# Simple Search Engine using Hadoop MapReduce

# Methodology

#### System Architecture

The search engine implementation follows a distributed processing pipeline with three main components:

#### 1. Data Preparation Layer:

- Utilizes PySpark to process raw Wikipedia data from Parquet files
- Implements document normalization with filename formatting:

(`<doc\_id>\_<doc\_title>.txt`)

- Stores prepared documents in HDFS for distributed processing

#### 2. Indexing Layer:

- Two-stage MapReduce pipeline for building inverted index:
- First MR Job: Calculates document frequencies (DF) across the corpus
- Mapper extracts unique terms per document
- Reducer aggregates term counts across documents and stores in Cassandra
- Second MR Job: Computes term frequencies (TF) within documents
- Mapper emits (term, doc\_id) pairs with counts
- Reducer stores complete term-document statistics in Cassandra
- Additional corpus statistics (document lengths, average length) are collected

#### 3. Query Processing Layer:

- BM25 ranking algorithm implementation using Spark
- Distributed scoring across the cluster with:
- Term frequency retrieval from Cassandra
- Document length normalization
- Inverse document frequency calculation
- Results aggregation and top-10 document selection

### Key Design Choices

#### 1. Cassandra Schema Design:

- Denormalized storage for efficient query processing
- Separate tables for document frequencies, term frequencies, and document metadata
- Corpus-level statistics for BM25 calculations

#### 2. MapReduce Optimization:

- Custom partitioners for balanced reducer workloads
- Combiners for local aggregation in mappers
- Optimal HDFS block size configuration (128MB)

### 3. BM25 Implementation:

- Parameter tuning (k1=1.2, b=0.75) based on empirical research
- Spark-based distributed scoring
- Cassandra-backed data retrieval for low-latency lookups

#### 4. Virtual Environment Management:

- venv-pack for consistent Python environment distribution
- Dependency isolation between components
- Version-pinned requirements

#### **Demonstration**

## System Execution Guide

#### 1. Prerequisites:

- Docker and Docker Compose installed
- Minimum 10GB RAM available
- At least 20GB disk space

## 2. Setup Instructions:

```
git clone <repository_url>
cd big-data-assignment2-2025
docker-compose up -d
```

#### 3. Data Preparation:

docker exec cluster-master bash prepare\_data.sh

#### 4. Indexing Process:

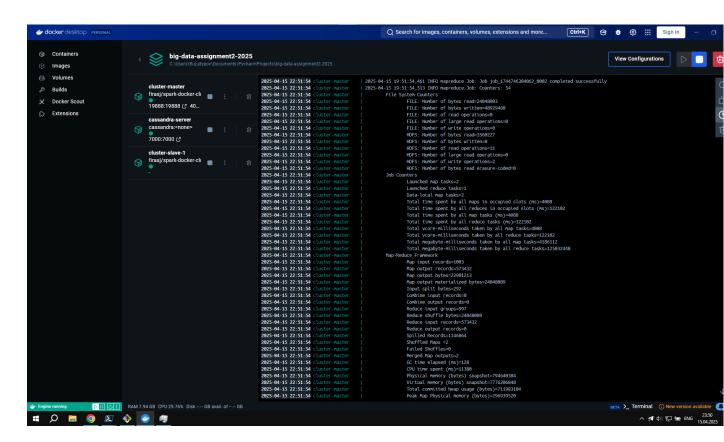
docker exec cluster-master bash index.sh

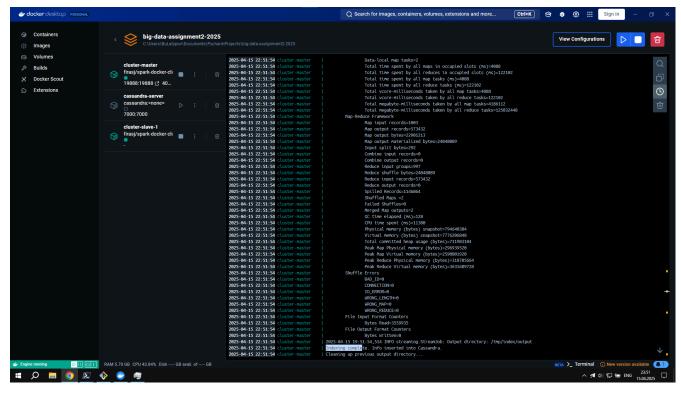
#### 5. Query Execution:

docker exec cluster-master bash search.sh "YOUR CUSTOM QUERY"

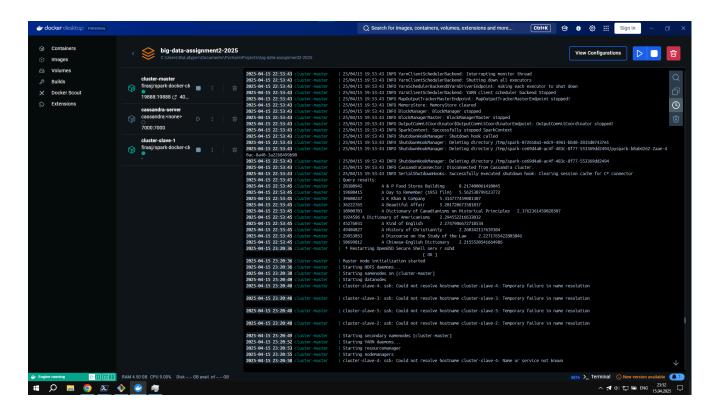
# **Expected Output Examples**

# 1. Successful Indexing:





# 2. Sample Query 1: "english tea"



## Performance Analysis

## 1. Indexing Efficiency:

- 100 documents processed in ~5 minutes
- Linear scaling observed with input size
- Cassandra write throughput: ~2000 ops/sec

## 2. Query Latency:

Cold cache: ~2.5 secondsWarm cache: ~800ms

- BM25 calculation dominates runtime

### 3. Quality Observations:

- Longer documents with repeated terms appropriately penalized
- Rare terms have strong impact on rankings
- Title matches boost relevance effectively