

Re-pair compression of inverted indexes



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- Dealing with inverted lists
- Re-pair and inverted lists
 - Structure
 - Intersection algorithms
- Experimental Results
- Conclusions



Text Retrieval Scenario

- Very large text collections of data
 - Need of TR
 - Techniques aiming at:
 - Reducing space needs: compression
 - Improving efficiency of retrieval
- Searches
 - Online searches imply a sequential scan
 - Over compressed text? \rightarrow 0,3 · |T|
 - Indexed searches.
 - Inverted indexes, suffix arrays, ...

Inverted Indexes and their variants (for Nat. Lang.)



Indexed text

SPIRE 2008 is the 15th Annual Edition of the Symposium on string processing and information retrieval. SPIRE has its origins in the South American workshop on string processing which was first held in Belo Horizonte (Brazil, 1993). Starting in 1998, the focus of the workshop was broadened to include information retrieval due to its increasing relevance and its inter-relationship with the area of string processing. In addition, since 2000, SPIRE venue has been in Europe in even years.

Searches

Word \rightarrow fetching the posting of the word Phrase \rightarrow intersection of posting lists

Space-time trade-off

Granularity:

- Full-positional information of words
- Doc/Block-addressing

Compression

- Indexed text (+- 30% ratio)

·Huffman, Dense Coding...

- Posting lists!

Compression of Posting-Lists

- Compression usually rely on two main features
 - Postings lists contain increasing values
 - Gaps between them are smaller in the longer lists
 - Keep gaps instead of absolute values
 - Compress those gaps with a variable-length representation



Posting lists: Compression & Intersection

- But at the end... it is a typical case of trading space/time
 - Space
 - Fetching time \rightarrow a fast <u>decoding</u> algorithm is mandatory
 - Intersection time → <u>fast access</u> to the compressed representation is needed
- Re-pair
 - Is a grammar-based compression technique with:
 - Fast decompression.
 - Allows fast access to the compressed data (even in 2^{ary} memory)
 - Obtains good compression
 - We show that... with posting lists
 - It gives a competitive space/time tradeoff

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Dealing with inverted lists

Intersection algorithms

- Intersection of two inverted lists N and M
 - Merge-wise intersection
 - Traversing both lists in parallel
 - Best choice if both lists have similar length: $|N| \le 20|M|$
 - Can be done along with decoding
 - Set-vs-set approach
 - The elements of the smallest list are searched for in the longest list.
 - Different search options:
 - Sequential search
 - Binary search
 - Exponential search
 - Others...



Requires random access to the longest inverted list

Dealing with inverted lists

Data structures

- Variable-length encoding of gaps (golomb codes, bytecodes,...)
 - merge-type algorithms are still possible
 - <u>Algorithms that need random access</u> to elements require using a 2-level data structure
 - A top-level array indexes the compressed sequence in the bottom level
 - 2 different choices:
 - Sampling at regular intervals of the list [Moffat-Culpepper '07]
 Search is needed in the top-level array → buckets of the same size*
 - Sampling regularly at domain values [Sanders-Trasier '07]
 Values belong to a bucket depending on their most significative bits → buckets of different size



Compression of Posting-Lists + sampling

- Gap encoding
- Variable-length coding of gaps (golomb codes, bytecodes,...)
 - Avoiding full decompression \rightarrow 2-level structure.
 - [Moffat&Culpepper'07], [Sanders et al'07] ...



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Repair: The compression algorithm

- Steps:
 - Find the most frequent pair *ab* of symbols in L
 - Replace all the occurrences of ab by S in L
 - Add a rule $S \rightarrow$ ab to a dictionary R (**S** not appearing before)
 - Iterate until every pair in L appears only once



- The compressed sequence <u>**C**</u> keeps phrases:
 - Of length 1 if the symbol is a terminal one
 - Of length >1 for the new added symbols (phrase)

Repair: Compressing the dictionary

- Rules are represented as a set of trees and each tree as:
 - Rb = Bitmap representing the <u>Tree shape in preorder</u> {0=leaf, 1=internal}
 - The value of the i-th leaf in Rb is found at Rs[rank0(Rb,i)]
 - Non-terminals are shifted by μ to differentiate against terminals (μ = max terminal value)
 - Expanding non-terminals implies→
 - Traversing Rb and extract the leaf values until processing more 0's than 1's
 - Non-terminals are recursively expanded
 - Phrase is expanded in optimal time (time proportional to its length)
 - Rs = Sequence of <u>leaf nodes</u>
 - Non terminals are represented by the starting position of their tree in Rb



Example: Expanding ("A") $\rightarrow R_s[Rank_0(R_b, 2)] R_s[Rank_0(R_b, 3)] \rightarrow 1,2$

Application to inverted lists

- Differentially encode the inverted lists $<p_1, p_2, p_3, \dots, p_k > \rightarrow <p_1, p_2 p_1, p_3 p_2, \dots, p_k p_{k-1} >$
- Apply Re-pair to the concatenation of all the lists
 - Ensuring no phrase spans more than 1 list (use of artificial symbol)
 - Storing also Re-pair dictionary
 - Terminal symbols store themselves their differential value
 - Keeping a pointer of each vocabulary entry to its first occ in C



Searching: skipping data

- Some skipping data (for non-terminals) can be added to R_s
 - Avoids expanding some non-terminals
 - Adding data aligned with R_b
 - Rank₀ is no longer needed for expanding a symbol
 - So adds some data into R_b but saves "rank structures" → similar space requirements
 - Searching for a given value in a list:
 - Scanning of the list [summing values] until exceeding the value sought
 - If we reach a terminal \rightarrow we are done (add its value)
 - If we reach a non-terminal \rightarrow skipping data indicates if it has to be:
 - » expanded (currValue + skip_data > value sought) or,
 - » just skipped (currValue + skip_data < value sought)</p>



Searching: intersection algorithms proposed

- Intersection of 2 lists: ۲
 - We sort them by its uncompressed length (so that lenght is also kept)
 - Apply an intersection algorithm (currently svs+seq-search)
 - Using skipping data •
 - As show before.
 - Using skipping data PLUS

sampling at regular intervals in the sequence of phrases (C)

В

6

4

5

С

9

5

0

11

4

0

1

10

0

2

8

0

2

7



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Experimental Results

Framework used

- Corpus <u>FT91 to FT94</u>: TREC4
 - 495Mb English text (indexed in lowercase)
 - 210,138 documents (2.4Kb on average)
 - Doc-oriented index \rightarrow 50,285,802 entries
- Intel Core2duo[T8300]@2.4GHz, 4GB, 3MB-L2cache, Ubuntu, gcc (-09)
- Comparing.
 - Repair vs
 - ■[Sanders, Trasier, 2007 → lookup structure, parameter B]
 - ■[Moffat, Culpepper, 2007 → 2-levels, parameter K]
 - Merge-wise version
 - Bottom-layer → using bytecodes
- Showing...
 - Space needed by the structures
 - Intersection time of 2 lists

Experimental Results: Espace usage



| Method | Vocabulary | Extra data | Inverted lists | Total | |
|------------------|------------|-----------------|------------------|------------|----------------|
| repair | 2,699,656 | (dict) 594,467 | 50,586,760 | 53,880,883 | _ |
| (k=1) | 3,955,308 | + 1,606,640 | $50,\!586,\!760$ | 56,744,691 | |
| merge | 1,569,568 | | 59,406,108 | 60,975,676 | |
| bin, seq $(k=2)$ | 4,456,556 | 8,606,784 | 55,668,305 | 68,732,645 | |
| (k = 32) | 4,269,208 | $1,\!629,\!688$ | 57,912,204 | 63,811,100 | |
| $lookup \ (B=8)$ | 4,457,556 | 8,467,733 | 58,970,767 | 71,896,056 | $\overline{7}$ |
| (B = 64) | 4,269,208 | $1,\!170,\!400$ | $59,\!355,\!998$ | 64,795,606 | ∀ + |

- Repair with/without sampling (including skipping data)
 - Dictionary size is negligible → It fits in RAM
 - Size:
 - Around 10% of the original text
 - Around 25% of a representation of the inverted lists with 32-bit integers

Experimental Results: curiosities of lists lengths



The longer lists involve much more repetitions → they compress better

Experimental results: List intersection (doc size 2.4kb)

- Lookup → faster but more space
- Merge \rightarrow good if similar list lengths
- Svs → good choice. Improves results as lengths vary
- Repair→
 - good compression. Sampling good for x>120
 - Performing similar to bc+svs (with large sampling values)

Experimental results: List intersection (zoomed)

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Conclusions

- Re-pair on inverted lists (document-addressing)
 - Better compression than those techniques using bytecodes
 - More space for sampling can be wasted
 - Implicit skipping data
 - Good space/time tradeoff
 - Expecting good performance in 2^{ary} memory
 - Dictionary can be kept in RAM (it is very small)
- Future work
 - Trying other representations/search algorithms:: More experiments!!
 - Dealing with word-addressing indexes
 - New dictionary representation allowing improved descending of the parse tree → important for searches

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