



# Day 3 Whitepaper Notes

## Context Engineering — Sessions & Memory

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### 1. Purpose & Core Idea

This paper focuses on how agents **retain, recall, and reuse knowledge** across interactions — the backbone of *continuity and adaptability* in agentic systems.

Agents don't become intelligent by remembering everything; they become intelligent by remembering *the right things at the right time*.

Context Engineering is the discipline of deciding **what information matters, when to forget, and how to recall efficiently**.

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### 2. The Anatomy of Context

An agent's context is a *living state* built from multiple layers — each with distinct purpose, scope, and lifetime.

Layer	Scope	Examples
<b>Session Context</b>	Active conversation state (unique ID, metadata, intent)	"User is configuring sales report filters."
<b>Short-Term Memory (STM)</b>	Recent steps within a session	Last 10 turns or API calls
<b>Long-Term Memory (LTM)</b>	Persistent knowledge across sessions	User preferences, past decisions, FAQs

Each layer is bounded by *cost, relevance, and privacy constraints*.






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### 3. Session Model and Lifecycle

Sessions act as the unit of continuity between the user and agent.

**Lifecycle:**

1.  *Create* — new session ID & metadata (user, goal, timestamp)
2.  *Active* — ongoing reason–act–observe loop
3.  *Suspended* — temporarily paused but retained in memory
4.  *Resumed* — context restored and continued
5.  *Archived* — stored for retrieval or long-term analytics

Sessions enable multi-turn reasoning with traceability, similar to how a thread in Slack or an HTTP session retains state.

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## 4. Memory Architectures Explained

### **a. Buffer Memory**

Stores recent N interactions (verbatim). Used for short, bounded contexts.

### **b. Summary Memory**

Compresses older turns into semantic summaries to prevent token overload while preserving intent.

### **c. Vector Memory**

Encodes past events into embeddings for semantic retrieval (using cosine similarity or ANN).

### **d. Hybrid Memory**

Combines summary + vector for optimal accuracy and cost control.

**Core principle:** *The agent should never re-read everything — it should recall what matters.*

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## 5. Context Management Loop

Each agent iteration follows this cycle:

1. **Retrieve Context** → collect session state + relevant memory entries.
2. **Generate Reasoning** → produce next action or tool call.
3. **Execute & Observe** → record outcomes.
4. **Update Memory** → append or summarize new facts.
5. **Trim Context** → remove stale or irrelevant details.

This loop keeps the agent's attention window focused and its reasoning grounded in past reality.

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## 6. Key Patterns in Session & Memory Design

Pattern	Purpose	Implementation Hint
<b>Session Token Chaining</b>	Maintain state across API calls	Use UUID per conversation thread
<b>Checkpointing</b>	Save intermediate reasoning steps	Periodic serialization
<b>Summarization Loop</b>	Prevent context overflow	Summarize every N turns
<b>Hierarchical Memory</b>	Speed retrieval	Split by topic or intent
<b>Memory Expiry Policies</b>	Data retention compliance	TTL for each memory type

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## 7. Engineering Challenges & Best Practices

- **Context Overflow** → **Trim Early**: Use summary and embedding indexes to control token growth.
- **State Isolation**: Separate user context from system context to avoid leaks across agents.
- **Memory Security**: Encrypt stored sessions & mask sensitive data (PII).

- **Cross-Session Learning:** Sync long-term memory for multi-agent collaboration.
  - **Observability:** Each session should be traceable via logs, spans, and telemetry IDs for debugging.
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## 8. Implementation in Google ADK (Notebooks Summary)

### Notebook 1: *Agent Sessions*

- Implemented a `SessionManager` class to create and track session objects.
- Added metadata (creation time, user ID, goal, status).
- Used ADK to resume/suspend sessions programmatically.
- Observed context continuity after pause/resume — agent picked up mid-conversation without losing intent.

### Notebook 2: *Agent Memory*

- Built a `MemoryProvider` class for short-term, long-term, and hybrid storage.
  - Integrated vector store retrieval for semantic recall.
  - Implemented context summarization pipelines using LLM compression.
  - Tested “reflective memory” — agent summarizes its own experience to learn patterns for next sessions.
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## 9. Privacy & Ethical Considerations

The paper warns that memory can become a “shadow database.” Key safeguards:

- Explicit user consent for long-term storage.

- Data minimization & retention limits.
  - Differential privacy for shared context.
  - Human-in-the-loop review for sensitive recall events.
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## 10. Real-World Applications

Use Case	Example
<b>Personalized Assistants</b>	Remember user preferences and projects across weeks.
<b>Enterprise Agents</b>	Maintain task state for tickets or sales analysis.
<b>Collaborative Systems</b>	Share context between Planner and Executor agents.
<b>Analytics Memory Agents</b>	Summarize daily insights from SQL queries.

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## 11. Core Takeaways

1. **Context = Intelligence.** A model without memory is a chatbot; with context, it becomes a colleague.
2. **Sessions are stateful threads** for continuity.
3. **Memory is not storage — it's strategy.**
4. **Summarization and retrieval** together enable scalable agentic reasoning.
5. **Security and Observability** must be baked into memory design.