An Experimental Study of Index Compression and DAAT Query Processing Methods

Antonio Mallia Michał Siedlaczek Torsten Suel

Department of Computer Science and Engineering Tandon School of Engineering New York University

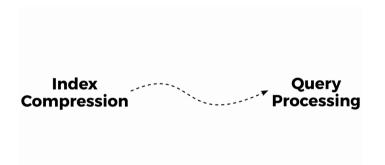
April 16th, 2019

Motivations

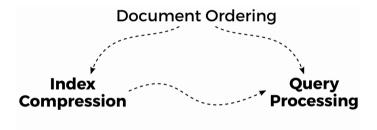


Index Compression Query Processing

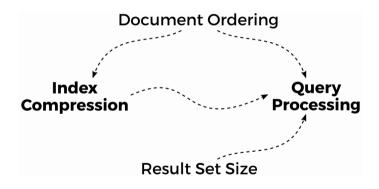












Main Contributions



- Confirmed some established results
- ▶ New important insights
- Modern and generic code base

Source Code

https://github.com/pisa-engine/pisa

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Index Compression



- Variable Byte Methods:
 - VarintGB [Dean 2009]
 - Varint-G8IU [Stepanov et al. 2011]
 - StreamVByte [Lemire et al. 2018]
- Word-Aligned Methods:
 - ▶ Simple16 [Zhang et al. 2008]
 - ► **Simple8b** [Anh and Moffat 2010]
 - ▶ SIMD-BP128 [Lemire and Boytsov 2015]
 - QMX [Trotman and Lin 2016]
- ▶ OptPForDelta [Yan et al. 2009]
- ▶ Partitioned Elias-Fano [Ottaviano and Venturini 2014]
- ▶ Binary Interpolative [Moffat and Stuiver 2000]
- ► Asymetric Numeral Systems [Moffat and Petri 2018]



Query Processing Algorithms



Top-k disjunctive Document-at-a-Time query processing algorithms with safe early-termination.

- ► MaxScore [Turtle and Flood 1995]
- ▶ **WAND** [Broder et al. 2003]
- ▶ Block-Max MaxScore [Chakrabarti et al. 2011]
- ▶ **Block-Max WAND** [Ding and Suel 2011]
- Variable Block-Max WAND [Mallia et al. 2017]

Document Ordering



The impact that document ID assignment has on index compression and query efficiency.

- ▶ Random baseline
- ▶ **URL** [Silvestri 2007]
- ► Recursive Graph Bisection (BP) [Dhulipala et al. 2016]

Result Set Size



- ▶ Typically small k in past top-k search studies
- Can be significantly larger for candidate retrieval for cascade ranking
- ► Recently shown that large *k* slow down retrieval [Crane et al. 2017]
- ► Thus, we experiment with values of *k* between 10 and 10,000

Experimental Setup

Implementation



Source Code

- ▶ https://github.com/pisa-engine/pisa
- ► Fork of ds2i: https://github.com/ot/ds2i

Third Party Libraries

- ▶ https://github.com/lemire/FastPFor
- ▶ https://github.com/andrewtrotman/JASSv2
- ▶ https://github.com/mpetri/partitioned_ef_ans

Testing Environment



- ► Implemented in C++17 and compiled with GCC 7.3 on highest optimization level
- ▶ Intel Core i7-4770 quad-core 3.40GHz CPU
- Haswell micro architecture supporting AVX2 instruction set
- ► CPUs L1, L2, and L3 cache sizes are 32KB, 256KB, and 8MB, respectively
- ▶ 32GiB RAM

Data Sets



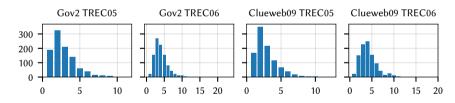
	Documents	Terms	Postings
GOV2	24,622,347	35,636,425	5,742,630,292
Clueweb09B	50,131,015	92,094,694	15,857,983,641

- ► HTML content parsed with Apache Tika
- ▶ Words stemmed with Porter2
- Stopwords kept

Queries



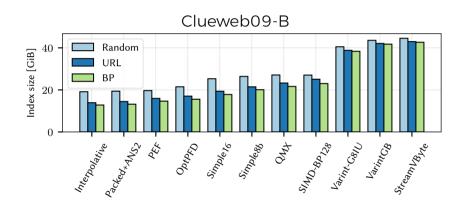
- ▶ TREC 2005 TREC 2006 from Terabyte Track Efficiency Task
- Queries with non-existent terms removed
- ► Initially sampled 1,000 queries for each query set and collection



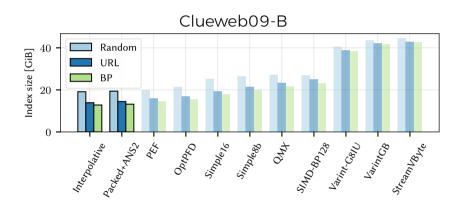
► Further sampled 1,000 queries for each query length from 2 to 6+

Results and Discussion

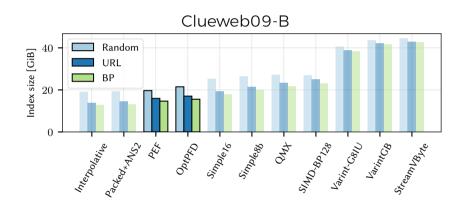




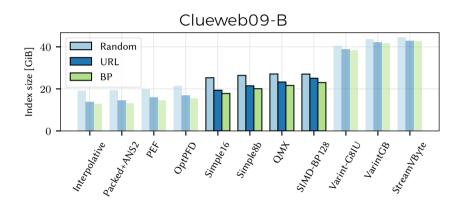




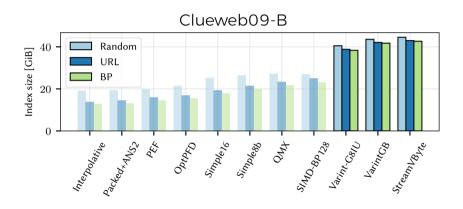






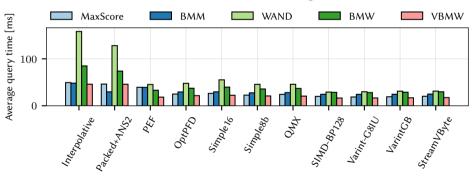






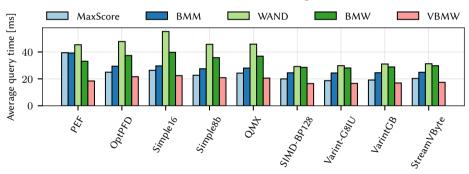






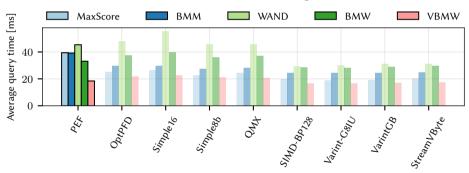






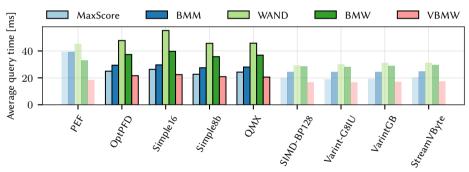






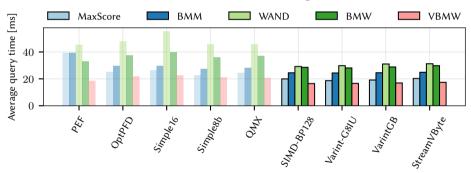








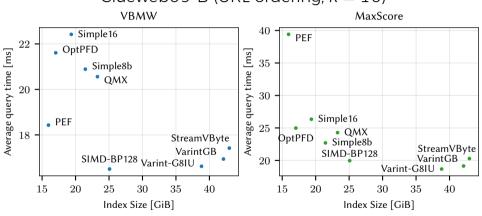




Query Speed v. Index Size

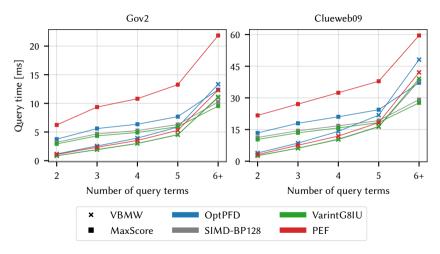


Clueweb09-B (URL ordering, k = 10)



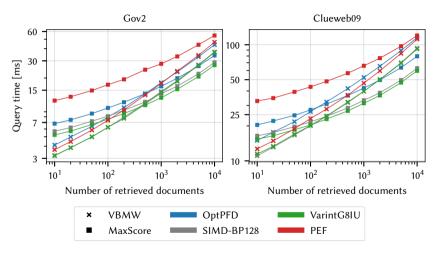
Query Length





Result Set Size





Conclusions



- Clear trade-off between speed and size
- Interesting compression insights
 - SIMD-BP128 matches speed of Varint methods while improving compression ratio
 - PEF's speed competitive when using VBMW
- Significant slowdown for large k
- MaxScore competitive with VBMW under certain circumstances
- Recursive Graph Bisection improves both compression and speed over URL ordering



Thank you for your time.

Any questions?

References I





Anh, V. N. and Moffat, A. (2010).

Index compression using 64-bit words.

Software: Practice and Experience, 40(2):131-147.



Broder, A. Z., Carmel, D., Herscovici, M., Soffer, A., and Zien, J. (2003).

Efficient query evaluation using a two-level retrieval process.

In Proc. of the 12th Intl. Conf. on Information and Knowledge Management, pages 426-434.



Chakrabarti, K., Chaudhuri, S., and Ganti, V. (2011).

Interval-based pruning for top-k processing over compressed lists.

In Proc. of the 2011 IEEE 27th Intl. Conf. on Data Engineering, pages 709-720.



Crane, M., Culpepper, J. S., Lin, J., Mackenzie, J., and Trotman, A. (2017).

A comparison of document-at-a-time and score-at-a-time query evaluation.

In Proc. of the 10th ACM Intl. Conf. on Web Search and Data Mining, pages 201-210.



Dean, J. (2009).

Challenges in building large-scale information retrieval systems: invited talk.

In Proc. of the 2nd ACM Intl. Conf. on Web Search and Data Mining, pages 1-1.



Dhulipala, L., Kabiljo, I., Karrer, B., Ottaviano, G., Pupyrev, S., and Shalita, A. (2016).

Compressing graphs and indexes with recursive graph bisection.

In Proceedings of the 22Nd ACM SIGKDD International Conference on Knowledge Discovery and Data Mining, pages 1535–1544.

References II





Ding, S. and Suel, T. (2011).

Faster top-k document retrieval using block-max indexes.

In Proc. of the 34th Ann. Intl. ACM SIGIR Conf. on Research and Development in Information Retrieval, pages 993-1002.



Lemire, D. and Boytsov, L. (2015).

Decoding billions of integers per second through vectorization.

Softw. Pract. Exper., 45(1):1-29.



Lemire, D., Kurz, N., and Rupp, C. (2018).

Stream vbyte: Faster byte-oriented integer compression. *Information Processing Letters*, 130:1–6.



Mallia, A., Ottaviano, G., Porciani, E., Tonellotto, N., and Venturini, R. (2017).

Faster blockmax WAND with variable-sized blocks.

In Proc. of the 40th Ann. Intl. ACM SIGIR Conf. on Research and Development in Information Retrieval, pages 625–634.



Moffat, A. and Petri, M. (2018).

Index compression using byte-aligned ANS coding and two-dimensional contexts.

In Proc. of the 11th ACM Intl. Conf. on Web Search and Data Mining, pages 405–413.

Moffat, A. and Stuiver, L. (2000).

Binary interpolative coding for effective index compression.

Inf. Retr., 3(1):25-47.

References III





Ottaviano, G. and Venturini, R. (2014).

Partitioned elias-fano indexes.

In Proc. of the 37th Ann. Intl. ACM SIGIR Conf. on Research and Development in Information Retrieval, pages 273–282



Silvestri, F. (2007).

Sorting out the document identifier assignment problem.

In Proc. of the 29th European Conf. on IR Research, pages 101-112.



Stepanov, A. A., Gangolli, A. R., Rose, D. E., Ernst, R. J., and Oberoi, P. S. (2011).

SIMD-based decoding of posting lists.

In Proc. of the 20th Intl. Conf. on Information and Knowledge Management, pages 317-326.



Trotman, A. and Lin, J. (2016).

In vacuo and in situ evaluation of SIMD codecs.

In Proc. of the 21st Australasian Document Computing Symposium, pages 1-8.



Turtle, H. and Flood, J. (1995).

Query evaluation: Strategies and optimizations.

Information Processina & Management, 31(6):831-850.



Yan, H., Ding, S., and Suel, T. (2009).

Inverted index compression and query processing with optimized document ordering.

In Proc. of the 18th Intl. Conf. on World Wide Web, pages 401-410.

References IV





Zhang, J., Long, X., and Suel, T. (2008).

Performance of compressed inverted list caching in search engines. In Proc. of the 17th Intl. Conf. on World Wide Web, pages 387–396.