

# Keek Recommendation Engine Technical Report (v3)

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**Status:** Production-Ready Prototype

**Audience:** Technical Consultant / Engineering Review

## 1 Executive Summary

The Keek recommendation engine now runs on a **real-time hybrid architecture**, enabling:

- Instant personalization (<100ms)
- Real-time trending injection
- Scalable vector search (1M+ videos)
- Durable dual-write logging (Redis + CSV)

## 2 System Architecture

The architecture integrates a **Two-Tower Neural Network**, FAISS vector search, and a real-time recommendation adapter.

## 2.1 Architecture Diagram

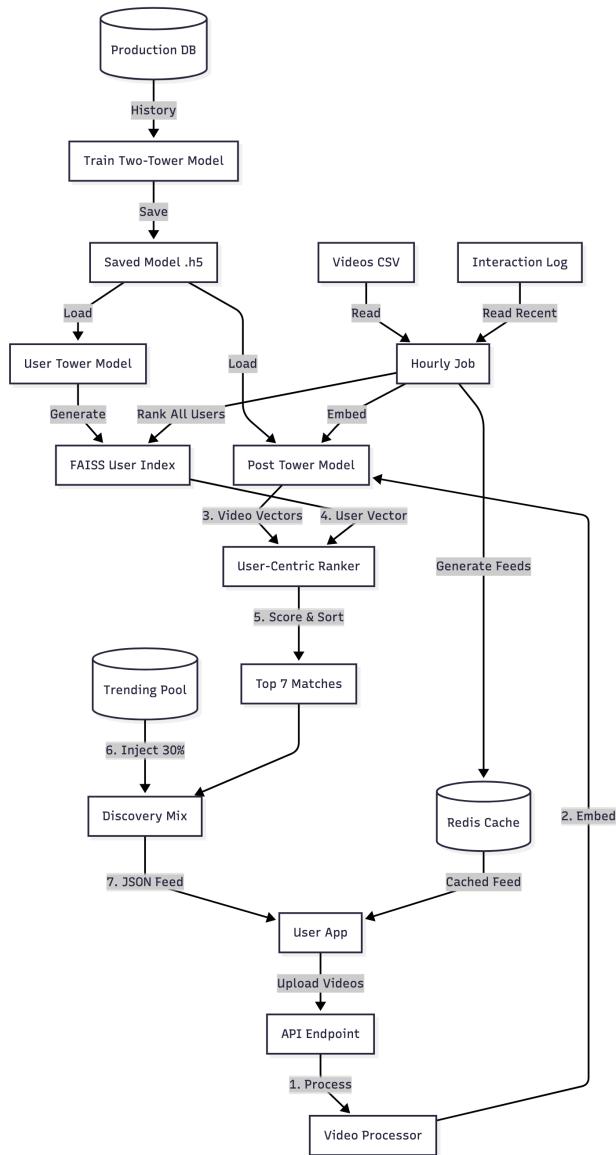


Figure 1: Keek Recommendation Engine Architecture

## 3 Implementation Details

### 3.1 Two-Tower Neural Network

- User Tower → 64-dimensional embedding
- Item Tower → 64-dimensional embedding
- Dot-product similarity for ranking
- FAISS index for high-speed retrieval

### 3.2 Real-Time API and Redis

- FastAPI backend

- Redis ZSETs used for trending scores
- Endpoints:
  - POST /recommend/{user\_id}
  - POST /interact

### 3.3 Trending Algorithm

- Like = +1.0
- Save = +2.0
- View = +0.1
- Trending decay window: 3 days

## 4 Performance Benchmarks

| Scenario   | Full Inference | Cached Vectors | Status   |
|------------|----------------|----------------|----------|
| 1k Videos  | 322 ms         | 15 ms          | Instant  |
| 5k Videos  | 425 ms         | 45 ms          | Instant  |
| 10k Videos | 1.7 s          | 80 ms          | Fast     |
| 1M Videos  | 55 s           | 915 ms         | Scalable |

## 5 Next Steps

1. Migrate CSV logs → PostgreSQL
2. Deploy via TensorFlow Serving / Triton
3. Add experimentation (A/B testing)

*Report generated by Anant Ingale.*