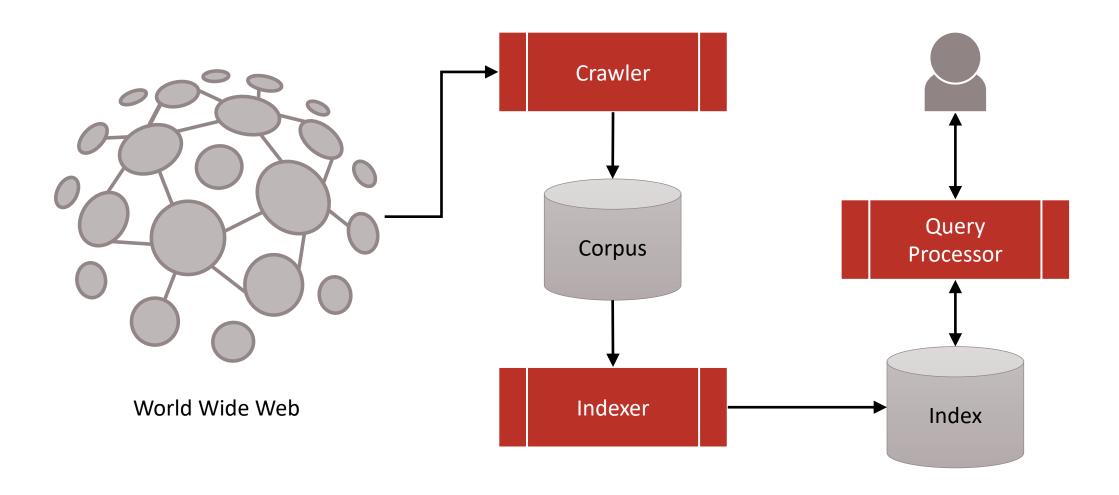


Information Retrieval

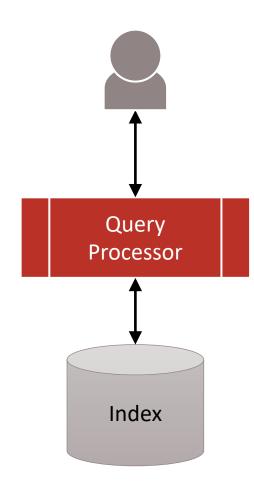
Document Matching

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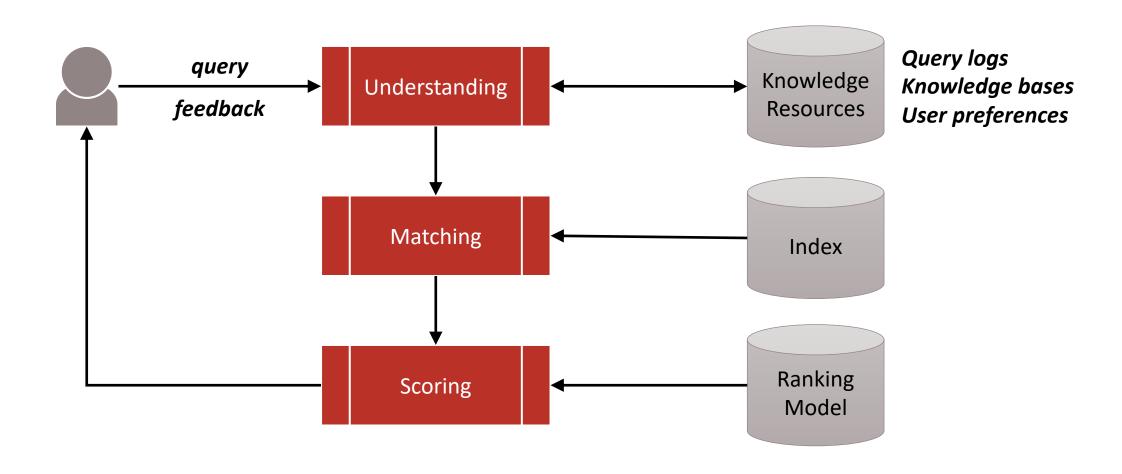
Search components



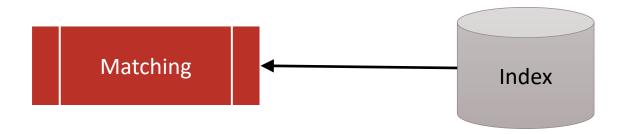
Search components



Query processing overview



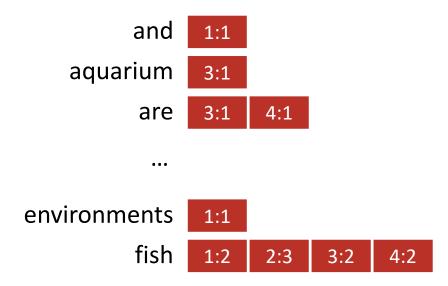
Query processing overview



Document matching

Scan postings lists for all query terms

[aquarium fish]



Document matching

Scan postings lists for all query terms

[aquarium fish]

Score matching documents

$$\circ f(q,d) = \sum_{t \in q} f(t,d)$$

Key challenge

Matching must operate under strict time constraints

- Even a slightly slower search (0.2s-0.4s) can lead to a dramatic drop in the perceived quality of the results
- What makes it so costly?
- Must score billions (trillions?) of documents
- Must answer thousands of concurrent queries

Solution #1: bypass scoring

Query distributions similar to Zipf

- Popular queries account for majority of traffic
 Caching can significantly improve efficiency
- Cache search results, or at least inverted lists

Problem: cache misses will happen eventually

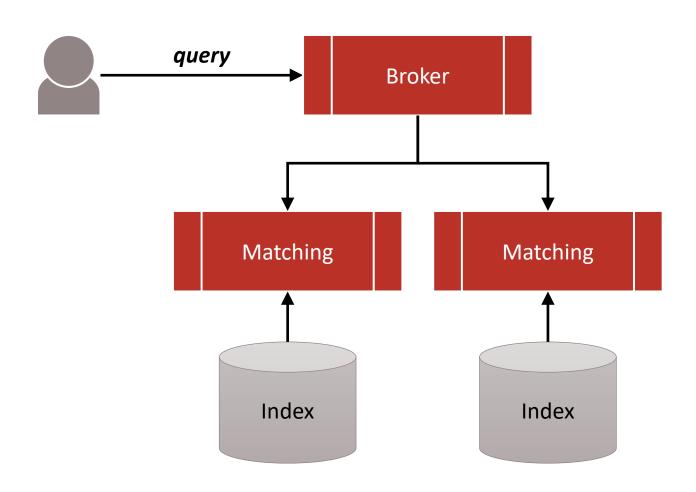
New queries, index updates

Solution #2: distribute the burden

Indexes are often distributed in a cluster

- Too large to fit in a single machine
- Replication helps load balancing

Solution #2: distribute the burden



Solution #2: distribute the burden

Indexes are often distributed in a cluster

- Too large to fit in one machine
- Replication helps load balancing

Problem: cannot scale indefinitely

- Costly resources (hardware, energy)
- Intra-node efficiency still crucial

Solution #3: score parsimoniously

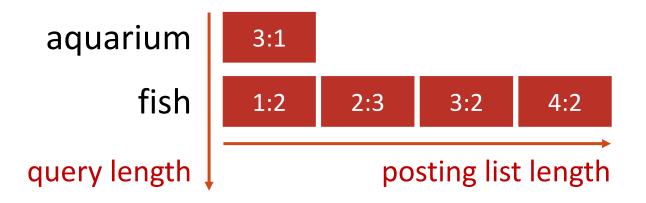
Some ranking models can be expensive

- Infeasible to score billions of documents
- Ranking as a multi-stage cascade
- Stage #1: Boolean matching (billions)
- Stage #2: Unsupervised scoring (millions)
- Stage #3: Supervised scoring (thousands)

Why is it still so costly?

Inherent cost of matching documents to queries

- Query length (number of posting lists)
- Posting lists length (number of postings per list)



How to traverse postings?

Inverted lists processed in sequence

```
      salt
      1:1
      4:1

      water
      1:1
      2:1
      4:1

      tropical
      1:2
      2:2
      3:1
```

Inverted lists processed in sequence

salt	1:1	4:1	
water	1:1	2:1	4:1
tropical	1:2	2:2	3:1
score	1:1	4:1	

Inverted lists processed in sequence



Inverted lists processed in sequence

```
      salt
      1:1
      4:1

      water
      1:1
      2:1
      4:1

      tropical
      1:2
      2:2
      3:1

      score
      1:4
      2:3
      3:1
      4:2
```

```
1: function taat(query, index, k)
2:
      scores = map()
3:
     results = heap(k)
 4:
     for term in tokenize(query)
        postings = index[term]
 5:
6:
        for (docid, weight) in postings
          if docid not in scores.keys()
7:
8:
            scores[docid] = 0
9:
          scores[docid] += weight
      for docid in scores.keys()
10:
11:
        results.add(docid, scores[docid])
12:
     return results
```

Inverted lists processed in parallel

```
      salt
      1:1
      4:1

      water
      1:1
      2:1
      4:1

      tropical
      1:2
      2:2
      3:1
```

Inverted lists processed in parallel

```
      salt
      1:1
      4:1

      water
      1:1
      2:1
      4:1

      tropical
      1:2
      2:2
      3:1

      score
      1:4
```

Inverted lists processed in parallel

```
      salt
      1:1
      4:1

      water
      1:1
      2:1
      4:1

      tropical
      1:2
      2:2
      3:1

      score
      1:4
      2:3
```

Inverted lists processed in parallel

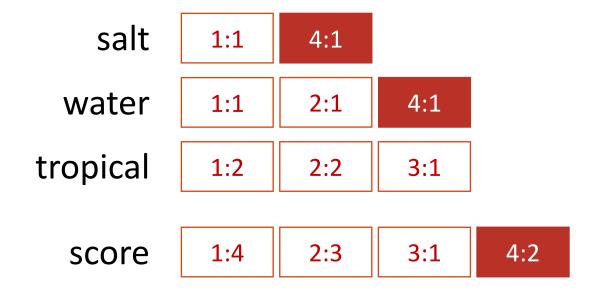
```
      salt
      1:1
      4:1

      water
      1:1
      2:1
      4:1

      tropical
      1:2
      2:2
      3:1

      score
      1:4
      2:3
      3:1
```

Inverted lists processed in parallel



```
1: function daat(query, index, k)
2:
     results = heap(k)
     targets = {docid for term in tokenize(query)
3:
                       for docid in index[term]}
     lists = [index[term] for term in tokenize(query)]
4:
 5:
      for target in targets
6:
       score = 0
7:
       for postings in lists
8:
          for (docid, weight) in postings
9:
            if docid == target
10:
              score += weight
11:
       results.add(target, score)
     return results
12:
```

Optimization techniques

No clear winner

- TAAT is more memory efficient (sequential access)
- DAAT uses less memory (no accumulators)

Naïve versions can be improved

- Calculate scores for fewer documents (conjunctions)
- Read less data from inverted lists (skipping)

Matching semantics

Disjunctive matching

- Documents must contain at least one query term
- More matches, lower precision

Problem: all lists must be fully traversed

 There may be a relevant document in the last position of a (very long) posting list for a common term

Matching semantics

Conjunctive matching

- Documents must contain all query terms
- Fewer matches, higher precision

Conjunctive matching preferred in practice

- Effective and efficient for short queries
- Combined with relaxation for long queries

Skipping

Search involves comparing lists of different lengths

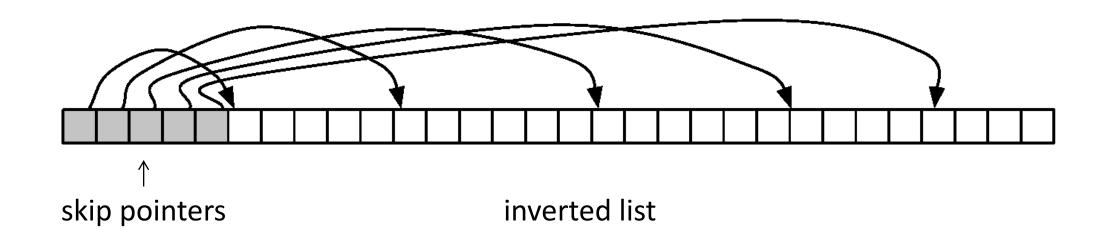
- Linear scanning can be very inefficient
- Skip ahead to document under comparison
- Compression makes jumping difficult (variable size encodings, only d-gaps stored)

Solution: store skip pointers in the index

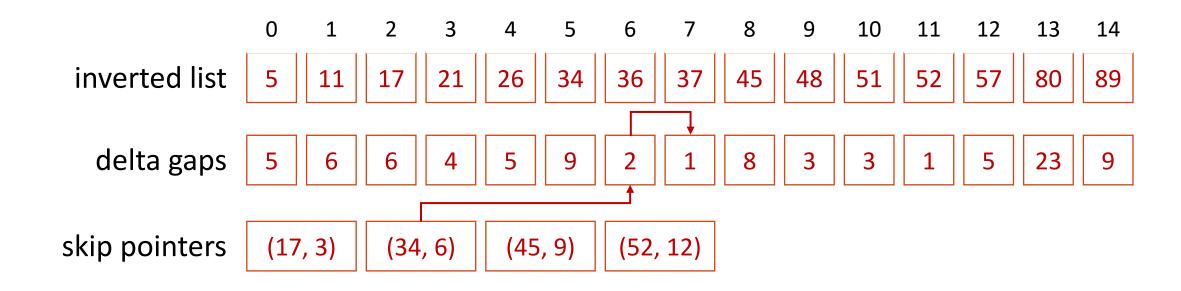
Skip pointers

A skip pointer (d, p) for document d and position p

 \circ Docid d precedes posting at (zero-based) position p

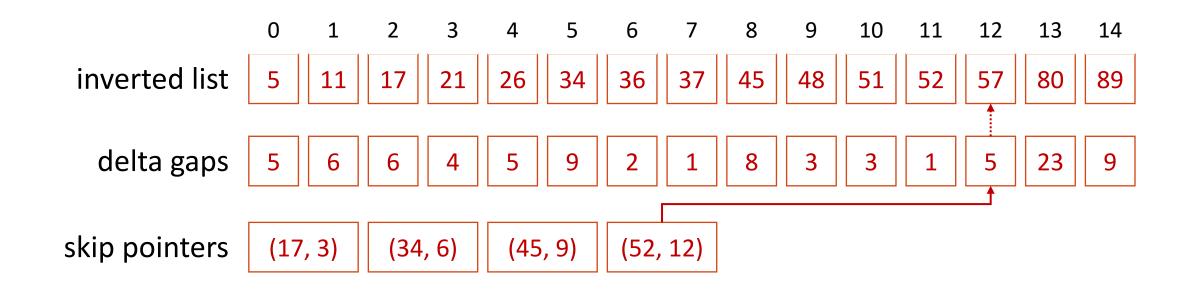


Skipping example



How to skip to docid 37?

Skipping example



How to skip to docid 54?

Other approaches

Unsafe early termination

- Ignore high-frequency word lists in TAAT
- Ignore documents at end of lists in DAAT

Can be improved with index tiering

- Postings ordered by quality (e.g., PageRank)
- Postings ordered by score (e.g., BM25)

Summary

Document matching can be challenging

- Traversing multiple, long inverted lists
- Several complementary approaches
- Caching, distribution, cascading
- Efficient index traversal still crucial
- Classic techniques, several optimizations

References

<u>Search Engines: Information Retrieval in Practice</u>, Ch. 5 Croft et al., 2009

Scalability Challenges in Web Search Engines, Ch. 4 Cambazoglu and Baeza-Yates, 2015

Efficient Query Processing Infrastructures

Tonellotto and Macdonald, SIGIR 2018



Coming next...

Efficient Matching

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