CS 5114: Theory of Algorithms

Clifford A. Shaffer

Department of Computer Science Virginia Tech Blacksburg, Virginia

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String Matching

Let $A = a_1 a_2 \cdots a_n$ and $B = b_1 b_2 \cdots b_m$, $m \le n$, be two strings of characters.

Problem: Given two strings *A* and *B*, find the first occurrence (if any) of *B* in *A*.

• Find the smallest k such that, for all i, $1 \le i \le m$, $a_{k+i} = b_i$.

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String Matching Example

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String Matching Worst Case

Brute force isn't too bad for small patterns and large alphabets.

However, try finding: yyyyyx in: yyyyyyyyyyyyx

Alternatively, consider searching for: xyyyyy

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Title page

Students should be familiar with inductive proofs, recursion, data structures, and programming at the CS3114 level.



no notes



O(mn) comparisons in worst case.



Our example was a little pessimistic... but it wasn't worst case!

In the second example, we can quickly reject a position - no backtracking.

Finding a Better Algorithm

Find B = xyxyyxyxyxx in

A = xyxxyxyxyxyxyxyxyxyxxx When things go wrong, focus on what the prefix might be.

xyxxyxyxyxyxyxyxyxxxxxxxxxyxy -- no chance for prefix until third x xyxyy -- xyx could be prefix xyxyyxyxxxx -- last xyxy could be prefix xyxyyxyxxxx -- success!

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Knuth-Morris-Pratt Algorithm

- Key to success:
 - Preprocess B to create a table of information on how far to slide B when a mismatch is encountered.
- Notation: B(i) is the first i characters of B.
- For each character:
 - We need the <u>maximum suffix</u> of B(i) that is equal to a prefix of B.
- next(i) =the maximum j (0 < j < i 1) such that $b_{i-j}b_{i-j+1}\cdots b_{i-1} = B(j)$, and 0 if no such j exists.
- We define next(1) = -1 to distinguish it.
- next(2) = 0. Why?

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Computing the table

B =

1 2 3 4 5 6 7 8 9 10 11 x y x y y x y x y x x x -1 0 0 1 2 0 1 2 3 4 3

- The third line is the "next" table.
- At each position ask "If I fail here, how many letters before me are good?"

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How to Compute Table?

- By induction.
- Base cases: next(1) and next(2) already determined.
- Induction Hypothesis: Values have been computed up to next(i - 1).
- Induction Step: For next(i): at most next(i-1) + 1.
 - ▶ When? $b_{i-1} = b_{next(i-1)+1}$.
 - ▶ That is, largest suffix can be extended by b_{i-1} .
- If $b_{i-1} \neq b_{next(i-1)+1}$, then need new suffix.
- But, this is just a mismatch, so use next table to compute where to check.

 Not only can we skip down several letters if we track the potential prefix, we don't need even to repeat the check of the prefix letters – just start that many characters down.

Knuth-Morris-Pratt Algorithm

**Top success

**Top

In all cases other than B[1] we compare current A value to appropriate B value. The test told us there was no match at that position. If B[1] does not match a character of A, that character is completely rejected. We must slide B over it.

Why? All that we know is that the 2nd letter failed to match. There is no value j such that 0 < j < i - 1. Conceptually, compare beginning of B to current character.



no notes



Induction step: Each step can only improve by 1.

While this is complex to understand, it is efficient to implement.

Complexity of KMP Algorithm

- A character of A may be compared against many characters of B.
 - For every mismatch, we have to look at another position in the table.
- How many backtracks are possible?
- If mismatch at b_k , then only k mismatches are possible.
- But, for each mismatch, we had to go forward a character to get to b_k.
- Since there are always n forward moves, the total cost is O(n).

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Example Using Table

Note: -x means don't actually compute on that character.

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Boyer-Moore String Match Algorithm

- Similar to KMP algorithm
- Start scanning B from end of B.
- When we get a mismatch, we can shift the pattern to the right until that character is seen again.
- Ex: If "Z" is not in B, can move m steps to right when encountering "Z".
- If "Z" in B at position i, move m i steps to the right.
- This algorithm might make less than *n* comparisons.
- Example: Find abc in

xbycabc abc abc

ab

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Probabilistic Algorithms

All algorithms discussed so far are deterministic.

<u>Probabilistic</u> algorithms include steps that are affected by <u>random</u> events.

Example: Pick one number in the upper half of the values in a set.

- **1** Pick maximum: n-1 comparisons.
- Pick maximum from just over 1/2 of the elements: n/2 comparisons.

Can we do better? Not if we want a guarantee.



no note



no note



Better for larger alphabets.



no notes

Probabilistic Algorithm

- Pick 2 numbers and choose the greater.
- This will be in the upper half with probability 3/4.
- Not good enough? Pick more numbers!
- For *k* numbers, greatest is in upper half with probability $1 - 2^{-k}$.
- Monte Carlo Algorithm: Good running time, result not guaranteed.
- Las Vegas Algorithm: Result guaranteed, but not the running time.

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Searching Linked Lists

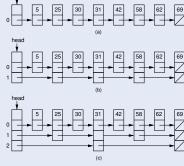
Assume the list is sorted, but is stored in a linked list.

Can we use binary search?

- Comparisons?
- "Work?"

What if we add additional pointers?

CS 5114: Theory of Algorithms "Perfect" Skip List

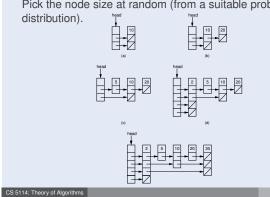


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Building a Skip List

Pick the node size at random (from a suitable probability distribution).



2014-02-28 CS 5114 Probabilistic Algorithm

Pick k big enough and the chance for failure becomes less than the chance that the machine will crash (i.e., probability of even getting an answer from a deterministic algorithm).

Rather have no answer than a wrong answer? If k is big enough, the probability of a wrong answer is less than any calamity with finite probability - with this probability independent of n.

2014-02-28 CS 5114 Searching Linked Lists

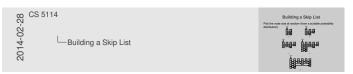
Same. Is this a good model? No.

Much higher since we must move around a lot (without comparisons) to get to the same position.

Might get to desired position faster.

2014-02-28 CS 5114 00000000 10101010 "Perfect" Skip List

What is the access time? $\log n$. We can insert/delete in log *n* time as well.



no notes

Skip List Analysis (1)

What distribution do we want for the node depths?

```
int randomLevel(void) { // Exponential distrib
  for (int level=0; Random(2) == 0; level++);
  return level;
}
```

What is the worst cost to search in the "perfect" Skip List?

What is the average cost to search in the "perfect" Skip List?

What is the cost to insert?

What is the average cost in the "typical" Skip List?

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Skip List Analysis (2)

How does this differ from a BST?

- Simpler or more complex?
- More or less efficient?
- Which relies on data distribution, which on basic laws of probability?

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Exponential decay. 1 link half of the time, 2 links one quarter, 3 links one eighth, and so on.

log n.

Close to log n.

log n.

log n.



About the same.

On average, about the same if data are well distributed.

BST relies on data distribution, while skiplist merely relies on chance.