

Inf 2B: Indexing and Sorting for the WWW

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Example 'Set of Documents'

Document	Text
1	Pease porridge hot, pease porridge cold,
2	Pease porridge in the pot,
3	Nine days old.
4	Some like it hot, some like it cold,
5	Some like it in the pot,
6	Nine days old.

A childrens rhyme, each line being treated as a document

Inverted Index

Large set D of documents (possibly from WWW).

We have a set of **terms** appearing in the documents.
The set of terms is called the **lexicon**.

Definition: An **inverted file entry** consists of a single term, followed by a list of the **locations** where the term appears in the set of documents.

Definition: An **Inverted Index** is a list of **inverted file entries**, one for each of the terms in the lexicon, presented in order of **term number**.

Inverted Index for our Example

Number	Term	Documents
1	cold	$\langle 2; 1, 4 \rangle$
2	days	$\langle 2; 3, 6 \rangle$
3	hot	$\langle 2; 1, 4 \rangle$
4	in	$\langle 2; 2, 5 \rangle$
5	it	$\langle 2; 4, 5 \rangle$
6	like	$\langle 2; 4, 5 \rangle$
7	nine	$\langle 2; 3, 6 \rangle$
8	old	$\langle 2; 3, 6 \rangle$
9	pease	$\langle 2; 1, 2 \rangle$
10	porridge	$\langle 2; 1, 2 \rangle$
11	pot	$\langle 2; 2, 5 \rangle$
12	some	$\langle 2; 4, 5 \rangle$
13	the	$\langle 2; 2, 5 \rangle$

Note: Frequency refers to number of documents.

Another Inverted Index for our Example

Number	Term	Documents;Words
1	cold	$\langle 2; (1; 6), (4; 8) \rangle$
2	days	$\langle 2; (3; 2), (6; 2) \rangle$
3	hot	$\langle 2; (1; 3), (4; 4) \rangle$
4	in	$\langle 2; (2; 3), (5; 4) \rangle$
5	it	$\langle 2; (4; 3, 7), (5; 3) \rangle$
6	like	$\langle 2; (4; 2, 6), (5; 2) \rangle$
7	nine	$\langle 2; (3; 1), (6; 1) \rangle$
8	old	$\langle 2; (3; 3), (6; 3) \rangle$
9	pease	$\langle 2; (1; 1, 4), (2; 1) \rangle$
10	porridge	$\langle 2; (1; 2, 5), (2; 2) \rangle$
11	pot	$\langle 2; (2; 5), (5; 6) \rangle$
12	some	$\langle 2; (4; 1, 5), (5; 1) \rangle$
13	the	$\langle 2; (2; 4), (5; 5) \rangle$

Inverted Index - Granularity

Granularity is the precision to which our Inverted Index locates terms in our set of documents.

First index for "Pease porridge" documents - granularity is **document-level** (this is the **default** through this lecture).

Second Index for "Pease porridge" - granularity is **word-level** (very fine).

Granularity of Index will affect quality of query results.

Inverted Index - Lexicon

1. Set of **all words** that appear in the set of Documents? **OR**
2. Set of **given keywords** forming the allowed vocabulary for search?

Option 1 is most common.

all words is misleading - after parsing a document, we will do some lexical analysis to

- ▶ remove "stop words" (for WWW documents, may be many).
- ▶ perform **case folding** (upper case/lower case letters)
- ▶ perform **stemming**

Inverted Index - Querying

Each **term** has a **term number**.

The **inverted file entries** in the Inverted index are stored in order of term number (in our examples, alphabetical).

Queries:

- ▶ A single term, eg "pease":
Binary search in Inverted Index for term number of "pease" (given by lexicon). return the **file entry** for this.
- ▶ Boolean queries, eg "pease" AND "cold":
Binary search for each of the **file entries**. Then perform **merge-like** linear scan of these lists (\cap for AND, \cup for OR).

Memory-Based Inversion

The “obvious” method for Inversion.

Work entirely in memory, as we have always done (till now).

Dictionary data structure stores items of the form $(term, list)$, where *term* is a term of the lexicon, and *list* is a list of $\langle d, f_{d,t} \rangle$ (document, frequency of *t* in document) entries.

AVL tree is a good choice for dictionary *S*.

Phase 1: consider each document *d*, recovering terms, and appending an entry for each term *t* in *d* into the list for *t* in *S*.

Phase 2: Read off $\langle t, d, f_{d,t} \rangle$ terms in order from *S* and into the [inverted file](#).

Running Time

Officially, $T_I(D)$ is the sum of:

- ▶ $T_p(D)$ (for work in line 3 for all documents)
- ▶ $T_q(D)$ (time for lines 4-7 over all $\langle t, d \rangle$ terms in Index)
- ▶ $T_w(D)$ (time for the loop in lines 8-12, linear in size of inverted index)

But [asymptotic analysis](#) is not relevant here.

[Our scenario](#): pack as many Documents as possible into memory.

Memory-Based Inversion

Algorithm `memoryBasedInversion(D)`

1. Create a *Dictionary* data structure *S*.
2. **for** $i \leftarrow 1$ **to** $|D|$ **do**
3. Take document $d_i \in D$ and parse it into index terms.
4. **for** each index term *t* in d_i **do**
5. Let $f_{d_i,t}$ be the frequency of *t* in d_i .
6. If *t* is not in *S*, insert it.
7. Append $\langle d_i, f_{d_i,t} \rangle$ to *t*'s list in *S*.
8. **for** each term $1 \leq t \leq T$ **do**
9. Make a new entry in the *inverted file*.
10. **for** each $\langle d, f_{d,t} \rangle$ in *t*'s list in *S* **do**
11. Append $\langle d, f_{d,t} \rangle$ to *t*'s *inverted file* entry.
12. Append *t*'s entry to the *inverted file*.

Disk space instead of memory

Could we implement `Algorithm memoryBasedInversion(D)` to keep some Documents (and part of the Index) [on disk](#) during the algorithm's execution?

... so as to pack more into memory.

NO! (lines 8-12 are the problem - need to “hop around” the disk)

[Sort-Based Inversion](#) uses **merge** to merge small sorted [runs](#) on disk (not in memory).

Careful (Non-sequential) Disk accesses are very expensive. Use two disks *A* and *B*.

- ▶ In phase 1 disk *A* is for input, disk *B* for output.
- ▶ Roles are reversed with each phase.

external MergeSort

Algorithm externalMergeSort(A)

1. **for** $i = 1$ **to** n/K **do**
2. read block- i of disk-A (K items) into memory;
3. sort block- i in memory using 'in-place' algorithm, output it.
4. /* disk-B now becomes current input-disk */
5. **for** $j = 1$ **to** $\lceil \lg(n/K) \rceil$ **do**
6. **for** $i = 1$ **to** $(n/2^{j+1}K)$ **do**
7. buffer $K/3$ entries of block- i and block- $i + 1$ from
 current input-disk into memory;
8. initialize the output buffer b (of size $K/3$);
9. **while** there are items left to sort **do**
10. do externalMerge on small in-memory blocks
11. /* output buffer b if full, stream block- i and $i + 1$. */
12. swap role of *current input-disk* between A and B.

Algorithm sortBasedInversion(D)

1. Call externalMergeSort on *temp file*, to sort in order of $\langle \tau, d \rangle$;
2. /* *temp file* now sorted. Output inverted file. */
3. **for** $1 \leq \tau \leq T$ **do**
4. Start a new *inverted file entry* for t (term number τ).
5. Read the triples $\langle \tau, d, f_{d,\tau} \rangle$ from *temp file* into t 's entry.
6. Append t 's entry to the *inverted file*.

Note that memory size is K above.

Sort-Based Inversion

Algorithm sortBasedInversion(D)

1. Create a *Dictionary* data structure S .
2. Create an empty *temp file* on disk.
3. **for** $i \leftarrow 1$ **to** $|D|$ **do**
4. Take document $d_i \in D$ and parse it into index terms.
5. **for** each index term t in d_i **do**
6. Let $f_{d_i,t}$ be the frequency of t in d_i .
7. Check whether $t \in S$ (and check term number τ).
8. If $t \notin S$, insert it (with the next free term number τ).
9. Write $\langle \tau, d_i, f_{d_i,\tau} \rangle$ to *temp file* (τ is t 's term number).

Further Reading

Managing Gigabytes by Ian. H. Witten, Alistair Moffat, and Timothy. C. Bell (Chapter 5 and Chapter 3).
Witten et al. give numbers (in terms of [hours](#), [Gigabytes](#)).

Lots on the web:

- ▶ Wikipedia
- ▶ Building a distributed Full-text Index for the Web, by S. Melnik, S. Raghavan, B. Yang, and H. Garcia-Molina. *ACM Transactions on Information Systems (TOIS)*, **19**(3). Online at: <http://www10.org/cdrom/papers/275/>
- ▶ Very Large Scale Information Retrieval, by David Hawking. Online at: <http://www.inf.ed.ac.uk/teaching/courses/tts/papers>