#### Full-text indexes

## Indexing sequence data

- Large sequence data should be stored into data structures (indexes) that support efficient and fast query and retrieval
- Example 1: to look up for specific patterns, we can use an inverted index
- Example 2: to look up for patterns (strings) of fixed length k (k-mers) we can use a hash table
- What if we need to search for patterns of arbitrary length?
- full-text indexes allow a search for patterns of any length occurring at any position of the text

#### What is a full-text index?

Text index = a data structure built from a given text (string, sequence) that supports certain type of (typically pattern look-up) queries

genomic sequences don't have word structure (cf inverted indexes) and query size is arbitrary (cf hash tables)

- other examples:
  - Chinese, Korean languages
  - agglutinative/inflectional languages (Finnish, German, ...):
     looking for particles
  - MIDI files, audio signals, program code, numerical sequences, ...
- **full-text indexes** allow a search for patterns of any length occurring at any position of the text

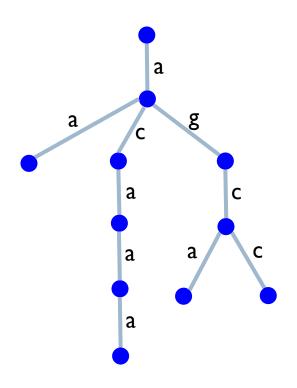
## "Ideal" index structure for T[1...n]

- $\blacktriangleright$  Takes space O(n)
- Can be constructed in time O(n)
- All occurrences of a query pattern P can be reported in time O(|P| + occ)

# Suffix tree

## Trie (aka digital tree)

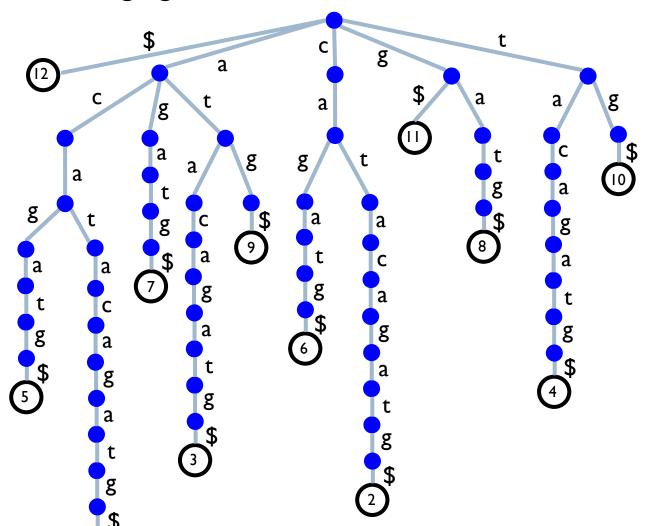
trie for {agca,acaaa,agcc,aa}



- every string of the set is "spelled" starting from root
- edges outgoing from a node are labeled by different characters
- trie can be viewed as an automaton recognizing the given set of strings (or all their prefixes)

#### Suffix trie

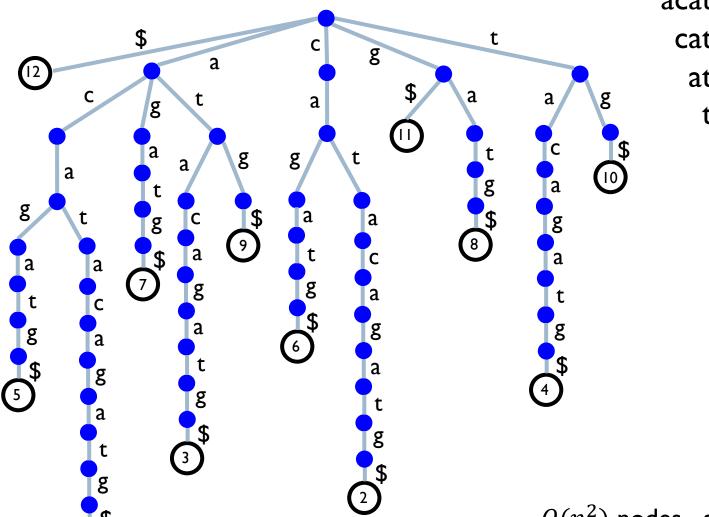
T=acatacagatg\$



acatacagatg\$ catacagatg\$ atacagatg\$ tacagatg\$ acagatg\$ cagatg\$ agatg\$ gatg\$ atg\$ 10 tg\$ g\$ 12

#### Suffix trie

T=acatacagatg\$

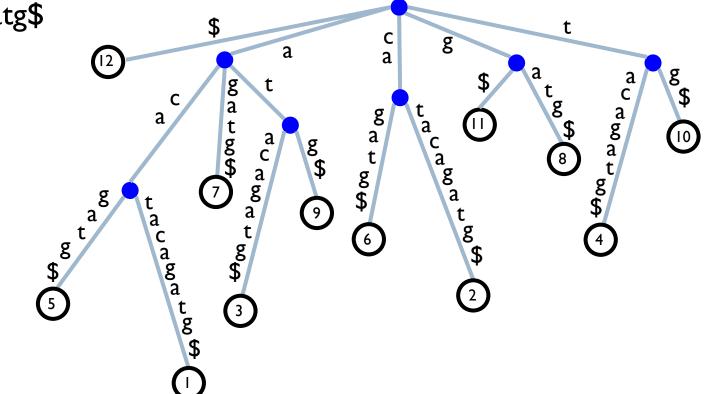


acatacagatg\$ catacagatg\$ atacagatg\$ tacagatg\$ acagatg\$ cagatg\$ agatg\$ gatg\$ atg\$ 10 tg\$ g\$ 12

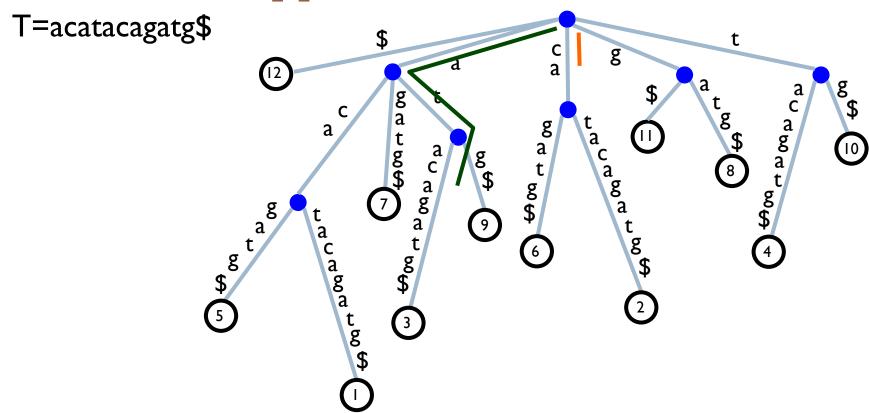
 $O(n^2)$  nodes  $a^nb^n$ 

#### Suffix tree

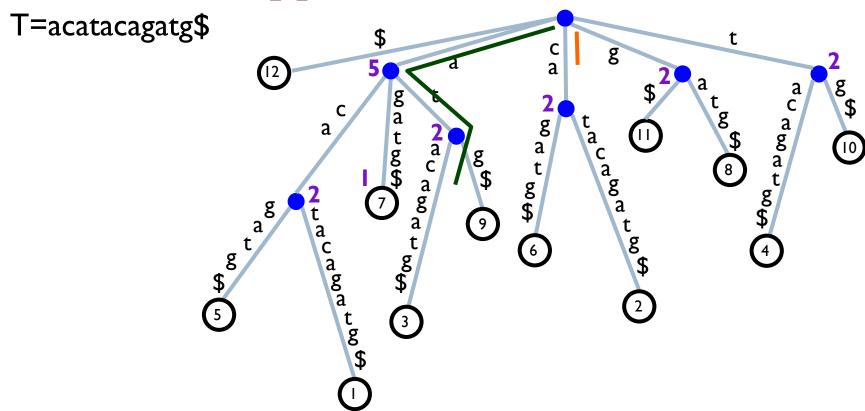
T=acatacagatg\$



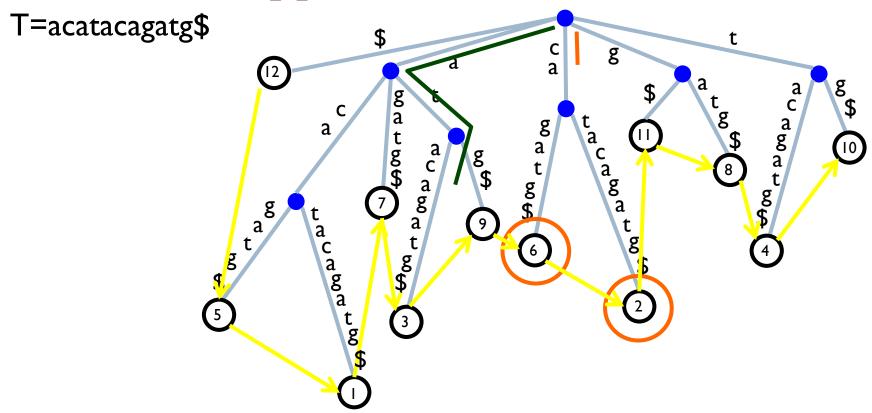
- explicit vs implicit nodes
- an edge label is start and end positions of corresponding substring (rather than substring itself)
- takes space O(n)



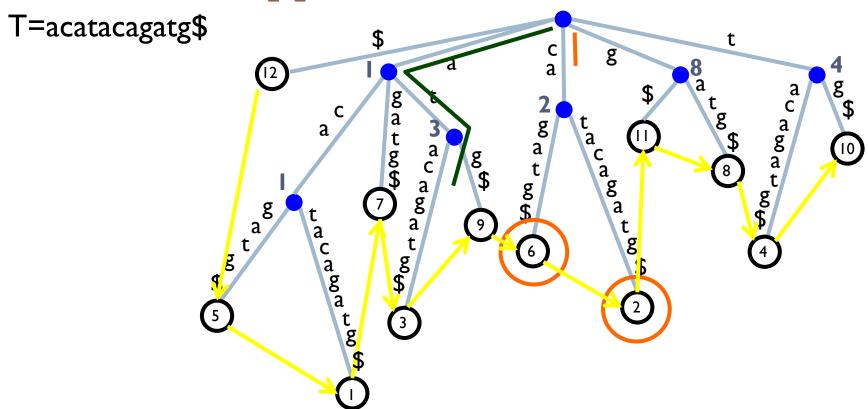
check if a pattern P occurs in T in time O(|P|). Ex:  $P_1$ =atac  $P_2$ =c



- check if a pattern P occurs in T in time O(|P|). Ex:  $P_1$ =atac  $P_2$ =c
- report the number of occurrences in  $O(|P|) \Rightarrow$  preprocess number of leaves



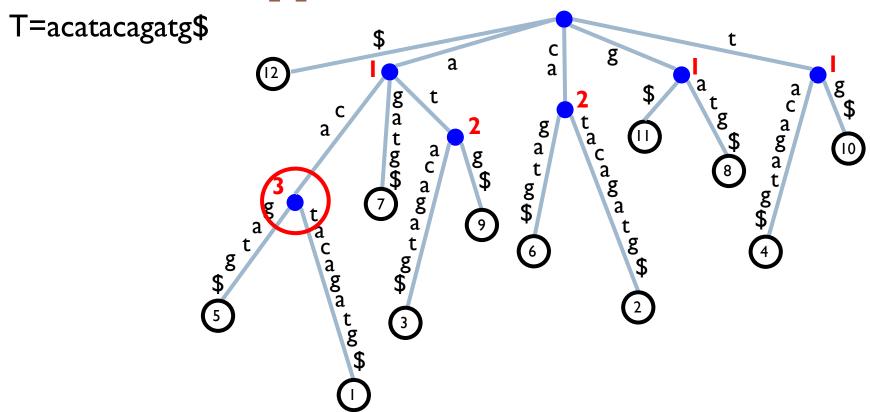
- check if a pattern P occurs in T in time O(|P|). Ex:  $P_1$ =atac  $P_2$ =Q
- report the number of occurrences in  $O(|P|) \Rightarrow$  preprocess number of leaves
- report all occurrences in  $O(|P| + occ) \Rightarrow$  chain leaves and preprocess leftmost and rightmost leaves



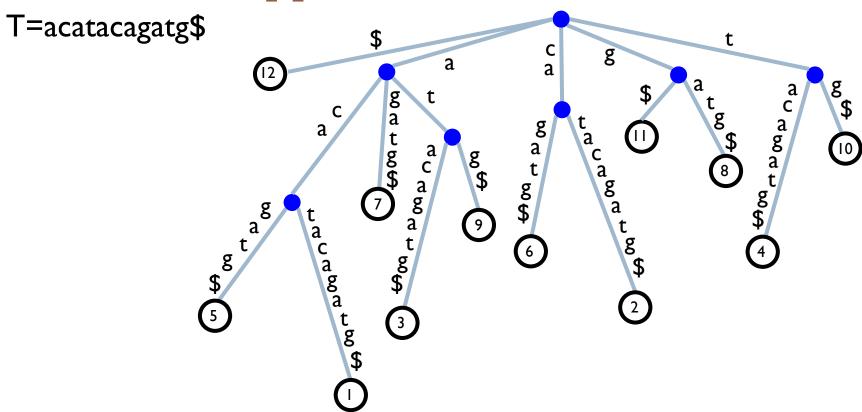
- check if a pattern P occurs in T in time O(|P|). Ex:  $P_1$ =atac  $P_2$ =Q
- report the number of occurrences in  $O(|P|) \Rightarrow$  preprocess number of leaves
- report all occurrences in  $O(|P| + occ) \Rightarrow$  chain leaves and preprocess leftmost and rightmost leaves
- report the first (leftmost) occurrence in  $O(|P|) \Rightarrow$  preprocess minimal leaf label

## Quiz 9.2

- Construct the suffix tree for string MISSISSIPPI\$
- ▶ How many internal nodes does it have?



 longest repeated substring ⇒ deepest (w.r.t. string depth) internal node in the suffix tree (aca)



- longest extension queries: given two positions i,j, output the length of the longest common substring starting at i,j
- reduces to lowest common ancestor (lca) queries
- · *lca* queries can be answered in O(1) time after linear-time preprocessing of the tree [Harel, Tarjan 84], [Bender, Farach-Colton 00]

## Suffix tree: some history

- Suffix tree can be constructed in time O(n)
  - Weiner 1973: right-to-left construction

#### LINEAR PATTERN MATCHING ALGORITHMS

Peter Weiner

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#### Abstract

In 1970, Knuth, Pratt, and Morris [1] showed how to do basic pattern matching in linear time. Related problems, such as those discussed in [4], have previously been solved by efficient but sub-optimal algorithms. In this paper, we introduce an interesting data structure called a bi-tree. A linear time algorithm for obtaining a compacted version of a bi-tree associated with a given string is presented. With this construction as the basic tool, we indicate how to solve several pattern match as problems, including some from [4], in linear time.

#### I. Introduction

In 1970, Knuth, Morris, and Pratt [1-2] showed how to match a given pattern into another given string in time proportional to the sum of the lengths of the pattern and string. Their algorithm was derived from a result of Cook [3] that the 2-way deterministic pushdown languages are recognizable on a random access machine in time 0(n). Since 1970, attention has been given to several related problems in pattern matching [4-6], but the algorithms developed in these investigations usually run in time which is slightly worse than linear, for example 0(n log n). It is of considerable interest to either establish that there exists a non-linear lower bound on the run time of all algorithms which solve a given pattern matching problem, or to exhibit an algorithm whose run time is of 0(n).

In the following sections, we introduce an interesting data structure, called a bi-tree, and show how an efficient calculation of a bi-tree can be applied to

done giving a formal definition of a bi-tree, we reversely decompositions and terminology concerning t-ary trees where Knuth [7] for further details.)

A transfer T over  $\Sigma = \{\sigma_1, \dots, \sigma_t\}$  is a set of

nodes N which is there empty or consists of a root,

note N, and t order d, definit t-ary trees.

Clearly, every note t \in N is the root of some

Clearly, every note  $C \in \mathbb{N}$  is the root of some t-ary tree  $T^1$  which iself this its of  $n_1$  and t ordered disjoint t-ary tree,  $(Y, T_1^1, \dots, T_n^1)$ . We call the tree  $T_1^1$  a sub-tree of  $T_1^1$ ,  $(D_1, 1)$  so tree of  $T_1^1$  are considered to be sub-trees of  $(T_1^1, T_1^1)$  is navir to associate with a tree T a successor T retor

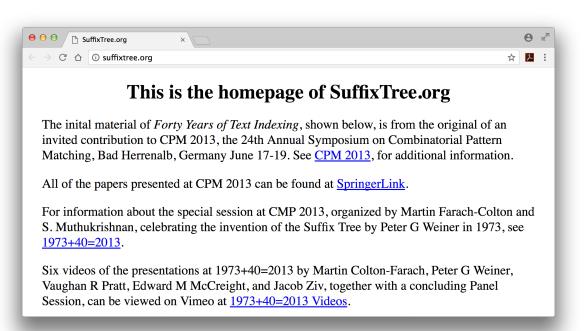
S: 
$$N \times \Sigma \rightarrow (N - \{n_0\}) \cup \{NIL\}$$

defined for all  $n_i \in N$  and  $\sigma_j \in \Sigma$  by

n the root of T if T is non-empty

#### Suffix tree: some history

- Suffix tree can be constructed in time O(n)
  - Weiner 1973: right-to-left construction
  - McCreight 1976: left-to-right
  - Ukkonen 1995: left-to-right and online
  - Farach 1997: for integer alphabets



#### review articles He first looked for the most frequent symbol and changed it into the most frequent letter of English, then simi-Tracing the first four decades in the life of suffix trees, their many incarnations, larly inferred the most frequent word, and their applications. then punctuation marks, and so on. Both before and after 1843, the BY ALBERTO APOSTOLICO, MAXIME CROCHEMORE, MARTIN FARACH-COLTON, ZVI GALIL, AND S. MUTHUKRISHNAN natural impulse when faced with some mysterious message has been to count frequencies of individual to kens or subassemblies in search of a clue. Perhaps one of the most intense 40 Years and fascinating subjects for this kind of scrutiny have been biosequences. As soon as some such sequences be came available, statistical analysts tried to link characters or blocks of of Suffix Trees characters to relevant biological functions. With the early examples of whole genomes emerging in the mid 1990s, it seemed natural to count the occurrences of all blocks of size 1, 2, and so on, up to any desired length looking for statistical characteriza-tions of coding regions, promoter regions, among others. This article is not about cryptogra WHEN WILLIAM LEGRAND finally decrypted the string, phy. It is about a data structure and its variants, and the many surprising and useful features it carries. Among it did not seem to make much more sense than it did before. these is the fact that, to set up a sta-tistical table of occurrences for all substrings (also called factors), of any 53###305))6\*,48264#.)4z);806",48#8P60))85;1# length, of a text string of n characters, it only takes time and space linear in (;: \$\pm\$\*8\pm\$83(88)5\*\pm\$,46(;88\*96\*?;8)\* \$\pm\$(;485);5\*\pm\$2:\* \$\pm\$ (;4956\*2(5\*Ñ4)8P8\*;4069285);)6‡8)4‡‡;1(‡9;48081;8; the length of the text string. While no-body would be so foolish as to solve 8\pmu1;4885;4)485\pmu528806\*81(ddag9;48;(88;4(\pmu734; the problem by first generating all exponentially many possible strings and then counting their occurrences 48)4‡;161;:188;‡?; one by one, a text string may still contain $\Theta(n^2)$ distinct substrings, so that The decoded message read: "A good glass in the tabulating all of them in linear space bishop's hostel in the devil's seat forty-one degrees

and thirteen minutes northeast and by north main

branch seventh limb east side shoot from the left eye

of the death's-head a bee line from the tree through

the shot fifty feet out." But at least it did sound more

main character of Edgar Allan Poe's "The Gold-Bug"36

solved a substitution cipher using symbol frequencies.

to discover the treasure he had been after. Legrand

like natural language, and eventually guided the

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never mind linear time, already seems

We dedicate this article to

who passed away on July 20.

our friend and colleague Alberto Apostolico (1948–2015),

He was a major figure in

the development of

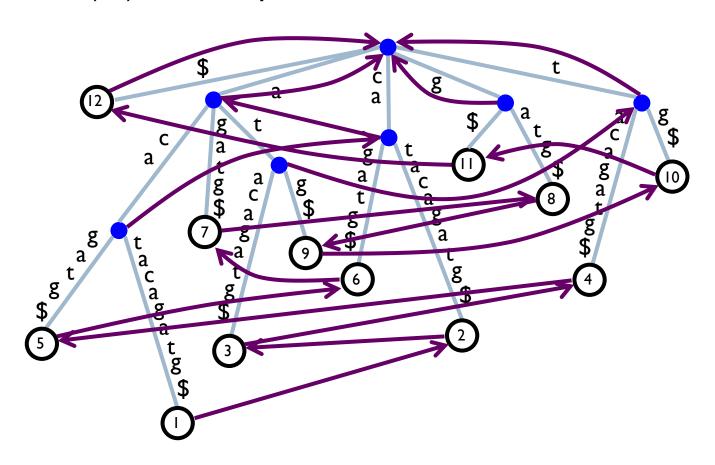
algorithms on strings

puzzling.

## Suffix tree augmented with suffix links

McCreight and Ukkonen use suffix links

suf-link( $\overline{au}$ )= $\overline{u}$  for explicit nodes  $\overline{au}$ 



## Suffix tree augmented with suffix links

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