

LHCbFinder: Semantic Search and Knowledge Discovery Framework

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Vision and Motivation

Current Knowledge Management Challenges at LHCb:

- Fragmented knowledge across multiple platforms (TWiki, Indico, arXiv, internal notes)
- Valuable institutional knowledge often undocumented or difficult to discover
- Steep learning curve for newcomers joining the collaboration

LHCbFinder Solution:

- Centralize scattered documentation in a semantic framework
- Enable intuitive natural language search across all resources
- Preserve and share institutional knowledge
- Reduce entry barriers for new members

Semantic Search Foundation

LHCbFinder employs a powerful semantic search pipeline that forms the foundation of our knowledge platform:

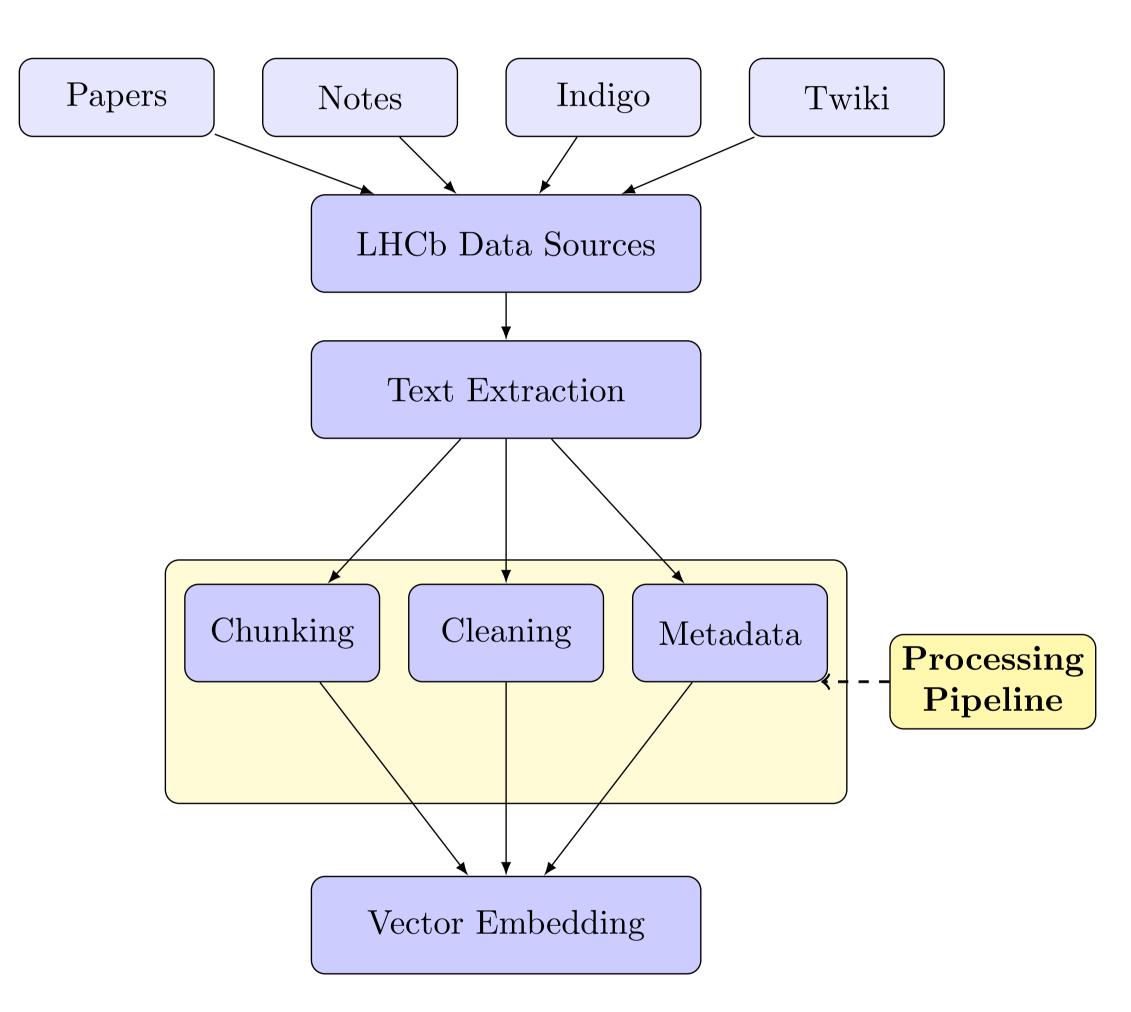


Figure 1. Text processing pipeline for creating vector embeddings

Strategic Implementation:

- Phase 1: Abstracts for rapid validation
- Phase 2: Full paper introductions for deeper context
- Phase 3: Comprehensive scaling to all document types

Key Features

Our semantic search system offers significant advantages over traditional keyword search:

- Understands meaning and context rather than just matching keywords
- Finds conceptually related documents using vector similarity
- Supports natural language queries for intuitive discovery

Understanding Embeddings and Vector Search

The heart of LHCbFinder is our embedding system that converts scientific text into semantic vectors:

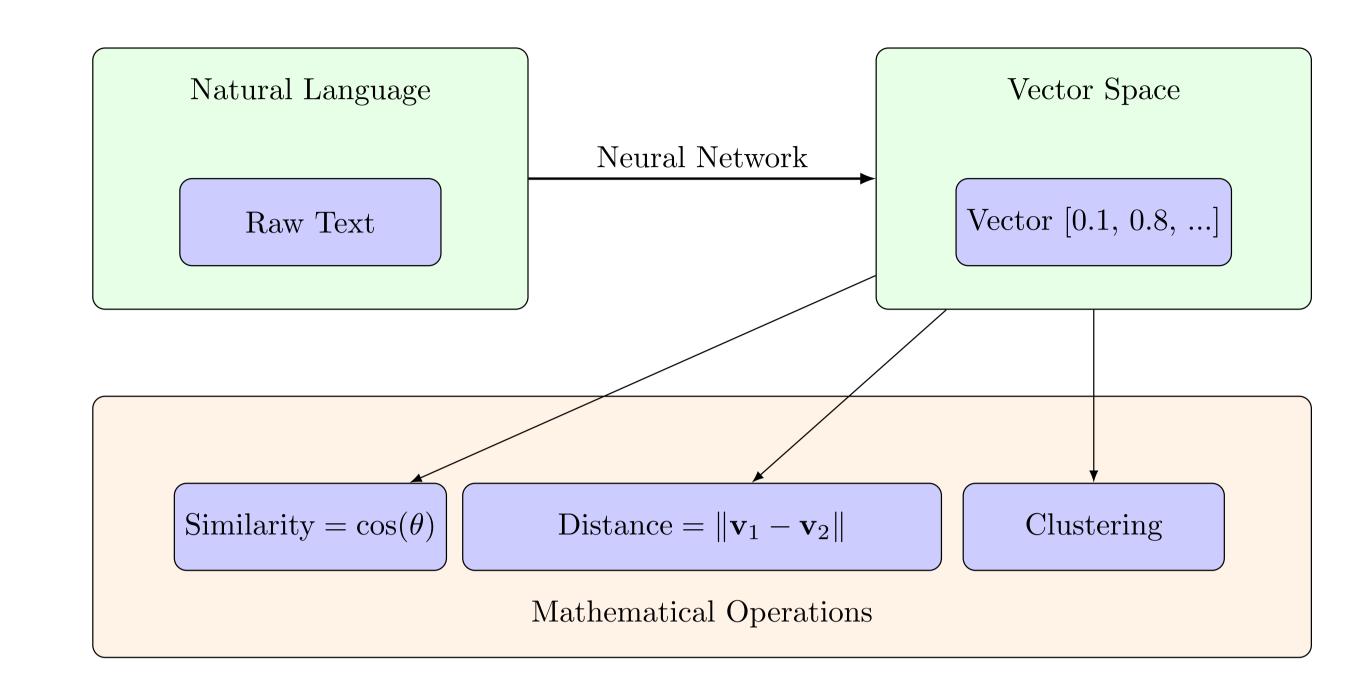


Figure 2. Converting text to vectors for semantic search

Embedding Model: BAAI/bge-large-en-v1.5

- 1024-dimensional vector embeddings
- Optimized for scientific literature
- Excellent technical term handling

Vector Database:

- Efficient similarity search with ChromaDB (local) and Pinecone (cloud)
- Scales to millions of vectors
- Supports hybrid deployment for flexibility

RAG: Retrieval-Augmented Generation Framework

Knowledge Base Platform

The next phase of LHCbFinder integrates LLMs with our vector knowledge base:

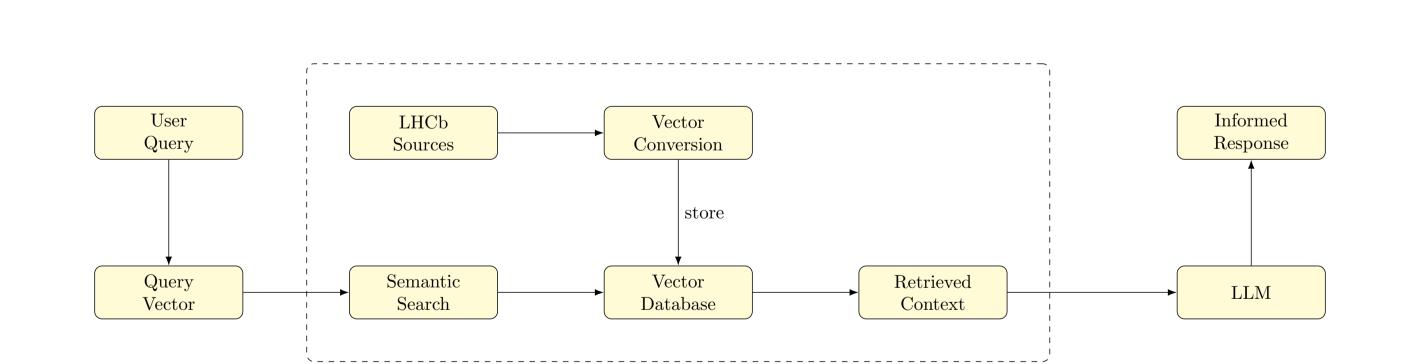


Figure 3. Retrieval-Augmented Generation architecture

Live Demo and Current Status

Ihcbfinder.net offers semantic search across LHCb papers with natural language queries. Try searches like "CP violation in B decays" or "Machine learning for tracking" to explore related papers ranked by semantic similarity.

Development Roadmap

Three-Phase Strategy

Phase 1: Founda-|Full semantic search for LHCb papers. Optimized embeddings and refined user experience. tion

Current

Phase 2: Integra- Developing document scraping/knowledge grabbing pipeline for diverse LHCb resources and intetion grating with LLM models. Next

Future

Phase 3: Expansion Framework-agnostic development to extend beyond LHCb, enabling knowledge discovery across different HEP experiments.

Current Priorities: Content acquisition, framework modularization, and LLM integration.

Technical Implementation

System Architecture:

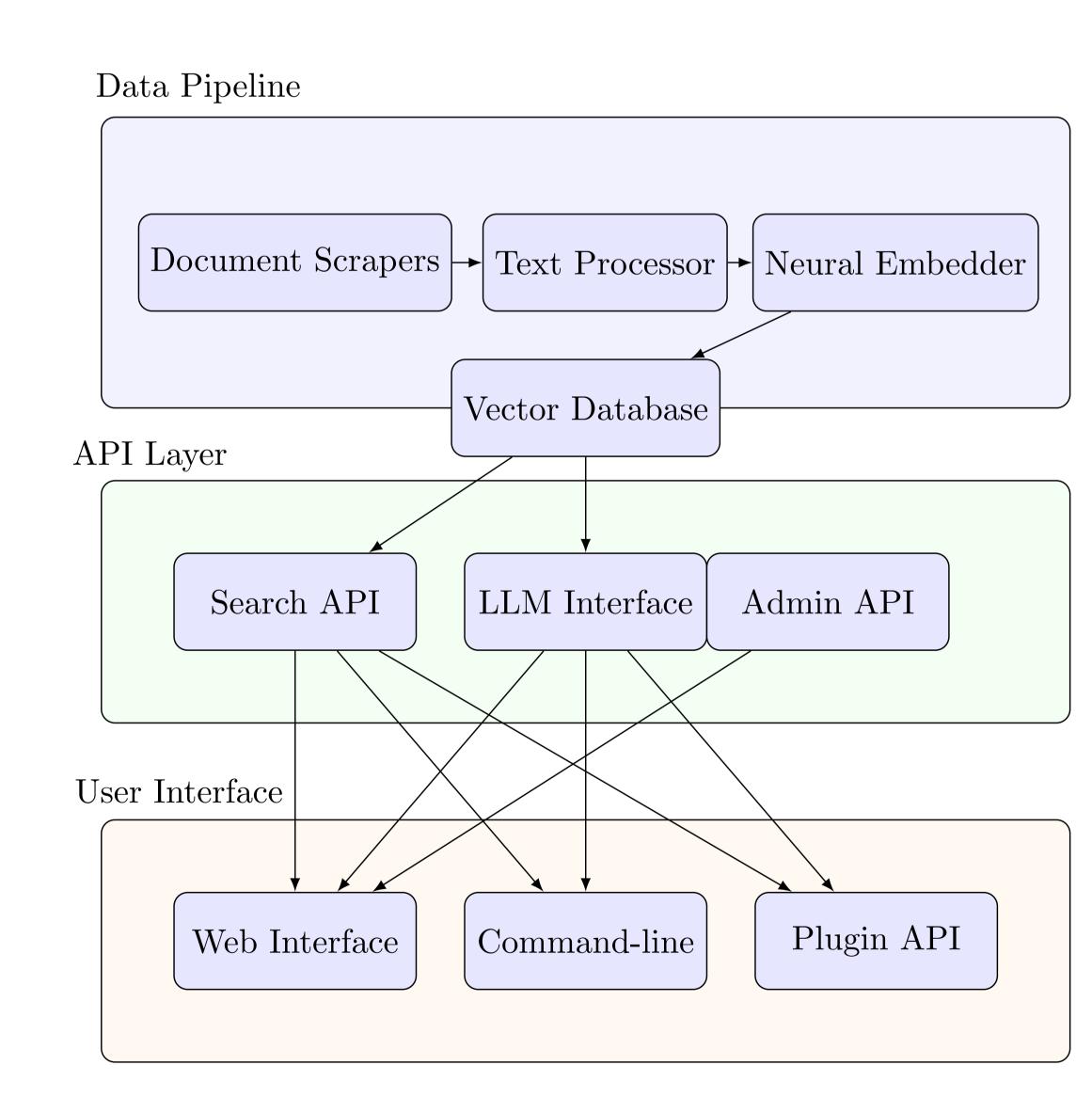


Figure 4. LHCbFinder system architecture

Future Research Directions

LHC-Specific Embedding Models: Training specialized vector embeddings on physics domain terminology to improve semantic understanding of technical concepts in HEP documentation.

LLM Integration Framework: Evaluating various open-source LLMs for compatibility with scientific knowledge retrieval, including benchmarking domain-specific response quality.

Experiment-Agnostic Architecture: Developing a modular design to extend beyond LHCb, enabling knowledge discovery across all LHC experiments with minimal adaptation.