## Information Retrieval WS 2015 / 2016

Lecture 3, Tuesday November 3<sup>rd</sup>, 2015 (Efficient List Intersection)

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### Overview of this lecture

#### Organizational

- Your experiences with ES 2
- Using built-in functions
- Contents
  - List Intersection
     Recap, Time Measurement
  - Non-algorithmic improvements Arrays, Branching, Sentinels
  - Algorithmic improvements
     Galloping Search, Skip Pointers

Ranking

Beware !!

**Exercise Sheet 3:** implement list intersection and make it as fast as possible on a small benchmark we have prepared

# 

#### Summary / excerpts

Time-intensive for many, mainly due to debugging
 This should become (much) better with experience

- Parameter tuning: waiting long for each index build
   Some problems with built-in functions ... see slide 5
- Master solutions for ES1 would have been nice

Always available in the course SVN under /solutions

- Feedback from the tutors very much appreciated
- In Python, 4-space indent + 80-char limit is annoying
   I agree, but it's absolutely standard in Python world

## Experiences with ES2 2/4

#### Results

Standard BM25 parameters gave sub-optimal results
 Smaller **b** worked better (less penalty for longer docs)
 Smaller **k** worked better (less boost for larger tf)

- Boosting popular documents helped a bit
- Boosting matches in title helped a bit
- Boost matches of most or all query words helped a bit
- Best results: P@3 ≈ 60%, P@R ≈ 40%, MAP ≈ 40%
   The last two results are typical: it's extremely hard to get most or even all relevant documents at the top
   For better P@3, sth like synonyms or query logs needed

Experiences with ES2 3/4

- Built-in functions / library functions
  - Using them is OK, if and only if:
    - You are aware of the complexity of the function
    - You are aware of the complexity of the code using them

- That complexity is ok for the task at hand
- Not doing this is one of the most common reasons for performance leaks in software

- Built-in functions / library functions
  - Example 1: merging two lists by concatenating them and then sorting the concatenated list

Takes time n · log n, versus linear time for "zipper" alg

 Example 2: use "in" or "find" to locate an element in a list, and doing this n times

Each call to "in" or "find" uses linear time, which gives quadratic time overall  $\rightarrow$  terrible running time

 Example 3: use std::set\_intersection to implement a linear-time intersect

Ok, provided that you convinced yourself that this works only on sorted lists and runs in linear time

#### Motivation (recap)

- In Lectures 1 & 2 we have merged the inverted lists
  - So that we also have a chance to find relevant docs that do not contain all of the query words
- For efficiency reasons, many search engines only return results which contain all the query words
  - Apache's Lucene, the most widely used open-source search engine, supports intersect (AND) and merge (OR)
  - In most applications, intersect is used by default
- Today we will focus on efficiency and therefore on list intersection

List Intersection 2/4

#### Time measurement

There can be significant variation, for example due to:
 Other jobs running on your machine
 The Java garbage collector running unpredictably
 Data is partly in disk cache / L1-cache / TBL cache

 Therefore, always repeat your time measurements, and take the average over all these

For ES3, repeat 10 times for each measurement

Note: repetition itself can also distort the truth because of caching effects ... but not an issue for us today List Intersection 3/4

#### Time measurement in **Java**

- For millisecond resolution

long time1 = System.currentTimeMillis();
// whatever code you want to time
long time2 = System.currentTimeMillis();
long millis = time2 - time1;

- For microsecond resolution

long time1 = System.nanoTime();
// whatever code you want to time
long time2 = System.nanoTime();
long micros = (time2 - time1) / 1000;

List Intersection 4/4

■ Time measurement in **C++** 

– For millisecond resolution (C-Style) #include <time.h>

clock\_t time1 = clock();

// whatever code you want to time
clock\_t time2 = clock();

size\_t millis = 1000 \* (time2 - time1) / CLOCKS\_PER\_SEC;

- For microsecond resolution (C++11) #include <chrono>
auto time1 = high\_resolution\_clock::now();
// whatever code you want to time
auto time2 = high\_resolution\_clock::now();
size t micros = duration\_cock::now();

size\_t micros = duration\_cast<microseconds>(...).count();

#### Native arrays

 Java: ArrayList much worse than native [] array
 Elements of an ArrayList cannot be basic data types (e.g. int), but have to be objects (e.g. Integer)
 This causes inefficient byte code / machine code - C++: std::vector is as good as [] with option –O3

Elements of an std::vector can be basic data types as well as objects

Due to C++'s templating mechanism, machine code for std::vector<int> is almost the same as for int[] Non-algorithmic improvements 2/3

#### Predictable branches

Branches = all conditional parts in your code

In particular, **if ... then ... else** parts

 Modern processors do pipelining = speculative execution of future instructions before the current ones are done 

- For conditional parts they have to guess the outcome
- So good to **minimize** amount of conditional parts

#### Sentinels

Special elements to avoid testing for index out of bound
 Less code + further reduction in number of branches

ZW

- For list intersection: id  $\infty$  at the end of both lists
  - For Java, take: Integer.MAX\_VALUE
  - For C++, take: std::numeric\_limits<int>::max()

## Algorithmic improvements 1/8

#### Preliminaries

- We have to two lists, which we want to intersect

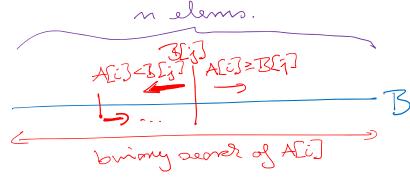
- Let **A** be the smaller list, with **k** elements
- Let **B** be the longer list, with **n** elements

List intersection is commutative, so we can always assume that the first list is A, and the second is B

- Binary search in the longer list
  - Search each element from A in B, using binary search
  - This has time complexity  $\Theta(k \cdot \log n)$ 
    - Good for small k ... but for  $k = \Theta(n)$  this is  $\Theta(n \cdot \log n)$ , and hence slower than the "zipper"-style linear intersect



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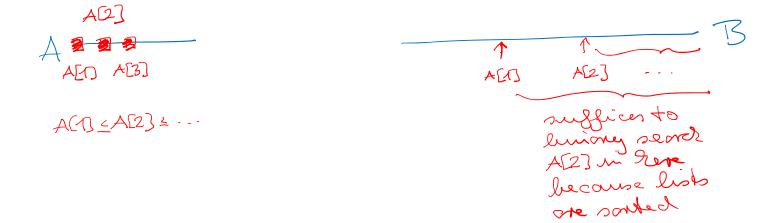
Algorithmic improvements 3/8

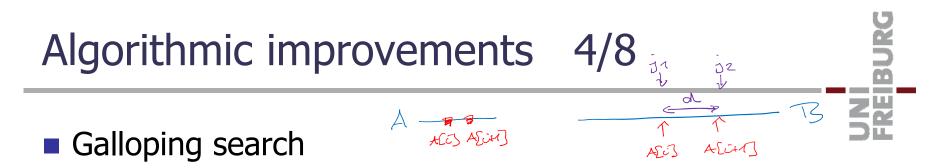
Binary search in remainder of longer list

- Time complexity in the best case  $\Theta(k + \log n)$ First element from A at the end of list B 

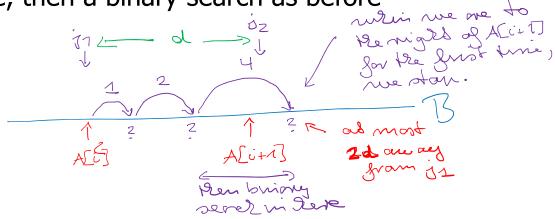
- Time complexity in the worst case  $\Theta(k \cdot \log n)$ All elements of A at the beginning of list B
- Time complexity in the "typical" case  $\Theta(k \cdot \log n)$

Elements of A "evenly distributed" over list B





- Goal: when elements A[i] and A[i+1] are located at positions j<sub>1</sub> and j<sub>2</sub> in B, then, with d:= j<sub>2</sub> j<sub>1</sub> ("gap"): spend only time O(log d) to locate element A[i+1]
- Idea: first do an exponential search, to get an upper bound on the range, then a binary search as before



## Algorithmic improvements 5/8

Galloping search, time complexity

- Let  $j_1, ..., j_k$  the positions of the elements of A in B
- Let  $d_i = j_i j_{i-1}$  for i > 1 and  $d_i = 1$  (the "gaps")

Note that  $\Sigma_i d_i \le n =$  the number of elements in B

– Then the time complexity is  $O(\Sigma_i \log d_i)$ 

Not a nice formula, so let's find the maximum value, independent of the particular  $d_1, ..., d_k$ 

- Lemma:  $\Sigma_i \log d_i$  is maximized when all  $d_i = n / k$
- Galloping search therefore takes time  $O(k \cdot \log (1 + n/k))$ this is always O(n)

**Proof of Lemma** ... max  $\Sigma_i \log d_i$  under constraint  $\Sigma_i d_i \le n$ 

- This is an instance of Lagrangian optimization:
  - 1. Write constraint as equation:  $\Sigma_i d_i n' = 0 \dots n' < n$

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- 2. Define  $\mathbf{L}(\mathbf{d_{1\prime}}, ..., \mathbf{d_{k\prime}}, \boldsymbol{\lambda}) = \Sigma_i \log d_i + \lambda \cdot (\Sigma_i d_i n')$
- 3. Set partial derivatives = 0 to find all local optima and check the objective function at the borders

Algorithmic improvements 7/8

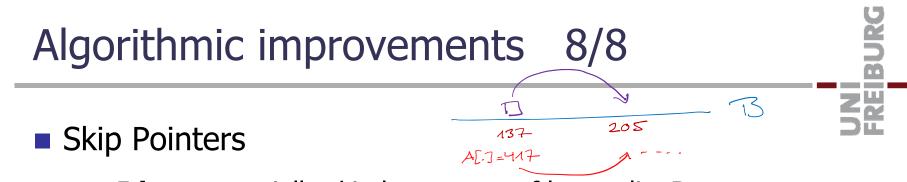
#### Comparison-based lower bound

- Recall the lower-bound for comparison-based sorting There are n! possible outputs, we have to differentiate between all of them, and only two choices per step Hence #steps required  $\geq \log_2(n!) = \Omega(n \cdot \log n)$
- We can use a similar argument for intersection / union:

There are n+k over k ways how the k elements from A can be placed within the n elements from B, ...

Hence #steps required  $\geq \log_2 (n/k)^k = \mathbf{k} \cdot \log_2 (n/k)$ 

Galloping search is hence asymptotically optimal



- Idea: potentially skip large parts of longer list B
- Skip pointer = special element in list B that points to an element B[j] further to the right

When intersecting, follow pointer if current  $A[i] \ge B[j]$ 

Placement of skip pointers is heuristic ... for ES3 you can investigate good placements experimentally

– Advantage: **very simple** to implement

In particular, simpler than galloping search and thus often more effective in practice, even if not "optimal"

## References

#### Textbook

Section 2.3: Faster intersection with skip pointers

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#### Literature

A simple algorithm for merging two linearly ordered setsF.K. Hwang and S. LinSICOMP 1(1):31–39, 1980A fast set intersection algorithm for sorted sequencesR. Baeza-YatesCPM, LNCS 3109, 31–39, 2004

#### Wikipedia

http://en.wikipedia.org/wiki/Lagrange multiplier