

Lecture 19,20: String Matching Algorithms / Tries

BT 3051 – Data Structures and Algorithms for Biology

Karthik Raman

Department of Biotechnology
Indian Institute of Technology Madras

KNUTH-MORRIS-PRATT ALGORITHM

Knuth–Morris–Pratt Algorithm

- ▶ What is a major inefficiency of Brute-Force/Boyer-Moore (in the worst cases)?
- ▶ For a certain alignment of the pattern, if we find several matching characters but then detect a mismatch, we ignore all the information gained by the successful comparisons after restarting with the next incremental placement of the pattern!
- ▶ *What can we do instead?*

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- Consider a partial match

$$\begin{array}{ccccccc}
 a_1 & a_2 & \dots & a_{j-r+1} & \dots & a_j & a_{j+1} \\
 & & & = & = & = & \neq \\
 & & & b_1 & \dots & b_r & b_{r+1}
 \end{array}$$

- In the next step, we will want to align

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- ▶ But, this information comes actually from the pattern itself, rather than the text!
- ▶ Can we intelligently pre-process the pattern, so that we can appropriately restart matching?
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- ▶ $O(m)$ for the pre-processing of the pattern
- ▶ The complexity analysis is non-trivial

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String Matching

Algorithm Design Summary (Skiena)

- ▶ Is the search pattern/text short? *Naïve matching*
- ▶ Is the search pattern/text very long? *Knuth–Morris–Pratt*
- ▶ Do we expect to find the pattern or not? *Boyer–Moore*
- ▶ Will we perform multiple queries on the same text? *Suffix trees*
- ▶ Will we search many texts using the same pattern? *Complex algorithms ...*
- ▶ What if the input contains a spelling error? *Approximate string matching*

QUERYING MULTIPLE PATTERNS

Genome Sequencing

Challenges

- ▶ **Fragment (read) assembly**
 - ▶ Inevitable for new species
 - ▶ Computationally challenging/expensive
- ▶ Read mapping (Fragment alignment)

Why is this important?

- ▶ Huge number of genomes being sequenced
- ▶ e.g. Personal Genome Project (UK): 10^5 genomes!

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How do we assemble individual genomes efficiently using a reference genome?

or

How do we match several patterns against a single text!?

Exact Pattern Matching: Multiple Patterns

Problem

Where do each of the reads match the reference genome exactly?

- ▶ Input: A collection of strings (reads: *Patterns*) and a string (genome: *Text*)
- ▶ Output: All positions in the genome *Text* where a string in *Patterns* appears as a substring

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Can we store the patterns intelligently, in an efficient data structure?

TRIES

Trie

- ▶ Let S be a set of s strings from alphabet Σ such that *no string in S is a prefix of another string*
- ▶ A standard trie^a for S is an ordered tree T with the following properties:
 - ▶ Each node of T , except the root, is labeled with a character of Σ
 - ▶ The children of an internal node of T have distinct labels
 - ▶ Each leaf node of T is associated with a string of Σ such that the label of each node on the path from the root to the leaf is a prefix of the string of Σ associated with a leaf
- ▶ Thus, a trie T represents the strings of S with paths from the root to the leaves of T
- ▶ Note the importance of assuming that no string in S is a prefix of another string
 - ▶ this ensures that each string of S is uniquely associated with a leaf of T

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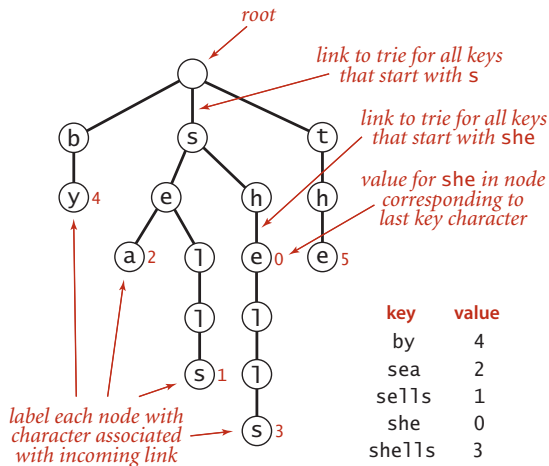
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Example Trie

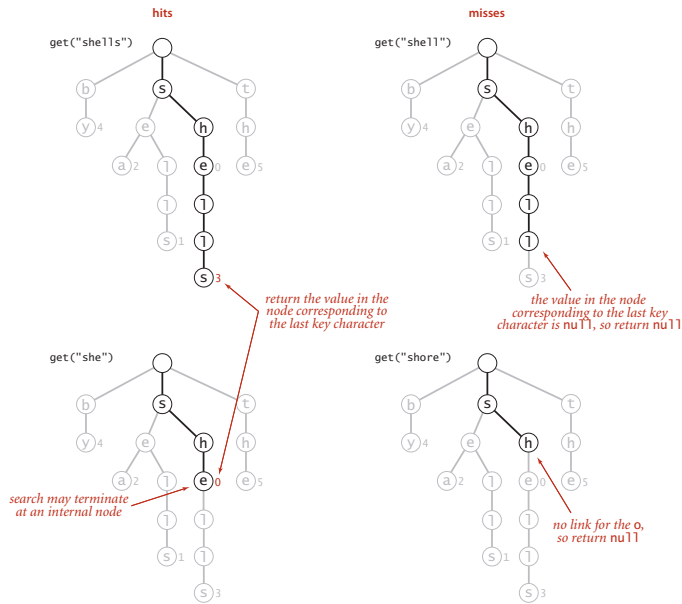
Figure Courtesy: Sedgewick



key	value
by	4
sea	2
sells	1
she	0
shells	3
the	5

Trie Searches

Figure Courtesy: Sedgewick



How to match patterns vs. text?

- ▶ Slide the trie down the text!
- ▶ Search for patterns at every step
- ▶ Memory usage is high!

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What is the time complexity?

