CS 600.226: Data Structures Michael Schatz

Nov 12, 2018 Lecture 31: Suffix Arrays



HW8

Assignment 8: Competitive Spelling Bee

Out on: November 2, 2018 Due by: November 9, 2018 before 10:00 pm Collaboration: None Grading:

Packaging 10%, Style 10% (where applicable), Testing 10% (where applicable), Performance 30% (where applicable), Functionality 40% (where applicable)

Overview

Your "one" task for this assignment is to take the simple spell checker we give you and to turn it into the fastest, most memory-efficient spell checker in the course, subject to the constraints detailed below. You are expected to do this by (once again) implementing the Map interface, this time using one of several hash table techniques (your choice, see below).

Not all CPU operations are created equal



ithare.com Operation Cost in CPU Cycles	10*	10'	10?	10' 1	04 1	05	10
"Simple" register-register op (ADD,OR,etc.)	41				-		1
Memory write	- 1					-	
Bypass delay: switch between integer and floating-point units	0-3						
"Right" branch of "if"	1-2						
Floating-point/vector addition	1-3						
Multiplication (integer/float/vector)	1.2						
Return error and check	1.2						
L1 read	3.4						
L2 read		10-12					
"Wrong" branch of "if" (branch misprediction)		10-20					
Floating-point division		10.40					
128-bit vector division		18-70	1				
C function direct call		15-30					
Integer division		15-40					
C function indirect call		20-50	2				
C++ virtual function call		30-88					
L3 read		30.70	I				
Main RAM read			100-158				
NUMA: different-socket L3 read			100-308				
Allocation+deallocation pair (small objects)			200-600	2			
NUMA: different-socket main RAM read			203.600	a harrist and		-	
Kernel call				1000-1500			
Thread context switch (direct costs)				2000			
C++ Exception thrown+caught				5000-10000		-	
Thread context switch (total costs,							_
including angle involution)					10000 - 1 million		

http://ithare.com/infographics-operation-costs-in-cpu-clock-cycles/

and the second

4

ŝ

300m

30km

Hashing

O(1)

Array["Mike"] = 10; *Array["Peter"]* = 15 BST:O(lg n) -> Hash:O(1)

Hash Function enables Map<K, V> as Map<Integer, V> as Array<V>

- h(): K -> Integer for any possible key K
- h() should distribute the keys uniformly over all integers
- if k_1 and k_2 are "close", $h(k_1)$ and $h(k_2)$ should be "far" apart

Typically want to return a small integer, so that we can use it as an index into an array

- An array with 4B cells in not very practical if we only expect a few thousand to a few million entries
- How do we restrict an arbitrary integer x into a value up to some maximum value n?

0 <= x % n < n

Compression function: c(i) = abs(i) % length(a)

Transforms from a large range of integers to a small range (to store in array a)

Collision Strategies

1. Separate chaining:

More object overhead, but degrades gracefully as load approaches 1



2. Linear Probing

Minimal overhead, easy to implement, but tends to form large clusters



3. Quadratic Probing

Slightly more complex, but skips over larger and larger amounts to find holes



Advanced Hashing Summary

Collisions

- Be careful of collisions!
- Perfect hashing guarantees we avoid collisions, universal hashing lets us pick a new hash function from a family as needed

Cuckoo Hashing

- Simple open addressing technique with O(1) lookup with expected O(1) insert
- Often outperforms other collision management schemes

• $\lambda = 1$ • $\lambda = 4$ • $\lambda = 10$ • $\lambda = 10$

Bloom Filters

- Store a huge set in a tiny amount of space allowing for a small rate of false positives.
- Used as a quick pre-filter to determine if the slow operation needs to take place



Part I: Intro to Genomics aka Data Structures on Strings

DNA: The secret of life



Your DNA, along with your environment and experiences, shapes who you are

- Height
- Hair, eye, skin color
- Broad/narrow, small/large features
- Susceptibility to disease
- Response to drug treatments
- Longevity and cognition

Physical traits tend to be strongly genetic, social characteristics tend to be strongly environmental, and everything else is a combination

Cells & DNA



Cells & DNA



The Origins of DNA Sequencing

Nature Vol. 265 February 24 1977

687

articles

Nucleotide sequence of bacteriophage $\Phi X174 DNA$

F. Sanger, G. M. Air^{*}, B. G. Barrell, N. L. Brown^{*}, A. R. Coulson, J. C. Fiddes, C. A. Hutchison III[‡], P. M. Slocombe[‡] & M. Smith^{*}

MRC Laboratory of Molecular Biology, Hills Road, Cambridge CB2 2QH, UK

A DNA sequence for the genome of bacteriophage $\Phi X 174$ of approximately 5,375 nucleotides has been determined using the rapid and simple 'plus and minus' method. The sequence identifies many of the features responsible for the production of the proteins of the nine known genes of the organism, including initiation and termination sites for the proteins and RNAs. Two pairs of genes are coded by the same region of DNA using different reading frames.

THE genome of bacteriophage Φ X174 is a single-stranded, circular DNA of approximately 5,400 nucleotides coding for nine known proteins. The order of these genes, as determined by genetic techniques¹⁻⁴, is A-B-C-D-E-J-F-G-H. Genes F, G and H code for structural proteins of the virus capsid, and gene J (as defined by sequence work) codes for a small basic protein strand DNA of ΦX has the same sequence as the mRNA and, in certain conditions, will bind ribosomes so that a protected fragment can be isolated and sequenced. Only one major site was found. By comparison with the amino acid sequence data it was found that this ribosome binding site sequence coded for the initiation of the gene G proteinth (positions 2,362-2,413).

At this stage sequencing techniques using primed synthesis with DNA polymerase were being developed* and Schott¹¹ synthesised a decanceleotide with a sequence complementary to part of the ribosome binding site. This was used to prime into the intercistronic region between the *F* and *G* genes, using DNA polymerase and ¹¹P-labelled triphosphates¹⁵. The ribo-substitution technique¹⁶ facilitated the sequence determination of the labelled DNA produced. This decanucleotide-primed system was also used to develop the plus and minus method¹. Suitable synthetic primers are, however, difficult to prepare and as

Sanger et al. (1977) Nature Ist Complete Organism Bacteriophage φX174; 5375 bp

Awarded Nobel Prize in 1980



Radioactive Chain Termination 5000bp / week / person

http://en.wikipedia.org/wiki/File:Sequencing.jpg http://www.answers.com/topic/automated-sequencer

Milestones in DNA Sequencing



(TIGR/Celera, 1995-2001)

Oxford Nanopore MinION





- Thumb drive sized sequencer powered over USB
- Capacity for 512 reads at once
- Senses DNA by measuring changes to ion flow



Nanopore Sequencing





Nanopore Basecalling



- Hidden Markov model
- Only four options per transition
- Pore type = distinct kmer length



- Form probabilistic path through measured states currents and transitions
 - e.g. Viterbi algorithm

Basecalling currently performed at Amazon with frequent updates to algorithm

Cost per Genome



How much is a zettabase?

Unit	Size
Base	
Kilobase	1,000
Megabase	1,000,000
Gigabase	1,000,000,000
Terabase	1,000,000,000,000
Petabase	I,000,000,000,000,000
Exabase	I,000,000,000,000,000,000
Zettabase	1,000,000,000,000,000,000,000

How much is a zettabase?

100 GB / Genome 4.7GB / DVD ~20 DVDs / Genome

Х

10,000,000,000 Genomes

=

1ZB Data 200,000,000,000 DVDs







150,000 miles of DVDs $\sim \frac{1}{2}$ distance to moon

Both currently ~100PB And growing exponentially

How much is a zettabase?



200,000,000,000 DVDs

150,000 miles of DVDs $\sim \frac{1}{2}$ distance to moon

Both currently ~100PB And growing exponentially

Schatzlab Overview



Human Genetics

Genetics of Autism Spectrum Disorders

Narzisi et al. (2015) lossifov et al. (2014)



Cancer Biology

Indels, CNVs, SVs, & Cell Phylogenetics

Nattestad et al. (2018) Goodwin et al. (2015)



Agricultural Genomics

Rice, Corn, Wheat and many others

Chin et al. (2016) Schatz et al. (2014)



Evolutionary Biology

Inflorescence diversity

Lemmon et al. (2016) Park et al. (2011)

Computational Biology @ CS



Alexis Battle



Ben Langmead



James Taylor



Liliana Florea



Mihaela Pertea



Steven Salzberg

Computational Biology @ JHU



Results Domain Knowledge

Machine Learning classification, modeling, visualization & data Integration

Scalable Algorithms Streaming, Sampling, Indexing, Parallel

Compute Systems CPU, GPU, Distributed, Clouds, Workflows

IO Systems Hardrives, Networking, Databases, Compression, LIMS

Sensors & Metadata Sequencers, Microscopy, Imaging, Mass spec, Metadata & Ontologies





Schatz, MC (2017) Nature Methods 14, 347–348 (2017) doi:10.1038/nmeth.4240

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Ultrafast and memory-efficient alignment of short DNA sequences to the human genome Langmead et al. (2009) *Genome Biology* 10:R25 doi: 10.1186/gb-2009-10-3-r25

Sequencers, Microscopy, Imaging, Mass spec, Metadata & Ontologies

Comparisonal Dialass

Reproducible RNA-seq analysis using recount2 Collado-Torres et al (2017) Nature Biotechnology. 35, 319–321

IO Systems Hardrives, Networking, Databases, Compression, LIMS

Sensors & Metadata Sequencers, Microscopy, Imaging, Mass spec, Metadata & Ontologies

Genetic effects on gene expression across human tissues. The GTEx Consortium (2017) Nature 550, 204–213

Sequencers, Microscopy, Imaging, Mass spec, Metadata & Ontologies

verworking, Databases, Compression, Env

Personal Genomics

How does your genome compare to the reference?

- Where is GATTACA in the human genome?
- Strategy I: Brute Force

No match at offset I

- Where is GATTACA in the human genome?
- Strategy I: Brute Force

Match at offset 2

- Where is GATTACA in the human genome?
- Strategy I: Brute Force

No match at offset 3...

- Where is GATTACA in the human genome?
- Strategy I: Brute Force

No match at offset 9 <- Checking each possible position takes time

Brute Force Analysis

- Brute Force:
 - At every possible offset in the genome:
 - Do all of the characters of the query match?

• Analysis

- Simple, easy to understand
- Genome length = n
- Query length = m
- Comparisons: (n-m+1) * m
- Overall runtime: O(nm)

[How long would it take if we double the genome size, read length?] [How long would it take if we double both?]

[3B]

|/|

[2|B]

Expected Occurrences

The expected number of occurrences (e-value) of a given sequence in a genome depends on the length of the genome and inversely on the length of the sequence

- I in 4 bases are G, I in 16 positions are GA, I in 64 positions are GAT, ...
- I in 16,384 should be GATTACA
- E=n/(4^m)

[183,105 expected occurrences] [How long do the reads need to be for a significant match?]

Brute Force Reflections

Why check every position?

– GATTACA can't possibly start at position 15

2 3 5 6 7 8 9 2 3 4 5 4 ••• Т G Т Т Α С Α G Α Т Т Α С С Α Α Т Т Α С G Α

[WHY?]

[3B + 7]

- Improve runtime to O(n + m)
 - If we double both, it just takes twice as long
 - Knuth-Morris-Pratt, 1977
 - Boyer-Moyer, 1977, 1991
- For one-off scans, this is the best we can do (optimal performance)
 - We have to read every character of the genome, and every character of the query
 - For short queries, runtime is dominated by the length of the genome

How can we make this go faster?

Hash Table Lookup

- By construction, multiple keys have the same hash value
 - Store elements with the same key in a bucket chained together
 - A good hash evenly distributes the values: R/H have the same hash value
 - Looking up a value scans the entire bucket
 - Slows down the search as a function of the hash table load

Full-Text Indexing: Suffix Arrays

- What if we need to check many queries?
 - We don't need to check every page of the phone book to find 'Schatz'
 - Sorting alphabetically lets us immediately skip 96% (25/26) of the book without any loss in accuracy
- Sorting the genome: Suffix Array (Manber & Myers, 1991)
 - Sort every suffix of the genome

- Strategy 2: Binary search
 - Compare to the middle, refine as higher or lower
- Searching for GATTACA ٠
 - Lo = I; Hi = 15;

Lo	#	Sequence	Pos
-	Ι	ACAGATTACC	6
	2	ACC	13
	3	AGATTACC	8
	4	ATTACAGATTACC	3
	5	ATTACC	10
	6	C	15
	7	CAGATTACC	7
	8	CC	14
	9	GATTACAGATTACC	2
	10	GATTACC	9
	11	TACAGATTACC	5
	12	TACC	12
	13	TGATTACAGATTACC	I
	14	TTACAGATTACC	4
Hi	15	TTACC	П

- Strategy 2: Binary search
 - Compare to the middle, refine as higher or lower
- Searching for GATTACA
 - Lo = I; Hi = I5; Mid = (I+I5)/2 = 8
 - Middle = Suffix[8] = CC

Lo	#	Sequence	Pos
\rightarrow	I	ACAGATTACC	6
	2	ACC	13
	3	AGATTACC	8
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	5	ATTACC	10
	6	C	15
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	8	CC	14
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	10	GATTACC	9
	11	TACAGATTACC	5
	12	TACC	12
	13	TGATTACAGATTACC	Ι
	14	TTACAGATTACC	4
Hi	15	TTACC	11

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 - Lo = I; Hi = I5; Mid = (I+I5)/2 = 8
 - Middle = Suffix[8] = CC
 => Higher: Lo = Mid + 1

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 - Middle = Suffix[8] = CC
 => Higher: Lo = Mid + 1
 - Lo = 9; Hi = 15; Mid = (9+15)/2 = 12
 - Middle = Suffix[12] = TACC

	#	Sequence	Pos
	I	ACAGATTACC	6
	2	ACC	13
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	4	ATTACAGATTACC	3
	5	ATTACC	10
	6	C	15
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	12	TACC	12
	13	TGATTACAGATTACC	I
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- Strategy 2: Binary search
 - Compare to the middle, refine as higher or lower
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 - Lo = I; Hi = I5; Mid = (1+15)/2 = 8
 - Middle = Suffix[8] = CC = Higher: Lo = Mid + I
 - Lo = 9; Hi = 15; Mid = (9+15)/2 = 12
 - Middle = Suffix[12] = TACC = Lower: Hi = Mid - I
 - Lo = 9; Hi = 11;

	#	Sequence	Pos
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	5	ATTACC	10
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	10	GATTACC	9
Hi	11	TACAGATTACC	5
-	12	TACC	12
	13	TGATTACAGATTACC	I
	14	TTACAGATTACC	4
	15	TTACC	11

Hi

- Strategy 2: Binary search
 - Compare to the middle, refine as higher or lower
- Searching for GATTACA
 - Lo = I; Hi = 15; Mid = (1+15)/2 = 8
 - Middle = Suffix[8] = CC = Higher: Lo = Mid + I
 - Lo = 9; Hi = 15; Mid = (9+15)/2 = 12
 - Middle = Suffix[12] = TACC = Lower: Hi = Mid - I
 - Lo = 9; Hi = |1|; Mid = (9+|1|)/2 = 10
 - Middle = Suffix[10] = GATTACC

	#	Sequence	Pos
	Ι	ACAGATTACC	6
	2	ACC	13
	3	AGATTACC	8
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	5	ATTACC	10
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 => Lower: Hi = Mid 1
 - Lo = 9; Hi = |1|; Mid = (9+|1|)/2 = 10
 - Middle = Suffix[10] = GATTACC
 => Lower: Hi = Mid 1
 - Lo = 9; Hi = 9;

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13	TGATTACAGATTACC	I
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15	TTACC	

Lo

Hi

- Strategy 2: Binary search
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 - Lo = I; Hi = I5; Mid = (I+I5)/2 = 8
 - Middle = Suffix[8] = CC
 => Higher: Lo = Mid + 1
 - Lo = 9; Hi = 15; Mid = (9+15)/2 = 12
 - Middle = Suffix[12] = TACC
 => Lower: Hi = Mid 1
 - Lo = 9; Hi = |1|; Mid = (9+|1|)/2 = 10
 - Middle = Suffix[10] = GATTACC
 => Lower: Hi = Mid 1
 - Lo = 9; Hi = 9; Mid = (9+9)/2 = 9
 - Middle = Suffix[9] = GATTACA...
 => Match at position 2!

	#	Sequence	Pos
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	4	ATTACAGATTACC	3
	5	ATTACC	10
	6	C	15
	7	CAGATTACC	7
Lo	8	CC	14
H	9	GATTACAGATTACC	2
	10	GATTACC	9
		TACAGATTACC	5
	12	TACC	12
	13	TGATTACAGATTACC	Ι
	14	TTACAGATTACC	4
	15	TTACC	

Binary Search Analysis

• Binary Search

Initialize search range to entire list mid = (hi+lo)/2; middle = suffix[mid] if query matches middle: done else if query < middle: pick low range else if query > middle: pick hi range Repeat until done or empty range

[WHEN?]

[32]

- Analysis
 - More complicated method
 - How many times do we repeat?
 - How many times can it cut the range in half?
 - Find smallest x such that: $n/(2^x) \le 1$; x = $lg_2(n)$
- Total Runtime: O(m lg n)
 - More complicated, but much faster!
 - Looking up a query loops 32 times instead of 3B

[How long does it take to search 6B or 24B nucleotides?]

Suffix Array Construction

How can we store the suffix array?

[How many characters are in all suffixes combined?]

$$S = 1 + 2 + 3 + \dots + n = \sum_{i=1}^{n} i = \frac{n(n+1)}{2} = O(n^2)$$

- Hopeless to explicitly store 4.5 billion billion characters
- Instead use implicit representation
 - Keep I copy of the genome, and a list of sorted offsets
 - Storing 3 billion offsets fits on a server (12GB)
- Searching the array is very fast, but it takes time to construct
 - This time will be amortized over many, many searches
 - Run it once "overnight" and save it away for all future queries

Quadratic Sorting Algorithms

Selection Sort Move next smallest into position

Bubble Sort Swap up bigger values over smaller

Insertion Sort Slide next value into correct position

These algorithms will work, but are very slow for 3B suffixes!

How can we go faster?

Divide and Conquer

- Selection sort is slow because it rescans the entire list for each element
 - How can we split up the unsorted list into independent ranges?
 - Hint I: Binary search splits up the problem into 2 independent ranges (hi/lo)
 - Hint 2: Assume we know the median value of a list

[How many times can we split a list in half?]

QuickSort Analysis

QuickSort(Input: list of n numbers) // see if we can quit if (length(list)) <= 1): return list

```
// split list into lo & hi
pivot = median(list)
lo = {}; hi = {};
for (i = 1 to length(list))
if (list[i] < pivot): append(lo, list[i])
else: append(hi, list[i])</pre>
```

```
// recurse on sublists
return (append(QuickSort(lo), QuickSort(hi))
```

Analysis (Assume we can find the median in O(n))

$$T(n) = \begin{cases} O(1) & \text{if } n \le 1\\ O(n) + 2T(n/2) & \text{else} \end{cases}$$
$$T(n) = n + 2(\frac{n}{2}) + 4(\frac{n}{4}) + \dots + n(\frac{n}{n}) = \sum_{i=0}^{lg(n)} \frac{2^{i}n}{2^{i}} = \sum_{i=0}^{lg(n)} n = O(n \lg n)$$

http://en.wikipedia.org/wiki/Quicksort

Picking the Median

• What if we miss the median and do a 90/10 split instead?

[How many times can we cut 10% off a list?]

Randomized Quicksort

- 90/10 split runtime analysis $T(n) = n + T(\frac{n}{10}) + T(\frac{9n}{10})$ $T(n) = n + \frac{n}{10} + T(\frac{n}{100}) + T(\frac{9n}{100}) + \frac{9n}{10} + T(\frac{9n}{100}) + T(\frac{81n}{100})$ $T(n) = n + n + T(\frac{n}{100}) + 2T(\frac{9n}{100}) + T(\frac{81n}{100})$ $T(n) = \sum_{i=0}^{\log_{10/9}(n)} n = O(n \lg n)$ Find smallest x s.t. (9/10)^x n \le 1 $(10/9)^{x} \ge n$ $x \ge \log_{10/9} n$
- If we randomly pick a pivot, we will get at least a 90/10 split with very high probability
 - Everything is okay as long as we always slice off a fraction of the list

[Challenge Question:What happens if we slice I element]

Exact Matching Review & Overview

Where is GATTACA in the human genome?

*** These are general techniques applicable to any search problem ***

Next Steps

- I. Reflect on the magic and power of Suffix Arrays!
- I. Assignment 8 due Friday November 16 @ 10pm