

# Ambient Knowledge — Product & Architecture Write-up

Timebox: built as a 4–8 hour take-home MVP.

## TL;DR

Ambient Knowledge is an **AI collaborator for workplace communication** that lives *inside the message composer*.

It proactively finds relevant internal context while you write, shows **grounded sources**, and lets you attach or link that context with one click — reducing back-and-forth without turning the product into “yet another chatbot.”

## 1) Product Vision

Workplace context is fragmented across docs, tickets, incident notes, and prior threads. As a result:

- Senders spend time searching (or skip it) and messages go out missing key background.
- Recipients ask clarifying questions, causing delays and context-switching.

Ambient Knowledge makes the *default* action “send a better message” by making context retrieval **automatic, calm, and auditable**.

## 2) Thesis (how AI should facilitate collaboration)

AI should act like **invisible infrastructure** that reduces coordination cost:

- Prefer **fewer steps and fewer screens**, not new workflows.
- Make improvements **legible** (why this context? what sources?) so teams can trust it.
- Optimize for **shared understanding** between people (sender + recipient), not just helping one user type faster.

This MVP fits that thesis by putting AI in the smallest useful surface: the composer and the message view — a collaborator that helps you include what your teammate will need.

## 3) What This Is (and isn't)

Ambient Knowledge surfaces relevant organizational context while a person writes or reads a message.

- It **is** an inline collaborator: optional, grounded in real artifacts, and designed for one-click usefulness.
- It **is not** a chatbot UI or a separate prompt destination.

## 4) How to run

From a shell, set `REPO_ROOT` to the folder that contains this repo, then:

```
export REPO_ROOT="/path/to/your/code"
cd "$REPO_ROOT/ambientKnowledge"
cd impl
npm install
npm run dev
```

Optional Claude synthesis:

- Copy `impl/.env.example` → `impl/.env.local`
- Set `ANTHROPIC_API_KEY=...`

Note: AI synthesis requires a valid key. If the key is missing/invalid, `/api/context` will error.

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## 5) Design Principles

- **AI as infrastructure, not destination.** The best AI interaction is invisible. Remove workflow steps rather than creating new surfaces.
  - **Calm by default.** Never interrupt typing. Suggestions are dismissible and optional. Low confidence → show less.
  - **Grounded and auditable.** Every summary is attributable to retrieved sources. No hallucination — generation is constrained to retrieved snippets.
  - **Operational controls.** Under load shedding or rate limiting, the system can choose retrieval-only instead of synthesis.
  - **Cost-aware intelligence.** Use expensive inference only when justified. Prefer deterministic retrieval and cached summaries first.
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## 6) Core User Journeys

**Adding context while composing:** User drafts a message → system retrieves related artifacts → proposes a context card (topic + summary + source links) → user attaches with one click → message sends with compact context block.

**Understanding an incoming message:** User receives a message with unfamiliar references → taps "What's this about?" → system retrieves and synthesizes background → user sees summary + source links inline.

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## 7) Engineering Decisions

This section exists to answer “why this stack/architecture, and what did you consider instead?”

### Stack choices

- **Next.js (App Router) + TypeScript:** fast iteration, server routes for retrieval/synthesis, strong typing for extensibility.
- **Tailwind CSS:** speed + consistency for a polished UI within the timebox.
- **Framer Motion:** small, intentional transitions to make AI suggestions feel calm instead of jumpy.

### Architecture choices

- **Tiered serving (cache → retrieval-only → synthesis):** keeps the composer responsive and cost predictable.
- **Retrieval-first design with sources:** maximizes trust and debuggability; LLM is a constrained layer on top.

### Alternatives considered (not implemented in MVP)

- Embeddings + vector DB (pgvector/Pinecone) and BM25 hybrid search.
  - Streaming synthesis and async enrichment via background workers/queues.
  - Redis-backed cache/rate limits for multi-instance deployments.
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## 8) Technical Stack

Layer	Choice
Framework	Next.js 15 App Router, React, TypeScript
Styling	Tailwind CSS + Framer Motion
API	/api/context route (retrieval + synthesis)
Retrieval	Glean-inspired local ingestion + chunk index (hybrid lexical + lightweight semantic proxy)
Synthesis	Anthropic Claude
Data	26 structured mock knowledge artifacts + 11 seeded channels/threads

## 9) What's Implemented

### Context Engine ( contextEngine.ts )

Three-tier serving pipeline that routes every request from cheapest to most expensive:

Tier	Mechanism	Latency
Cache	Reuse prior context for similar intent + channel	~10ms
Retrieval-only	Deterministic summary from top snippets, no LLM	~50ms
Synthesis	Full LLM generation from retrieved context	~1–3s

Supporting infrastructure:

- **In-memory cache** with 45s TTL, 5000 entry max, freshness metadata.
- **Request deduplication** — identical in-flight queries share one response.
- **Rate limiting** — fixed-window (20 req/10s per client).
- **Load shedding** — max 24 concurrent synthesis calls; excess downgrades to retrieval.
- **Mode-aware scoring** — `compose` mode biases toward recipient relevance; `incoming_lookup` mode biases toward message topic.

### Retrieval Engine ( retrieval.ts )

Glean-inspired retrieval flow:

- **Ingest** knowledge items and normalize fields.
- **Chunk** bodies into paragraph/sentence-ish segments.
- **Hybrid score** chunks (lexical TF-IDF-ish + deterministic semantic proxy) and **collapse** best chunks back to document-level hits.
- **ACL filter** by viewer principals (demo: `group:all` + current user).

This is implemented in `gleanIndex.ts` and called by `retrieval.ts`, while preserving the same `retrieve()` interface used by the context engine.

### Knowledge Base (26 artifacts)

Covers multiple project areas (platform, mobile, design system, onboarding, security/ops, release) to demonstrate retrieval differentiation:

Domain	Example Artifacts
Platform Migration	Monolith decomposition, billing DB dual-write strategy
Mobile App v2	React Native rewrite, transaction list performance
API Gateway	Kong vs Envoy benchmark, rollout release notes
Design System	Cross-platform component library, Figma token pipeline
Customer Onboarding	Guided setup wizard, A/B test results (34% activation lift)
Security & Ops	Q1 audit findings (IDOR, JWT), Feb 18 incident postmortem

Each channel surfaces a distinct top artifact — billing migration, design system tokens, onboarding metrics, security findings, gateway release notes, RN performance.

Channel & DM Navigation

- 11 channels (including incident/security/release/QA/leadership) + multiple DM threads.
- Discriminated union routing ( `ActiveView = channel | dm` ) with independent message stores.
- Sidebar highlights correct selection for both channels and DMs.

Browser History & State Preservation

- `history.pushState()` with serialized `NavState` (view, attached context, draft text) on every navigation.
- `replaceState()` snapshots current state before leaving, so mid-composition drafts survive.
- Back/forward restores the full view, attached context, and draft text. The context suggestion hook re-fires immediately on restore.

Context Lookup Gating

Heuristic determines when to show the "What's this about?" button on incoming messages:

- **Suppress** for simple replies/acknowledgements ("Perfect, got it, I'll do X").
- **Show** for messages with clarification signals (questions, approvals, decisions) or technical content (specific terms, metrics, version refs).

Wiki Pages

Dynamic `/wiki/[id]` route renders any knowledge artifact as a full page with related items. Source links in context cards link directly to these pages.

10) Tradeoffs (and what I'd do with more time)

Tradeoffs made for speed and clarity in a 4–8 hour MVP:

- **Mock connectors/data:** knowledge comes from a seeded JSON store instead of real integrations.
- **No real auth/ACL graph:** permission checks are simplified to keep the demo coherent.
- **Deterministic “semantic proxy”:** avoids embedding infra; keeps the demo reproducible offline.
- **In-memory caching/rate limits:** good for the prototype; not production-grade multi-instance.

With more time:

- Replace semantic proxy with real embeddings + hybrid retrieval, preserve citations.
  - Add async enrichment + streaming so retrieval appears immediately and synthesis upgrades in-place.
  - Add Redis + queues for true multi-tenant scaling and fairness.
  - Expand “incoming message understanding” into a first-class workflow (reply suggestions, shared context snippets).
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## 11) Scaling to 1000+ Users

At organizational scale, “LLM per keystroke” fails. The core risks:

- **Request volume explosion** — typing-triggered requests × concurrent users = high QPS spikes.
- **LLM saturation** — throughput, latency, and cost destabilize during peaks.
- **Retrieval fan-out** — slow sources block the whole suggestion pipeline.
- **Multi-tenant fairness** — heavy teams starve others without controls.

### Architecture Strategy

**Tiered inference (already implemented):** Cache → retrieval-only → synthesis. The agent is never the bottleneck for baseline usefulness.

**Admission control:** Per-user/workspace token budgets, QPS caps, and queue depth thresholds. Downgrade to lower tier rather than block.

**Async enrichment (planned):** Return retrieval result immediately; upgrade with synthesis when ready. User sees context improve in-place.

**Hybrid retrieval (partially implemented):** The demo already uses lexical + a lightweight semantic proxy with ACL filtering. At scale, you’d replace the semantic proxy with real embeddings + vector search (and likely keep BM25 in the mix).

### AmbientKnowledge vs Glean (explicitly)

What this demo borrows from Glean’s *shape*:

- **Connector → ingest → index → retrieve → synthesize** mental model (even though the demo’s “connector” is a JSON seed).
- **Chunk-level retrieval** with doc-level collapsing.
- **Permission filtering** (ACL principals) applied during retrieval.
- **Hybrid scoring** (lexical + semantic signal) rather than pure keyword.

What Glean does that this demo intentionally does not:

- **Real connectors + incremental crawls** (Slack/Drive/Jira/GitHub/etc), change detection, retries.
- **True embeddings + vector infrastructure** (model serving, vector DB, ANN search tuning).
- **Enterprise ACL graph** (groups, nested groups, external shares, doc-level vs field-level permissions).
- **Personalization at scale** (click feedback loops, recency/affinity graphs, per-user ranking).
- **Hard multi-tenant isolation** (per-tenant indexes, quotas, compliance controls).

**Context precomputation:** Background jobs pre-generate context bundles for high-traffic channels and recurring topics.

**Model routing:** Smaller models for first-pass synthesis; escalate to larger models for low-confidence/high-importance cases.

**UX Under Load**

Users never experience a blocked composer. Under constrained conditions: show deterministic context quickly, mark confidence clearly, allow immediate attach/send, upgrade opportunistically when richer synthesis arrives.

**Target SLOs**

Tier	p95 Latency	Availability
Cache / Retrieval	< 700ms	99.9%
Synthesis (async)	< 2.5s	99.5%

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**12) API Contract**

**Request:** { draftText, recipientId, channelId?, mode?: "compose" | "incoming\_lookup" }

**Response:** { topic?, summary?, openQuestions?, sources?, confidence?, servingTier?, freshnessMs?, debug? }

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**13) What's Next**

Priority	Improvement
1	Replace in-memory cache/rate-limit with Redis for multi-instance support
2	Async synthesis enrichment — return retrieval immediately, stream upgrade
3	Production hybrid retrieval (BM25 + embeddings) with permission filtering
4	Per-tenant fairness quotas and priority classes
5	Observability dashboards (tier mix, cache hit rate, queue depth, p95 latency)

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**14) Success Metrics**

**Product:** Context attach rate · "What's this about?" usage · Reduction in follow-up clarification messages · Time-to-send improvement.

**System:** p95 latency by tier · Cache hit rate · Synthesis fallback rate · Cost per context card.

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**15) How this meets the evaluation criteria**

- **Extensibility/readability:** clear separation of retrieval, synthesis, and UI; typed contracts and small modules.
- **Creative AI-human collaboration:** AI helps people send *better, more complete* messages by attaching shared context (not just "help me write").
- **Polish & UX:** calm, optional suggestions; one-click attach/insert; sources always visible.

- **Beautiful simplicity:** AI reduces screens (no chatbot surface) and reduces work (automatic retrieval).
  - **Robustness:** tiered serving + load shedding makes failure modes graceful and cost-aware.
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## 16) Graceful Degradation

If LLM synthesis is unavailable (missing key, rate limiting, transient errors), the system downgrades to **retrieval-only** instead of blocking the composer. The user still gets relevant context with visible sources.