

Appendix: HQF-DE Implementation Details

A System Architecture

Figure 1 shows the overall HQF-DE system architecture with the four document variants flowing through both retrieval paths.

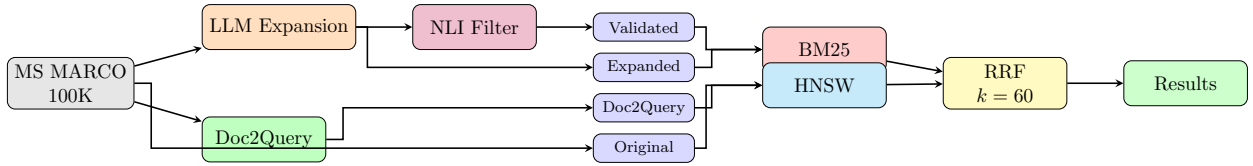


Figure 1: HQF-DE system architecture: document expansion variants flow through dual retrieval paths.

B Pipeline Stages

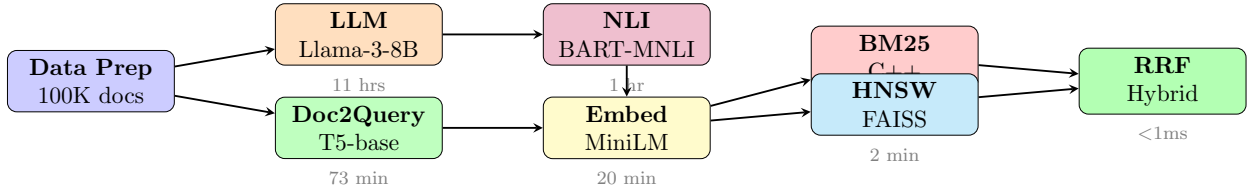


Figure 2: Pipeline stages with processing times. GPU stages run on Colab A100; CPU stages run locally.

C Infrastructure Adaptation

Due to GPU memory constraints on local hardware, we adopted a hybrid execution approach.

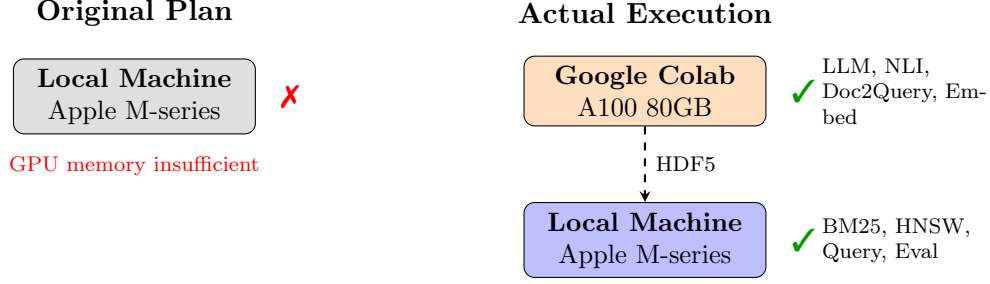


Figure 3: Infrastructure adaptation: GPU tasks on Colab, CPU tasks locally.

Table 1: Time and Space Complexity of Key Components

| Component | Time | Space |
|-------------------|--------------------------|-----------------------------|
| BM25 Indexing | $O(N \cdot L)$ | $O(V + P)$ |
| BM25 Query | $O(\sum_t L_t)$ | $O(k)$ |
| HNSW Construction | $O(N \log N \cdot M)$ | $O(N \cdot M \cdot \log N)$ |
| HNSW Query | $O(\log N + k \cdot ef)$ | $O(ef)$ |
| RRF Fusion | $O(k_1 + k_2)$ | $O(k_1 + k_2)$ |

N =docs, L =doc length, V =vocab, P =postings, $|L_t|$ =posting list, M =connections, ef =beam width

D Complexity Summary

E Complete Results

E.1 HNSW Dense Retrieval

E.2 BM25 Sparse Retrieval

E.3 Hybrid Retrieval (RRF)

F Key Findings Visualization

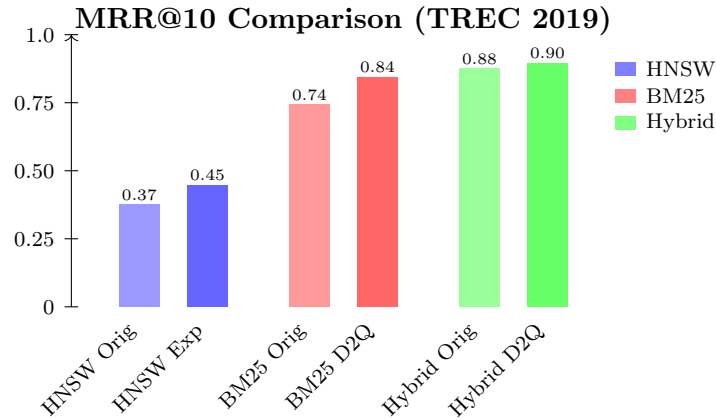


Figure 4: MRR@10 comparison. Hybrid + Doc2Query achieves best performance (0.90).

Table 2: HNSW Results (MRR@10 / nDCG@10 / Recall@100)

| Variant | TREC 2019 | TREC 2020 | Dev Set |
|-----------|-------------------------------------|------------------------------|-----------------------|
| Original | 0.374 / 0.154 / 0.485 | 0.275 / 0.105 / 0.551 | 0.011 / 0.002 / 0.876 |
| Expanded | 0.447 / 0.154 / 0.471 | 0.273 / 0.094 / 0.550 | 0.012 / 0.003 / 0.866 |
| Validated | 0.381 / 0.149 / 0.472 | 0.267 / 0.092 / 0.551 | 0.013 / 0.004 / 0.869 |
| Doc2Query | 0.305 / 0.134 / 0.472 | 0.307 / 0.085 / 0.540 | 0.013 / 0.003 / 0.865 |

Table 3: BM25 Results (MRR@10 / nDCG@10 / Recall@100)

| Variant | TREC 2019 | TREC 2020 | Dev Set |
|-----------|--|--|--|
| Original | 0.744 / 0.419 / 0.568 | 0.688 / 0.428 / 0.590 | 0.689 / 0.712 / 0.933 |
| Expanded | 0.752 / 0.428 / 0.579 | 0.713 / 0.412 / 0.602 | 0.658 / 0.685 / 0.908 |
| Validated | 0.736 / 0.415 / 0.578 | 0.703 / 0.411 / 0.601 | 0.662 / 0.689 / 0.911 |
| Doc2Query | 0.844 / 0.492 / 0.596 | 0.741 / 0.465 / 0.628 | 0.746 / 0.767 / 0.944 |

G Expansion Method Comparison

| | |
|--|--|
| LLM Expansion Best for: Dense retrieval Gain: +19% MRR (HNSW) Time: 11 hours | Doc2Query Best for: BM25 & Hybrid Gain: +13% MRR (BM25) Time: 73 minutes |
|--|--|

Recommendation

Use Doc2Query for production
 (faster, better for BM25/Hybrid).
 Use LLM expansion for dense-only retrieval.

Figure 5: Expansion method comparison with practical recommendations.

H AI Assistance

Claude and Gemini were used to assist with code development, debugging, and L^AT_EX formatting throughout this project.

Table 4: Hybrid Results (MRR@10 / nDCG@10 / Recall@100)

| Variant | TREC 2019 | TREC 2020 | Dev Set |
|-----------|------------------------------|------------------------------|------------------------------|
| Original | 0.877 / 0.475 / 0.594 | 0.757 / 0.431 / 0.641 | 0.701 / 0.730 / 0.945 |
| Expanded | 0.810 / 0.459 / 0.588 | 0.723 / 0.412 / 0.636 | 0.676 / 0.707 / 0.923 |
| Validated | 0.810 / 0.451 / 0.587 | 0.719 / 0.410 / 0.635 | 0.680 / 0.709 / 0.926 |
| Doc2Query | 0.896 / 0.512 / 0.613 | 0.779 / 0.468 / 0.653 | 0.762 / 0.783 / 0.951 |