

# Text Indexing

## Lecture 10: Inverted Index

Florian Kurpicz

The slides are licensed under a Creative Commons Attribution-ShareAlike 4.0 International License © ⓘ ⓘ: [www.creativecommons.org/licenses/by-sa/4.0](http://www.creativecommons.org/licenses/by-sa/4.0) | commit 224e27c compiled at 2022-01-10-09:33



<https://pingo.scc.kit.edu/569917>

# The Inverted Index

## Definition: Inverted Index

Given a set of documents and terms that are contained in the documents, an inverted index stores the terms and associated with each term  $t$

- the number of documents  $f_t$  that contain  $t$  and
- an ordered list  $L(t)$  of documents containing  $t$

- 1 The old night keeper keeps the keep in the town
- 2 In the big old house in the big old gown
- 3 The house in the town had the big old keep
- 4 Where the old night keeper never did sleep
- 5 The night keeper keeps the keep in the night
- 6 And keeps in the dark and sleeps in the light

# The Inverted Index

## Definition: Inverted Index

Given a set of documents and terms that are contained in the documents, an inverted index stores the terms and associated with each term  $t$

- the number of documents  $f_t$  that contain  $t$  and
- an ordered list  $L(t)$  of documents containing  $t$

- 1 The old night keeper keeps the keep in the town
- 2 In the big old house in the big old gown
- 3 The house in the town had the big old keep
- 4 Where the old night keeper never did sleep
- 5 The night keeper keeps the keep in the night
- 6 And keeps in the dark and sleeps in the light

term $t$	$f_t$	$L(t)$
and	1	[6]
big	2	[2, 3]
dark	1	[6]
...	...	...
had	1	[3]
house	2	[2, 3]
in	5	[1, 2, 3, 5, 6]
...	...	...

# The Inverted Index

## Definition: Inverted Index

Given a set of documents and terms that are contained in the documents, an inverted index stores the terms and associated with each term  $t$

- the number of documents  $f_t$  that contain  $t$  and
- an ordered list  $L(t)$  of documents containing  $t$

- 1 The old night keeper keeps the keep **in** the town
- 2 **In** the **big** old **house** **in** the **big** old gown
- 3 The **house** **in** the town **had** the **big** old keep
- 4 Where the old night keeper never did sleep
- 5 The night keeper keeps the keep **in** the night
- 6 **And** keeps **in** the **dark** **and** sleeps **in** the light

term $t$	$f_t$	$L(t)$
<b>and</b>	1	[6]
<b>big</b>	2	[2, 3]
<b>dark</b>	1	[6]
...	...	...
<b>had</b>	1	[3]
<b>house</b>	2	[2, 3]
<b>in</b>	5	[1, 2, 3, 5, 6]
...	...	...

# The Inverted Index: Queries

## Conjunctive Queries

- Given two lists  $M$  and  $N$ , return all documents contained in both lists:  $M \cap N$

1 The old night keeper keeps the keep in the town  
2 In the big old house in the big old gown  
3 The house in the town had the big old keep  
4 Where the old night keeper never did sleep  
5 The night keeper keeps the keep in the night  
6 And keeps in the dark and sleeps in the light

# The Inverted Index: Queries

## Conjunctive Queries

- Given two lists  $M$  and  $N$ , return all documents contained in both lists:  $M \cap N$

## Disjunctive Queries

- Given two lists  $M$  and  $N$ , return all documents contained in either list:  $M \cup N$

- The old night keeper keeps the keep **in** the town
- In** the **big** old **house** **in** the **big** old gown
- The **house** **in** the town **had** the **big** old keep
- Where the old night keeper never did sleep
- The night keeper keeps the keep **in** the night
- And** keeps **in** the **dark** **and** sleeps **in** the light

term $t$	$f_t$	$L(t)$
<b>and</b>	1	[6]
<b>big</b>	2	[2, 3]
<b>dark</b>	1	[6]
...	...	...
<b>had</b>	1	[3]
<b>house</b>	2	[2, 3]
<b>in</b>	5	[1, 2, 3, 5, 6]
...	...	...

# The Inverted Index: Queries


## Conjunctive Queries

- Given two lists  $M$  and  $N$ , return all documents contained in both lists:  $M \cap N$

## Disjunctive Queries

- Given two lists  $M$  and  $N$ , return all documents contained in either list:  $M \cup N$

## Phrase Queries

- Given two terms  $t_1$  and  $t_2$ , return all documents containing  $t_1 t_2$   all previous discussed indices can do so

- The old night keeper keeps the keep **in** the town
- In** the **big** old **house** **in** the **big** old gown
- The **house** **in** the town **had** the **big** old keep
- Where the old night keeper never did sleep
- The night keeper keeps the keep **in** the night
- And** keeps **in** the **dark** **and** sleeps **in** the light

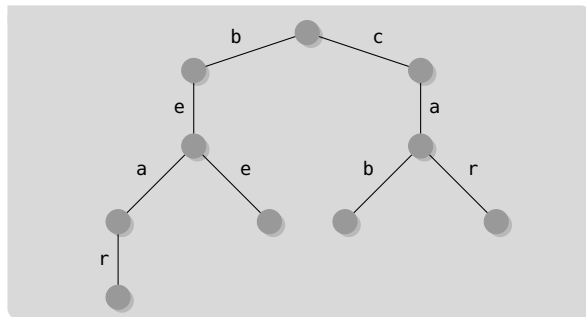
term $t$	$f_t$	$L(t)$
<b>and</b>	1	[6]
<b>big</b>	2	[2, 3]
<b>dark</b>	1	[6]
...	...	...
<b>had</b>	1	[3]
<b>house</b>	2	[2, 3]
<b>in</b>	5	[1, 2, 3, 5, 6]
...	...	...



# Inverted Index: Representing the Terms (1/2)

- terms can be represented using tries
- in each leaf, store pointer to list for term

- simple representation
- easy to add and remove terms



## Inverted Index: Representing the Words (2/2)

- use multiplicative hash function
  - $h(t[1] \dots t[\ell]) = ((\sum_{i=1}^{\ell} a_i \cdot t[i]) \bmod p) \bmod m$
  - for prime  $p < m$  and
  - fixed random integers  $a_i \in [1, p]$
- 
- good worst case guarantee
  - $\text{Prob}[h(x) = h(y)] = O(1/m)$  for  $x \neq y$

# Inverted Index: Document Lists

- document ids are sorted
- if ids are in  $[1, U]$ , storing them requires  $\lceil \lg U \rceil$  bits per id

## Binary Codes

- an integer  $x$  can be represented as binary  $(x)_2$
- for fast access, all binary representations must have the same width

## Now

- different ideas on how to better store ids
- not all ideas work with all algorithms
- different space usage and complexity

# Difference Encoding

- given a document list  $N = [d_1, \dots, d_{|N|}]$
- the document ids are sorted:  $d_1 < \dots < d_{|N|}$
- store first id
- represent other ids by difference:  $\delta_i = d_i - d_{i-1}$

## Definition: $\Delta$ -Encoding

A  **$\Delta$ -encoded** document list  $N = [d_1, \dots, d_{|N|}]$  is  
 $N = [d_1, d_2 - d_1, \dots, d_{|N|} - d_{|N|-1}]$

# Difference Encoding

- given a document list  $N = [d_1, \dots, d_{|N|}]$
- the document ids are sorted:  $d_1 < \dots < d_{|N|}$
- store first id
- represent other ids by difference:  $\delta_i = d_i - d_{i-1}$

## Definition: $\Delta$ -Encoding

A  **$\Delta$ -encoded** document list  $N = [d_1, \dots, d_{|N|}]$  is  $N = [d_1, d_2 - d_1, \dots, d_{|N|} - d_{|N|-1}]$

Just ids:

- $N = [4, 11, 12, 30, 42, 54]$

$\Delta$ -encoded

- $N = [4, 7, 1, 18, 12, 12]$

# Difference Encoding

- given a document list  $N = [d_1, \dots, d_{|N|}]$
- the document ids are sorted:  $d_1 < \dots < d_{|N|}$
- store first id
- represent other ids by difference:  $\delta_i = d_i - d_{i-1}$

## Definition: $\Delta$ -Encoding

A  **$\Delta$ -encoded** document list  $N = [d_1, \dots, d_{|N|}]$  is  
 $N = [d_1, d_2 - d_1, \dots, d_{|N|} - d_{|N|-1}]$

- can this be compressed further?
- accessing id requires scanning

Just ids:

- $N = [4, 11, 12, 30, 42, 54]$

$\Delta$ -encoded

- $N = [4, 7, 1, 18, 12, 12]$

# Unary Encoding

## Definition: Unary Codes

Given an integer  $x > 0$ , its unary code  $(x)_1$  is  $1^{x-1}0$

- $|(x)_1| = x$  bits
- encoded integers can be accessed using rank and select queries
- if 0 has to be encoded, all codes require an additional bit

# Unary Encoding

## Definition: Unary Codes

Given an integer  $x > 0$ , its unary code  $(x)_1$  is  $1^{x-1}0$

- $|(x)_1| = x$  bits
- encoded integers can be accessed using rank and select queries
- if 0 has to be encoded, all codes require an additional bit

Just ids:

- $N = [4, 11, 12, 30, 42, 54]$

$\Delta$ -encoded

- $N = [4, 7, 1, 18, 12, 12]$

Unary Codes:

- $N = [1110111111001^{17}01^{11}0111111111110]$



# Ternary Encoding

## Definition: Ternary Codes

Given an integer  $x > 0$ , represent it in ternary using

- 00 to represent 0
- 01 to represent 1
- 10 to represent 2

and append 11 to each code to obtain its ternary code  $(x)_3$

- $|(x)_3| = 2 \lfloor \lg_3(x - 1) \rfloor + 2$

# Ternary Encoding

## Definition: Ternary Codes

Given an integer  $x > 0$ , represent it in ternary using

- 00 to represent 0
- 01 to represent 1
- 10 to represent 2

and append 11 to each code to obtain its ternary code  $(x)_3$

$$\lfloor (x)_3 \rfloor = 2 \lfloor \lg_3(x - 1) \rfloor + 2$$

Just ids:

- $N = [4, 11, 12, 30, 42, 54]$

$\Delta$ -encoded

- $N = [4, 7, 1, 18, 12, 12]$

Unary Codes:

- $N = [1110111111001^{17}01^{11}0111111111110]$

Ternary Codes:

- $N = [100011\ 100011\ 01101011\ 01001011\ 01001011]$

# Fibonacci Encoding

## Lemma: Zeckendorf's Theorem

Let  $f_i$  be the  $i$ -th Fibonacci number, then each integer  $x > 0$  can be represented as

$$n = \sum_{i=2}^k c_i f_i$$

with  $c_i \in \{0, 1\}$  and  $c_i + c_{i+1} < 2$

## Definition: Fibonacci Code

Given an integer  $x > 0$  use the sequence of  $c_i$ 's followed by a 1 as its Fibonacci code  $(x)_\phi$

# Fibonacci Encoding

## Lemma: Zeckendorf's Theorem

Let  $f_i$  be the  $i$ -th Fibonacci number, then each integer  $x > 0$  can be represented as

$$n = \sum_{i=2}^k c_i f_i$$

with  $c_i \in \{0, 1\}$  and  $c_i + c_{i+1} < 2$

## Definition: Fibonacci Code

Given an integer  $x > 0$  use the sequence of  $c_i$ 's followed by a 1 as its Fibonacci code  $(x)_\phi$

- 11 does not occur in any sequence
- to compute find largest Fibonacci number  $f_i < x$  and repeat process for  $x - f_i$
- Fibonacci codes are smaller than ternary codes for smaller integers

# Fibonacci Encoding

## Lemma: Zeckendorf's Theorem

Let  $f_i$  be the  $i$ -th Fibonacci number, then each integer  $x > 0$  can be represented as

$$n = \sum_{i=2}^k c_i f_i$$

with  $c_i \in \{0, 1\}$  and  $c_i + c_{i+1} < 2$

## Definition: Fibonacci Code

Given an integer  $x > 0$  use the sequence of  $c_i$ 's followed by a 1 as its Fibonacci code  $(x)_\phi$

- 11 does not occur in any sequence
- to compute find largest Fibonacci number  $f_i < x$  and repeat process for  $x - f_i$
- Fibonacci codes are smaller than ternary codes for smaller integers

- $f_2 = 1, f_3 = 2, f_4 = 3, f_5 = 5, f_6 = 8, f_7 = 13$
- 4:  $f_2 + f_4 = 1011$
- 7:  $f_3 + f_5 = 01011$
- 1:  $f_2 = 11$
- 18:  $f_5 + f_7 = 0001011$
- 12:  $f_2 + f_4 + f_6 = 101011$

## Elias- $\gamma$ -Encoding [Eli75]

### Definition: Elias- $\gamma$ -Code

Given an integer  $x > 0$ , its Elias-*gamma*-code  $(x)_\gamma$  is

$$(x)_\gamma = 0^{\lfloor \lg x \rfloor} (x)_2$$

- $|(x)_\gamma| = 2 \lfloor \lg x \rfloor + 1$  bit
- first part gives length of binary representation
- first bit of  $(x)_2$  is one bit

# Elias- $\gamma$ -Encoding [Eli75]

## Definition: Elias- $\gamma$ -Code

Given an integer  $x > 0$ , its Elias-*gamma*-code  $(x)_\gamma$  is

$$(x)_\gamma = 0^{\lfloor \lg x \rfloor} (x)_2$$

- $|(x)_\gamma| = 2 \lfloor \lg x \rfloor + 1$  bit
- first part gives length of binary representation
- first bit of  $(x)_2$  is one bit

- 4: 00 100
- 7: 00 111
- 1: 1
- 18: 0000 10010
- 12: 000 1000

# Elias- $\delta$ -Encoding [Eli75]

## Definition: Elias- $\delta$ -Code

Given an integer  $x > 0$ , its Elias- $\delta$ -code  $(x)_\delta$  is

$$(x)_\delta = (\lfloor \lg x \rfloor + 1)_\gamma(x)_2[2^{\lfloor \lg x \rfloor} \cdot (x)_2]$$

- encode length of binary representation using Elias- $\gamma$  code
- first bit of binary representation not required anymore
- $|(x)_\delta| = 2\lfloor \lg(\lfloor \lg x \rfloor + 1) \rfloor + 1 + \lfloor \lg x \rfloor$  bits



# Elias- $\delta$ -Encoding [Eli75]

## Definition: Elias- $\delta$ -Code

Given an integer  $x > 0$ , its Elias- $\delta$ -code  $(x)_\delta$  is

$$(x)_\delta = (\lfloor \lg x \rfloor + 1)_\gamma(x)_2[2^{\lfloor \lg(x)_2 \rfloor}]$$

- encode length of binary representation using Elias- $\gamma$  code
- first bit of binary representation not required anymore
- $|(x)_\delta| = 2\lfloor \lg(\lfloor \lg x \rfloor + 1) \rfloor + 1 + \lfloor \lg x \rfloor$  bits

## Elias- $\gamma$

- 4: 00 100
- 7: 00 111
- 1: 1
- 18: 0000 10010
- 12: 000 1000

## Elias- $\delta$

- 4: 0 11 00
- 7: 0 1111
- 1: 1
- 18: 00 101 0010
- 12: 00 100 100

# Golomb Encoding [Gol66]

## Definition: Golomb Code

Given an integer  $x > 0$  and a constant  $b > 0$ , the Golomb code consists of

- $q = \lfloor \frac{x}{b} \rfloor$
- $r = x - qb = x \% b$
- $c = \lceil \lg b \rceil$

with

$$(x)_{\text{Gol}(b)} = (q)_1(r)_2$$

where  $(r)_2$  depends on its size

- $r < 2^{\lceil \lg b \rceil - 1}$ :  $r$  requires  $\lceil \lg b \rceil$  bits and starts with a 0
- $r \geq 2^{\lceil \lg b \rceil - 1}$ :  $r$  requires  $\lceil \lg b \rceil$  bits and starts with a 1 and it encodes  $r - 2^{\lceil \lg b \rceil - 1}$

# Golomb Encoding [Gol66]

## Definition: Golomb Code

Given an integer  $x > 0$  and a constant  $b > 0$ , the Golomb code consists of

- $q = \lfloor \frac{x}{b} \rfloor$
- $r = x - qb = x \% b$
- $c = \lceil \lg b \rceil$

with

$$(x)_{\text{Gol}(b)} = (q)_1(r)_2$$

where  $(r)_2$  depends on its size

- $r < 2^{\lceil \lg b \rceil - 1}$ :  $r$  requires  $\lceil \lg b \rceil$  bits and starts with a 0
- $r \geq 2^{\lceil \lg b \rceil - 1}$ :  $r$  requires  $\lceil \lg b \rceil$  bits and starts with a 1 and it encodes  $r - 2^{\lceil \lg b \rceil - 1}$

- $b$  has to be fixed for all codes
- still variable length

# Golomb Encoding [Gol66]

## Definition: Golomb Code

Given an integer  $x > 0$  and a constant  $b > 0$ , the Golomb code consists of

- $q = \lfloor \frac{x}{b} \rfloor$
- $r = x - qb = x \% b$
- $c = \lceil \lg b \rceil$

with

$$(x)_{\text{Gol}(b)} = (q)_1(r)_2$$

where  $(r)_2$  depends on its size

- $r < 2^{\lceil \lg b \rceil - 1}$ :  $r$  requires  $\lceil \lg b \rceil$  bits and starts with a 0
- $r \geq 2^{\lceil \lg b \rceil - 1}$ :  $r$  requires  $\lceil \lg b \rceil$  bits and starts with a 1 and it encodes  $r - 2^{\lceil \lg b \rceil - 1}$

- $b$  has to be fixed for all codes
- still variable length

- for  $b = 5$ , there are 4 remainders: 00, 01, 100, 101, and 110
- $2^{\lceil \lg 5 \rceil - 1} = 2$
- $0, 1 < 2$ : 00 and 01 require 2 bits
- $2, 3, 4 \geq 2$ : require 3 bits and encode 0, 1, 2 starting with 1

# Comparison of Codes

# Back to Queries: Conjunctive Queries

## Task

- given terms  $t_1, \dots, t_k$
  - intersect  $L(t_1) \cap L(t_2) \cap \dots \cap L(t_k)$
- 
- pairwise intersection usually works best
  - intersection of two lists is of interest
  - start with two shortest and continue like that

# Back to Queries: Conjunctive Queries

## Task

- given terms  $t_1, \dots, t_k$
  - intersect  $L(t_1) \cap L(t_2) \cap \dots \cap L(t_k)$
- 
- pairwise intersection usually works best
  - intersection of two lists is of interest
  - start with two shortest and continue like that

## Setting

- two lists  $M$  and  $N$  with
  - $|M| = m$  and  $|N| = n$  and
  - $m \leq n$
- 
- different algorithms to intersect lists
  - assuming lists are  $\Delta$  encoded

# Naive Scanning

## Zipper

- scan both lists as in binary merging



# Naive Scanning

## Zipper

- scan both lists as in binary merging

## Lemma: Running Time Zipper

Intersecting two sorted lists of sizes  $m$  and  $n$  using zipper requires  $O(m + n)$  time.

# Naive Scanning

## Zipper

- scan both lists as in binary merging

## Lemma: Running Time Zipper

Intersecting two sorted lists of sizes  $m$  and  $n$  using zipper requires  $O(m + n)$  time.

## Proof (Sketch)

- compare entries until one list is empty
- if  $\max\{id : id \in N\} < \text{some element in } M$ , then all elements in  $N$  are compared
- resulting in  $O(n + m)$  time

# Naive Scanning

## Zipper

- scan both lists as in binary merging

## Lemma: Running Time Zipper

Intersecting two sorted lists of sizes  $m$  and  $n$  using zipper requires  $O(m + n)$  time.

## Proof (Sketch)

- compare entries until one list is empty
- if  $\max\{id : id \in N\} < \text{some element in } M$ , then all elements in  $N$  are compared
- resulting in  $O(n + m)$  time

- works well with  $\Delta$ -encoding
- in real implementations zipping is good until  $n > 20m$  [BS05]

# Naive Scanning

## Zipper

- scan both lists as in binary merging


## Lemma: Running Time Zipper

Intersecting two sorted lists of sizes  $m$  and  $n$  using zipper requires  $O(m + n)$  time.

## Proof (Sketch)

- compare entries until one list is empty
- if  $\max\{id : id \in N\} < \text{some element in } M$ , then all elements in  $N$  are compared
- resulting in  $O(n + m)$  time

- works well with  $\Delta$ -encoding
- in real implementations zipping is good until  $n > 20m$  [BS05]

- example on the board 

# Binary Search (1/2)

## Simple Binary Search

- search each document in  $M$  in  $N$  using binary search

# Binary Search (1/2)

## Simple Binary Search

- search each document in  $M$  in  $N$  using binary search

## Lemma: Running Time Simple Binary Search

Intersecting two sorted lists of sizes  $m$  and  $n$  using a simple binary search requires  $O(m \lg n)$  time.

# Binary Search (1/2)

## Simple Binary Search

- search each document in  $M$  in  $N$  using binary search

## Lemma: Running Time Simple Binary Search

Intersecting two sorted lists of sizes  $m$  and  $n$  using a simple binary search requires  $O(m \lg n)$  time.

## Proof (Sketch)

- binary search on  $N$  because  $n \geq m$
- for each id in  $N$  binary search in  $O(\lg n)$  time
- resulting in  $O(m \lg n)$  total time

# Binary Search (1/2)

## Simple Binary Search


- search each document in  $M$  in  $N$  using binary search

## Lemma: Running Time Simple Binary Search

Intersecting two sorted lists of sizes  $m$  and  $n$  using a simple binary search requires  $O(m \lg n)$  time.

## Proof (Sketch)

- binary search on  $N$  because  $n \geq m$
- for each id in  $N$  binary search in  $O(\lg n)$  time
- resulting in  $O(m \lg n)$  total time

- example on the board 



# Binary Search (1/2)

## Simple Binary Search


- search each document in  $M$  in  $N$  using binary search

## Lemma: Running Time Simple Binary Search

Intersecting two sorted lists of sizes  $m$  and  $n$  using a simple binary search requires  $O(m \lg n)$  time.

## Proof (Sketch)

- binary search on  $N$  because  $n \geq m$
- for each id in  $N$  binary search in  $O(\lg n)$  time
- resulting in  $O(m \lg n)$  total time

- example on the board 

- binary search not work with  $\Delta$ -encoding

## Binary Search (2/2)

### Double Binary Search

- let  $p_m = \lfloor \frac{m}{2} \rfloor$
- search for  $M[p_m]$  in  $N$  using binary search
- let result be position  $p_n$
- if  $M[p_m] = N[p_n]$  add  $M[p_m]$  to result
- continue recursively by intersecting
  - $M[1, p_m] \cap N[1, p_n]$  and
  - $M[1 + p_m, |M|] \cap N[1 + p_n, |N|]$

## Binary Search (2/2)

### Double Binary Search

- let  $p_m = \lfloor \frac{m}{2} \rfloor$
- search for  $M[p_m]$  in  $N$  using binary search
- let result be position  $p_n$
- if  $M[p_m] = N[p_n]$  add  $M[p_m]$  to result
- continue recursively by intersecting
  - $M[1, p_m] \cap N[1, p_n]$  and
  - $M[1 + p_m, |M|] \cap N[1 + p_n, |N|]$

### Lemma: Running Time Double Binary Search

Intersecting two sorted lists of sizes  $m$  and  $n$  using a double binary search requires  $O(m \lg \frac{n}{m})$  time.

# Binary Search (2/2)

## Double Binary Search

- let  $p_m = \lfloor \frac{m}{2} \rfloor$
- search for  $M[p_m]$  in  $N$  using binary search
- let result be position  $p_n$
- if  $M[p_m] = N[p_n]$  add  $M[p_m]$  to result
- continue recursively by intersecting
  - $M[1, p_m] \cap N[1, p_n]$  and
  - $M[1 + p_m, |M|] \cap N[1 + p_n, |N|]$

## Lemma: Running Time Double Binary Search

Intersecting two sorted lists of sizes  $m$  and  $n$  using a double binary search requires  $O(m \lg \frac{n}{m})$  time.

## Proof (Sketch)

- look at running time of binary search at each recursion depth
- depth 0:  $\lg n$
- depth 1:  $2 \lg \frac{n}{2}$
- depth 2:  $4 \lg \frac{n}{4}$
- depth  $m$ :  $m \lg \frac{n}{m}$

Since depth of recursion is at most  $m$ , this results in

- $\sum_{i=0}^{\lg m} \frac{m}{2^i} (\lg \frac{n}{m} + i) = m (\lg \frac{n}{m} \sum_{i=0}^{\lg m} \frac{1}{2^i} + \sum_{i=0}^{\lg m} \frac{1}{2^i})$
- total:  $O(m \lg \frac{n}{m})$

# Binary Search (2/2)

## Double Binary Search

- let  $p_m = \lfloor \frac{m}{2} \rfloor$
- search for  $M[p_m]$  in  $N$  using binary search
- let result be position  $p_n$
- if  $M[p_m] = N[p_n]$  add  $M[p_m]$  to result
- continue recursively by intersecting
  - $M[1, p_m] \cap N[1, p_n]$  and
  - $M[1 + p_m, |M|] \cap N[1 + p_n, |N|]$

## Lemma: Running Time Double Binary Search


Intersecting two sorted lists of sizes  $m$  and  $n$  using a double binary search requires  $O(m \lg \frac{n}{m})$  time.

## Proof (Sketch)

- look at running time of binary search at each recursion depth
- depth 0:  $\lg n$
- depth 1:  $2 \lg \frac{n}{2}$
- depth 2:  $4 \lg \frac{n}{4}$
- depth  $m$ :  $m \lg \frac{n}{m}$

Since depth of recursion is at most  $m$ , this results in

- $\sum_{i=0}^{\lg m} \frac{m}{2^i} (\lg \frac{n}{m} + i) = m (\lg \frac{n}{m} \sum_{i=0}^{\lg m} \frac{1}{2^i} + \sum_{i=0}^{\lg m} \frac{i}{2^i})$
- total:  $O(m \lg \frac{n}{m})$

- example on board 

# Exponential Search

## Exponential Search

- assume that  $M[1..i]$  have been processed and
- $M[i]$  is closest to  $N[j]$  for some  $j$
- now find  $M[i + 1]$  in  $N$  by comparing it to  $N[j], N[j + 1], N[j + 2], N[j + 4], \dots$  until
- $N[j + 2^k] \geq M[i + 1]$  if  $N[j + 2^k] = M[i + 1]$ , we are done with this iteration
- binary search for  $M[i + 1]$  in  $N[j + 2^{k-1}..j + 2^k]$

# Exponential Search

## Exponential Search

- assume that  $M[1..i]$  have been processed and
- $M[i]$  is closest to  $N[j]$  for some  $j$
- now find  $M[i + 1]$  in  $N$  by comparing it to  $N[j], N[j + 1], N[j + 2], N[j + 4], \dots$  until
- $N[j + 2^k] \geq M[i + 1]$  if  $N[j + 2^k] = M[i + 1]$ , we are done with this iteration
- binary search for  $M[i + 1]$  in  $N[j + 2^{k-1}..j + 2^k]$

## Lemma: Running Time Exponential Search

Intersecting two sorted lists of sizes  $m$  and  $n$  using an exponential search requires  $O(m \lg \frac{n}{m})$  time.

# Exponential Search

## Exponential Search

- assume that  $M[1..i]$  have been processed and
- $M[i]$  is closest to  $N[j]$  for some  $j$
- now find  $M[i + 1]$  in  $N$  by comparing it to  $N[j], N[j + 1], N[j + 2], N[j + 4], \dots$  until
- $N[j + 2^k] \geq M[i + 1]$  if  $N[j + 2^k] = M[i + 1]$ , we are done with this iteration
- binary search for  $M[i + 1]$  in  $N[j + 2^{k-1}..j + 2^k]$

## Lemma: Running Time Exponential Search

Intersecting two sorted lists of sizes  $m$  and  $n$  using an exponential search requires  $O(m \lg \frac{n}{m})$  time.

## Proof

- searching for each element  $M[i]$  requires  $O(\lg d_i)$  time
- $d_i$  is distance between  $M[i - 1]$  and  $M[i]$  in  $N$
- $O(\sum_i^m \lg d_i)$ , which is maximal if  $d_i = \frac{n}{m}$
- total:  $O(m \frac{n}{m})$



# Exponential Search

## Exponential Search


- assume that  $M[1..i]$  have been processed and
- $M[i]$  is closest to  $N[j]$  for some  $j$
- now find  $M[i + 1]$  in  $N$  by comparing it to  $N[j], N[j + 1], N[j + 2], N[j + 4], \dots$  until
- $N[j + 2^k] \geq M[i + 1]$  if  $N[j + 2^k] = M[i + 1]$ , we are done with this iteration
- binary search for  $M[i + 1]$  in  $N[j + 2^{k-1}..j + 2^k]$

## Lemma: Running Time Exponential Search

Intersecting two sorted lists of sizes  $m$  and  $n$  using an exponential search requires  $O(m \lg \frac{n}{m})$  time.

## Proof

- searching for each element  $M[i]$  requires  $O(\lg d_i)$  time
- $d_i$  is distance between  $M[i - 1]$  and  $M[i]$  in  $N$
- $O(\sum_i^m \lg d_i)$ , which is maximal if  $d_i = \frac{n}{m}$
- total:  $O(m \frac{n}{m})$

- example on board 

# Exponential Search

## Exponential Search


- assume that  $M[1..i]$  have been processed and
- $M[i]$  is closest to  $N[j]$  for some  $j$
- now find  $M[i + 1]$  in  $N$  by comparing it to  $N[j], N[j + 1], N[j + 2], N[j + 4], \dots$  until
- $N[j + 2^k] \geq M[i + 1]$  if  $N[j + 2^k] = M[i + 1]$ , we are done with this iteration
- binary search for  $M[i + 1]$  in  $N[j + 2^{k-1}..j + 2^k]$

## Lemma: Running Time Exponential Search

Intersecting two sorted lists of sizes  $m$  and  $n$  using a exponential search requires  $O(m \lg \frac{n}{m})$  time.

## Proof

- searching for each element  $M[i]$  requires  $O(\lg d_i)$  time
- $d_i$  is distance between  $M[i - 1]$  and  $M[i]$  in  $N$
- $O(\sum_i^m \lg d_i)$ , which is maximal if  $d_i = \frac{n}{m}$
- total:  $O(m \frac{n}{m})$

- example on board 

- works well if lists do not fit into main memory
- still not working with  $\Delta$ -encoding

# Engineered Representations

## Two-Level Representation

- store every  $B$ -th element of the list in top-level
- in addition to  $\Delta$ -encoded ids
- store original id for each sampled value in id-list


# Engineered Representations

## Two-Level Representation

- store every  $B$ -th element of the list in top-level
- in addition to  $\Delta$ -encoded ids
- store original id for each sampled value in id-list

## Binary Search

- binary search on top-level
- scan on list in relevant interval

- example on board 


# Engineered Representations

## Two-Level Representation

- store every  $B$ -th element of the list in top-level
- in addition to  $\Delta$ -encoded ids
- store original id for each sampled value in id-list

## Binary Search


- binary search on top-level
- scan on list in relevant interval

- example on board 

## Skipper [MZ96]

- scan top-level and
- go down in  $\Delta$ -encoded list as soon as possible

- avoids complex binary search control structure


- example on board 

# Intersection with Randomized Inverted Indices [ST07]

- assume ids are in  $[0, U)$  with  $U = 2^{2^u}$
- ids have to be random ⓘ more details in paper
- choose tuning parameter  $B$  ⓘ determine average bucket size
- given a list  $N = [d_1, \dots, d_n]$  and  $k_N = \lceil \lg \frac{UB}{n} \rceil$
- per list, represent ids in
  - buckets  $b_i^N$  containing
  - partial ids  $\{d_j \bmod 2^{k_N} : d_j / 2^{k_N} = i\}$
- due to randomization, average bucket size is between  $B/2$  and  $B$
- elements in buckets can be  $\Delta$ -encoded


# Intersection with Randomized Inverted Indices [ST07]

- assume ids are in  $[0, U)$  with  $U = 2^{2^u}$
- ids have to be random ⓘ more details in paper
- choose tuning parameter  $B$  ⓘ determine average bucket size
- given a list  $N = [d_1, \dots, d_n]$  and  $k_N = \lceil \lg \frac{UB}{n} \rceil$
- per list, represent ids in
  - buckets  $b_i^N$  containing
  - partial ids  $\{d_j \bmod 2^{k_N} : d_j / 2^{k_N} = i\}$
- due to randomization, average bucket size is between  $B/2$  and  $B$
- elements in buckets can be  $\Delta$ -encoded

- example on board 

# Intersection with Randomized Inverted Indices [ST07]

- assume ids are in  $[0, U)$  with  $U = 2^{2^u}$
- ids have to be random ⓘ more details in paper
- choose tuning parameter  $B$  ⓘ determine average bucket size
- given a list  $N = [d_1, \dots, d_n]$  and  $k_N = \lceil \lg \frac{UB}{n} \rceil$
- per list, represent ids in
  - buckets  $b_i^N$  containing
  - partial ids  $\{d_j \bmod 2^{k_N} : d_j / 2^{k_N} = i\}$
- due to randomization, average bucket size is between  $B/2$  and  $B$
- elements in buckets can be  $\Delta$ -encoded

- example on board 


## Intersection

- for each element  $M[i]$  find bucket of  $N$
- can be same bucket as for  $M[i - 1]$ , if so, continue at position of  $M[i - 1]$  in bucket
  - ⓘ continuing is important
- scan bucket until element  $\geq M[i]$  is found
- if equal, output  $M[i]$



# Intersection with Randomized Inverted Indices [ST07]

- assume ids are in  $[0, U)$  with  $U = 2^{2^u}$
- ids have to be random ⓘ more details in paper
- choose tuning parameter  $B$  ⓘ determine average bucket size
- given a list  $N = [d_1, \dots, d_n]$  and  $k_N = \lceil \lg \frac{UB}{n} \rceil$
- per list, represent ids in
  - buckets  $b_i^N$  containing
  - partial ids  $\{d_j \bmod 2^{k_N} : d_j / 2^{k_N} = i\}$
- due to randomization, average bucket size is between  $B/2$  and  $B$
- elements in buckets can be  $\Delta$ -encoded

- example on board 

## Intersection

- for each element  $M[i]$  find bucket of  $N$
- can be same bucket as for  $M[i - 1]$ , if so, continue at position of  $M[i - 1]$  in bucket
  - ⓘ continuing is important
- scan bucket until element  $\geq M[i]$  is found
- if equal, output  $M[i]$

## Lemma: Running Time

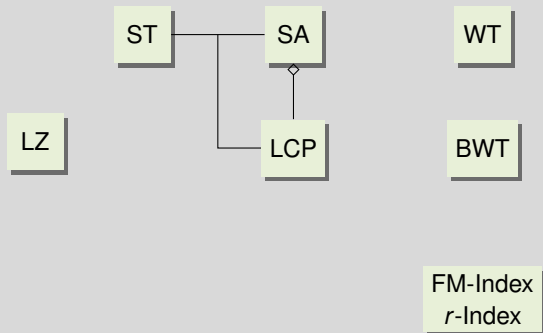
Intersecting two sorted lists of sizes  $m$  and  $n$  using a randomized inverted indices requires  $O(m + \min\{n, Bm\})$  time.

# Conclusion and Outlook

## This Lecture

- inverted index
- space efficient encodings of document lists
- efficient intersection algorithms

## Linear Time Construction



# Conclusion and Outlook

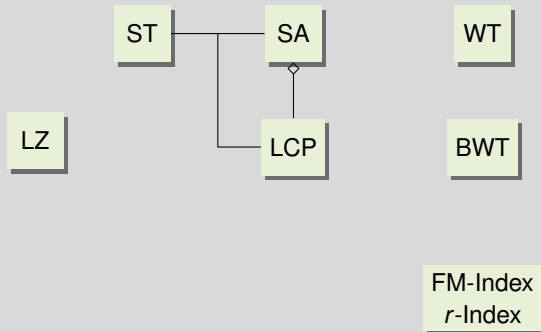
## This Lecture

- inverted index
- space efficient encodings of document lists
- efficient intersection algorithms

## Next Lecture

- top- $k$  retrieval

## Linear Time Construction



# Oral Exam

## Remaining Lectures

- 17.01. top- $k$  retrieval
- 24.01. longest common extensions
- 31.01. TBD & Q&A
- 07.02. project presentation

- 20 minute long oral exam
- conducted by Prof. Sanders and me
- most likely virtual ⓘ technic check

- 09.03. is default date for all(?) exams
- 08.02. possible, but needs good arguments

- any questions

# Bibliography I

- [BS05] Ricardo A. Baeza-Yates and Alejandro Salinger. “Experimental Analysis of a Fast Intersection Algorithm for Sorted Sequences”. In: *SPIRE*. Volume 3772. Lecture Notes in Computer Science. Springer, 2005doi10.1007/11575832\_2, pages 13–24.
- [Eli75] Peter Elias. “Universal Codeword Sets and Representations of the Integers”. In: *IEEE Trans. Inf. Theory* 21.2 (1975), pages 194–203. DOI: [10.1109/TIT.1975.1055349](https://doi.org/10.1109/TIT.1975.1055349).
- [Gol66] Solomon W. Golomb. “Run-length Encodings (Corresp.)”. In: *IEEE Trans. Inf. Theory* 12.3 (1966), pages 399–401. DOI: [10.1109/TIT.1966.1053907](https://doi.org/10.1109/TIT.1966.1053907).
- [MZ96] Alistair Moffat and Justin Zobel. “Self-Indexing Inverted Files for Fast Text Retrieval”. In: *ACM Trans. Inf. Syst.* 14.4 (1996), pages 349–379.
- [ST07] Peter Sanders and Frederik Transier. “Intersection in Integer Inverted Indices”. In: *ALENEX*. SIAM, 2007. DOI: [10.1137/1.9781611972870.7](https://doi.org/10.1137/1.9781611972870.7).