(query)

    # Execute code in sandbox
    result = python_repl.run(code)

    return result

# Example: "Calculate factorial of 10"
result = code_agent("Calculate factorial of 10"),
    javascript: `// JavaScript with VM2 sandbox
const { VM } = require('vm2');

function codeAgent(query) {
    const code = llm.generateCode(query);

    const vm = new VM({ timeout: 1000 });
    const result = vm.run(code);

    return result;
}
```,
 links: [
 { title: "CodeAct Paper", url: "https://arxiv.org/abs/2402.01030" },
 { title: "Manus AI", url: "https://manus.im" }
]
},
{
 id: "self-reflection",
 name: "Self-Reflection",
 category: "Self-Improving",
 description: "Self-Reflection or Reflexion involves self-evaluating outputs, using feedback loops to identify errors, and iteratively refining them. It's often used in AI development to improve model performance and handle complex tasks like natural language processing and decision-making.",
 usedBy: "Open Serve AI",
 whenToUse: "High-quality outputs needed, writing tasks, complex problem solving",
 workflow: ["Query", "LLM", "Memory/Tools", "First draft", "Critique", "Generator", "Result"],
 codeExamples: {
 python: `from langchain.prompts import PromptTemplate
def self_reflection_agent(query):
 # Generate initial response
 draft = llm.invoke(query)
```

```

```

# Self-critique
critique_prompt = f"Critique this response: {draft}"
critique = llm.invoke(critique_prompt)

# Improve based on critique
improve_prompt = f"Improve this based on critique:\\n{draft}\\n\\nCritique: {critique}"
final = llm.invoke(improve_prompt)

return final
}

},
{
id: "multi-agent",
name: "Multi-Agent Workflow",
category: "Collaborative",
description: "A Multi-Agent Workflow is a collaborative system where multiple specialized agents work together to build a usedBy: "Gemini Deep Research",
whenToUse: "Complex tasks requiring specialized expertise, parallel processing",
workflow: ["Query", "Agent 1 (Manager)", "Agent 2/3/n (Specialists)", "Aggregator LLM", "Output"],
codeExamples: {
python: `from crewai import Agent, Task, Crew

# Define specialized agents
researcher = Agent(
    role='Researcher',
    goal='Gather comprehensive information',
    tools=[search_tool]
)

analyst = Agent(
    role='Data Analyst',
    goal='Analyze and synthesize findings',
    tools=[analysis_tool]
)

writer = Agent(
    role='Writer',
    goal='Create final report',
    tools=[]
)

# Coordinate agents
crew = Crew(
    agents=[researcher, analyst, writer],
```

```

```
 tasks=[research_task, analysis_task, writing_task]
)

result = crew.kickoff()
}

},
{
 id: "agentic-rag",
 name: "Agentic RAG",
 category: "Advanced RAG",
 description: "Agentic RAG involves AI agents retrieving and evaluating relevant data to generate context-aware and well-reasoned responses.",
 usedBy: "Perplexity",
 whenToUse: "Question answering with citations, research tasks, fact-checking",
 workflow: ["Query", "Planning Agent", "Retrieval Agent", "Generator Agent", "Response Judge"],
 codeExamples: {
 python: `from langchain.agents import create_agent
from langchain.tools.retriever import create_retriever_tool

Create retrieval tool
retriever_tool = create_retriever_tool(
 retriever=vector_store.as_retriever(),
 name="knowledge_base",
 description="Search the knowledge base"
)

Planning agent decides queries
planner = create_agent(llm, [retriever_tool])

Execute agentic RAG
result = planner.invoke({
 "input": "Compare quantum computing approaches"
})`}
 },
 {
 id: "planning",
 name: "Planning Pattern",
 category: "Goal-Oriented",
 description: "The AI breaks a high-level goal into smaller steps, executes them with the help of agents, and revises the plan based on feedback.",
 whenToUse: "Multi-step projects, task decomposition, complex workflows",
 workflow: ["High-Level Goal", "Planning", "Generate Task", "Single Task Agent", "Adjusted Task", "Replan"],
 codeExamples: {
 python: `def planning_agent(goal):
 # Break down goal`}
 }
}
```

```
plan = llm.invoke(f"Break this into steps: {goal}")
```

```
tasks = parse_plan(plan)
```

```
results = []
```

```
for task in tasks:
```

```
 result = execute_task(task)
```

```
 if not result.success:
```

```
 # Replan
```

```
 tasks = replan(task, result.error)
```

```
 else:
```

```
 results.append(result)
```

```
return aggregate_results(results)
```

```
}
```

```
}
```

```
];
```

## Frameworks Data (Top 10)

```
javascript
```

```
const frameworks = [
 {
 id: "langchain",
 name: "LangChain",
 logo: "langchain-logo.png",
 category: "Modular",
 tagline: "Build custom LLM agents using reusable components",
 description: "Modular framework to build complex, multi-turn conversational agents using LLMs and custom tools. Indust",
 features: ["Agent executors", "Memory modules", "Tool integration", "Built-in templates"],
 bestFor: "Building complex, multi-turn conversational agents using LLMs and custom tools",
 pricing: "Free (Open Source)",
 languages: ["Python", "JavaScript"],
 stars: "90k+",
 difficulty: "Intermediate",
 officialLinks: {
 docs: "https://python.langchain.com",
 github: "https://github.com/langchain-ai/langchain",
 tutorials: "https://youtube.com/c/langchain"
 },
 codeExample: "// See design patterns section",
 pros: ["Extensive ecosystem", "Great documentation", "Large community"],
 cons: ["Can be complex", "Steep learning curve for beginners"],
 useCases: ["Chatbots", "RAG systems", "Document QA", "Agent workflows"]
 },
 // ... (add other 9 frameworks similarly)
];
```

## Tech Stack Data (by layer)

javascript

```
const techStack = {
 deployment: [
 { name: "Groq", category: "Infrastructure", description: "Ultra-fast LLM inference", url: "https://groq.com", pricing: "Pay-as-you-go" },
 { name: "AWS", category: "Cloud", description: "Amazon cloud services", url: "https://aws.amazon.com", pricing: "Pay-as-you-go" },
 { name: "Together.ai", category: "Inference", description: "Hosted model inference", url: "https://together.ai", pricing: "Pay-as-you-go" },
 // ... more tools
],
 evaluation: [
 { name: "LangSmith", category: "Monitoring", description: "LangChain observability", url: "https://smith.langchain.com", pricing: "Free" },
 { name: "Weights & Biases", category: "MLOps", description: "Experiment tracking", url: "https://wandb.ai", pricing: "Free" },
 // ... more tools
],
 // ... other layers
};
```

## TECHNICAL IMPLEMENTATION REQUIREMENTS

### HTML Structure

```
html
```

```
<!DOCTYPE html>
<html lang="en">
<head>
 <meta charset="UTF-8">
 <meta name="viewport" content="width=device-width, initial-scale=1.0">
 <title>Building Agentic Systems Guide</title>
 <style>
 /* Color Variables */
 :root {
 --primary-purple: #7C3AED;
 --accent-blue: #3B82F6;
 --teal: #06B6D4;
 --orange: #F97316;
 --dark-bg: #1E293B;
 --card-bg: #334155;
 --text-primary: #F1F5F9;
 --text-secondary: #94A3B8;
 --success: #10B981;
 --warning: #FBBF24;
 }

 *
 {
 margin: 0;
 padding: 0;
 box-sizing: border-box;
 }

 body {
 font-family: 'Inter', -apple-system, BlinkMacSystemFont, sans-serif;
 background: var(--dark-bg);
 color: var(--text-primary);
 line-height: 1.6;
 }

 /* Tab Navigation */
 .tabs {
 display: flex;
 gap: 1rem;
 padding: 1rem 2rem;
 background: linear-gradient(135deg, var(--primary-purple), var(--accent-blue));
 overflow-x: auto;
 position: sticky;
 top: 0;
 }
```

```
z-index: 100;
}

.tab {
 padding: 0.75rem 1.5rem;
 border: none;
 background: rgba(255, 255, 255, 0.1);
 color: white;
 cursor: pointer;
 border-radius: 8px;
 transition: all 0.3s;
 white-space: nowrap;
}

.tab:hover {
 background: rgba(255, 255, 255, 0.2);
 transform: translateY(-2px);
}

.tab.active {
 background: white;
 color: var(--primary-purple);
}

/* Main Content Area */
.content {
 max-width: 1400px;
 margin: 0 auto;
 padding: 2rem;
}

.tab-content {
 display: none;
}

.tab-content.active {
 display: block;
 animation: fadeIn 0.3s;
}

@keyframes fadeIn {
 from { opacity: 0; transform: translateY(10px); }
 to { opacity: 1; transform: translateY(0); }
}
```

```
/* Card Styles */
.card {
 background: var(--card-bg);
 border-radius: 12px;
 padding: 2rem;
 margin-bottom: 2rem;
 border: 1px solid rgba(255, 255, 255, 0.1);
 transition: all 0.3s;
}

.card:hover {
 transform: translateY(-5px);
 box-shadow: 0 10px 30px rgba(124, 58, 237, 0.3);
}

/* Code Block Styles */
.code-block {
 background: #0f172a;
 border-radius: 8px;
 padding: 1rem;
 margin: 1rem 0;
 position: relative;
}

.language-tabs {
 display: flex;
 gap: 0.5rem;
 margin-bottom: 1rem;
}

.language-tab {
 padding: 0.5rem 1rem;
 background: rgba(255, 255, 255, 0.05);
 border: none;
 color: var(--text-secondary);
 cursor: pointer;
 border-radius: 4px;
}

.language-tab.active {
 background: var(--primary-purple);
 color: white;
}
```

```
.copy-button {
 position: absolute;
 top: 1rem;
 right: 1rem;
 padding: 0.5rem 1rem;
 background: var(--accent-blue);
 border: none;
 color: white;
 border-radius: 4px;
 cursor: pointer;
}

/* Wizard Styles */

.wizard {
 max-width: 900px;
 margin: 0 auto;
}

.wizard-step {
 display: none;
}

.wizard-step.active {
 display: block;
}

.progress-bar {
 height: 4px;
 background: rgba(255, 255, 255, 0.1);
 border-radius: 2px;
 margin-bottom: 2rem;
}

.progress-fill {
 height: 100%;
 background: linear-gradient(90deg, var(--primary-purple), var(--teal));
 border-radius: 2px;
 transition: width 0.3s;
}

/* Search */

.search-box {
 width: 100%;
```

```
padding: 1rem;
background: var(--card-bg);
border: 1px solid rgba(255, 255, 255, 0.1);
border-radius: 8px;
color: white;
font-size: 1rem;
margin-bottom: 2rem;
}
```

*/\* Modal for Graphics \*/*

```
.modal {
 display: none;
 position: fixed;
 top: 0;
 left: 0;
 width: 100%;
 height: 100%;
 background: rgba(0, 0, 0, 0.9);
 z-index: 1000;
 justify-content: center;
 align-items: center;
}
```

```
.modal.active {
 display: flex;
}
```

```
.modal-content {
 max-width: 90%;
 max-height: 90%;
}
```

*/\* Comparison Matrix \*/*

```
.comparison-table {
 width: 100%;
 border-collapse: collapse;
 margin: 2rem 0;
}
```

```
.comparison-table th,
.comparison-table td {
 padding: 1rem;
 border: 1px solid rgba(255, 255, 255, 0.1);
 text-align: left;
```

```
}

.comparison-table th {
 background: var(--primary-purple);
 position: sticky;
 top: 0;
}

/* Responsive */

@media (max-width: 768px) {
 .tabs {
 padding: 1rem;
 }
}

.content {
 padding: 1rem;
}

.card {
 padding: 1rem;
}

}
}

</style>
</head>
<body>

<!-- Navigation Tabs -->
<div class="tabs">
 <button class="tab active" data-tab="home">Home</button>
 <button class="tab" data-tab="builder">★ Agent Builder</button>
 <button class="tab" data-tab="fundamentals">Fundamentals</button>
 <button class="tab" data-tab="patterns">Design Patterns</button>
 <button class="tab" data-tab="frameworks">Frameworks</button>
 <button class="tab" data-tab="advanced">Advanced Patterns</button>
 <button class="tab" data-tab="techstack">Tech Stack</button>
 <button class="tab" data-tab="best-practices">Best Practices</button>
 <button class="tab" data-tab="tools">Free Tools</button>
 <button class="tab" data-tab="graphics">Reference Graphics</button>
</div>

<!-- Main Content -->
<div class="content">
 <!-- HOME TAB -->
 <div id="home" class="tab-content active">
 <!-- Home content here -->
 </div>
</div></pre>
```

```

</div>

<!-- AGENT BUILDER TAB -->
<div id="builder" class="tab-content">
 <div class="wizard">
 <div class="progress-bar">
 <div class="progress-fill" style="width: 0%"></div>
 </div>

 <div class="wizard-step active" data-step="1">
 <!-- Step 1 content -->
 </div>

 <!-- More steps... -->
 </div>
</div>

<!-- Other tabs... -->
</div>

<!-- Graphics Modal -->
<div class="modal" id="graphicModal">
 <button class="close-btn">x</button>
 <button class="nav-btn prev"></button>

 <button class="nav-btn next"></button>
</div>

<script>
 // Tab switching
 const tabs = document.querySelectorAll('.tab');
 const tabContents = document.querySelectorAll('.tab-content');

 tabs.forEach(tab => {
 tab.addEventListener('click', () => {
 const targetId = tab.dataset.tab;

 tabs.forEach(t => t.classList.remove('active'));
 tab.classList.add('active');

 tabContents.forEach(content => {
 content.classList.remove('active');
 if (content.id === targetId) {
 content.classList.add('active');
 }
 });
 });
 });
</script>

```

```
 }

 });

 // Save progress
 localStorage.setItem('lastTab', targetId);
}

});

// Load graphics from CSV
async function loadGraphics() {
 // Read graphics.csv
 // Display as thumbnail grid
 // Add click handlers for modal
}

// Copy code functionality
function setupCopyButtons() {
 document.querySelectorAll('.copy-button').forEach(btn => {
 btn.addEventListener('click', async (e) => {
 const codeBlock = e.target.closest('.code-block');
 const code = codeBlock.querySelector('code').textContent;
 await navigator.clipboard.writeText(code);
 btn.textContent = 'Copied!';
 setTimeout(() => btn.textContent = 'Copy', 2000);
 });
 });
}

// Initialize app
document.addEventListener('DOMContentLoaded', () => {
 loadGraphics();
 setupCopyButtons();

 // Load last viewed tab
 const lastTab = localStorage.getItem('lastTab');
 if (lastTab) {
 document.querySelector(`[data-tab="${lastTab}"]`).click();
 }
});

</script>
</body>
</html>
```

## READING GRAPHICS FROM CSV

The app must read `graphics.csv` which has this structure:

- `FileName`
- `FilePath` (relative path)
- `Title`

Use Papaparse to read the CSV:

```
javascript
```

```
async function loadGraphicsFromCSV() {
 const response = await fetch('graphics.csv');
 const csvText = await response.text();

 Papa.parse(csvText, {
 header: true,
 skipEmptyLines: true,
 complete: (results) => {
 const graphics = results.data;
 renderGraphicsGallery(graphics);
 }
 });
}

function renderGraphicsGallery(graphics) {
 const gallery = document.getElementById('graphics-gallery');

 graphics.forEach((graphic, index) => {
 const thumbnail = document.createElement('div');
 thumbnail.className = 'graphic-thumbnail';
 thumbnail.innerHTML = `

 <h4>${graphic.Title}</h4>
 `;

 thumbnail.addEventListener('click', () => {
 openGraphicModal(graphic, index, graphics);
 });
 gallery.appendChild(thumbnail);
 });
}

function openGraphicModal(graphic, index, allGraphics) {
 const modal = document.getElementById('graphicModal');
 const img = document.getElementById('modalImage');

 img.src = graphic.FilePath;
 img.alt = graphic.Title;
 modal.classList.add('active');

 // Navigation
 document.querySelector('.prev').onclick = () => {
```

```
const newIndex = (index - 1 + allGraphics.length) % allGraphics.length;
openGraphicModal(allGraphics[newIndex], newIndex, allGraphics);
};

document.querySelector('.next').onclick = () => {
 const newIndex = (index + 1) % allGraphics.length;
 openGraphicModal(allGraphics[newIndex], newIndex, allGraphics);
};
}
```

---

## PERSISTENCE WITH localStorage

Track user progress:

javascript

```
const AppState = {
 progress: {
 completedSections: [],
 bookmarks: [],
 journalEntries: [],
 lastTab: 'home'
 },
}

save() {
 localStorage.setItem('agenticSystemsProgress', JSON.stringify(this.progress));
},

load() {
 const saved = localStorage.getItem('agenticSystemsProgress');
 if (saved) {
 this.progress = JSON.parse(saved);
 }
},

markComplete(sectionId) {
 if (!this.progress.completedSections.includes(sectionId)) {
 this.progress.completedSections.push(sectionId);
 this.save();
 }
},

addBookmark(item) {
 this.progress.bookmarks.push(item);
 this.save();
},

addJournalEntry(entry) {
 this.progress.journalEntries.push({
 date: new Date().toISOString(),
 ...entry
 });
 this.save();
}
};
```

## FINAL CHECKLIST

- All 10 tabs implemented
  - Interactive Agent Builder Wizard with 14 steps
  - Code examples in Python, JavaScript, TypeScript, Go, Java
  - Language toggle for all code blocks
  - Framework comparison matrix
  - All 44 graphics displayed as clickable thumbnails
  - Modal with prev/next navigation
  - Read graphics from CSV file
  - Progress tracking with localStorage
  - Learning journal
  - Search functionality
  - Bookmarks/favorites
  - Copy code buttons
  - Resource links for all frameworks/tools
  - Responsive design
  - Dark mode with infographic color scheme
- 

END OF PROMPT PART 2