### **ENGINEERING FAST INDEXES**

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Joint work with lots of super smart people



# Our recent work: Roaring Bitmaps

http://roaringbitmap.org/

#### Used by

- Apache Spark,
- Netflix Atlas,
- LinkedIn Pinot,
- Apache Lucene,
- Whoosh,
- Metamarket's Druid
- eBay's Apache Kylin

#### Further reading:

 Frame of Reference and Roaring Bitmaps (at Elastic, the company behind Elasticsearch)



#### Set data structures

We focus on sets of integers:  $S=\{1,2,3,1000\}$ . Ubiquitous in database or search engines.

- tests:  $x \in S$ ?
- intersections:  $S_2 \cap S_1$
- unions:  $S_2 \cup S_1$
- ullet differences:  $S_2\setminus S_1$
- Jaccard Index (Tanimoto similarity)  $|S_1 \cap S_1|/|S_1 \cup S_2|$



#### "Ordered" Set

- iterate
  - o in sorted order,
  - in reverse order,
  - $\circ$  skippable iterators (jump to first value  $\geq x$ )
- Rank: how many elements of the set are smaller than k?
- Select: find the k<sup>th</sup> smallest value
- Min/max: find the maximal and minimal value



## Let us make some assumptions...

- Many sets containing more than a few integers
- Integers span a wide range (e.g., [0, 100000))
- Mostly immutable (read often, write rarely)



## How do we implement integer sets?

Assume sets are *mostly* imutable.

```
sorted arrays ( std::vector<uint32_t> )
```

```
hash sets (java.util.HashSet<Integer>,
std::unordered_set<uint32_t>)
```

- . . .
- bitsets ( java.util.BitSet )
- ♥ ♥ compressed bitsets ♥ ♥



#### What is a bitset???

Efficient way to represent a set of integers.

E.g., 0, 1, 3, 4 becomes 0b11011 or "27".

Also called a "bitmap" or a "bit array".



#### Add and contains on bitset

Most of the processors work on 64-bit words.

Given index  $\times$ , the corresponding word index is  $\times/64$  and withinword bit index is  $\times\%64$ .

```
add(x) {
  array[x / 64] |= (1 << (x % 64))
}

contains(x) {
  return array[x / 64] & (1 << (x % 64))
}</pre>
```



## How fast can you set bits in a bitset?

Very fast! Roughly three instructions (on x64)...

(Or can use BMI's bts.)

On recent x64 can set one bit every pprox 1.65 cycles (in cache)

Recall: Modern processors are superscalar (more than one instruction per cycle)



### Bit-level parallelism

Bitsets are efficient: intersections

Intersection between {0, 1, 3} and {1, 3} can be computed as AND operation between 0b1011 and 0b1010.

Result is 0b1010 or {1, 3}.

Enables Branchless processing.



## Bitsets are efficient: in practice

```
for i in [0...n]
  out[i] = A[i] & B[i]
```

Recent x64 processors can do this at a speed of  $\approx 0.5$  cycles per pair of input 64-bit words (in cache) for n=1024.

0.5

memcpy runs at pprox 0.3 cycles.

0.3



### Bitsets can be inefficient

Relatively wasteful to represent {1, 32000, 64000} with a bitset. Would use 1000 bytes to store 3 numbers.

So we use compression...



# Memory usage example

dataset : census1881\_srt

format	bits per value
hash sets	200
arrays	32
bitsets	900
compressed bitsets (Roaring)	2



# Performance example (unions)

dataset : census1881\_srt

format	CPU cycles per value
hash sets	200
arrays	6
bitsets	30
compressed bitsets (Roaring)	1



# What is happening? (Bitsets)

Bitsets are often best... except if data is very sparse (lots of 0s). Then you spend a lot of time scanning zeros.

- Large memory usage
- Bad performance

Threshold? ~1:100



# Hash sets are not always fast

Hash sets have great one-value look-up. But they have poor data locality and non-trivial overhead...

```
h1 <- some hash set
h2 <- some hash set
...
for(x in h1) {
   insert x in h2 // "sure" to hit a new cache line!!!!
}</pre>
```



#### Want to kill Swift?

Swift is Apple's new language. Try this:

```
var d = Set<Int>()
for i in 1...size {
   d.insert(i)
}
//
var z = Set<Int>()
for i in d {
    z.insert(i)
}
```

This blows up! Quadratic-time.

Same problem with Rust.



# What is happening? (Arrays)

Arrays are your friends. Reliable. Simple. Economical.

But... binary search is *branchy* and has *bad locality*...

```
while (low <= high) {</pre>
  int middleIndex = (low + high) >>> 1;
  int middleValue = array.get(middleIndex);
  if (middleValue < ikey) {</pre>
    low = middleIndex + 1;
  } else if (middleValue > ikey) {
    high = middleIndex - 1;
  } else {
    return middleIndex;
return -(low + 1);
```



# Performance: value lookups ( $x \in S$ )

dataset: weather\_sept\_85

format	CPU cycles per query
hash sets ( std::unordered_set )	
arrays	900
bitsets	4
compressed bitsets (Roaring)	80



## How do you compress bitsets?

- We have long runs of 0s or 1s.
- Use run-length encoding (RLE)

Example: 0000000111111111100 can be coded as

$$00000000 - 111111111 - 00$$

or

using the format < number of repetitions >< value being repeated >



### **RLE-compressed bitsets**

- Oracle's BBC
- WAH (FastBit)
- EWAH (Git + Apache Hive)
- Concise (Druid)
- . . .

#### Further reading:

http://githubengineering.com/counting-objects/



# **Hybrid Model**

Decompose 32-bit space into 16-bit spaces (chunk).

Given value x, its chunk index is  $x \div 2^{16}$  (16 most significant bits).

For each chunk, use best container to store least 16 significant bits:

- a sorted array ({1,20,144})
- a bitset (0b10000101011)
- a sequences of sorted runs ([0,10],[15,20])

That's Roaring!

Prior work: O'Neil's RIDBit + BitMagic



# Roaring

- All containers fit in 8 kB (several fit in L1 cache)
- Attempts to select the best container as you build the bitmaps
- Calling run0ptimize will scan (quickly!) non-run containers and try to convert them to run containers



# Performance: union (weather\_sept\_85)

format	CPU cycles per value
bitsets	0.6
WAH	4
EWAH	2
Concise	5
Roaring	0.6



### What helps us...

- All modern processors have fast population-count functions
   (popcnt) to count the number of 1s in a word.
- Cheap to keep track of the number of values stored in a bitset!
- Choice between array, run and bitset covers many use cases!



# Go try it out!

- Java, Go, C, C++, C#, Rust, Python... (soon: Swift)
- http://roaringbitmap.org
- Documented interoperable serialized format.
- Free. Well-tested. Benchmarked.
- Peer reviewed
  - Consistently faster and smaller compressed bitmaps with Roaring. Softw., Pract. Exper. (2016)
  - Better bitmap performance with Roaring bitmaps. Softw.,
     Pract. Exper. (2016)
  - Optimizing Druid with Roaring bitmaps, IDEAS 2016, 2016
- Wide community (dozens of contributors).