

PRD — Content Concierge AI

Hybrid RAG-Enabled MVP (Traditional RAG + GraphRAG)

1. Purpose of This PRD

This document defines the **full AI technical blueprint** for a Content Concierge MVP that actively leverages **Hybrid RAG**, combining: - Traditional vector-based RAG - GraphRAG powered by TigerGraph

This PRD is designed to be **directly codeable**, with explicit components, data flows, and agent responsibilities.

2. AI System Goals

- Deepen personalization beyond APIs
 - Capture long-term thematic intelligence
 - Improve insight relevance via relationship-aware retrieval
 - Maintain explainability and compliance
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3. High-Level Architecture

Pattern: Agentic Hybrid Intelligence System

- LangGraph → orchestration
 - Vector Store → semantic retrieval
 - TigerGraph → relational reasoning
 - LLM → synthesis only
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4. Knowledge Layers

4.1 Traditional RAG Layer

Ingested Content: - Bank-approved research - Educational explainers - Historical market commentary

Pipeline: - Chunk → embed → store - Metadata: asset class, risk type, horizon, compliance tags

4.2 GraphRAG Layer (TigerGraph)

Core Entities: - User - Asset - Sector - Strategy - Market Theme - Risk Factor

Edges: - OWNS - EXPOSED_TO - ALIGNS_WITH - IMPACTED_BY

Graph enables reasoning such as:

"Users with X exposure and Y goal often care about Z themes"

5. LangGraph Agent Design

5.1 Agent State

```
{
  "user_context": {},
  "vector_context": [],
  "graph_context": [],
  "candidate_insights": [],
  "final_insights": []
}
```

5.2 Graph Nodes

1. **User Context Node**
2. Same as PoC 1
3. **Graph Context Expansion Node**
4. Query TigerGraph for related themes, risks, strategies
5. **Vector Retrieval Node**
6. Fetch semantically relevant documents
7. **Insight Planning Node (LLM)**
8. Merge graph + vector signals into insight candidates
9. **Synthesis Node (LLM)**

10. Generate text insights
 11. **Citation Resolution Node**
 12. Attach doc + graph provenance
 13. **Ranking Node**
 14. Select 2-3 best insights
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6. Citation Model

Each insight must include: - Source document(s) - Graph path explanation (optional, internal)

Example:

Insight derived from [Doc A] + exposure relationship (User → ETF → Sector → Theme)

7. Guardrails & Explainability

- LLM cannot traverse graph directly
 - All graph queries are deterministic
 - Insight text references only retrieved facts
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8. Observability

- Graph queries
 - Retrieved documents
 - Prompt inputs/outputs
 - Insight lineage
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9. Tradeoffs & Risks

- Higher latency
 - Increased infra complexity
 - Stronger long-term personalization
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10. Success Criteria

- Higher insight relevance vs API-only baseline
- Reduced repetition over time
- Clear provenance for every insight