Suffix Trees

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Algorithms on Strings Data Structures and Algorithms

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For the complete set of frames,

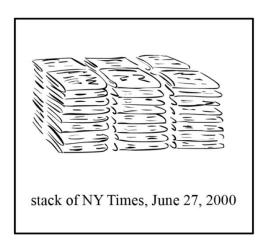
please see our videos in the

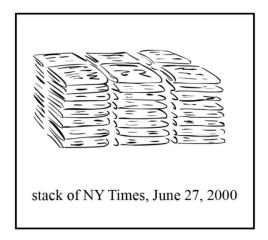
Algorithms on Strings course on Coursera

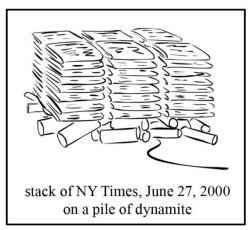
(Algorithms and Data Structures Specialization)

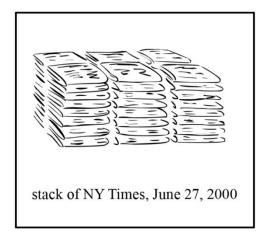
Outline

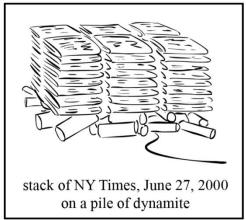
- From Genome Sequencing to Pattern Matching
- Brute Force Approach to Pattern Matching
- Herding Patterns into Trie
- Herding Text into Suffix Trie
- From Suffix Tries to Suffix Trees

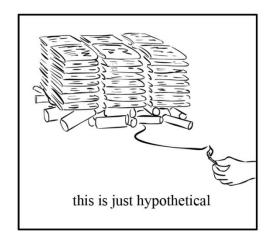




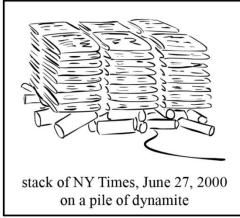


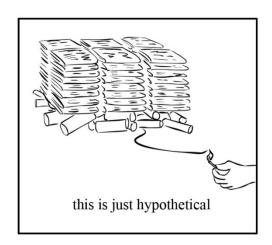






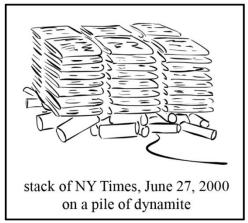


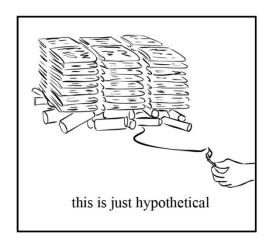




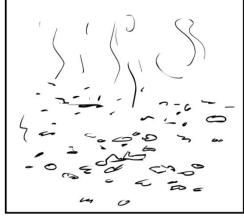




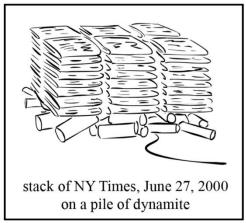


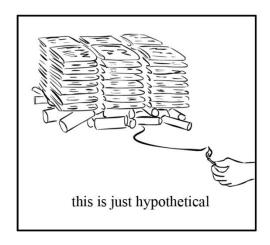




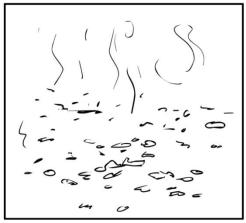


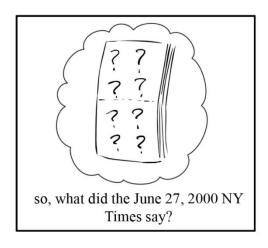




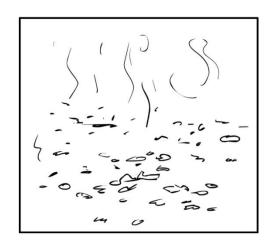








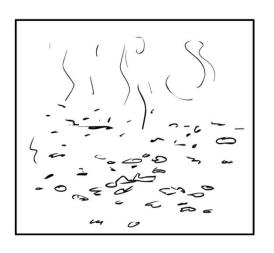
The Newspaper Problem as an Overlapping Puzzle



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yet named any suspects, alt is welc

The Newspaper Problem as an Overlapping Puzzle



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Millions of Copies of a Genome



Breaking the Genomes at Random Positions

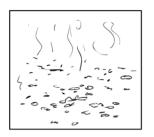


Generating Short Substring (Reads)



CTGATGA TGGACTACGCTAC TACTGCTAG CTGTATTACG ATCAGCTACCACA TCGTAGCTACG ATGCATTAGCAA GCTATCGGA TCAGCTACCA CATCGTAGC
CTGATGATG GACTACGCT ACTACTGCTA GCTGTATTACG ATCAGCTACC ACATCGTAGCT ACGATGCATTA GCAAGCTATC GGATCAGCTAC CACATCGTAGC
CTGATGATGG ACTACGCTAC TACTGCTAGCT GTATTACGATC AGCTACCAC ATCGTAGCTACG ATGCATTAGCA AGCTATCGG A TCAGCTACCA CATCGTAGC
CTGATGATGGACT ACGCTACTACT GCTAGCTGTAT TACGATCAGC TACCACATCGT AGCTACGATGCA TTAGCAAGCT ATCGGATCA GCTACCACATC GTAGC

Burning Some Reads



CTGATGA TGGACTACGCTAC TACTGCTAG CTGTATTACG ATCAGCTACCACA TCGTAGCTACG ATGCATTAGCAA GCTATCGGA TCAGCTACCA CATCGTAGC
CTGATGATG GACTACGCT ACTACTGCTA GCTGTATTACG ATCAGCTACC ACATCGTAGCT ACGATGCATTA GCAAGCTATC GGATCAGCTAC CACATCGTAGC
CTGATGATGG ACTACGCTAC TACTGCTAGCT GTATTACGATC AGCTACCAC ATCGTAGCTACG ATGCATTAGCA AGCTATCGG A TCAGCTACCA CATCGTAGC
CTGATGATGGACT ACGCTACTACT GCTAGCTGTAT TACGATCAGC TACCACATCGT AGCTACGATGCA TTAGCAAGCT ATCGGATCA GCTACCACATC GTAGC

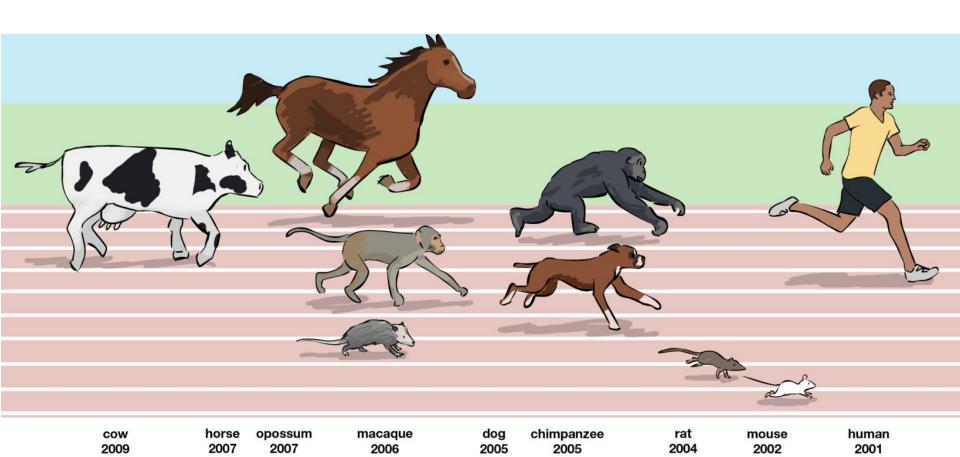
Assembling Genome



CTGATGA TGGACTACGCTAC TACTGCTAG CTGTATTACG ATCAGCTACCACA TCGTAGCTACG ATGCATTAGCAA GCTATCGGA TCAGCTACCA CATCGTAGC
CTGATGATG GACTACGCT ACTACTGCTA GCTGTATTACG ATCAGCTACC ACATCGTAGCT ACGATGCATTA GCAAGCTATC GGATCAGCTAC CACATCGTAGC
CTGATGATGG ACTACGCTAC TACTGCTAGCT GTATTACGATC AGCTACCAC ATCGTAGCTACG ATGCATTAGCA AGCTATCGG A TCAGCTACCA CATCGTAGC
CTGATGATGGACT ACGCTACTACT GCTAGCTGTAT TACGATCAGC TACCACATCGT AGCTACGATGCA TTAGCAAGCT ATCGGATCA GCTACCACATC GTAGC







Why Do We Sequence 1000s of Species?















Applications in

- medicine (mouse genome)
- agriculture (rice genome)
- biotechnology (genomes of energy-producing bacteria),
- -etc., etc., etc.

Few Mutations Can Make a Big Difference...

- Different people have slightly different genomes: on average, roughly 1 mutation in 1000 nucleotides.
- The 1 in 1000 nucleotides difference accounts for height, high cholesterol susceptibility, and 7000+ genetic diseases.

CTGATGATGGACTACGCTACTACTGCTAGCTGTATTACGA
TCAGCTACCACATCGTAGCTACGATGCATTAGCAAGCTAT
CGATCGATCGATCGATCGATCGATCGATCA
CTATACGAGCTACTACGTACGATCGCGGGACTATTA
TCGACTACAGATAAAACATGCTAGTACAACAGTATACATA
GCTGCGGGATACGATTAGCTAATAGCTGACGATATCCGAT

CTGATGATGGACTACGCTACTACTGCTAGCTGTATTACGA
TCAGCTACAACATCGTAGCTACGATGCATTAGCAAGCTAT
CGATCGATCGATCGATCGATCGATCGATCA
CTATACGAGCTACTACGTACGATCGCTGACTATTA
TCGACTACAGATGAAACATGCTAGTACAACAGTATACATA
GCTGCGGGATACGATTAGCTAATAGCTGACGATATCCGAT

Emergence of Personalized Medicine

- 2010: Nicholas Volker became the first child to be saved by genome sequencing.
 - Doctors could not diagnose his condition; he went through dozens of surgeries.
 - Sequencing revealed a mutation in a gene linked to a defect in his immune system.
 - This led doctors to use immunotherapy,
 which saved the child.

From Reference Genome to Personal Genomes

Reference human genome assembled in 2000.

CTGAGGATGGACTACGCTACTGATAGCTGTTT						reference	
						genome	
GAGGA	CCACG		TGA-AG			•	
CTGA	GGAC	C	ACTAC	A-AGCT	I	reads	
GATGG ACGCT			Τ	Т	GTTT		



CTGAGGATGGAC**C**ACGCTACTACTGA-AGCTGTTT

personal genome

Exact Pattern Matching

 Where does a read match the reference genome exactly?

Pattern Matching Problem:

- Input: A string Pattern (read) and a string Text (genome).
- Output: All positions in *Text* where *Pattern* appears as a substring.

Approximate Pattern Matching

 Where does a read match the reference genome approximately?

Approximate Pattern Matching Problem:

- Input: A string Pattern, a string Text, and an integer d
- Output: All positions in *Text* where *Pattern* appears as a substring with at most *d* mismatches.

Multiple Pattern Matching

 Where do billions of reads match the reference genome?

Multiple Pattern Matching Problem:

- Input: A set of strings Patterns and a string Text.
- Output: All positions in *Text* where a string from *Patterns* appears as a substring.

Outline

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A Brute Force Approach to Exact Pattern Matching

Genome

CTGATGATGGACTACGCTACTACTGCTAGCTGTATTACGATCAGCTAC**C**ACATCGTAGCTAC



Pattern drives along Text

panamabananas nana

Text



panamabananas

Text

nana



panamabananas

Text

nana



pa**n**amabananas **n**ana

Text



panamabananas nana

Text



panamabananas nana

Text



panamabananas

nana

Pattern

Text

pana**m**abananas

nana

Pattern

Text

panamabananas Text nana Pattern



panama**b**ananas

Text

nana



panamabananas

n a n a

5 ...

Text





panamaba**n**anas **n**ana

Text



panamaba**na**nas **na**na

Text



panamaba**nan**as **nan**a

Text



Pattern Found!

panamaba**nana**s **nana**

Text

Pattern



•

panamaban**a**nas

Text

n a n a



Brute Force Approach Is Fast!

• single Pattern: O(|Text| • |Pattern|)



The runtime of the **Knuth-Morris-Pratt** algorithm: O(|*Text*|) (wait for the lecture on algorithmic challenges in pattern matching)



Brute Force Approach is Slow for **Billions** of Patterns

- single Pattern: O(|Text| |Pattern|)
- multiple Patterns:

```
O(\sum_{\text{all strings } Pattern} | Text | \bullet | Pattern |)
```

Brute Force Approach is Slow for Billions of Patterns

- single Pattern: O(|Text| |Pattern|)
- multiple Patterns:

$$O(\sum_{\text{all strings } Pattern} | Text | \bullet | Pattern |) = \lim_{\text{In } Patterns}$$

O(|Text| • |Patterns|)

For human genome:

- $|Text| \approx 3*10^9$
- $|Patterns| \approx 10^{12}$



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Patterns Travel One at a Time

Genome

CTGATGATGGACTACGCTACTACTGCTAGCTGTATTACGATCAGCTAC**C**ACATCGTAGCTAC











Herding Patterns onto a Bus

Genome

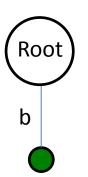
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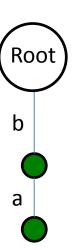




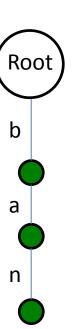
banana



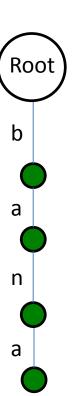
banana



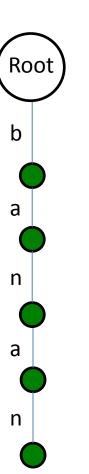
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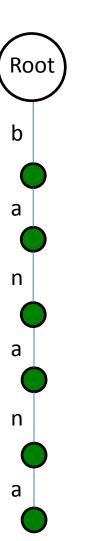
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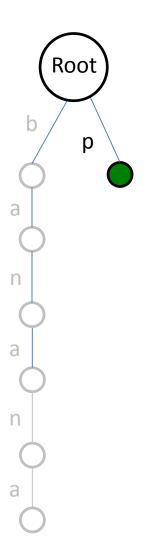
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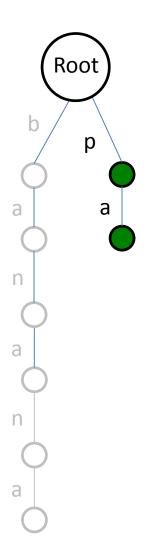


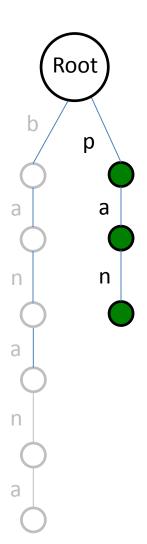
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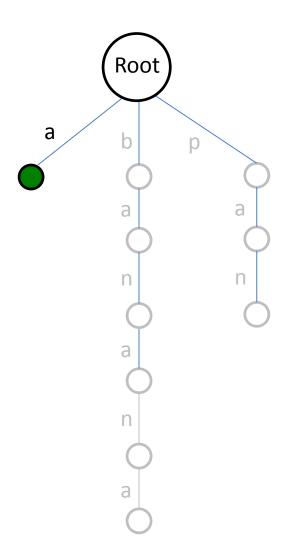


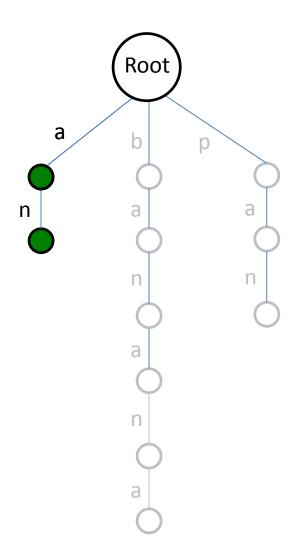
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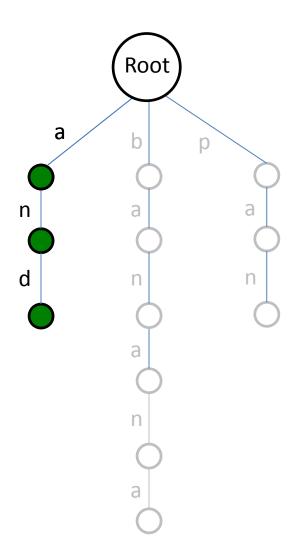


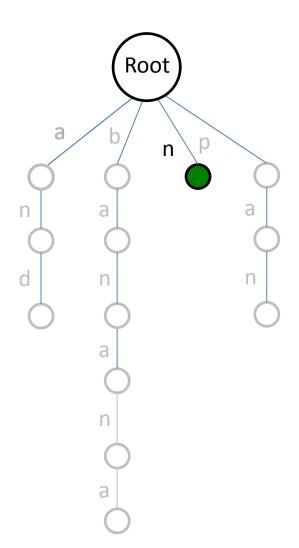


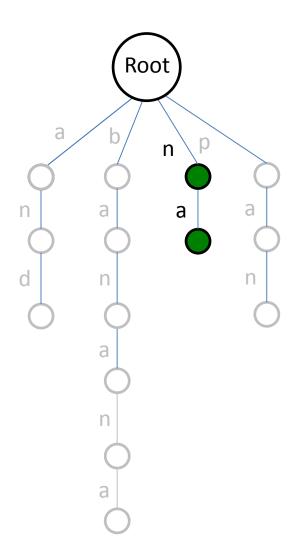


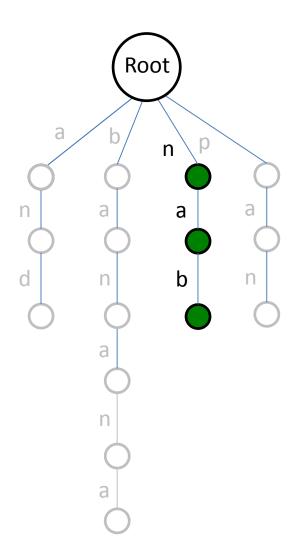


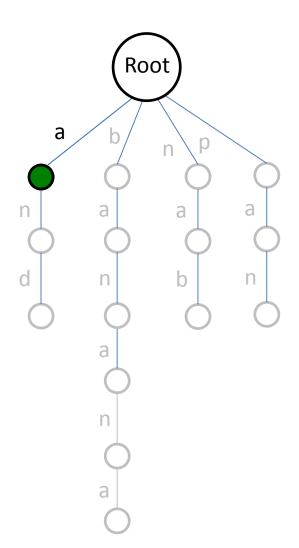


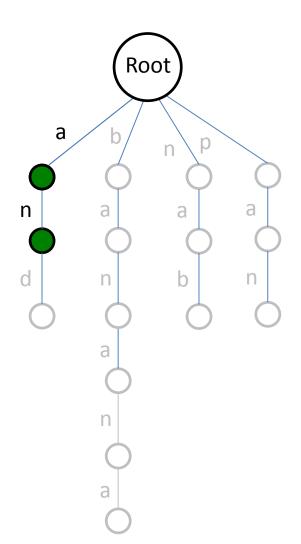


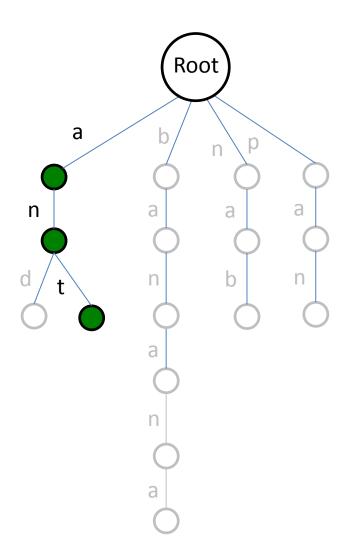


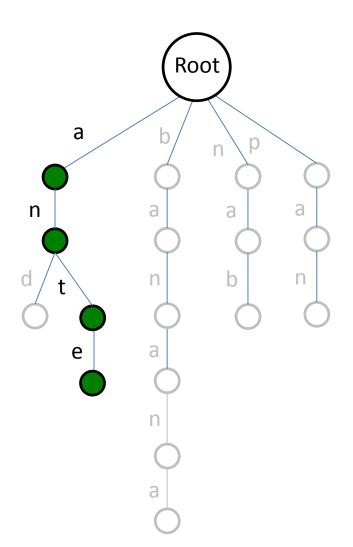


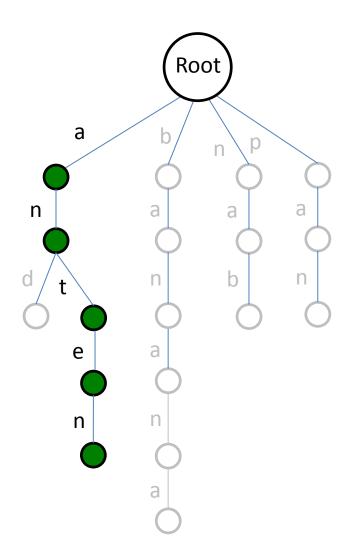


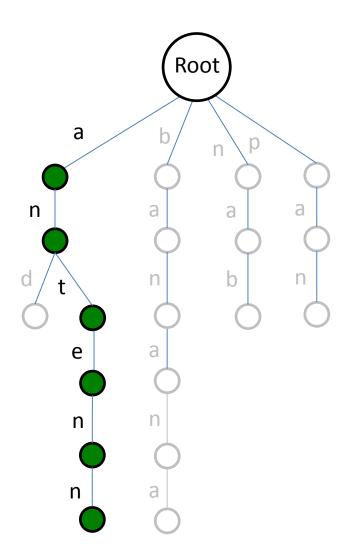


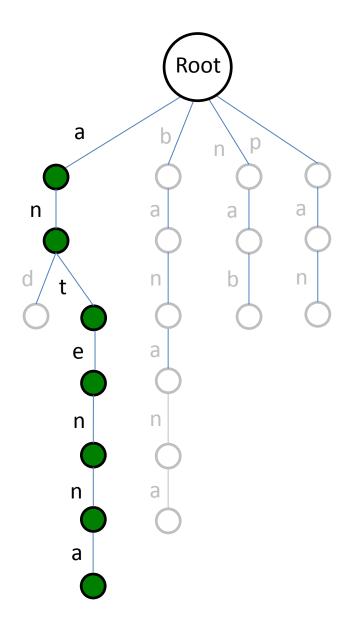


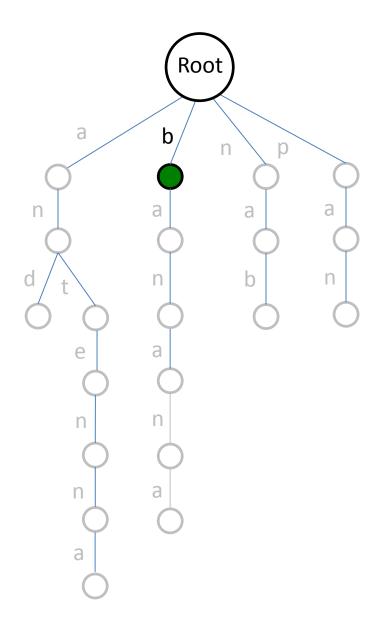


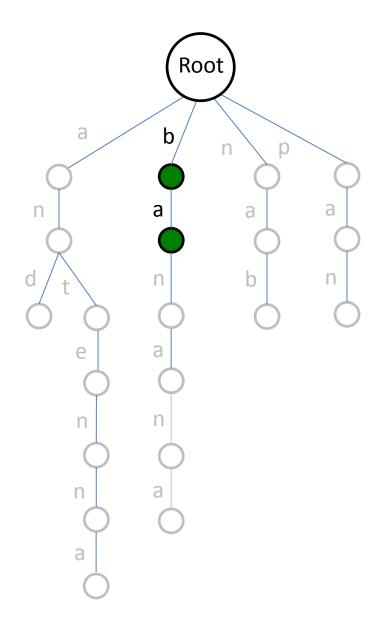


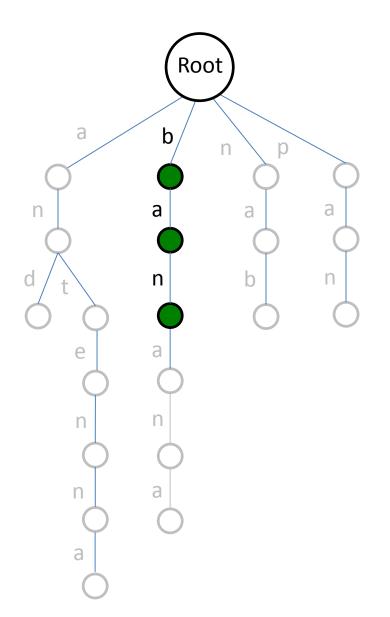


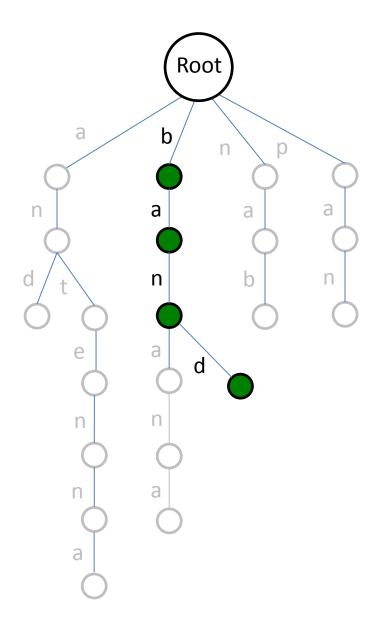


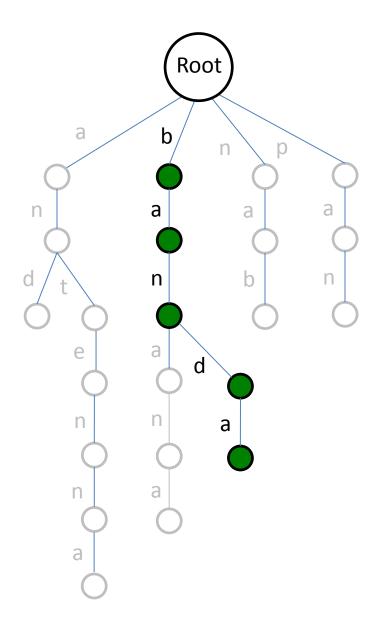


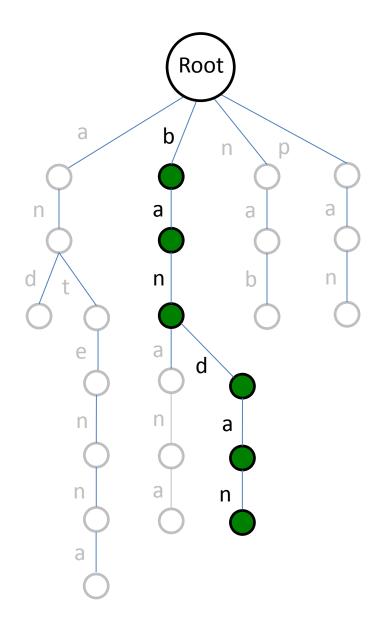


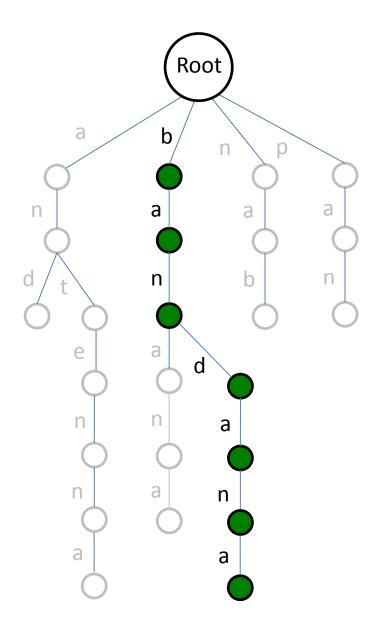


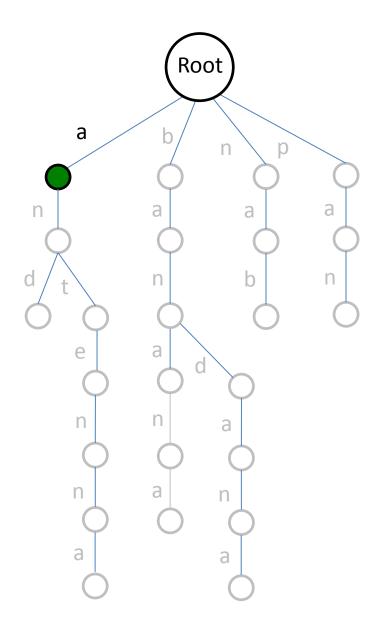


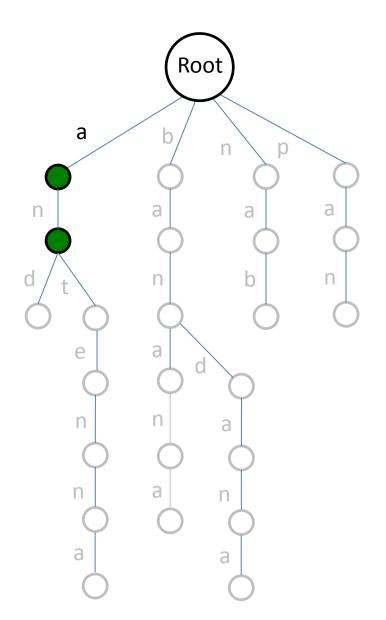


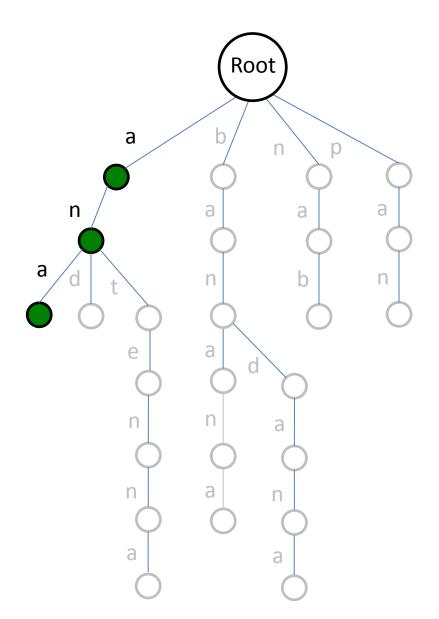


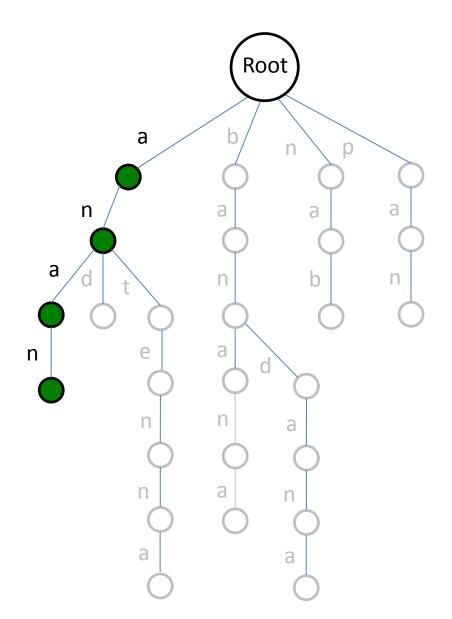


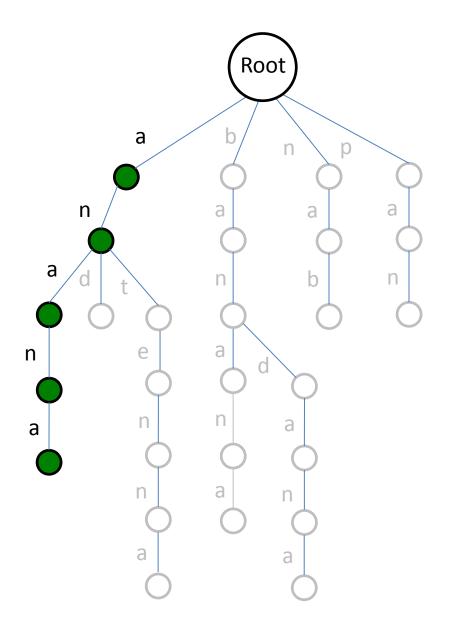


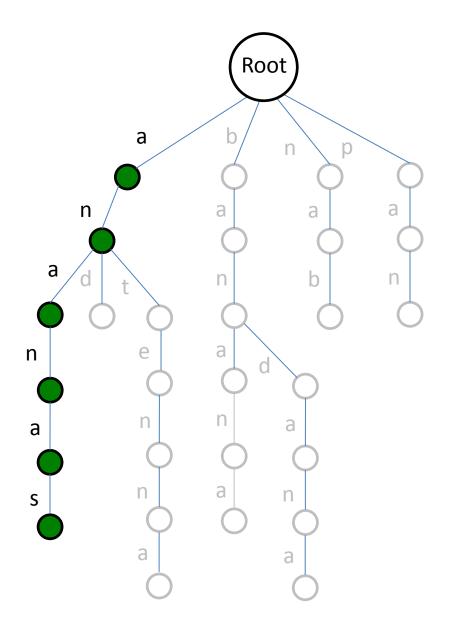


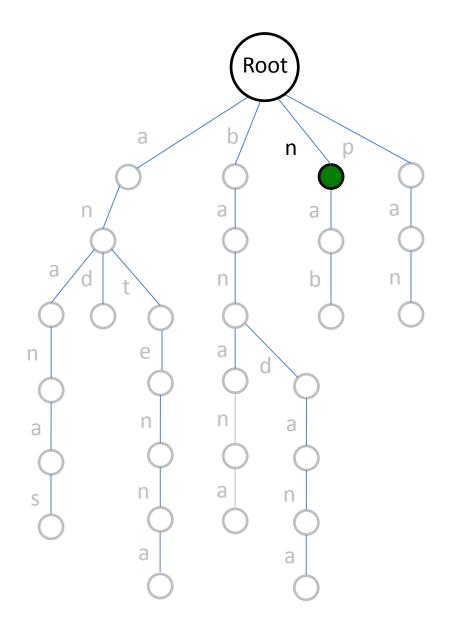


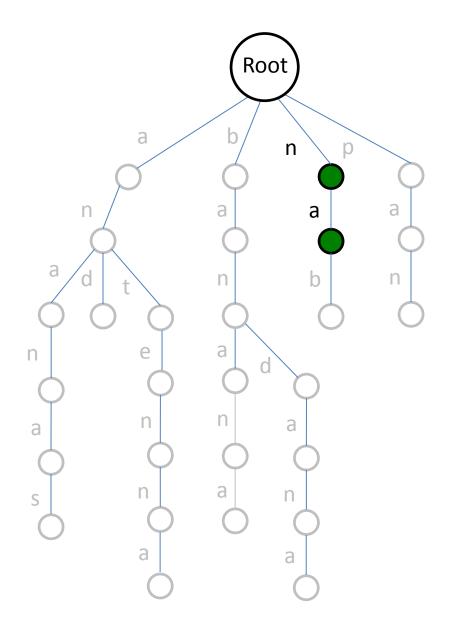


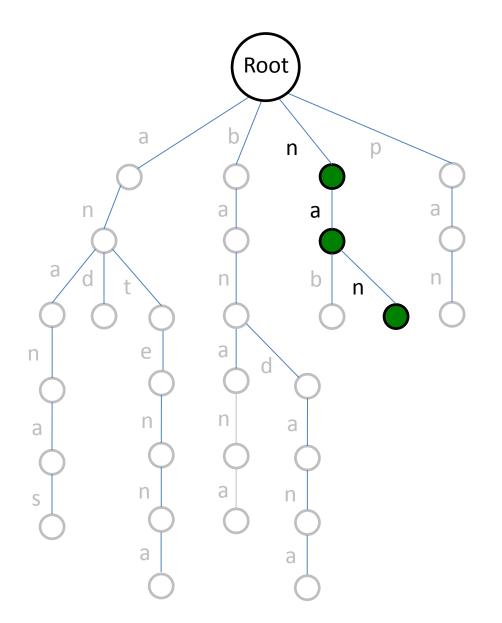






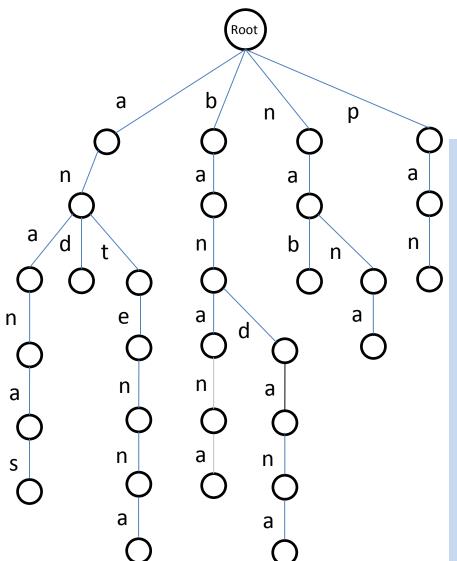






Root n n n а a а а S а **Trie**(Patterns)

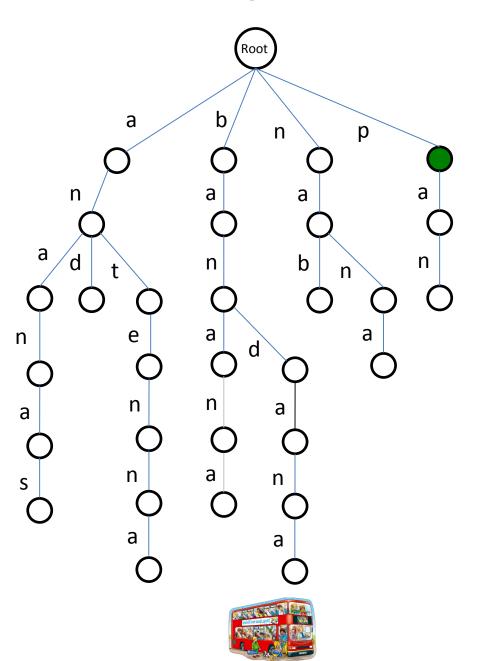
Patterns

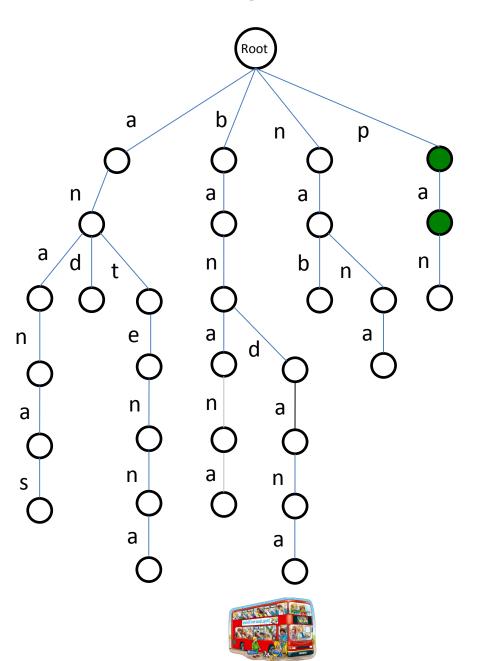


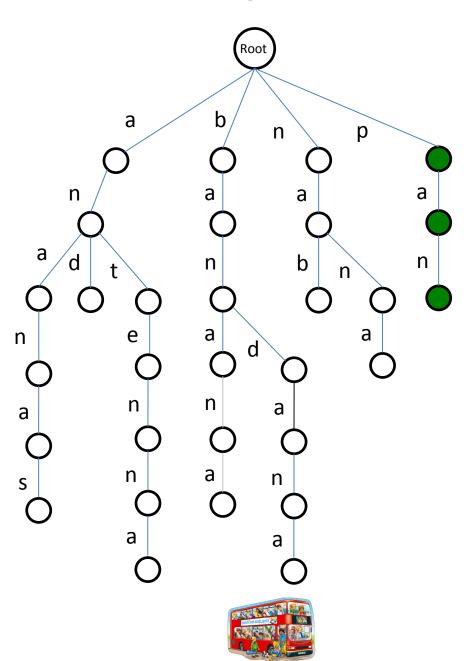
TrieMatching(*Text, Patterns*): drive Trie(*Patterns*) along *Text* at each position of *Text*

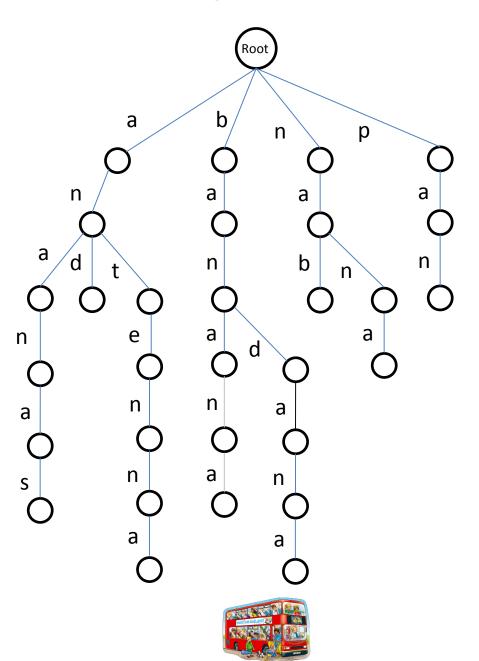
- walk down Trie(Patterns)
 by spelling symbols of Text
- a pattern from Patterns matches Text each time you reach a leaf!

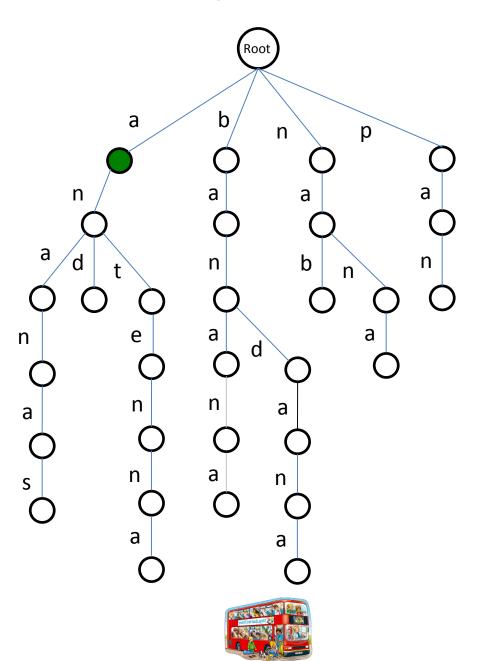
For simplicity, we assume that no pattern is a substring of another pattern

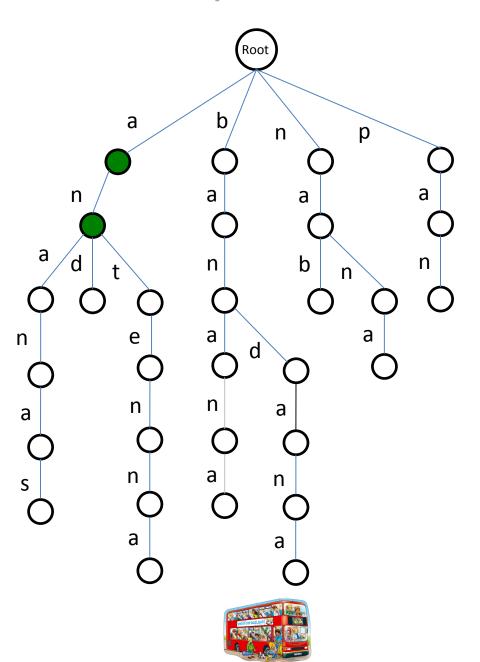


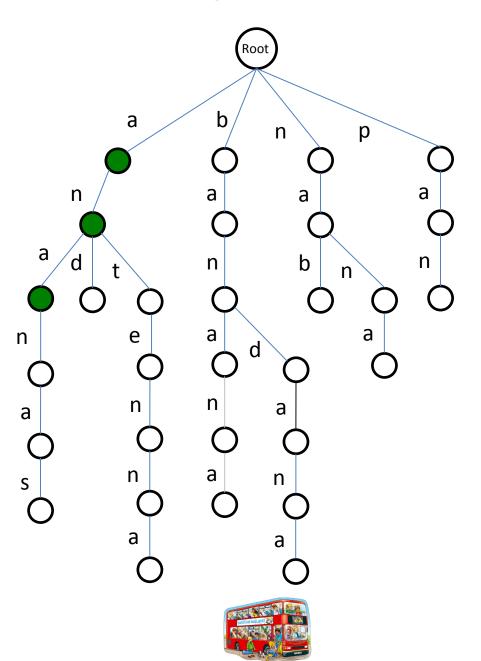


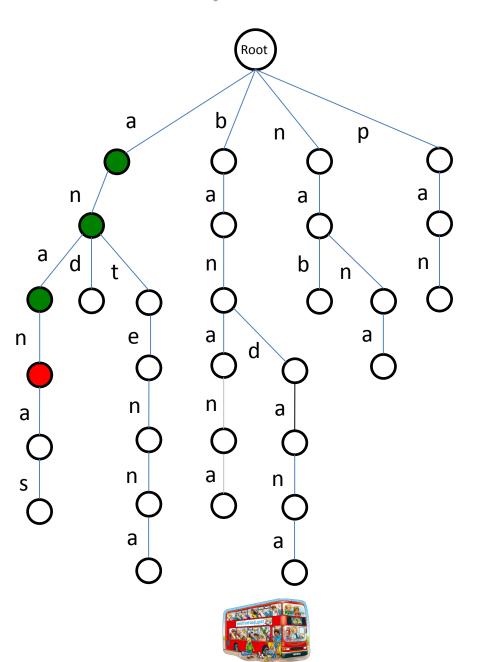


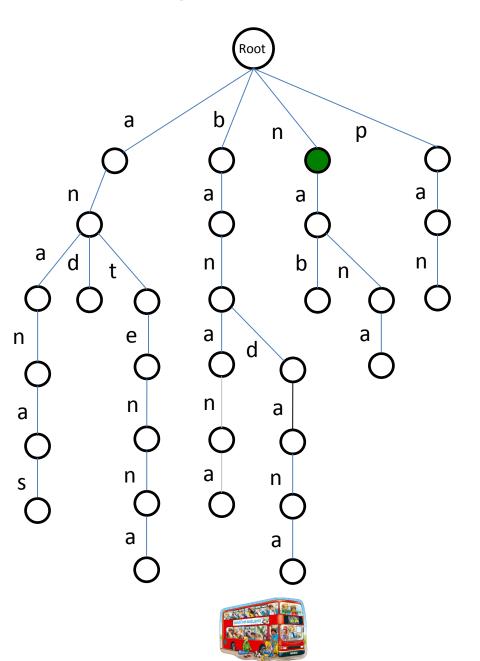


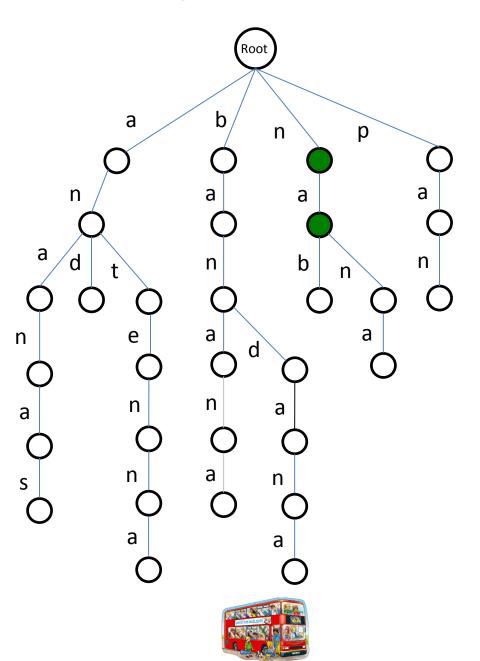


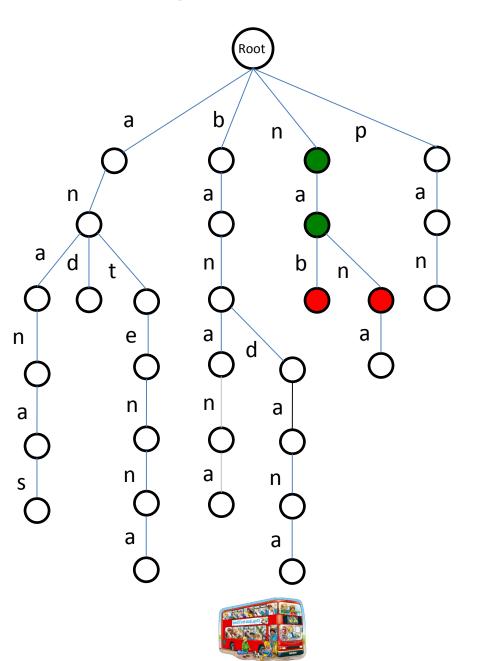


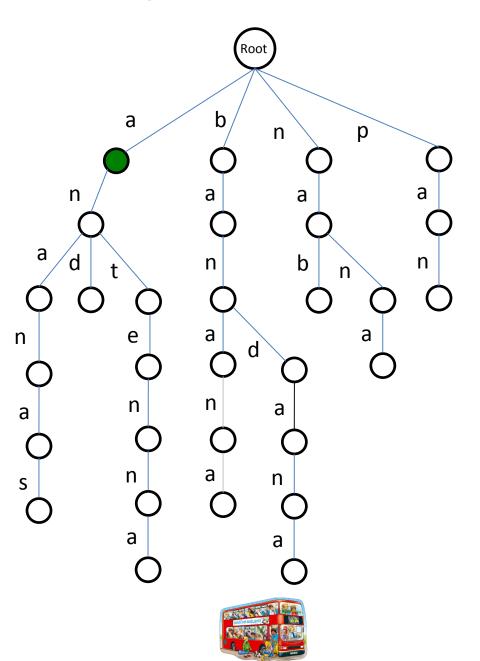


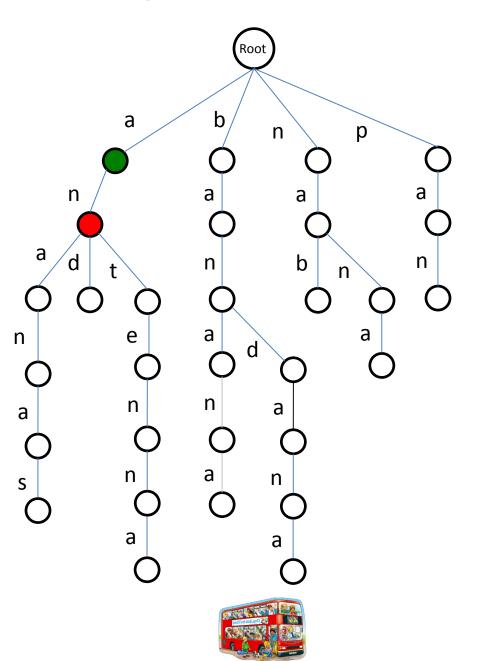


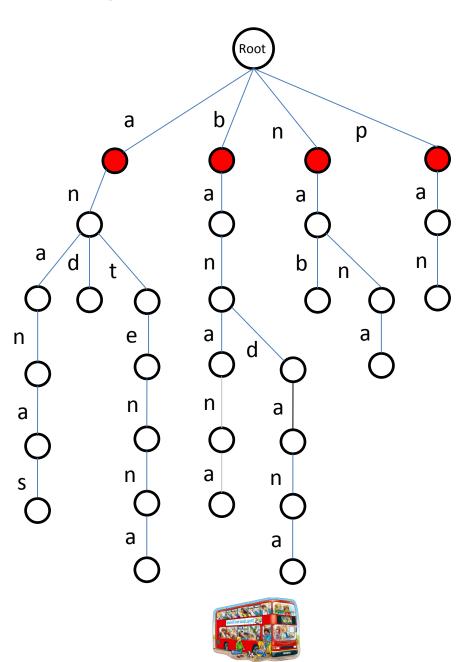


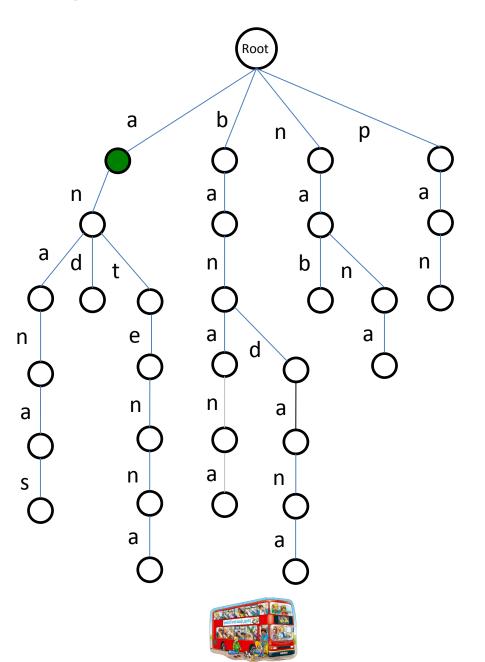


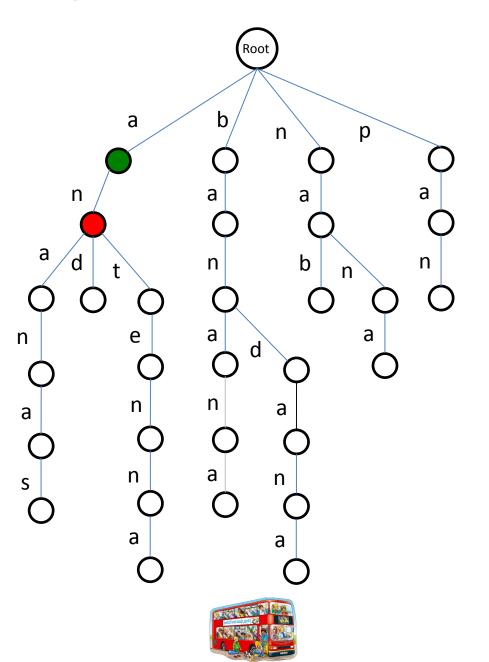


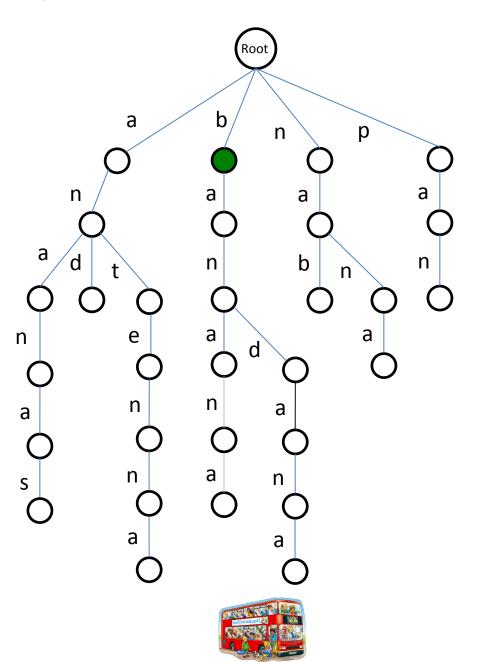


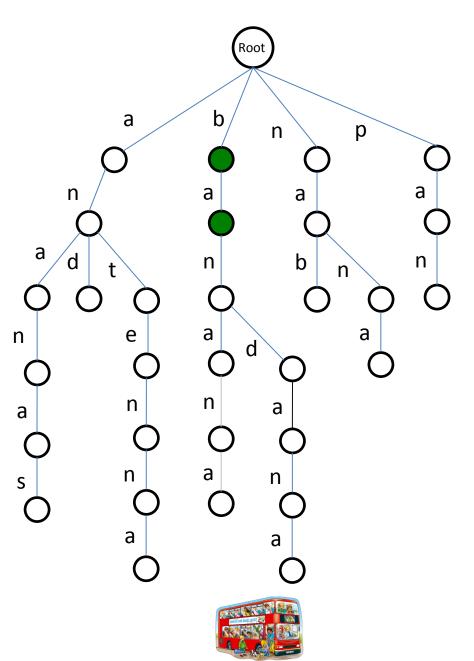


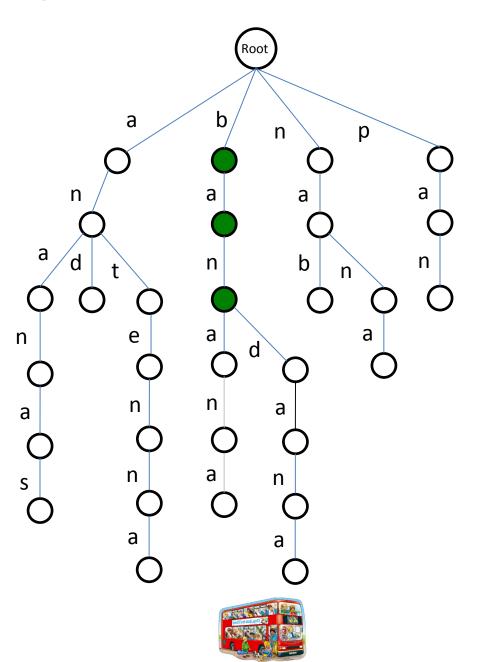


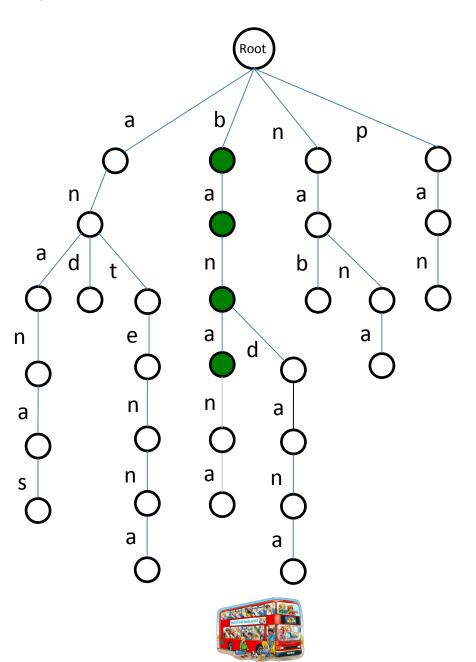


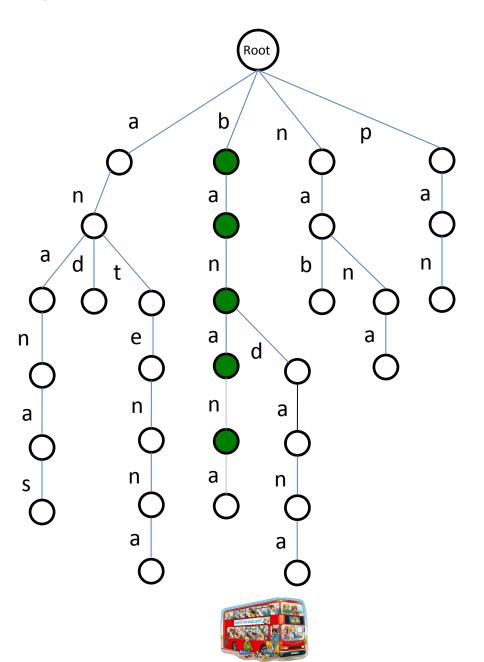


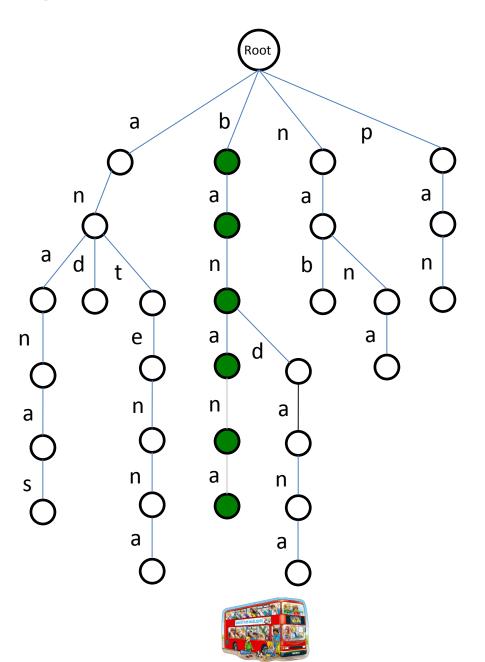


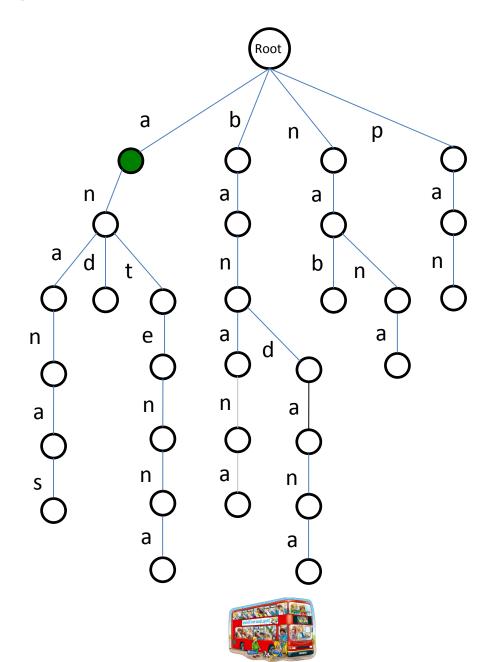


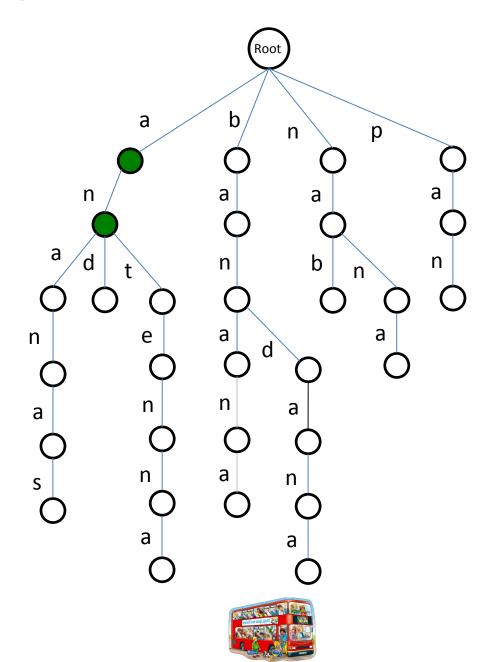


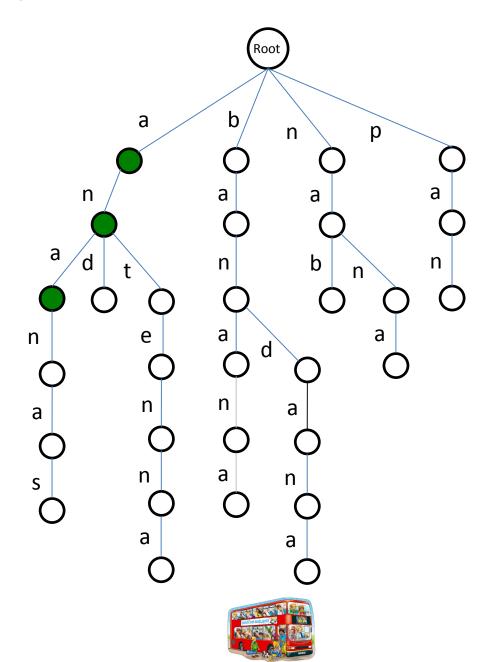


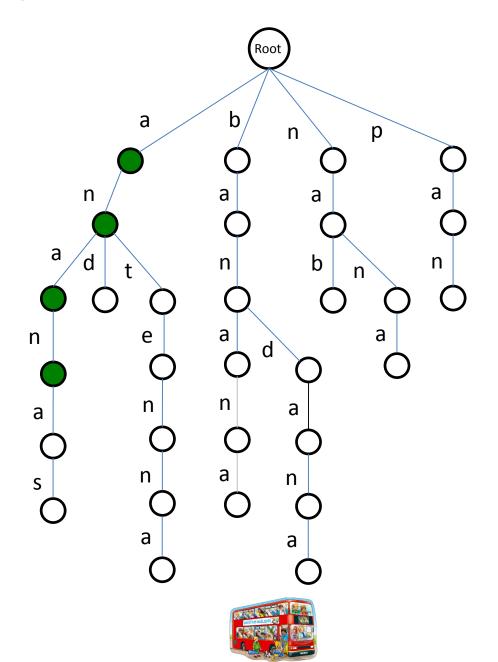


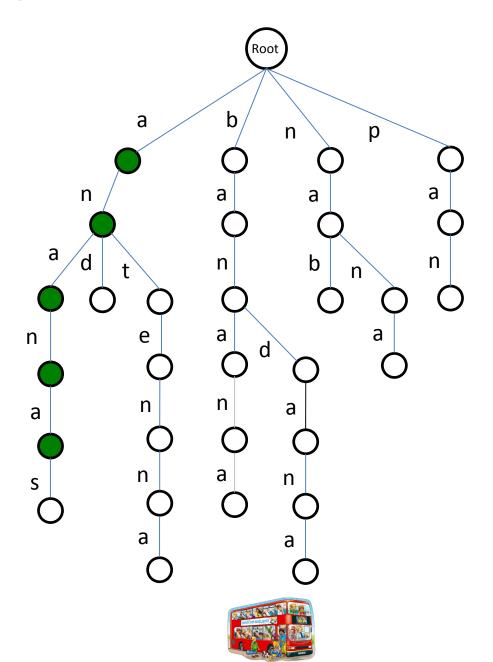




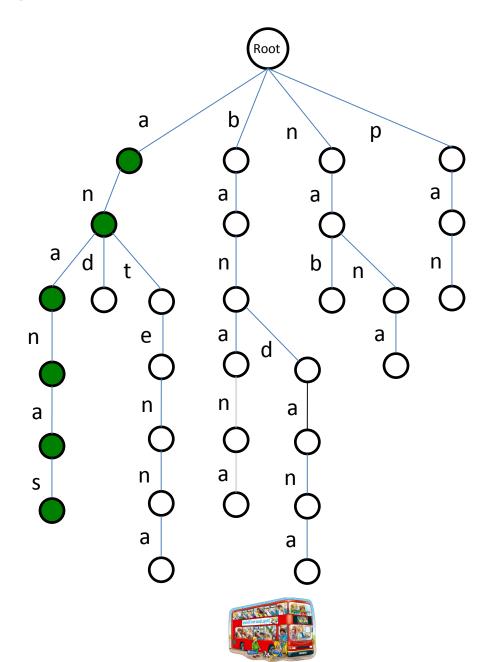


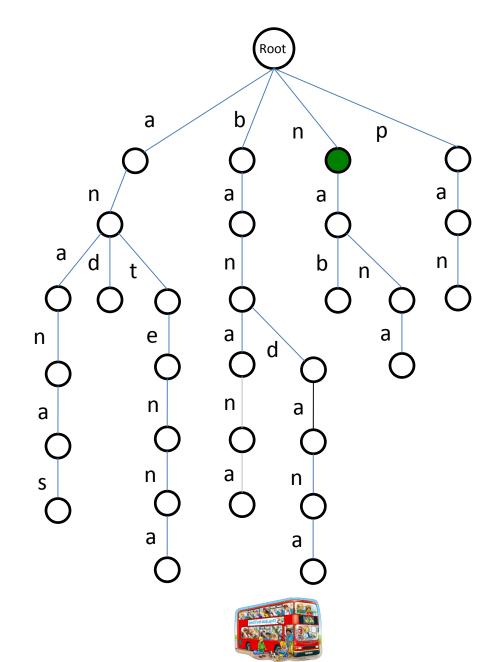


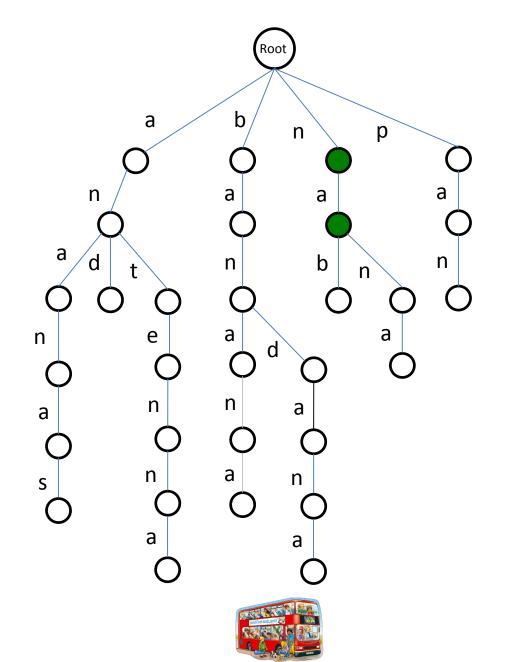


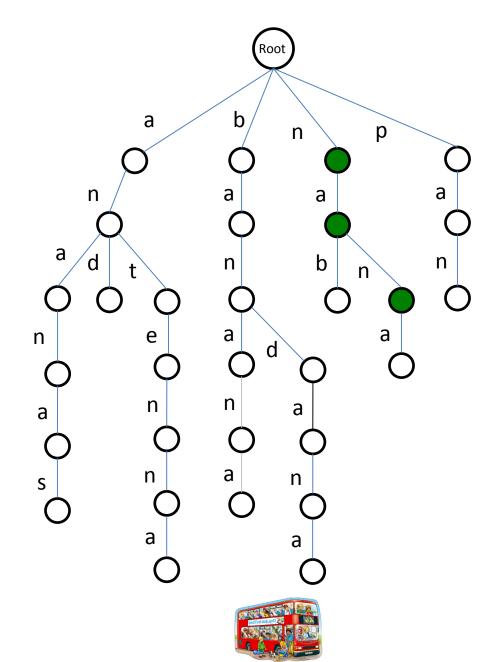


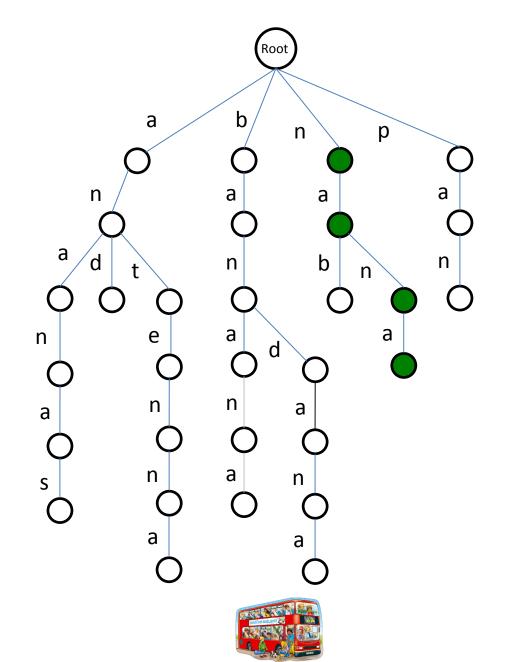
p a n a m a b a n a n a s

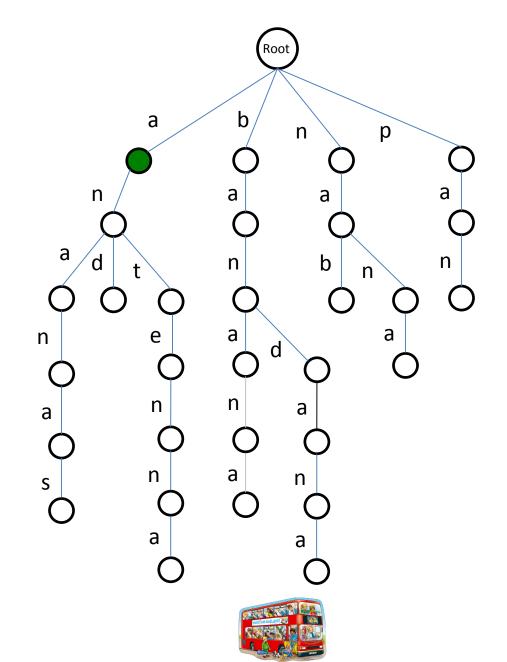


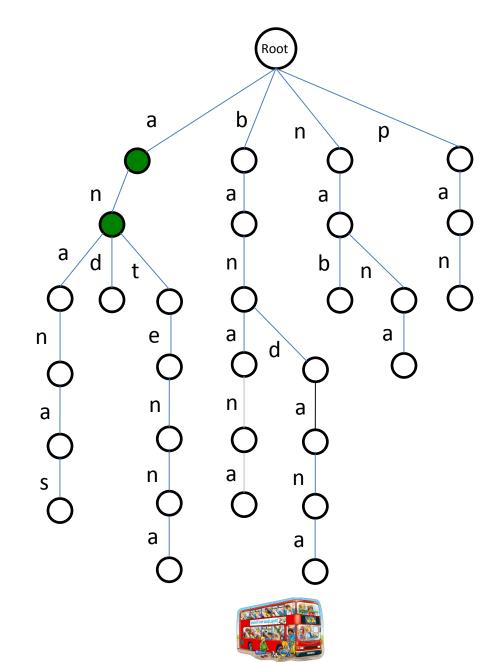


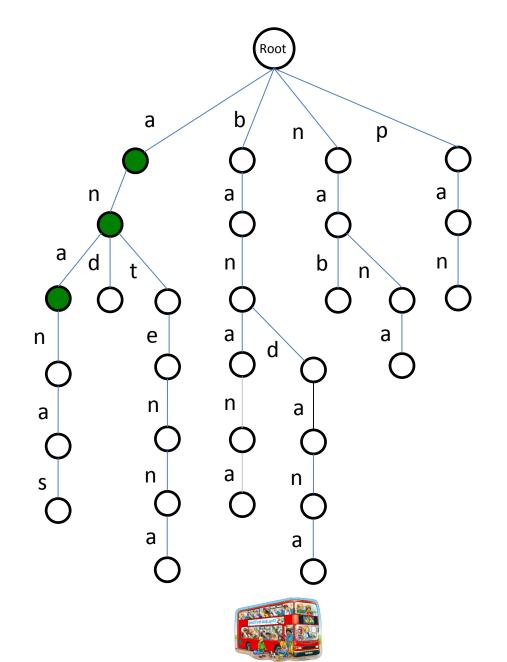


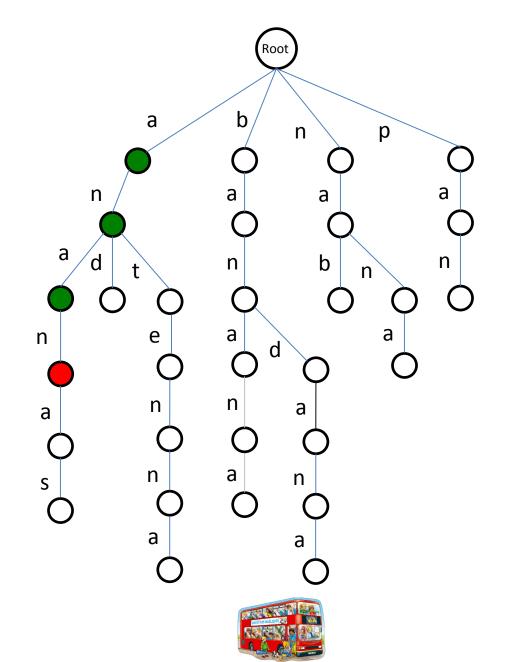


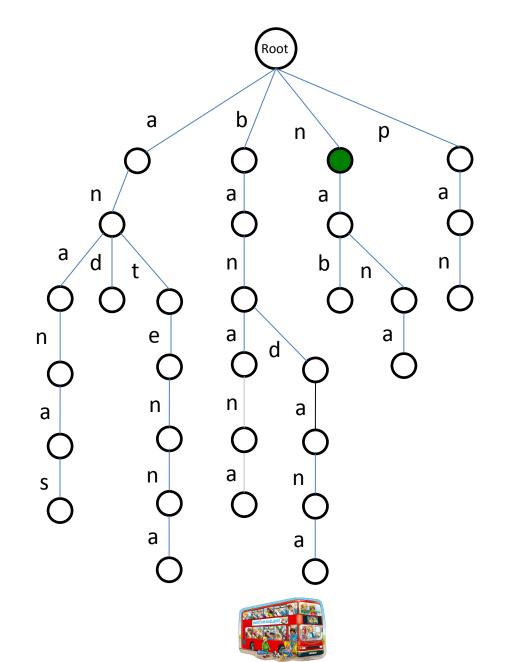


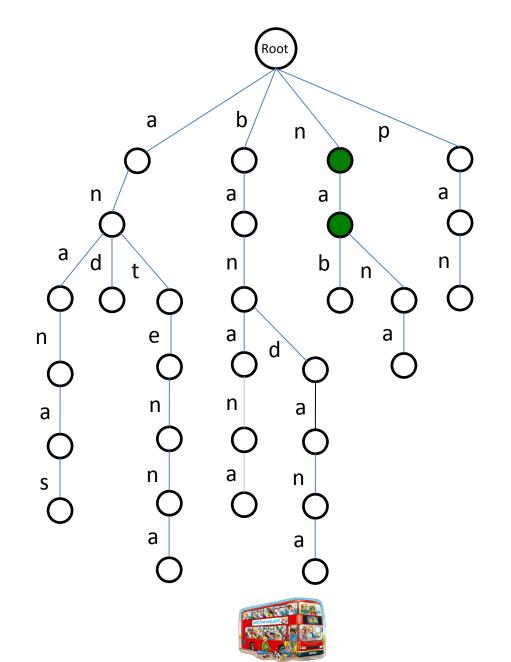


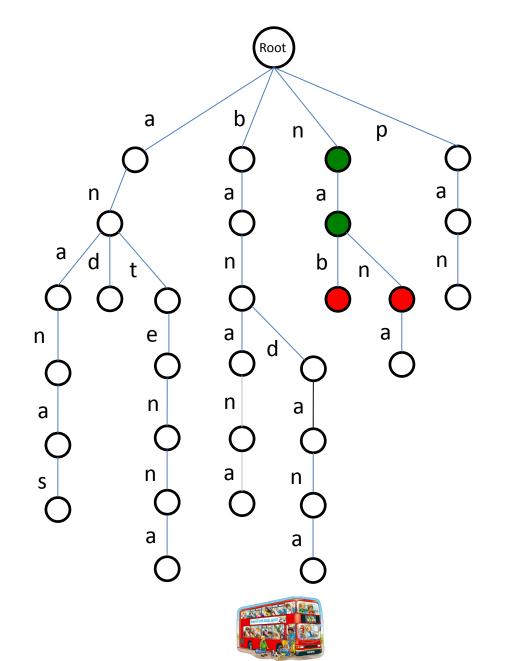


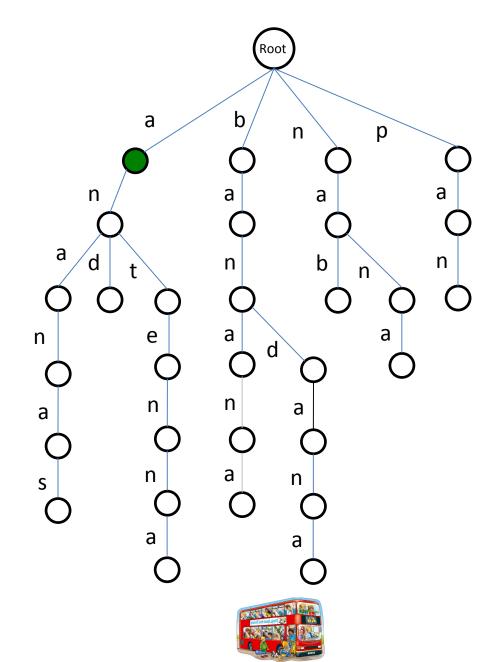


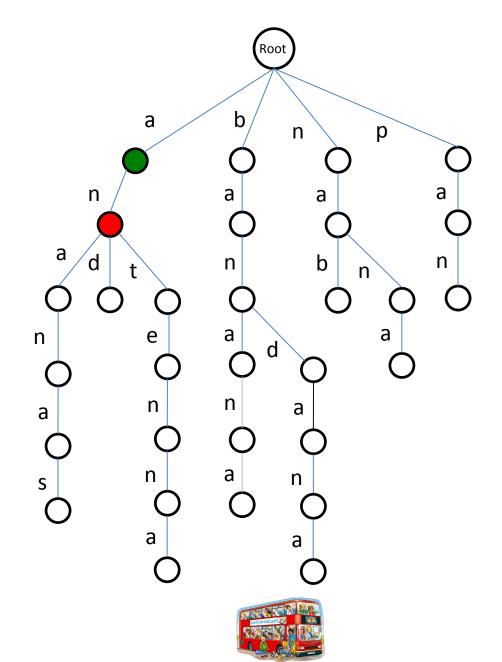


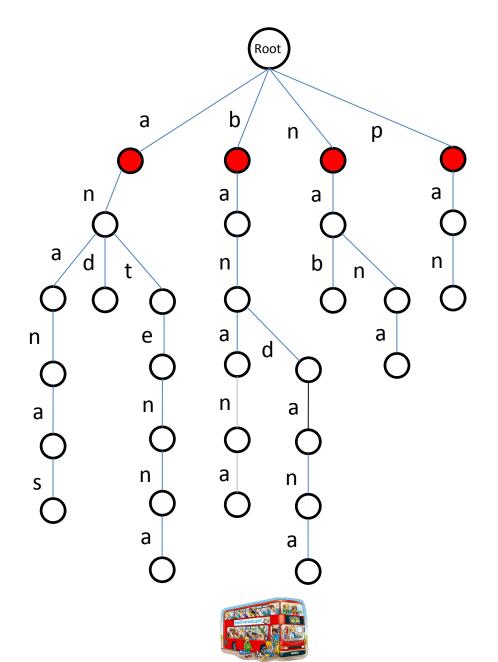












Our Bus Is Fast!

- Runtime of brute force approach:
 - O(|Text| |Patterns|)

- Runtime of TrieMatching:
 - O(|Text| * |LongestPattern|)



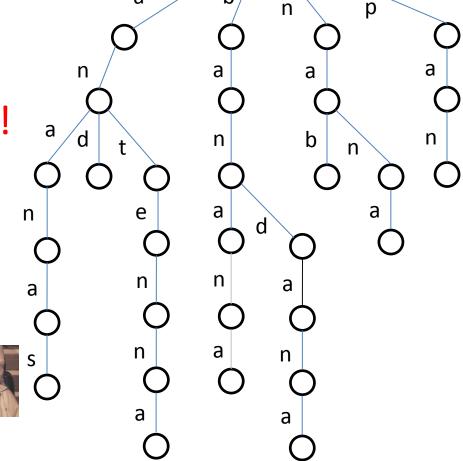
Trie construction takes O(|Patterns|) time

Memory Footprint of TrieMatching

Our trie has 30 edges

edges = O(|Patterns|)!

• For human genome: $|Patterns| \approx 10^{12}$



Root

Outline

- From Genome Sequencing to Pattern Matching
- Brute Force Approach to Pattern Matching
- Herding Patterns into Trie
- Herding Text into Suffix Trie
- From Suffix Tries to Suffix Trees

New Idea: Packing *Text* onto a Bus

Generate all suffixes of Text

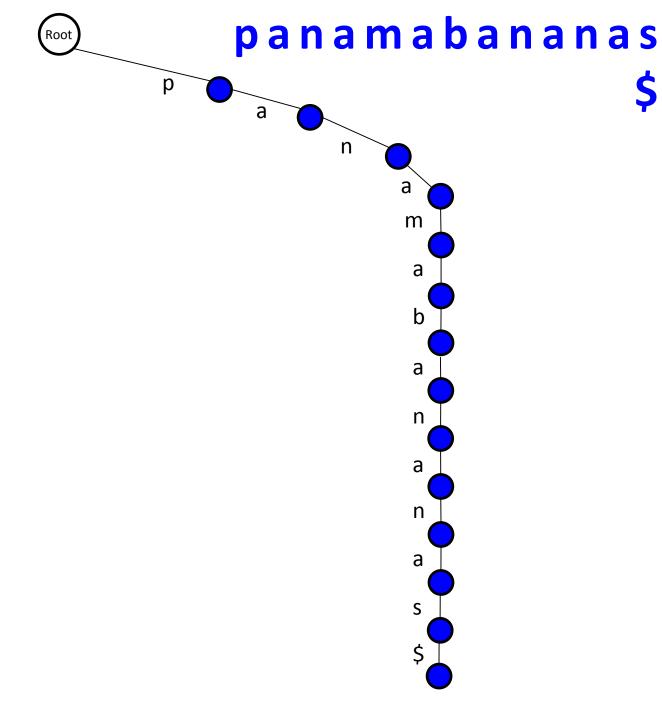
Form a trie out of these suffixes (suffix trie)

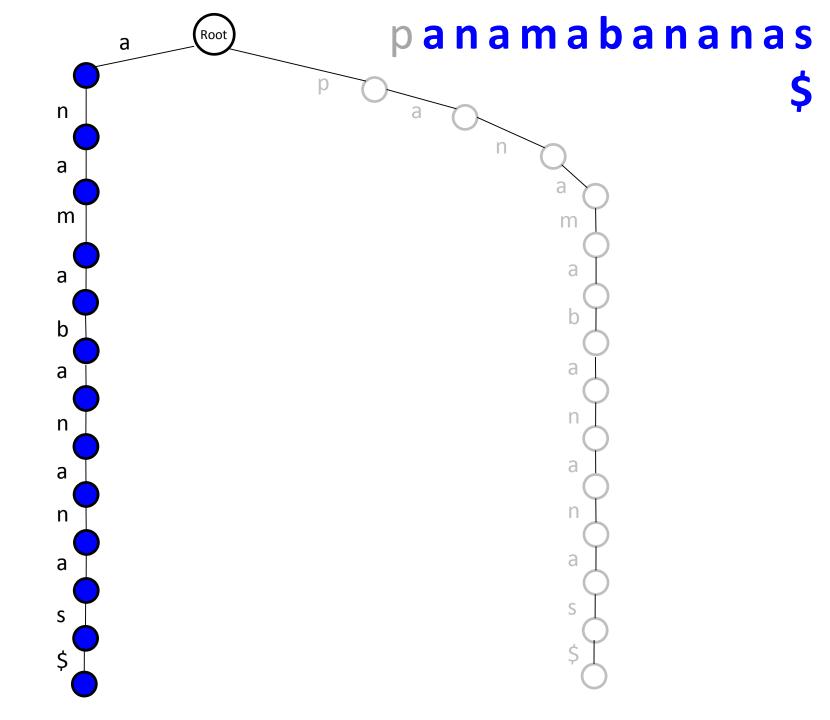
 For each Pattern, check if it can be spelled out from the root downward in the suffix trie

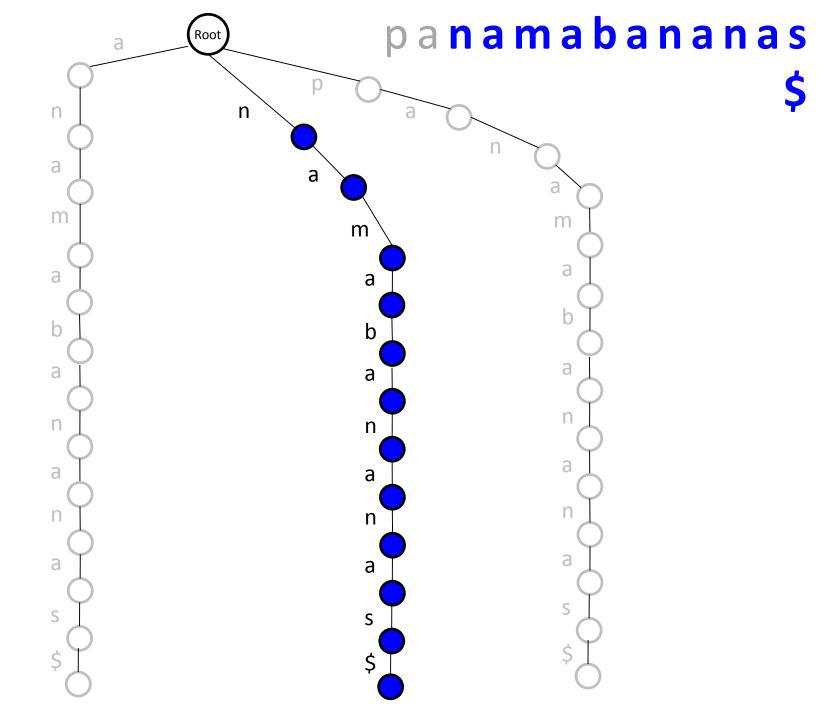


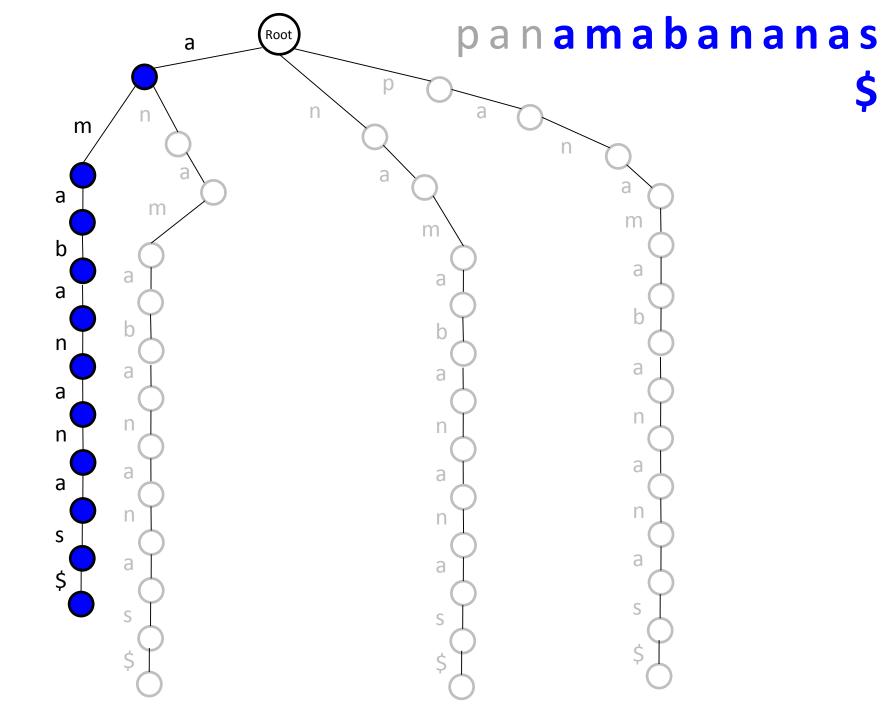


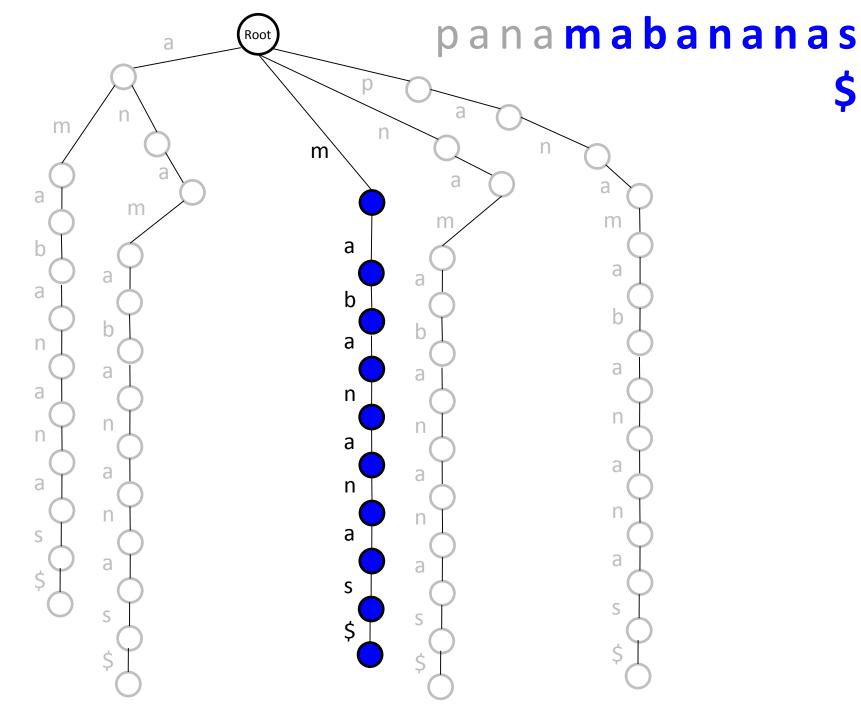
Adding "\$" sign in the end (we'll explain later why)

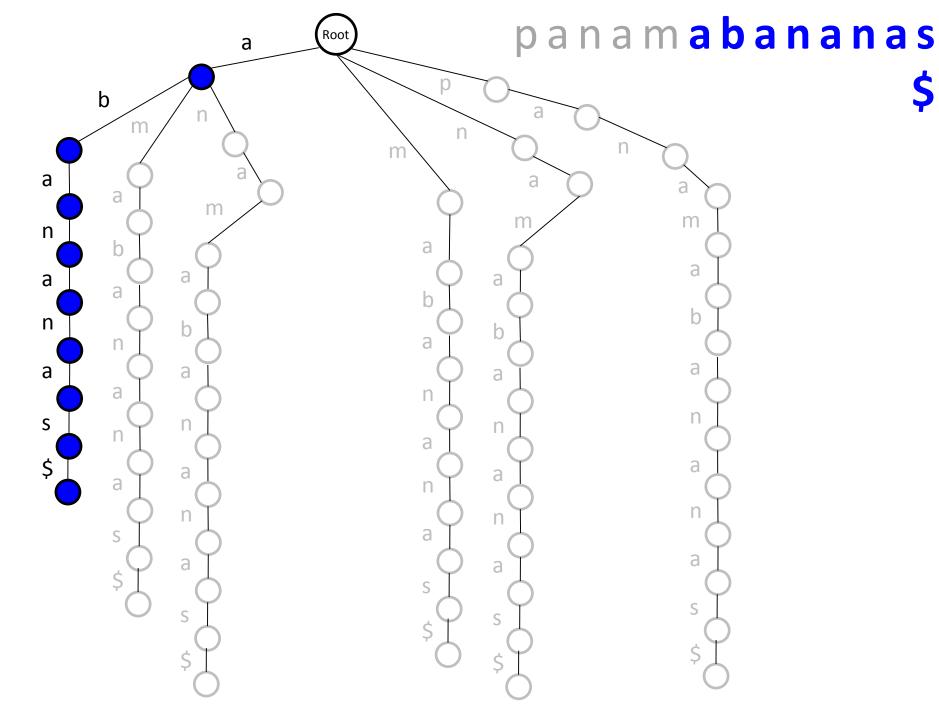


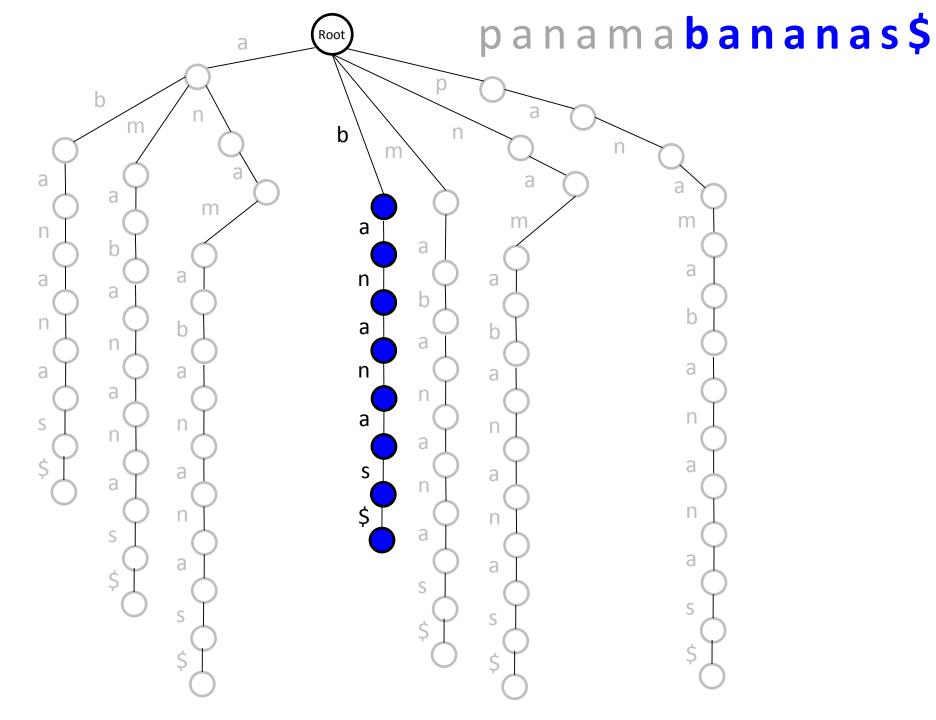


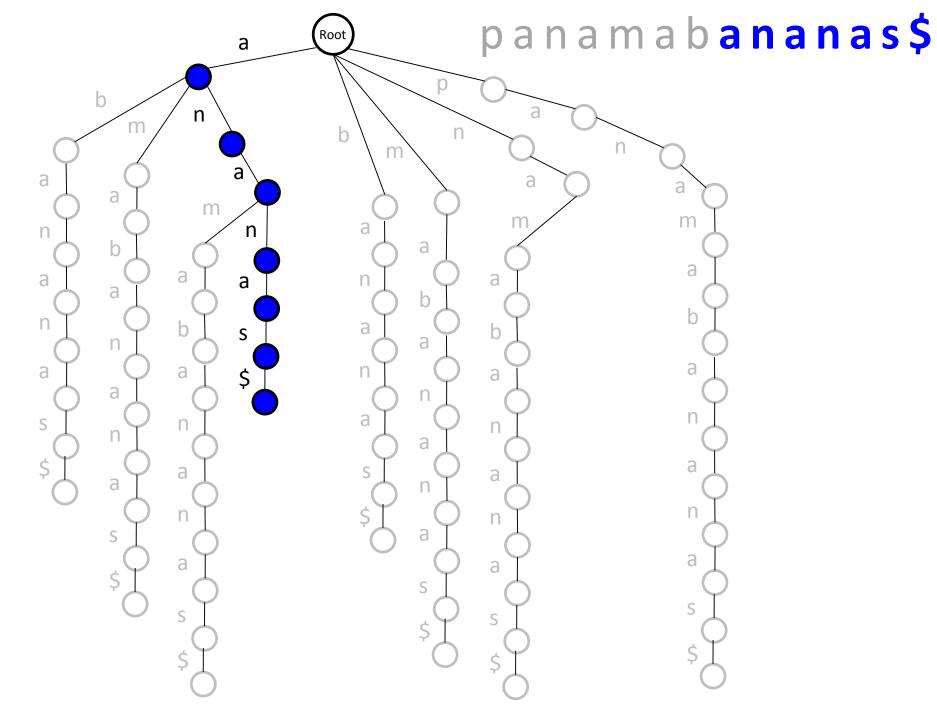


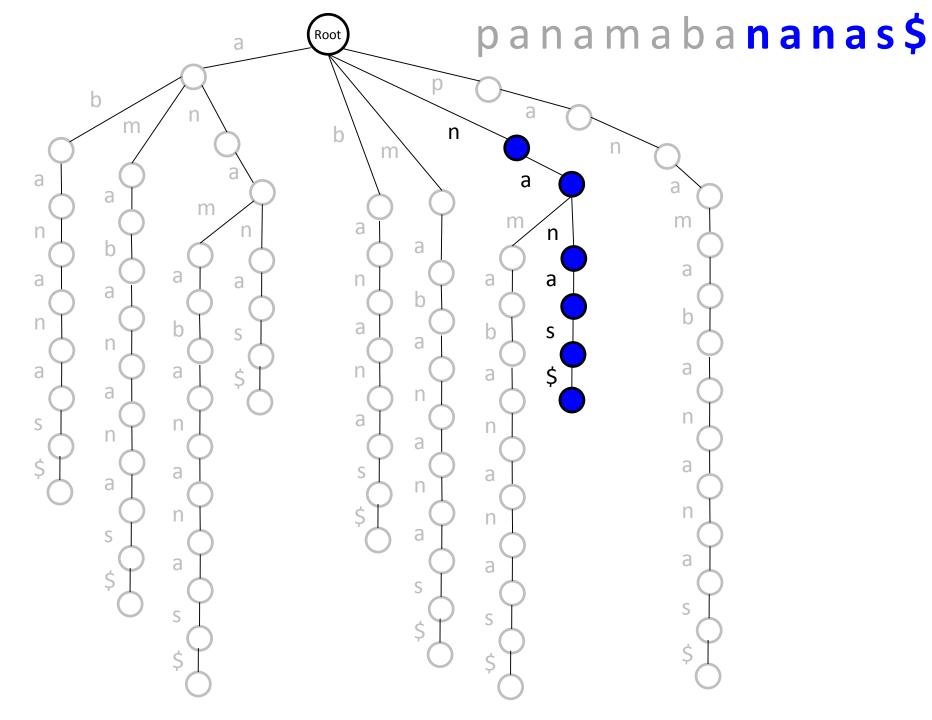


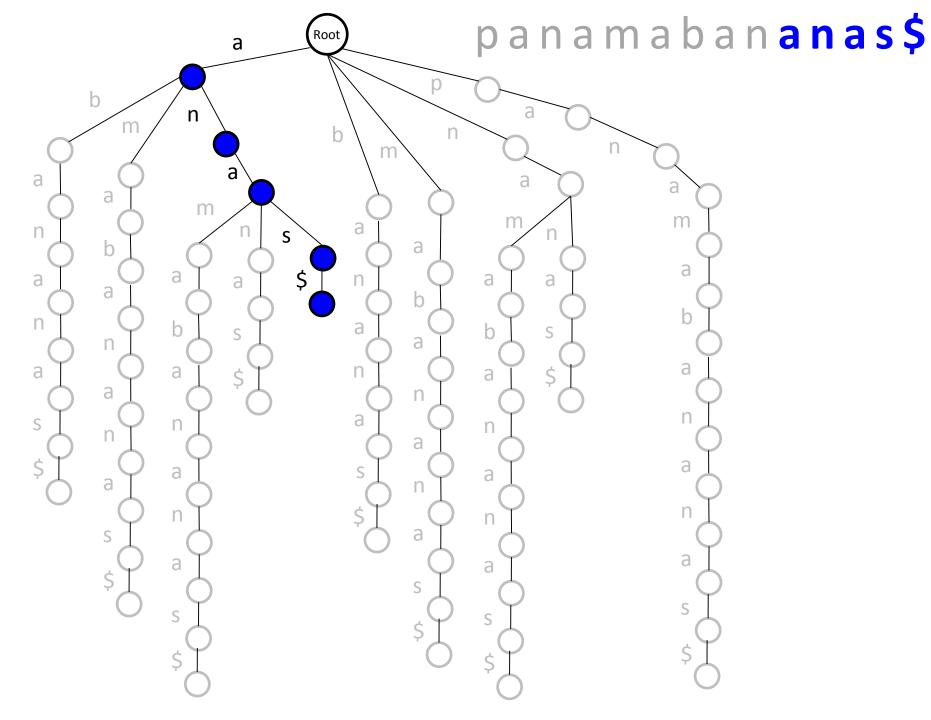


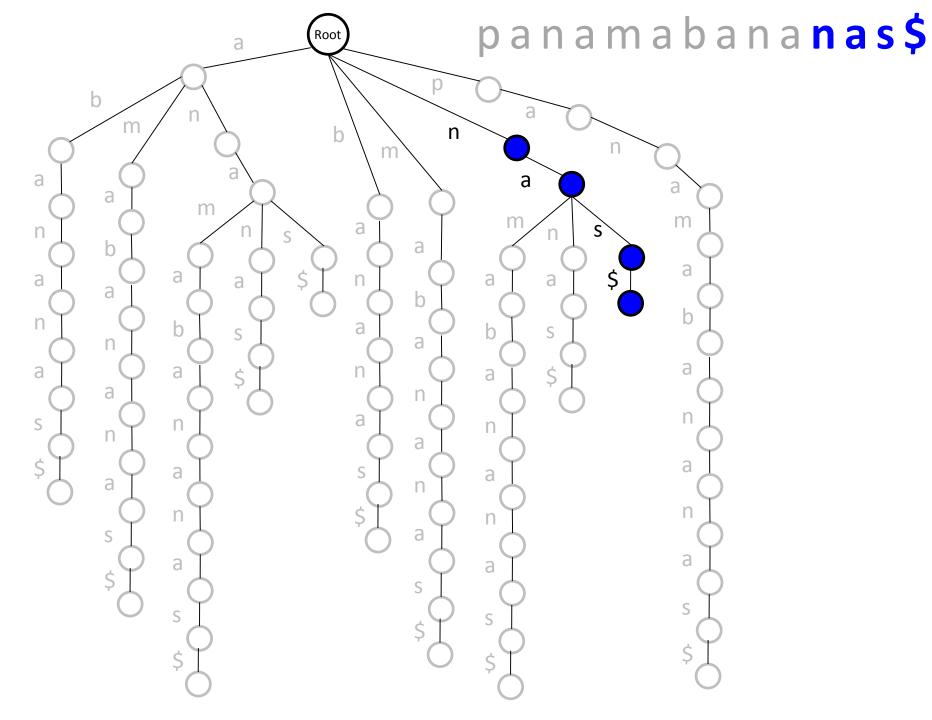


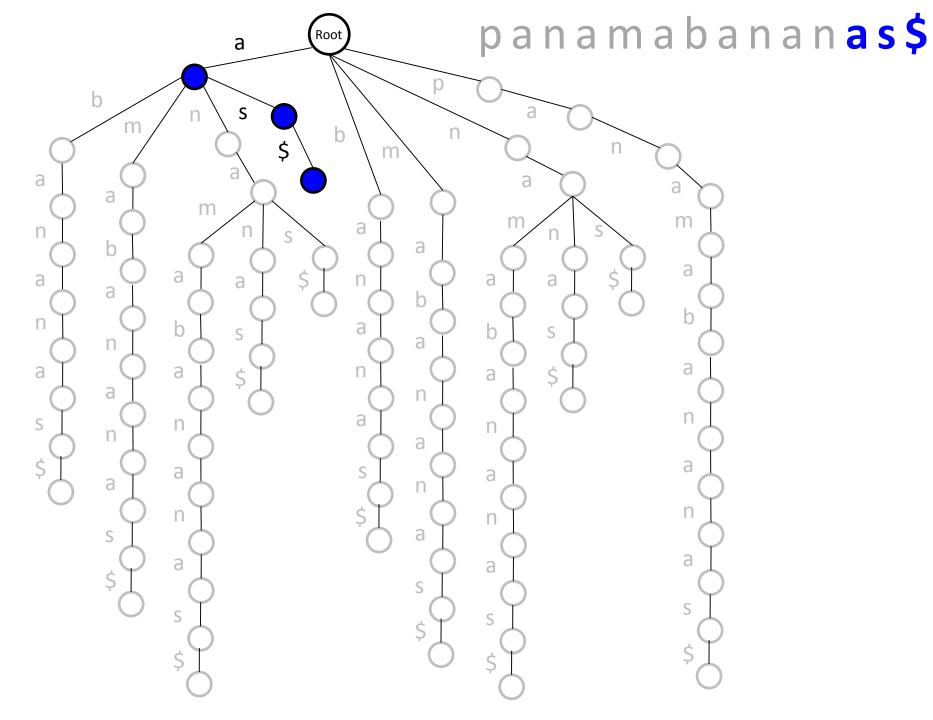


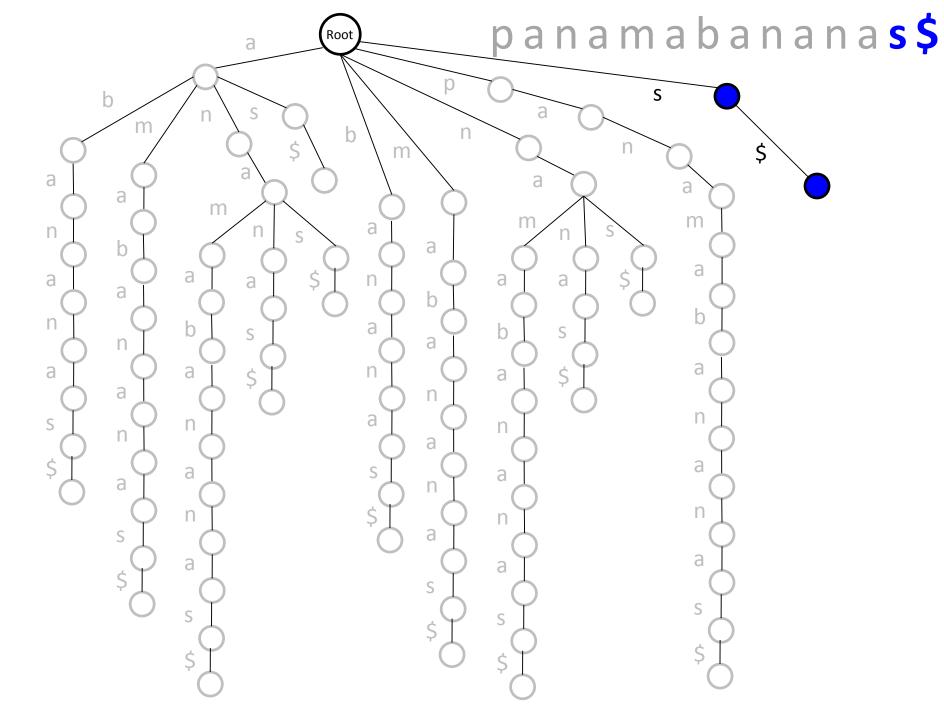


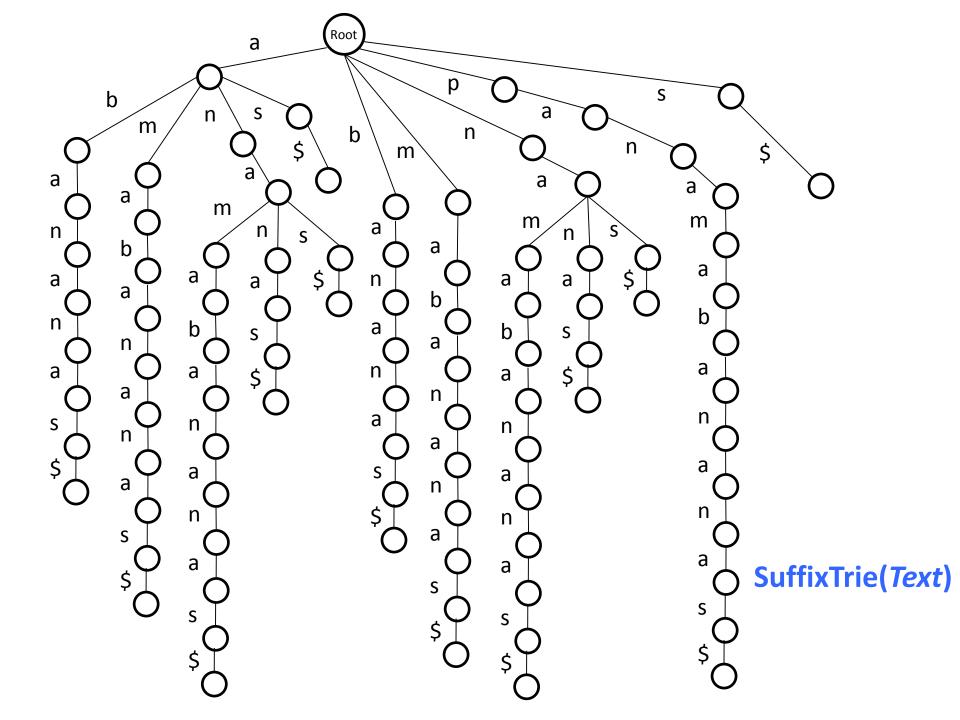


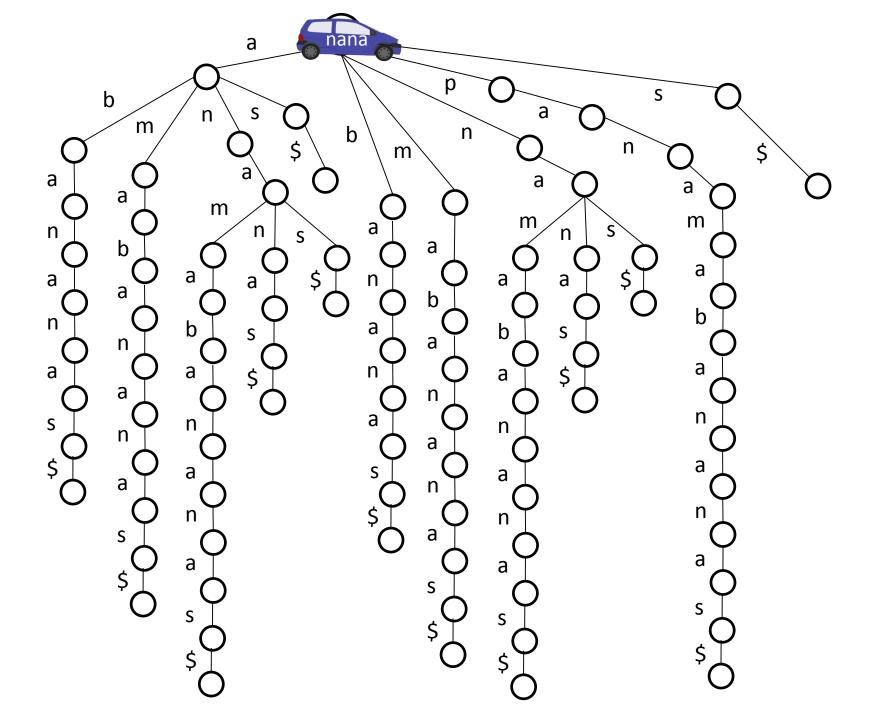


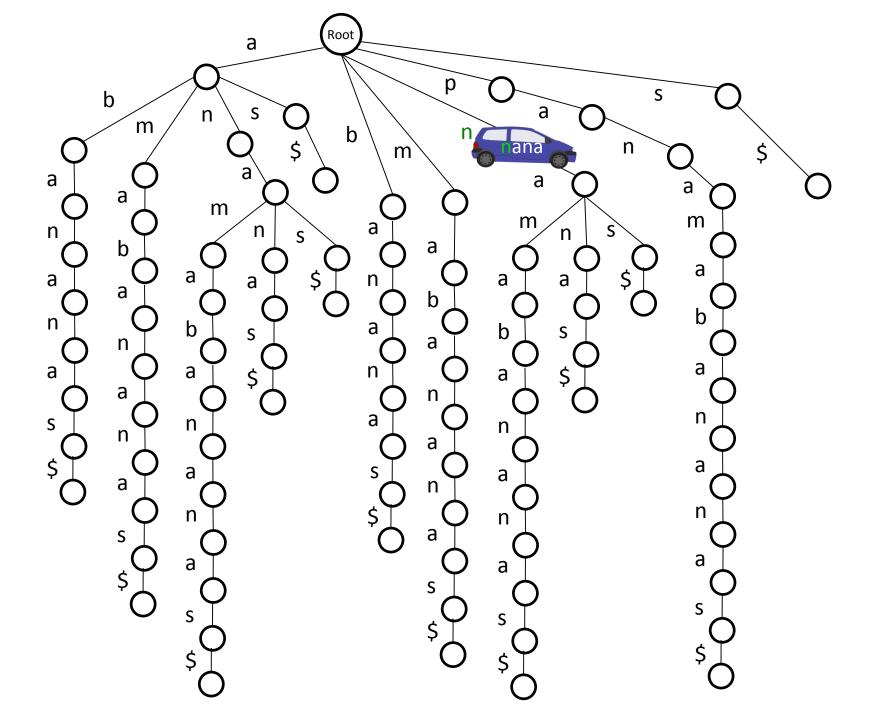


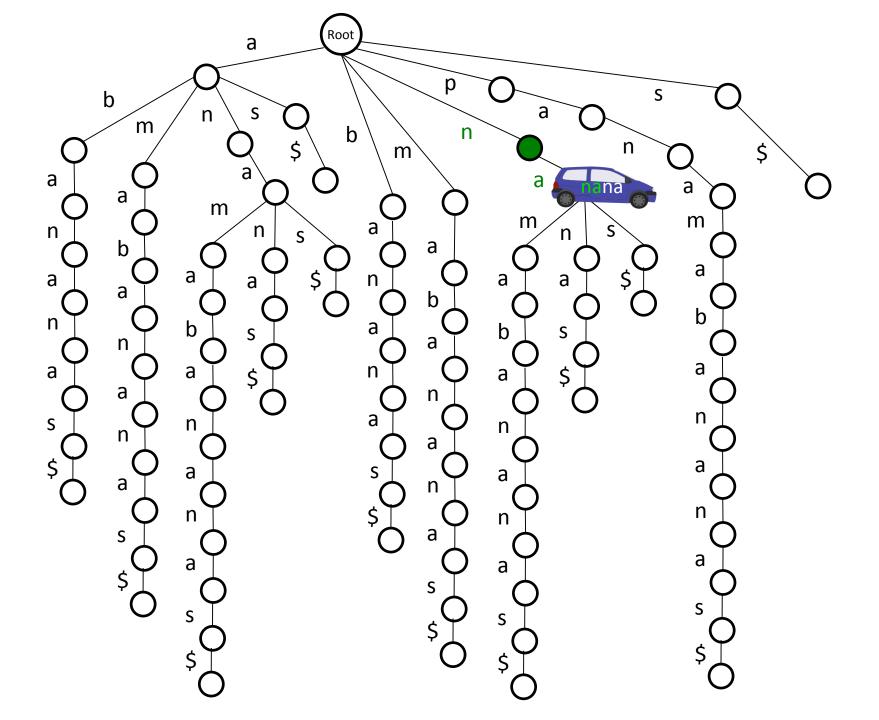


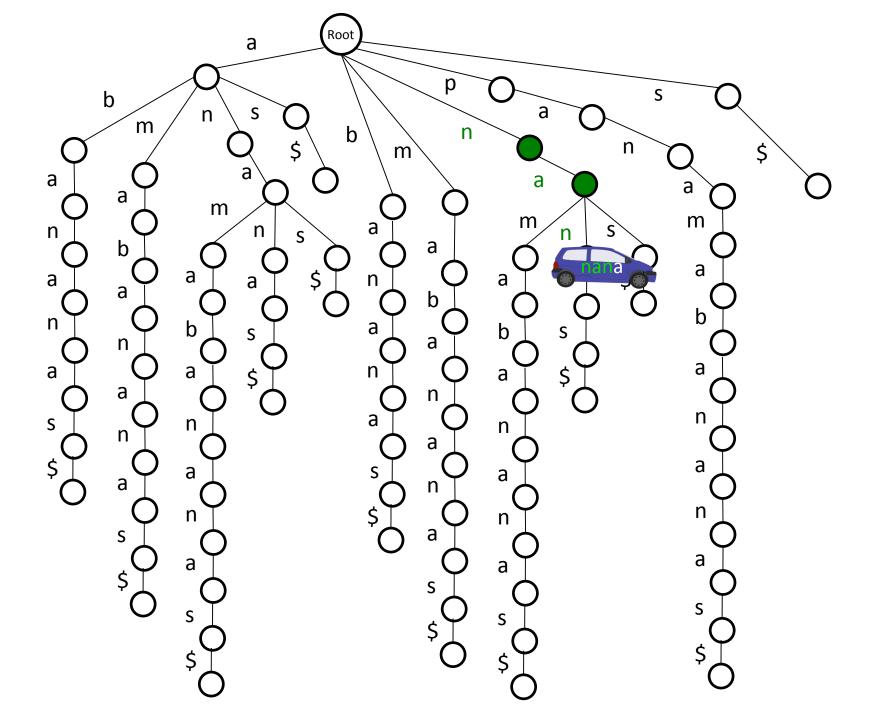


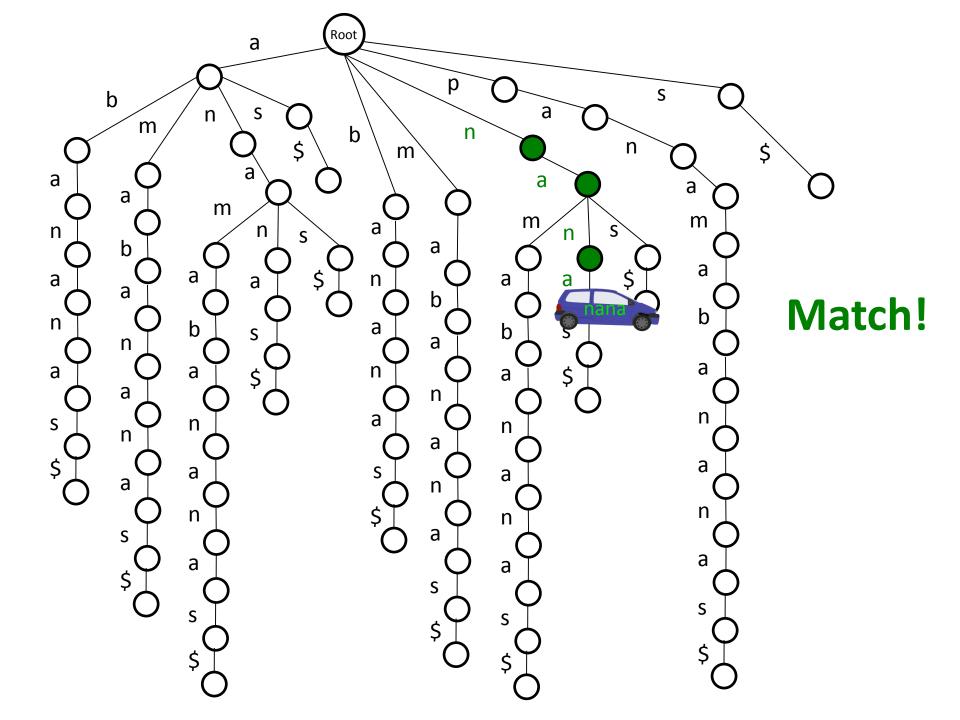


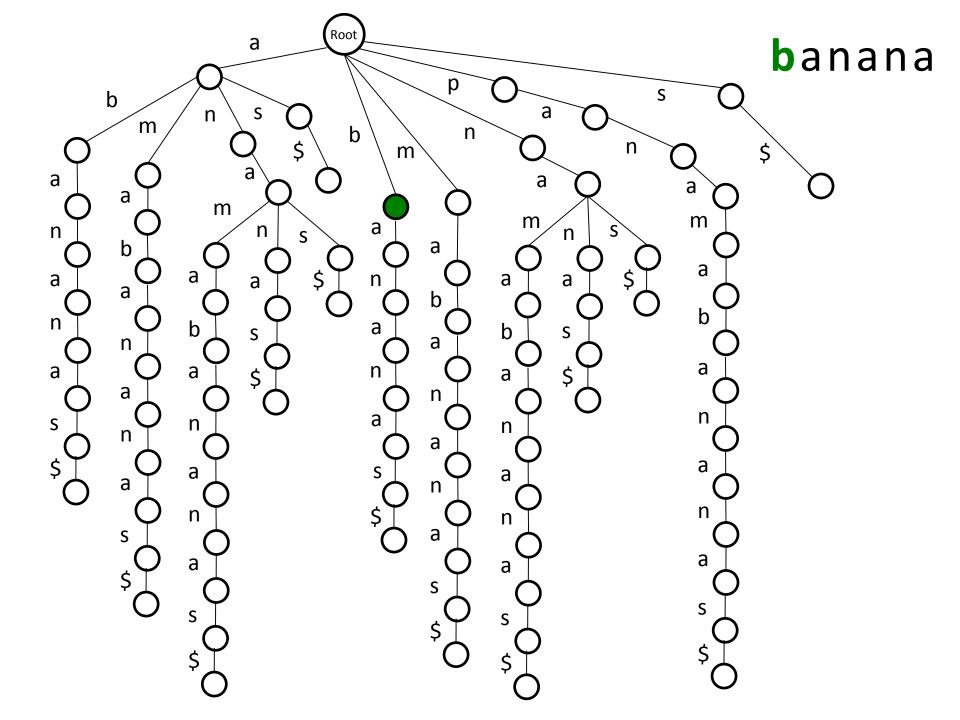


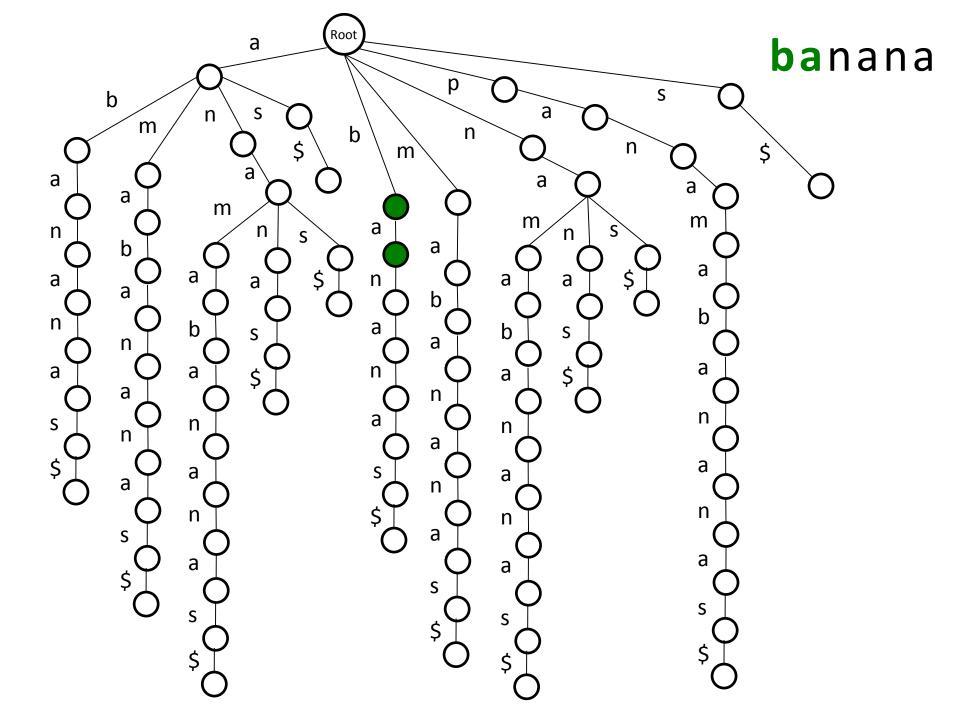


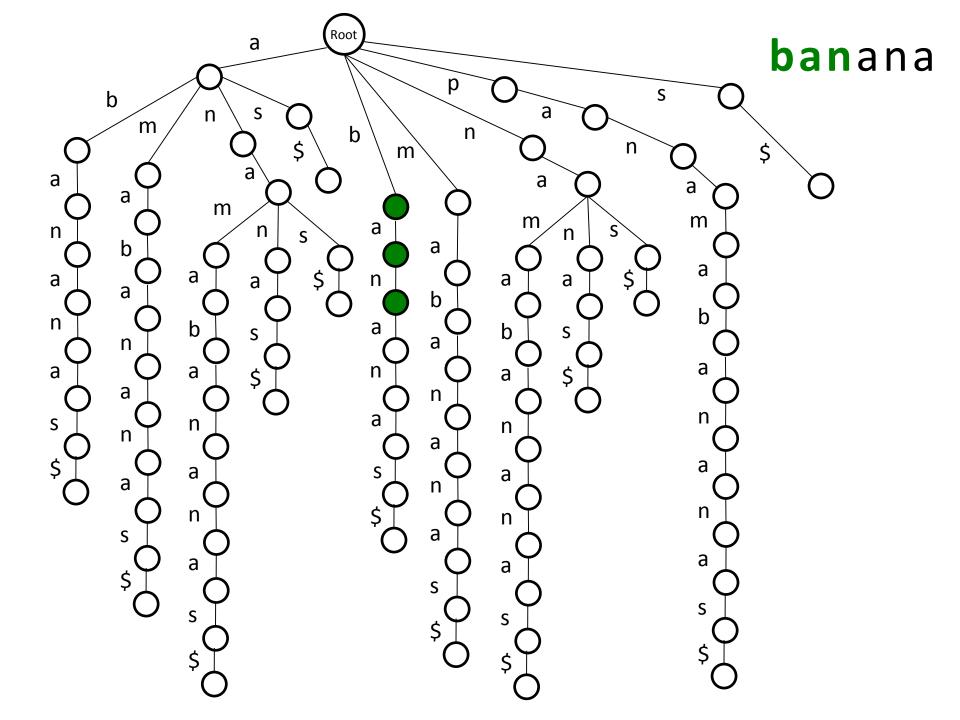


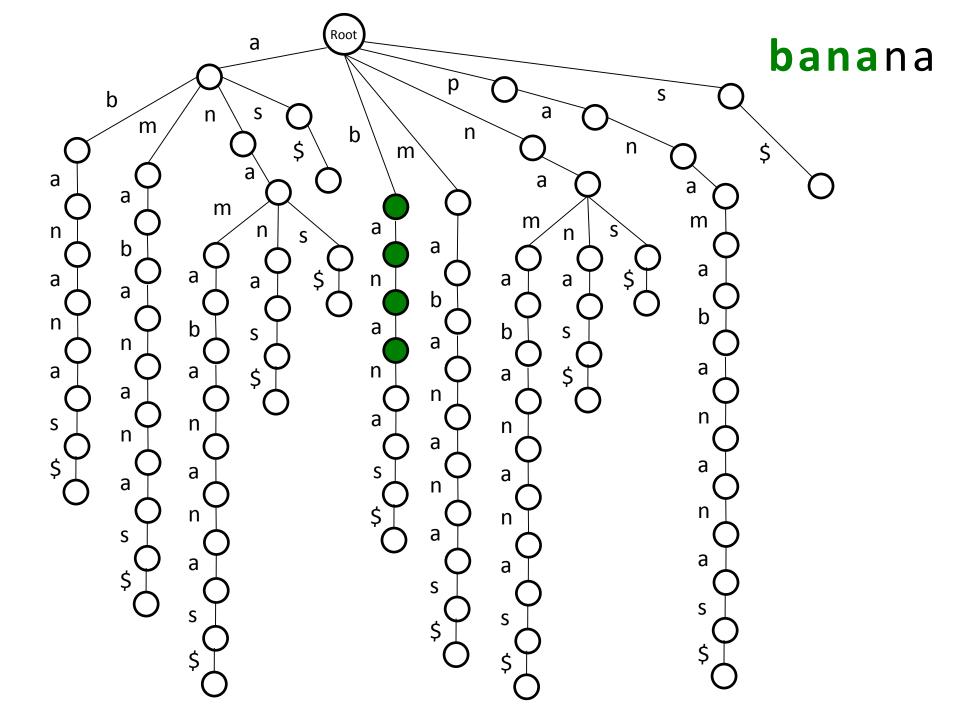


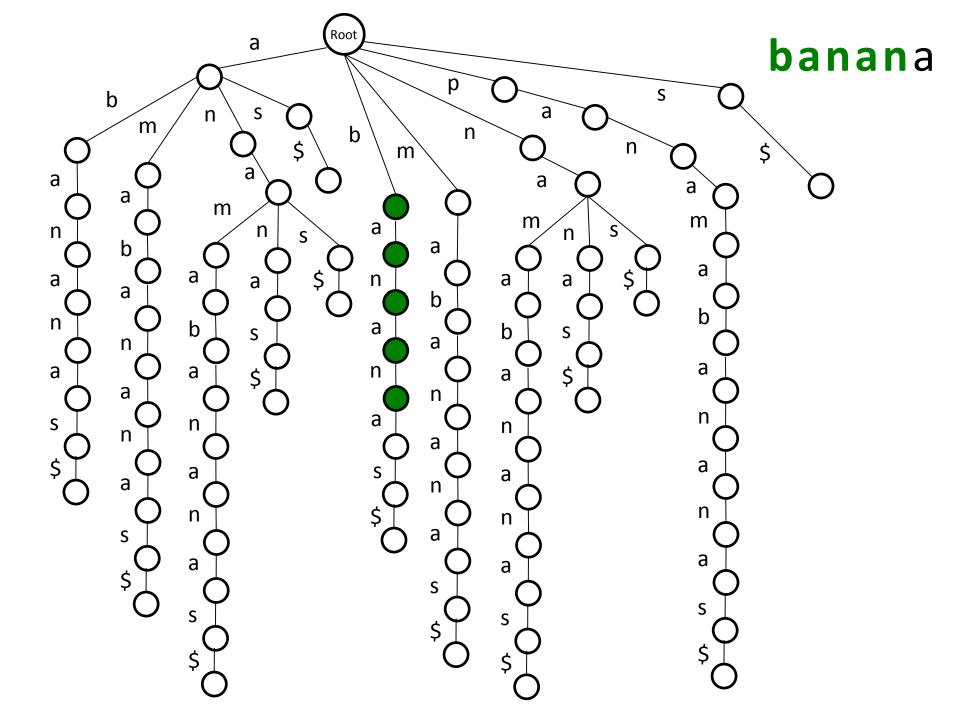


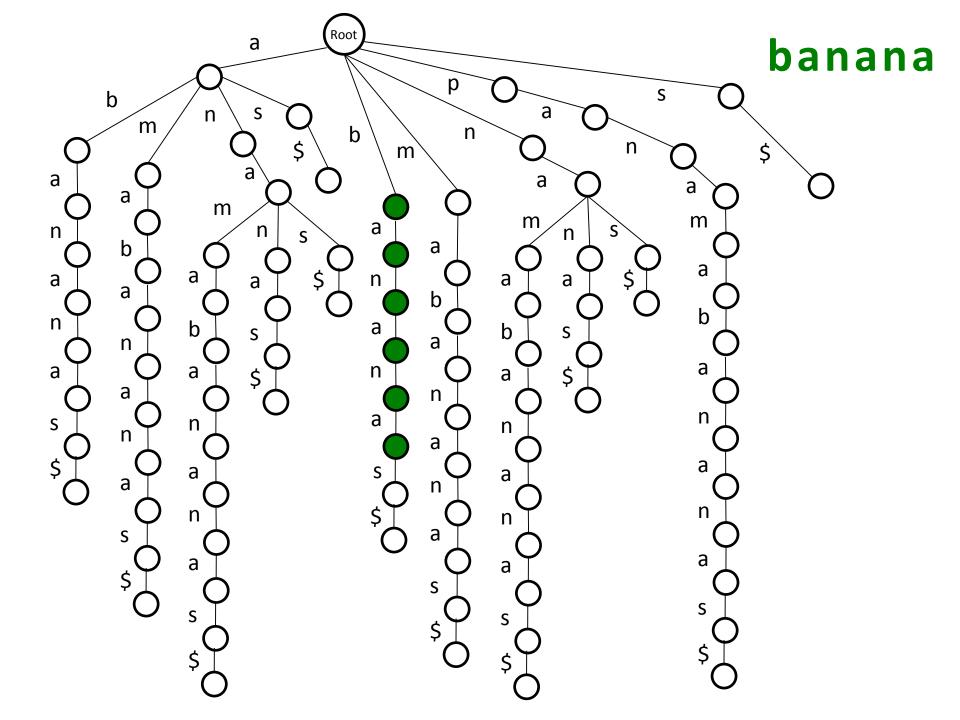




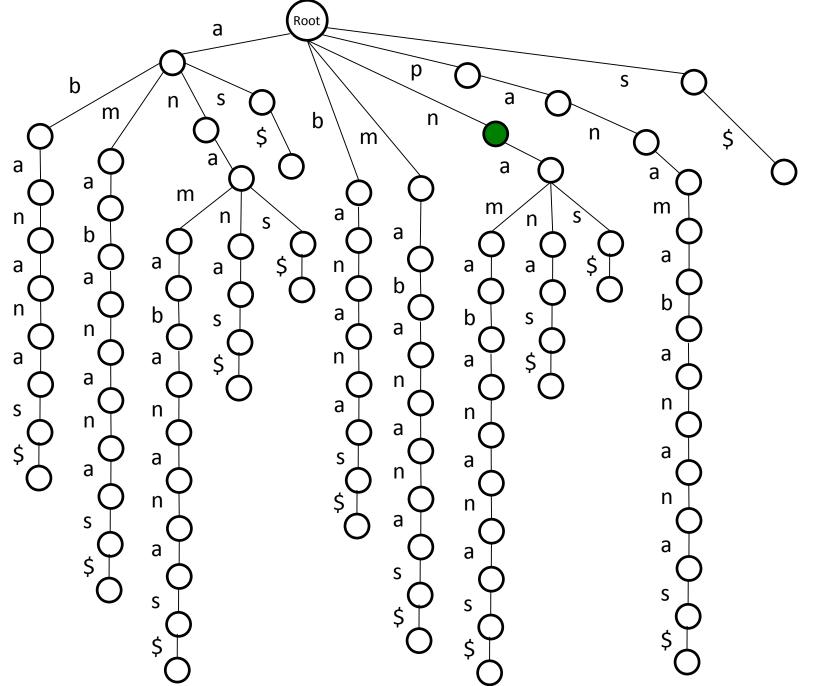




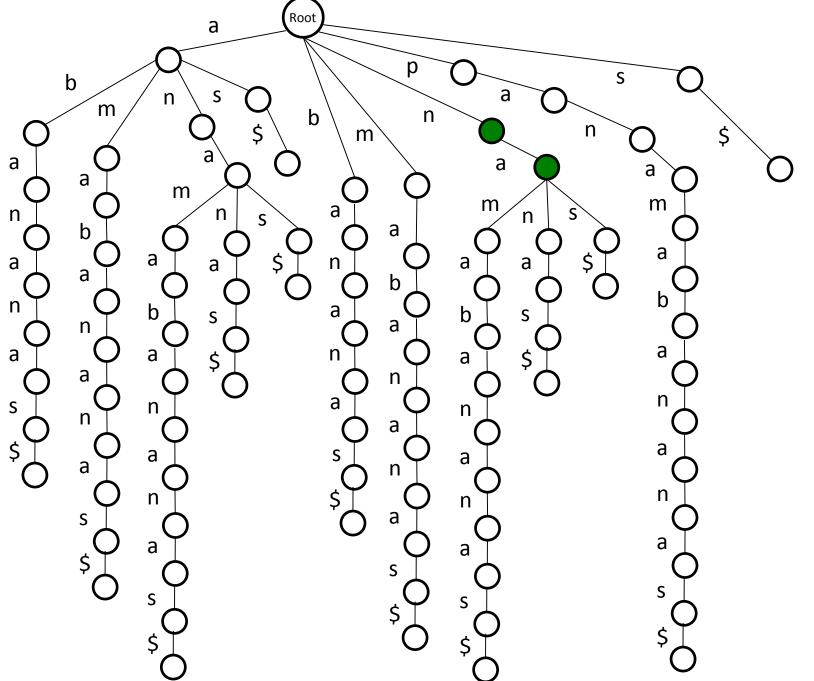




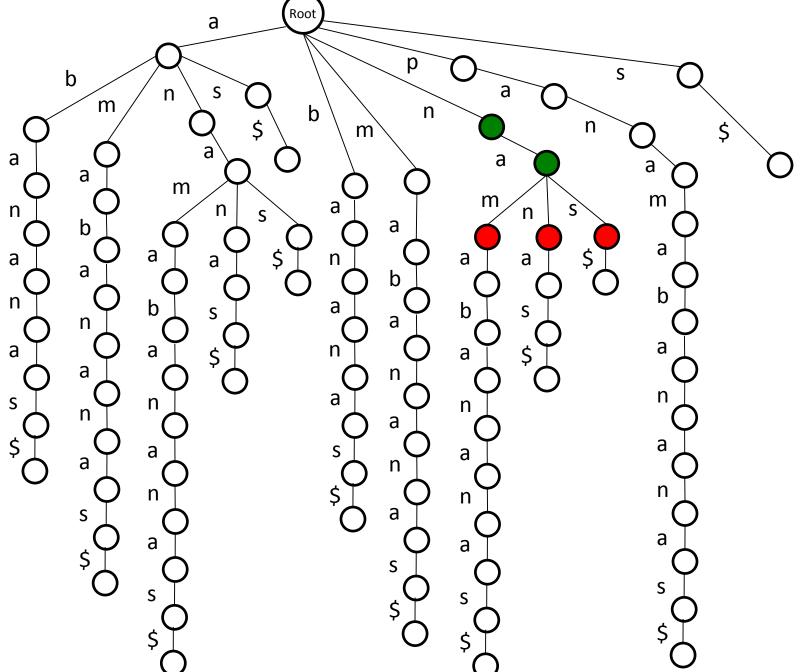
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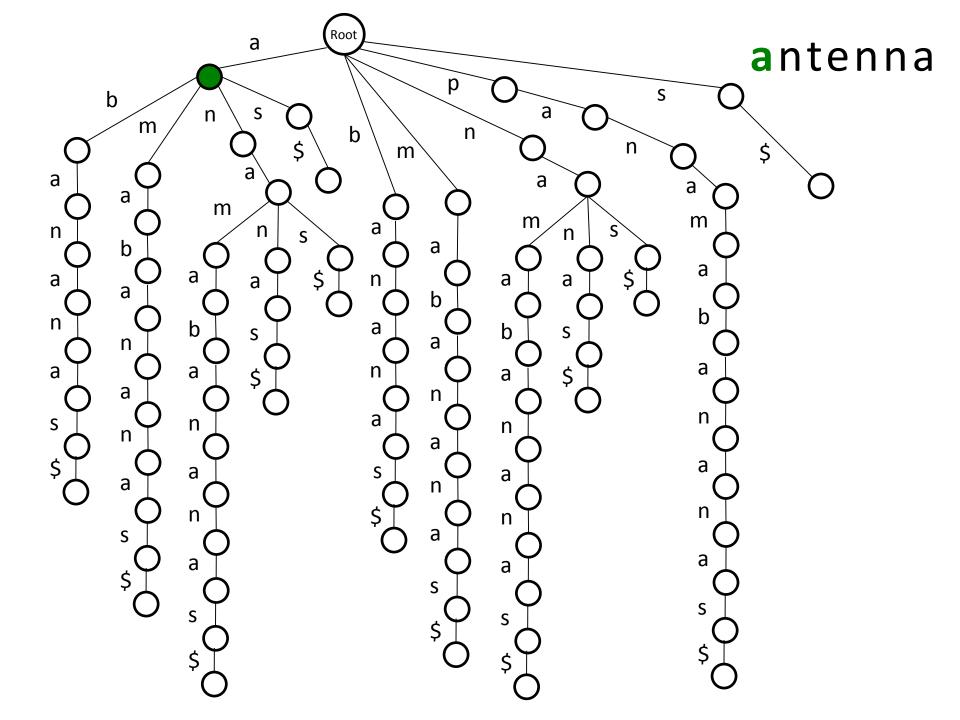


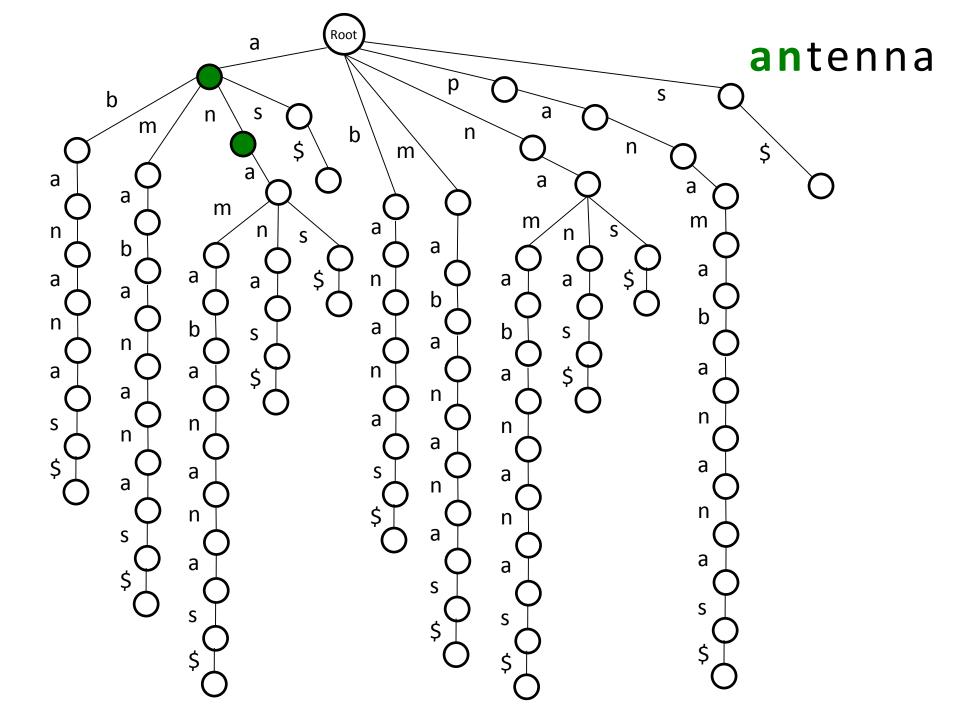
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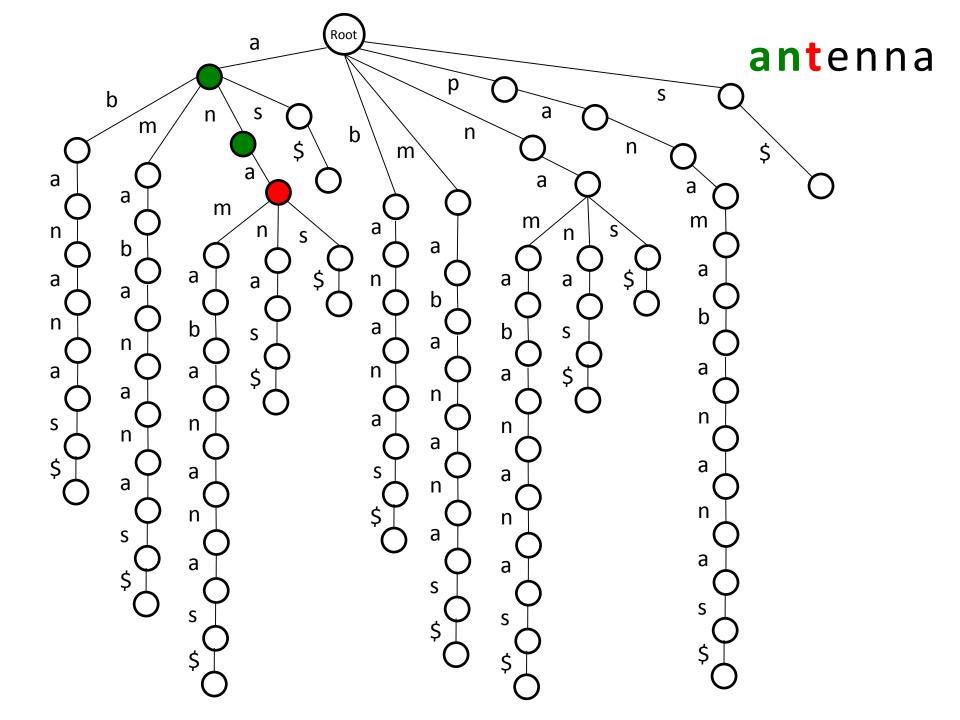


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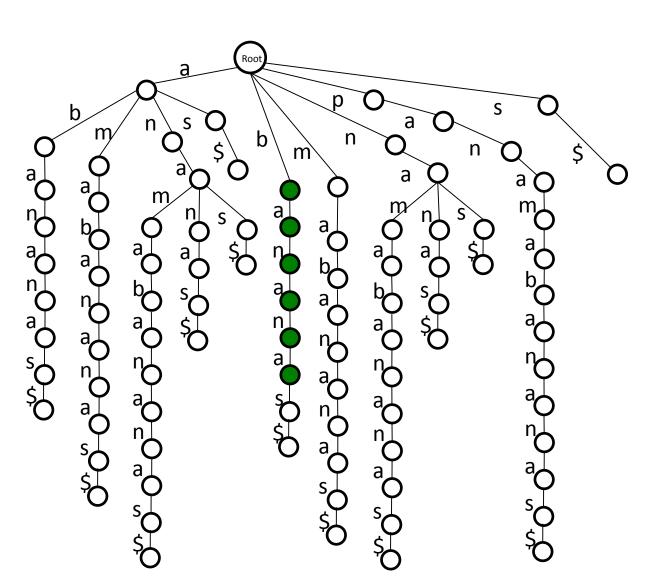






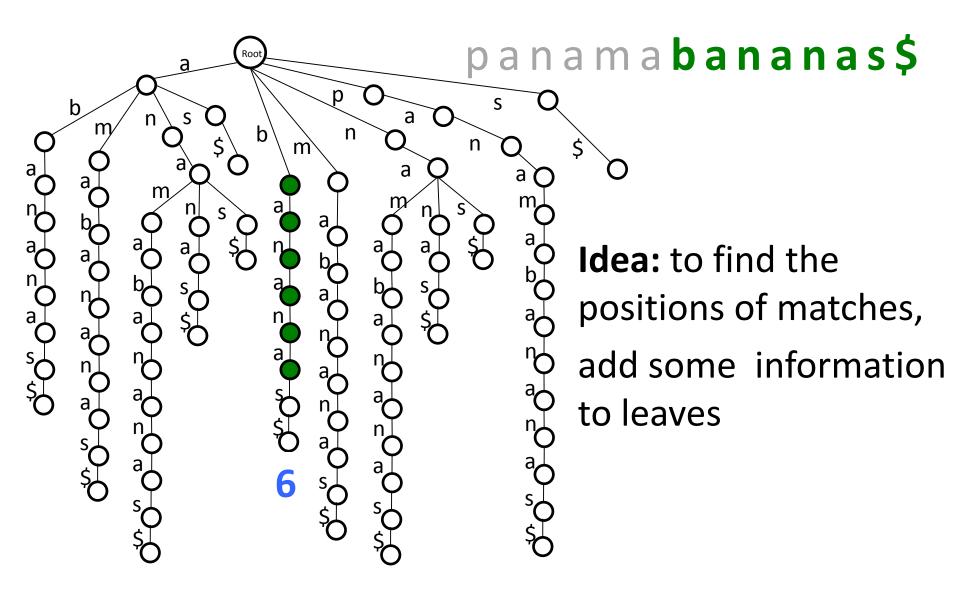


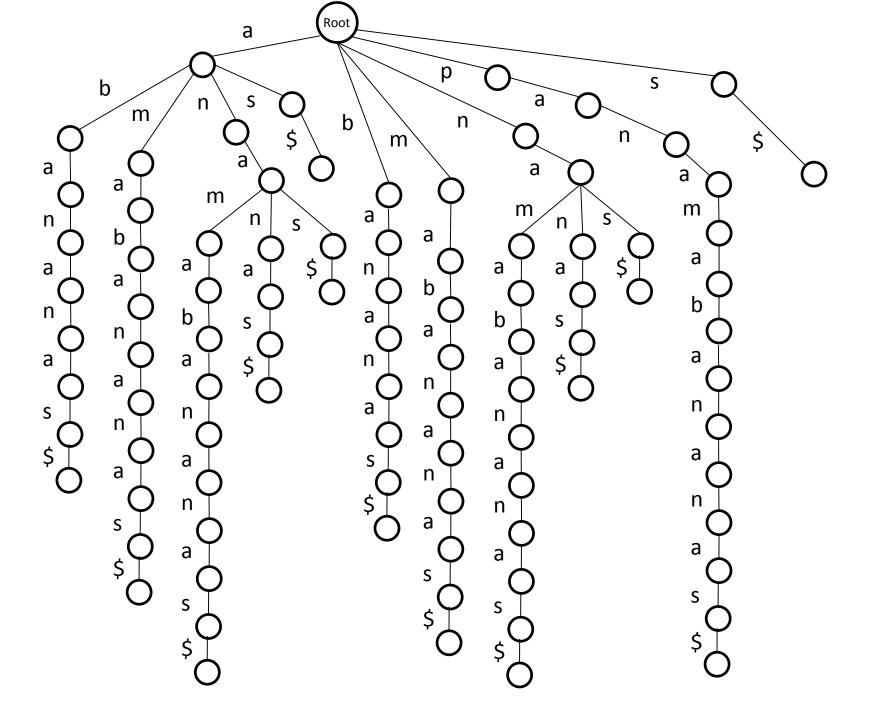
Where Are the Matches???

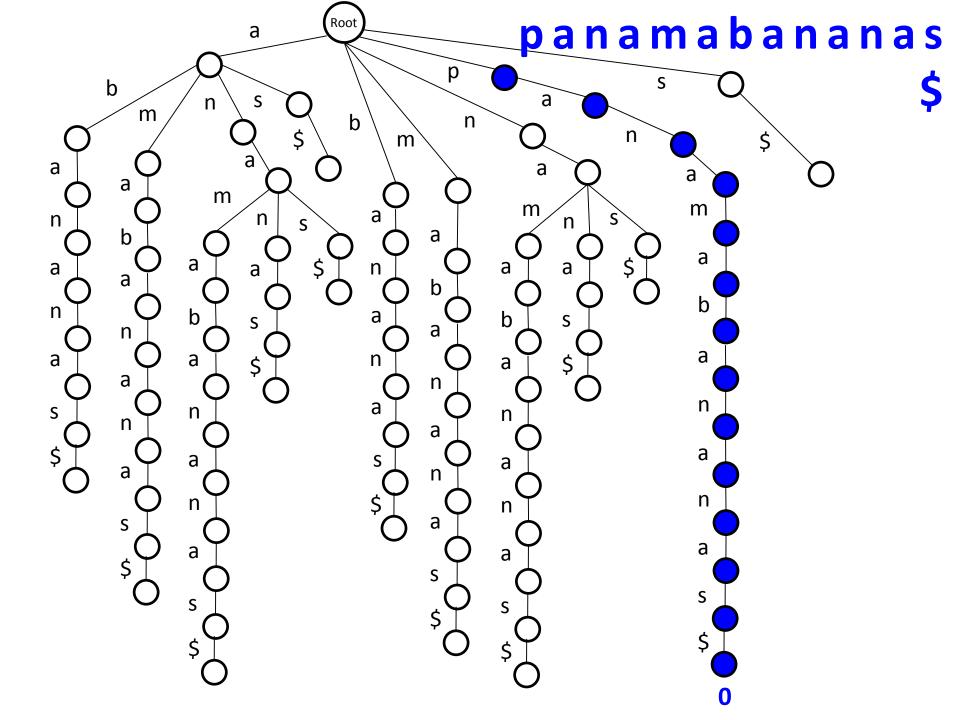


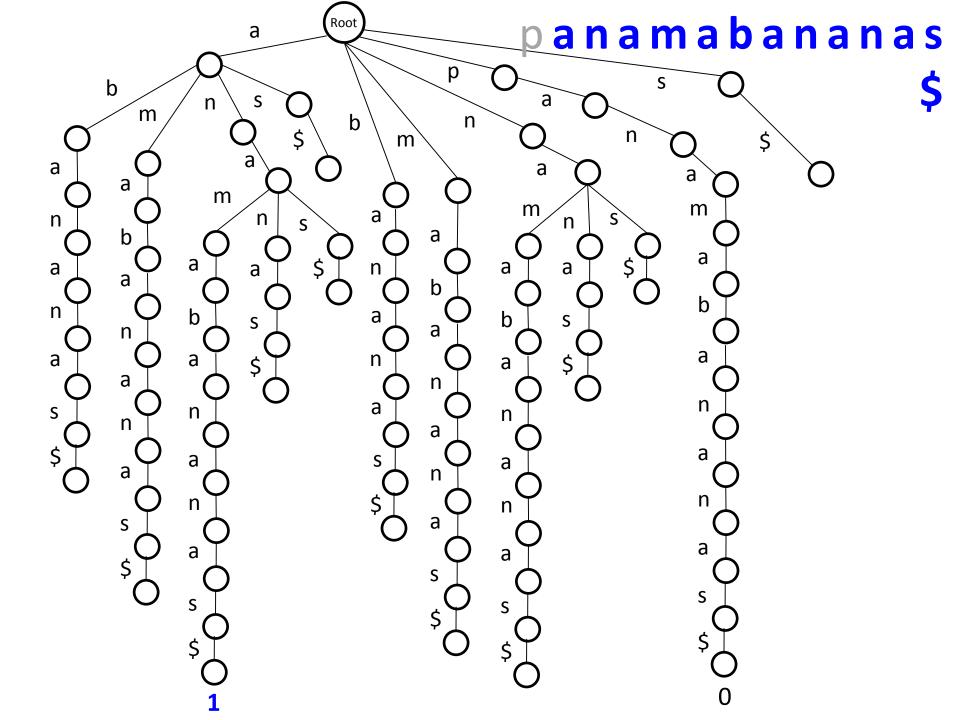
bananas\$

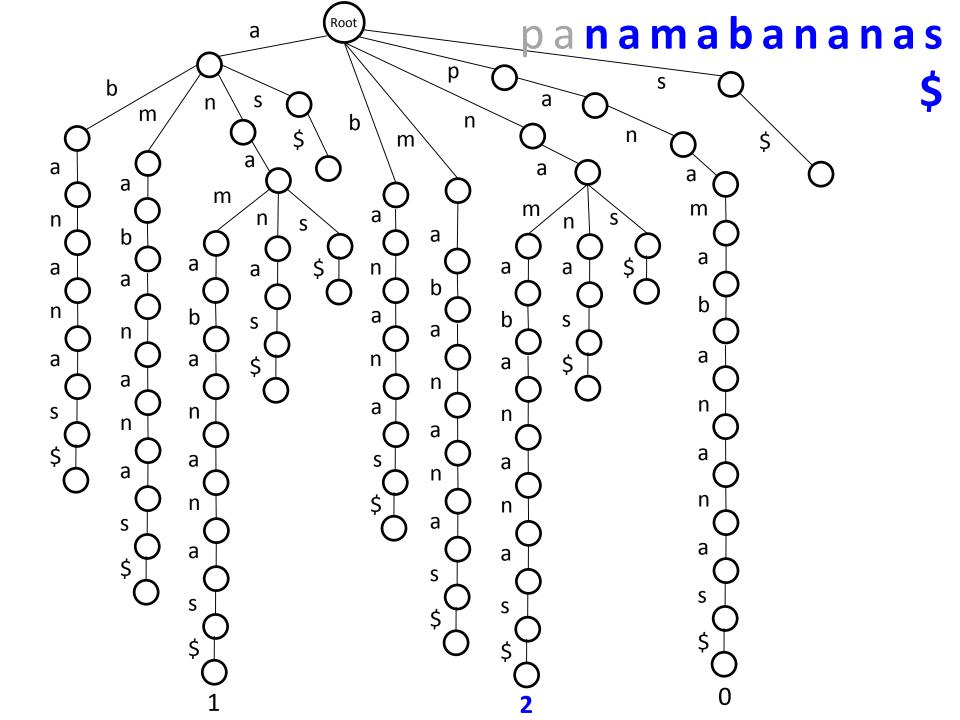
Where Are the Matches???

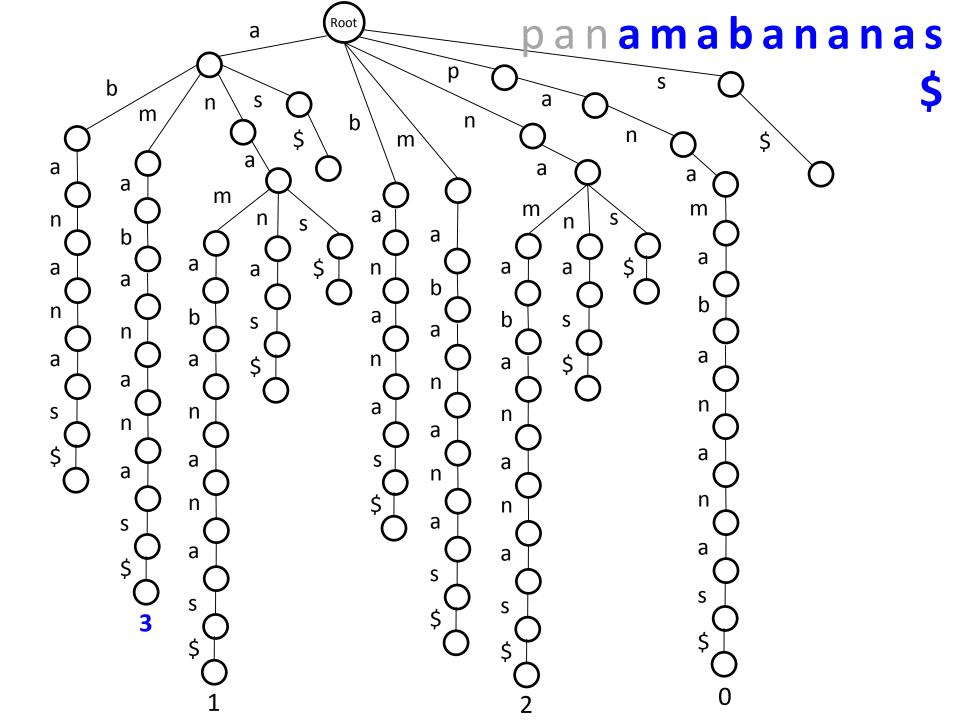


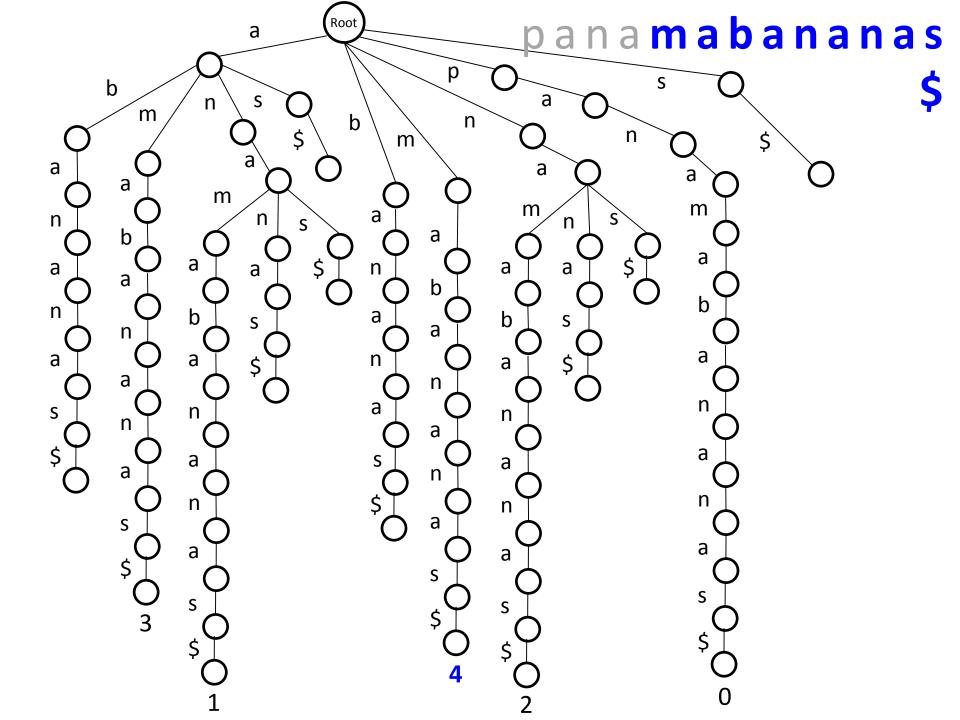


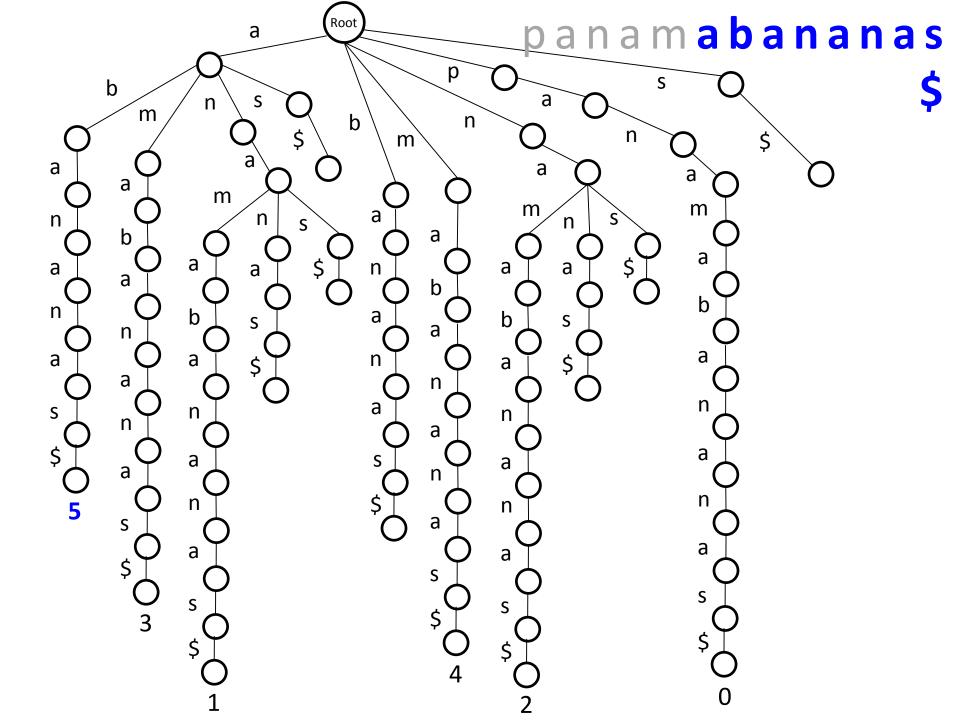


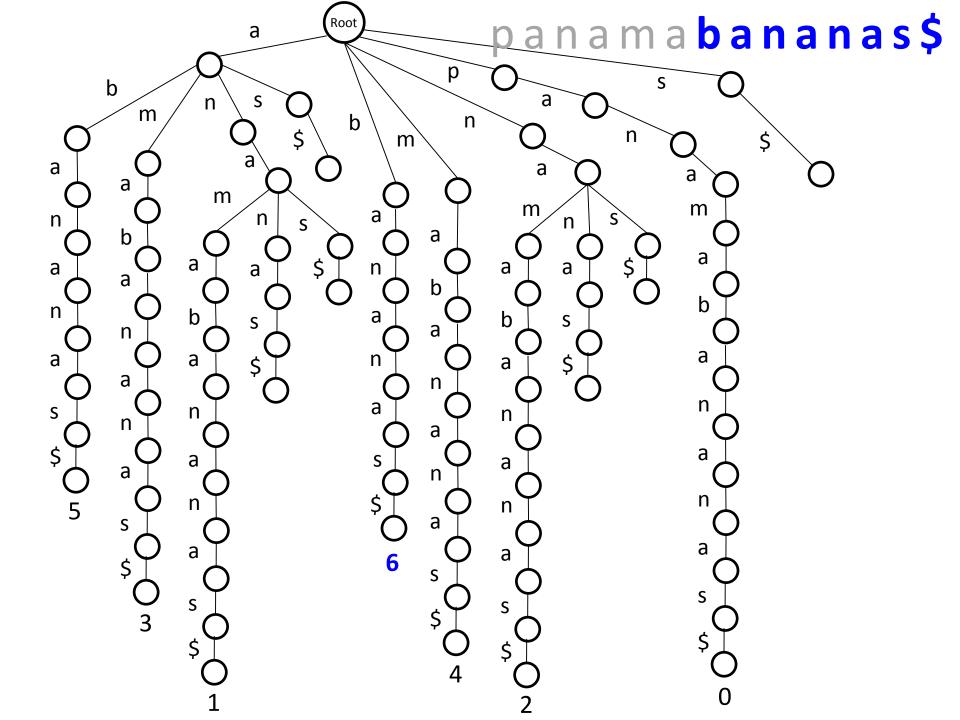


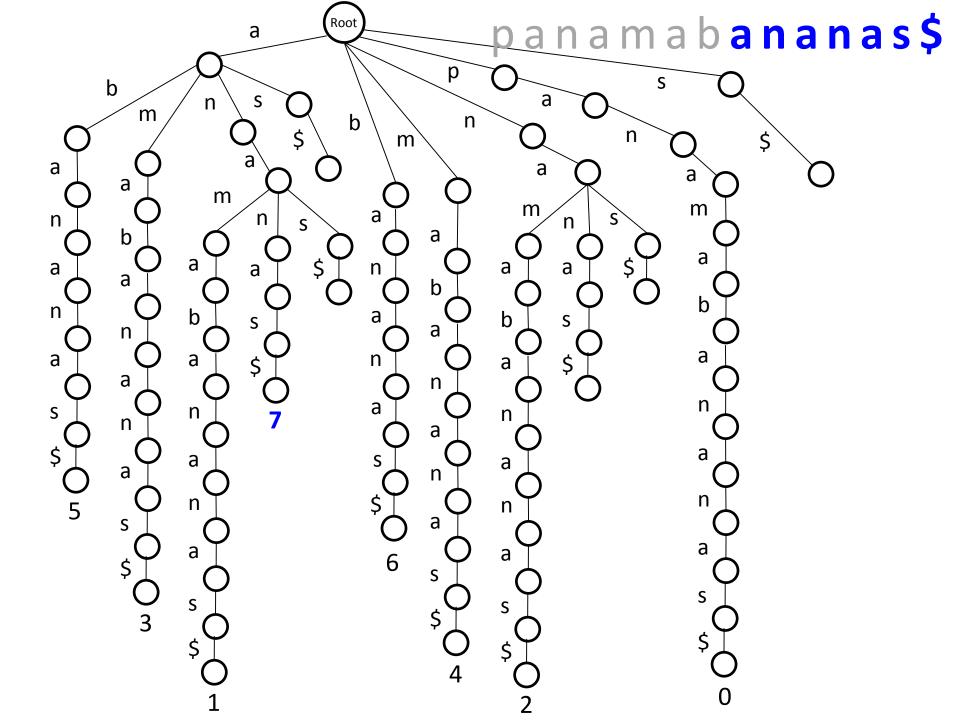


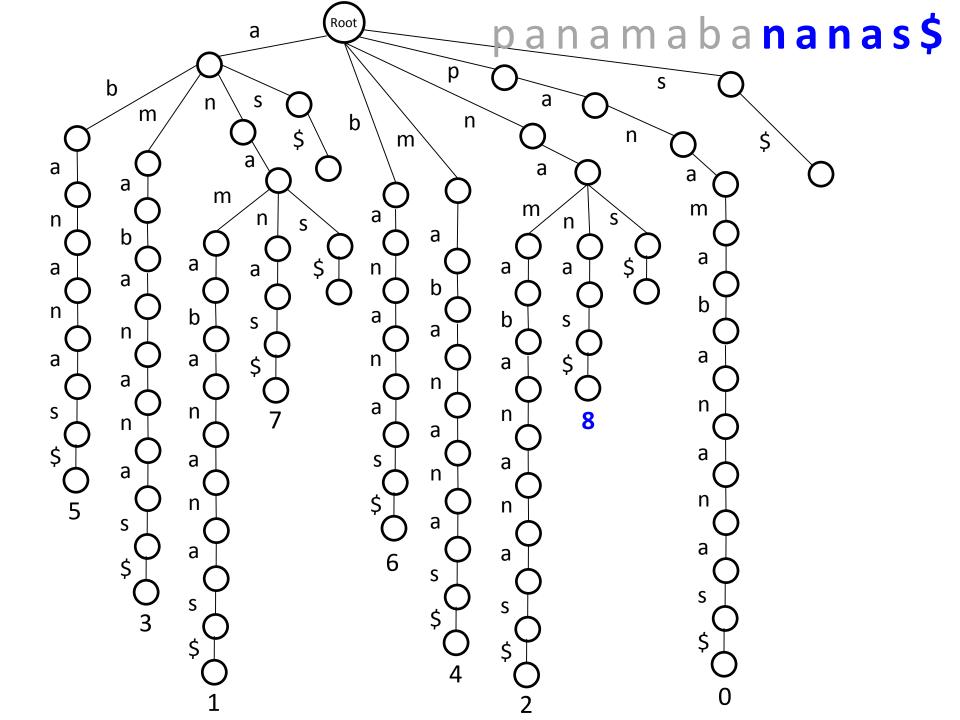


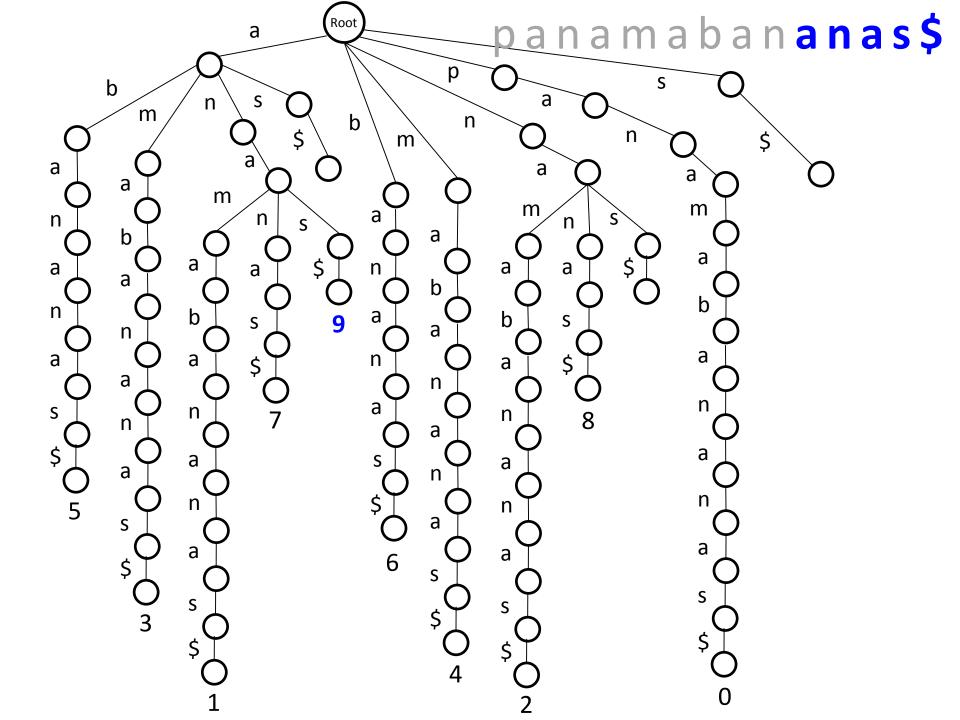


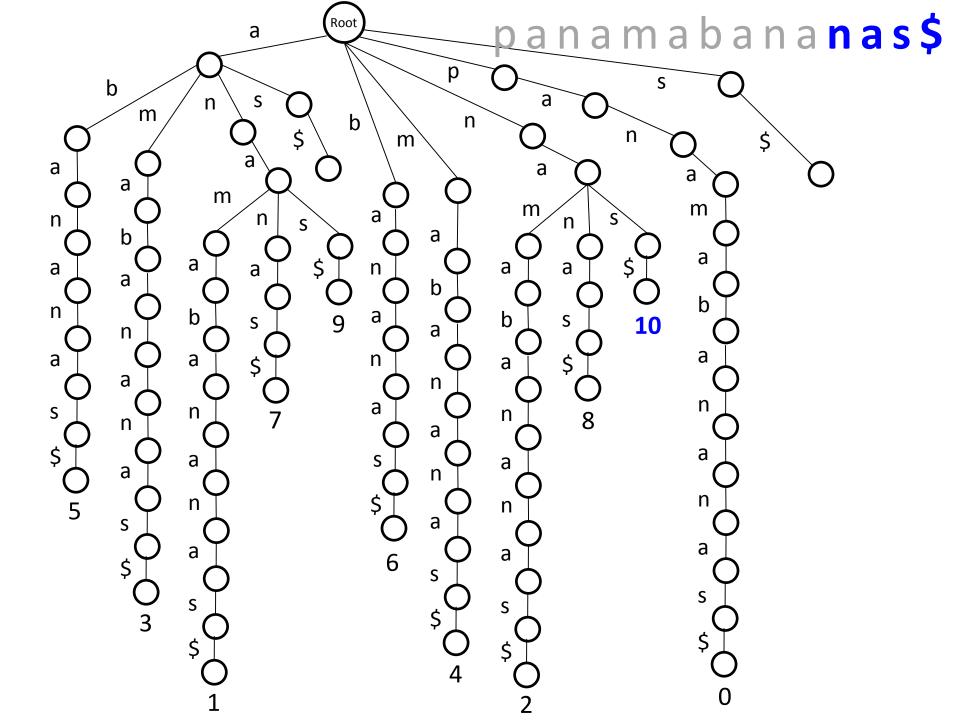


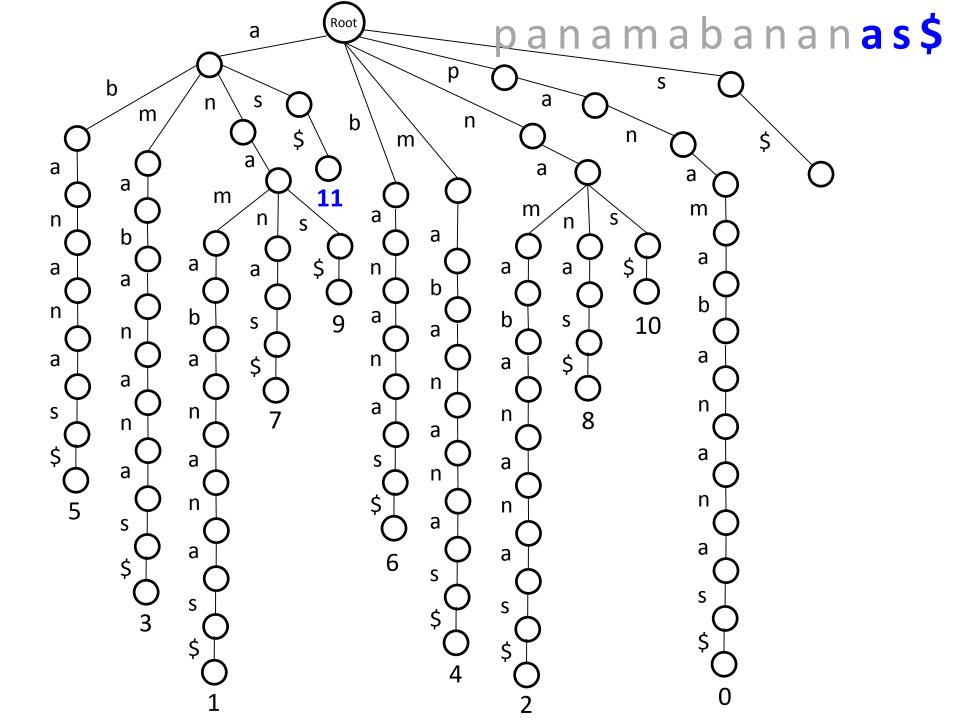


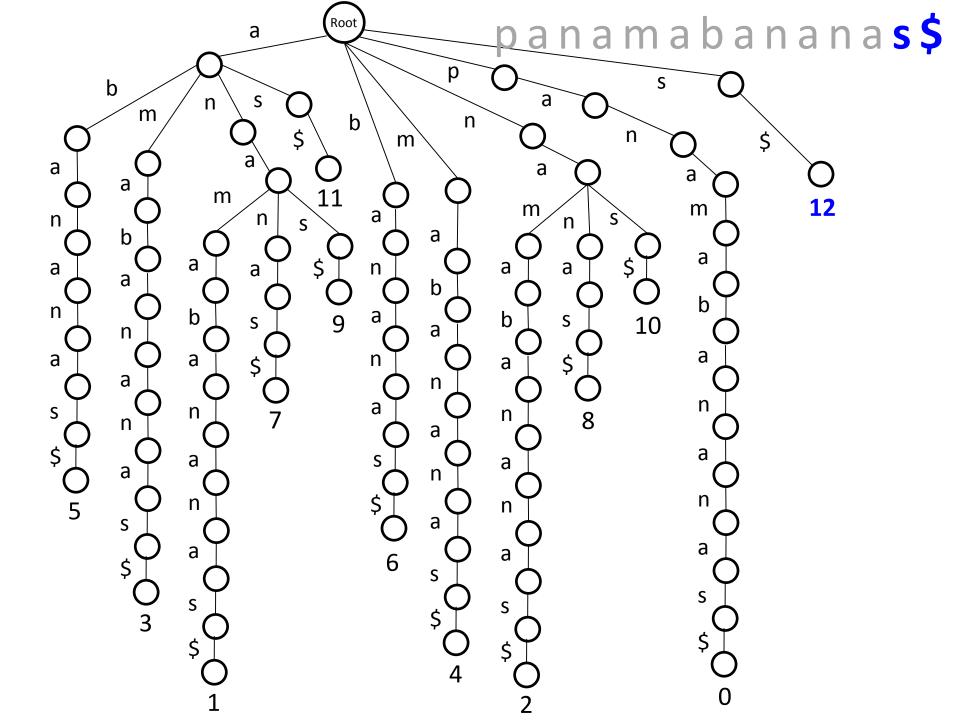


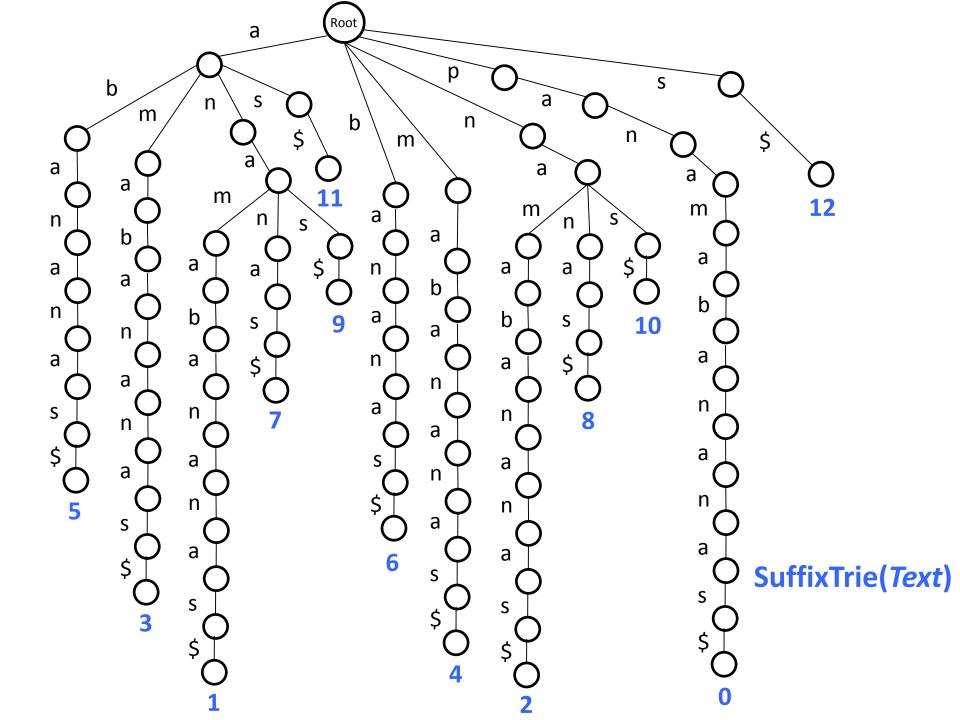






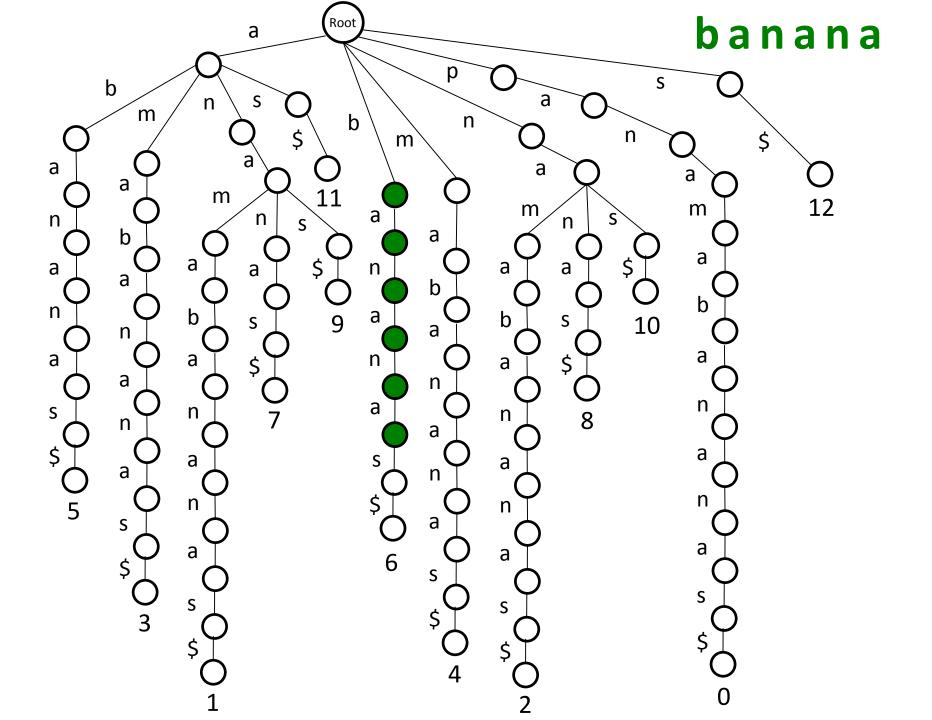


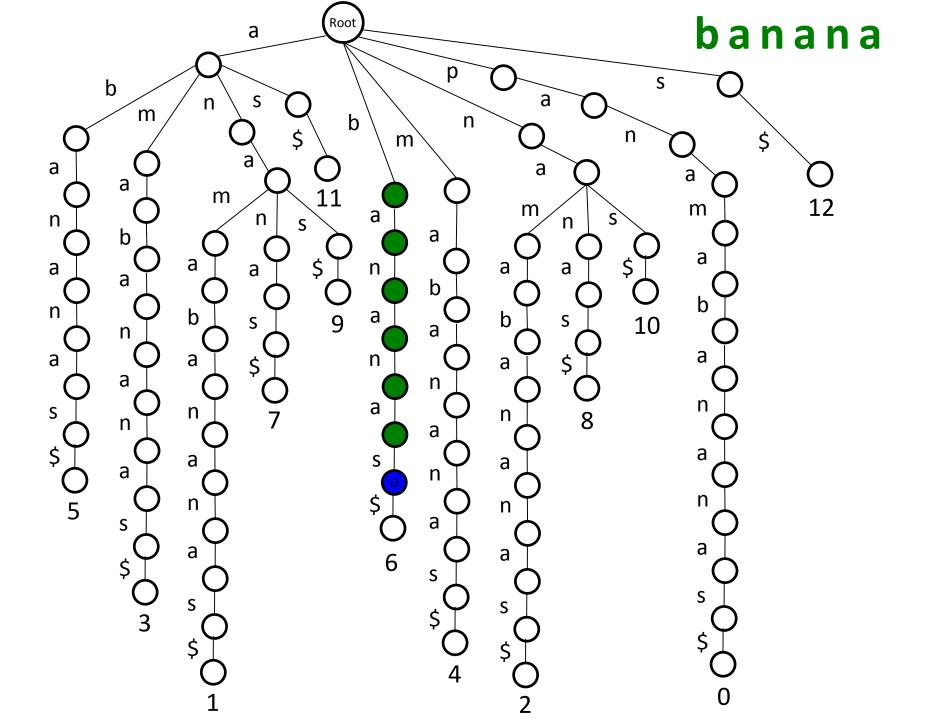


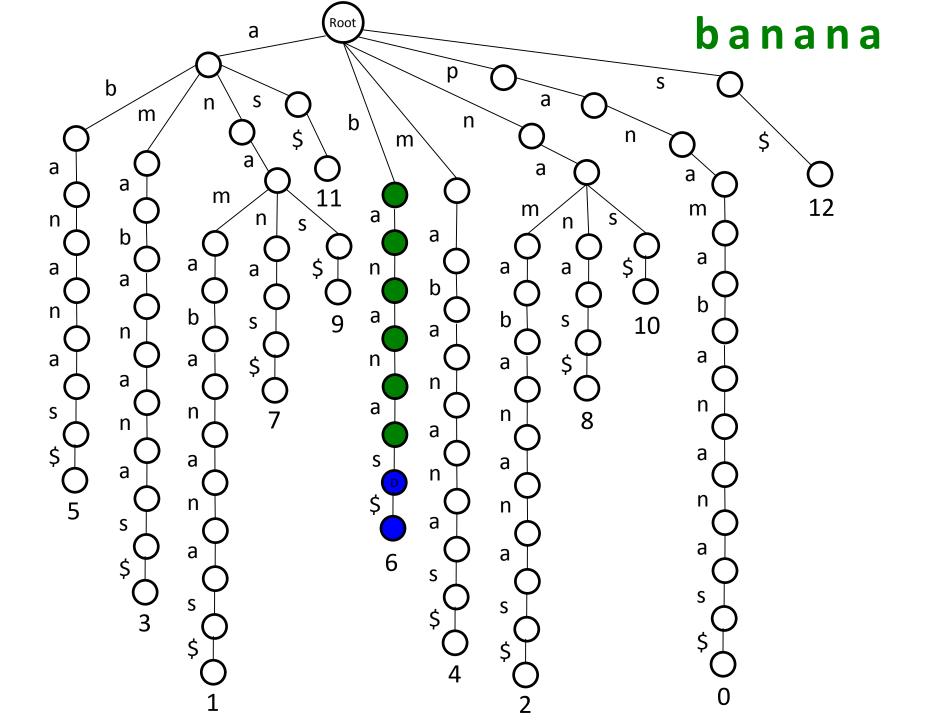


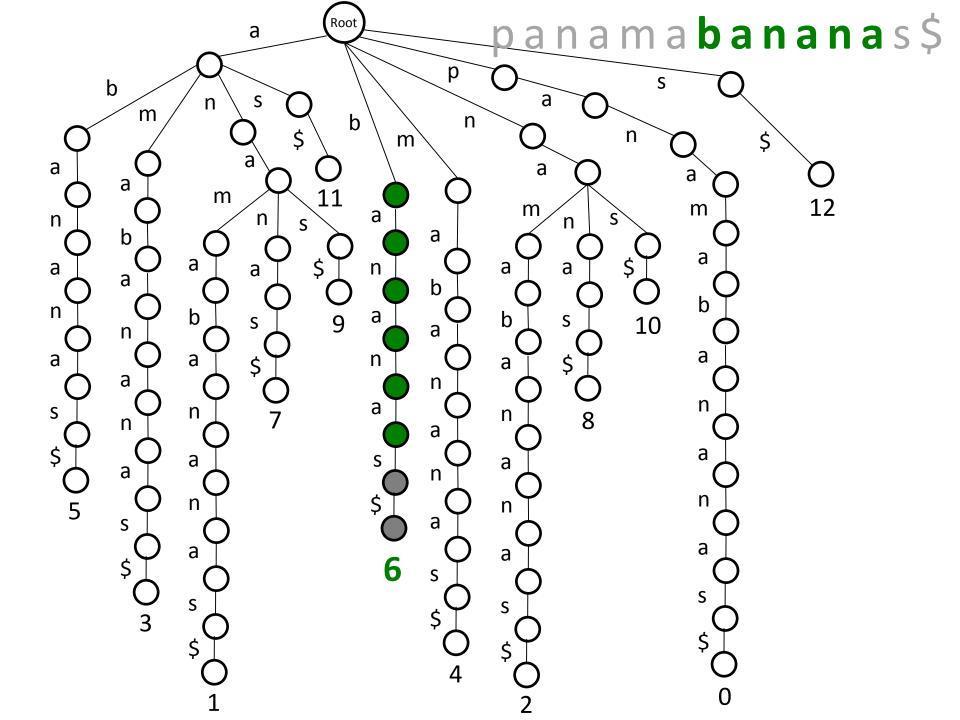
Walking Down to the Leaves to Find Matches

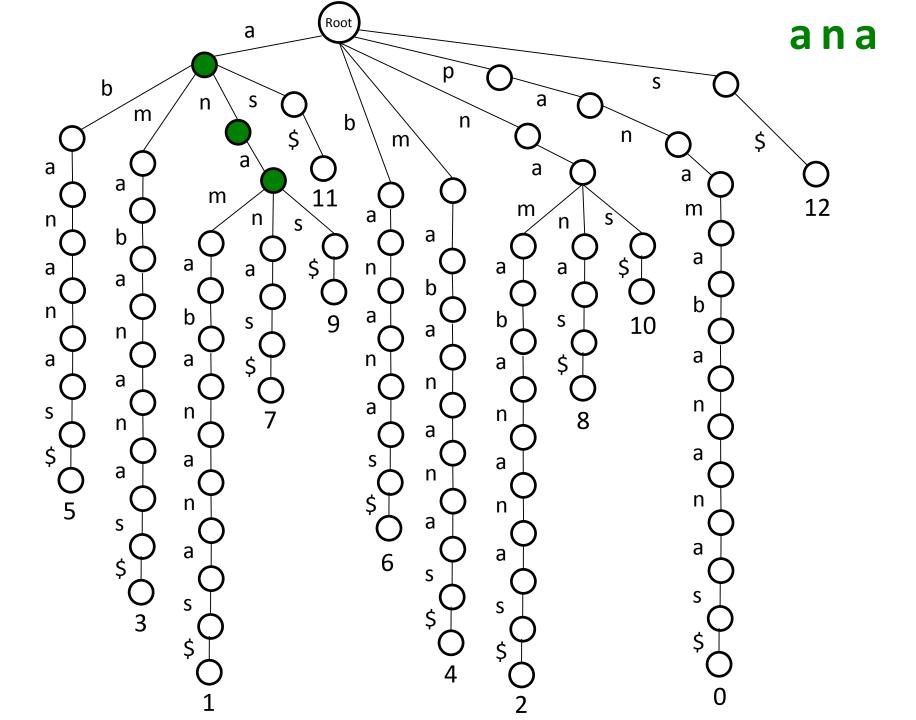
 Once we find a match, we "walk down" to the leaf (or leaves) in order to find the starting position of the match.

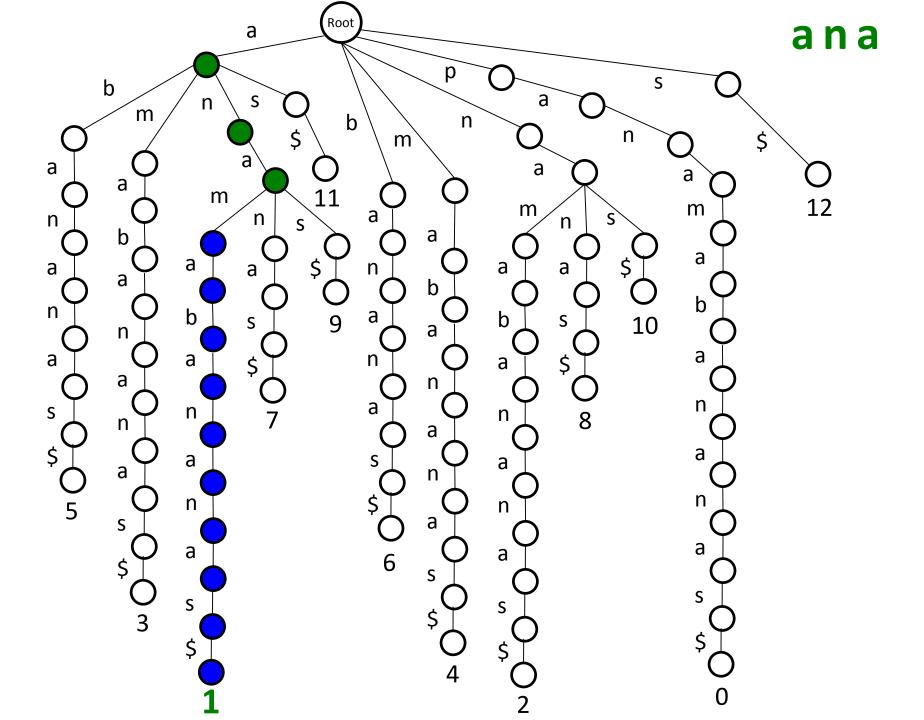


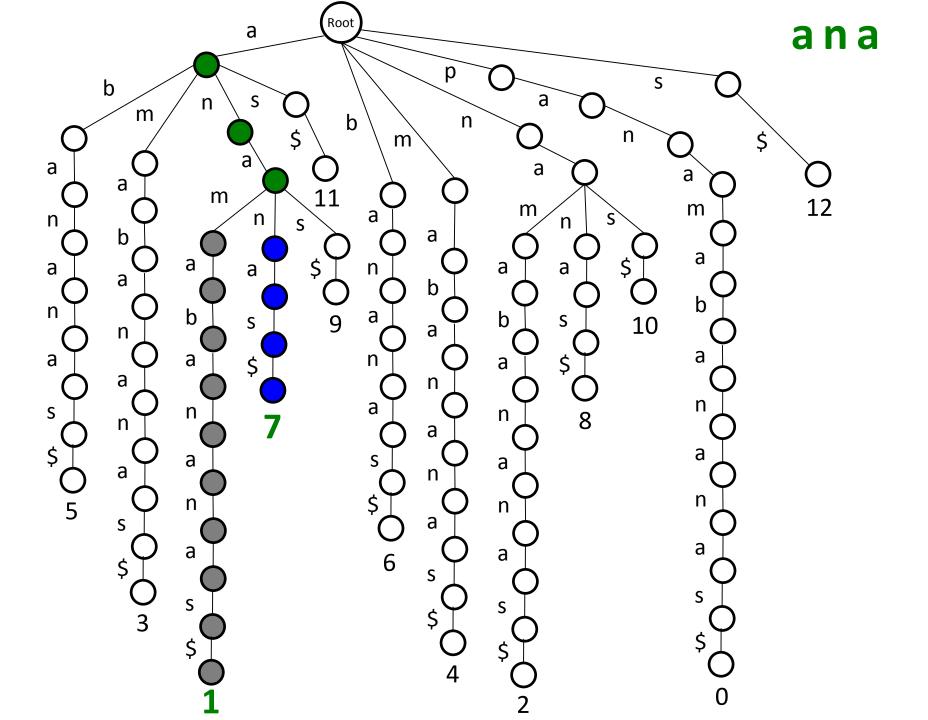


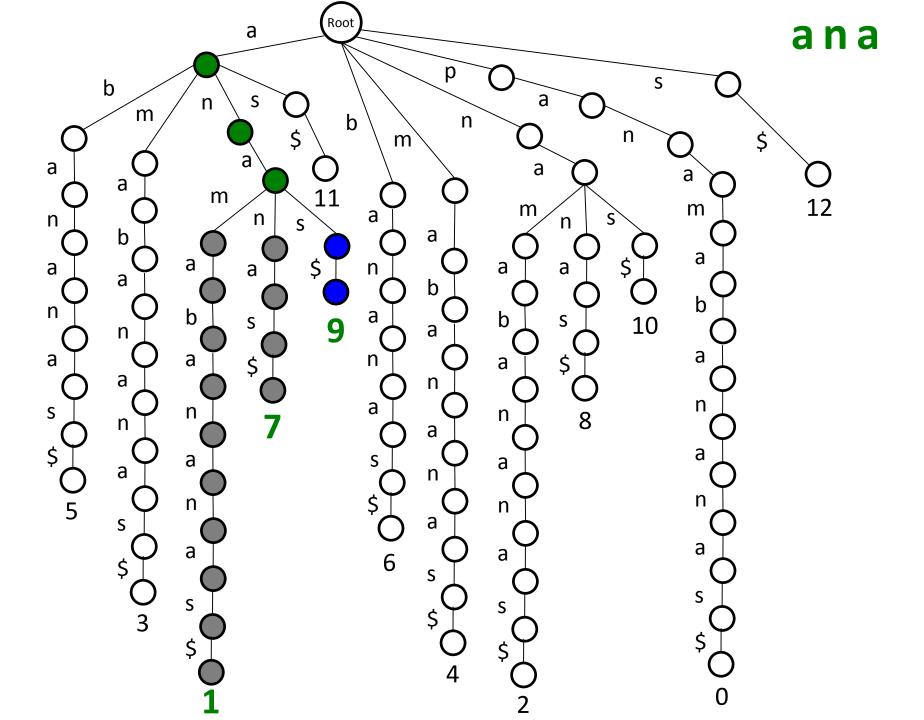


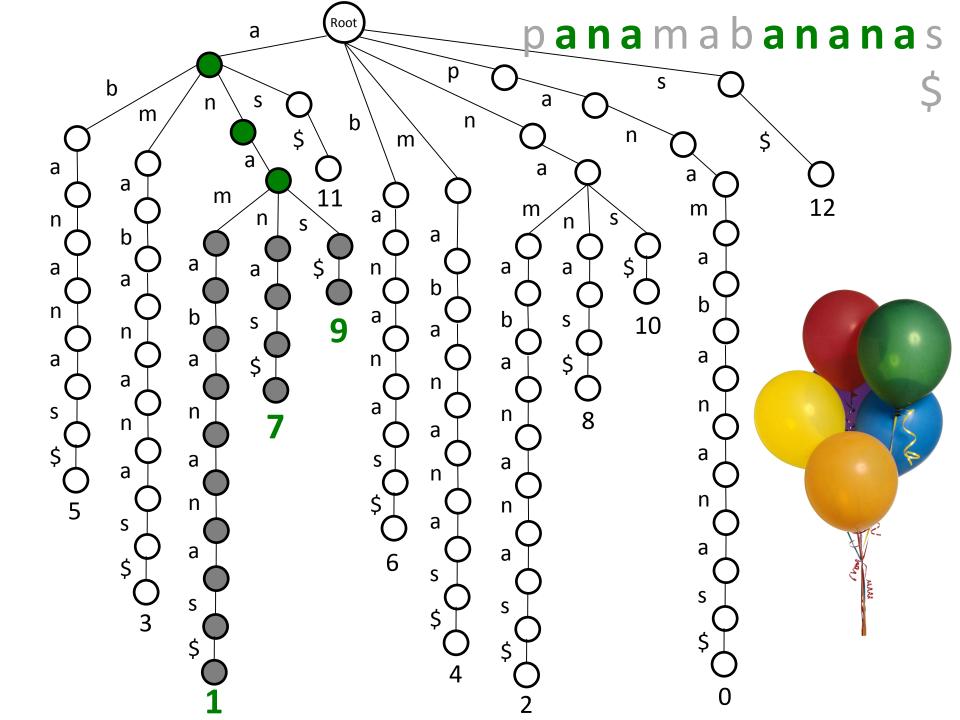












|*Text*| suffixes

Memory Footprint of Suffix Trie

The suffix trie is formed from | *Text* | suffixes with total length:

|Text|*(|Text|-1)/2

For human genome:

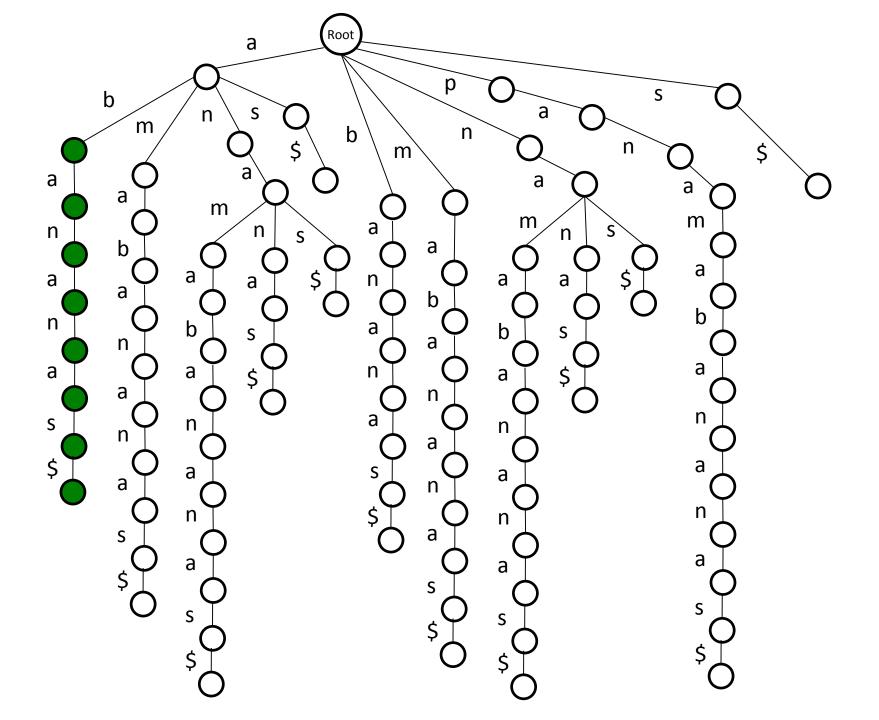
• $|Text| \approx 3*10^9$

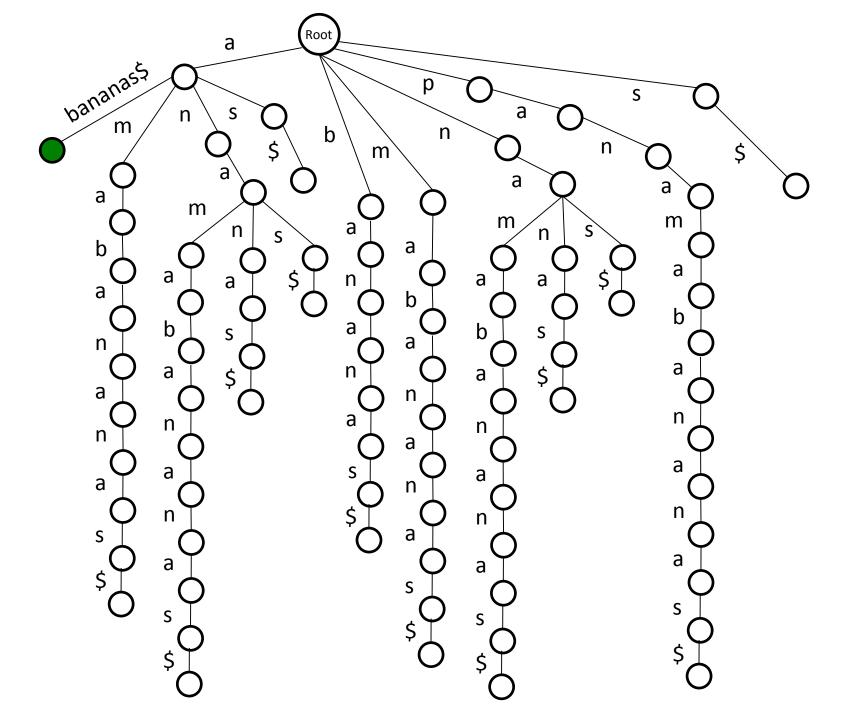


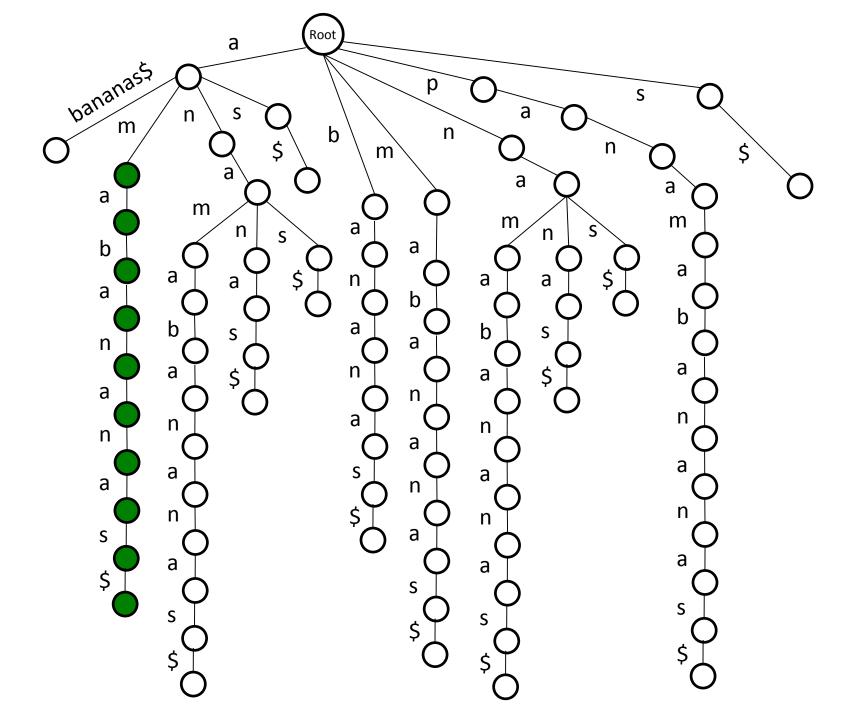
```
|Text| symbols
```

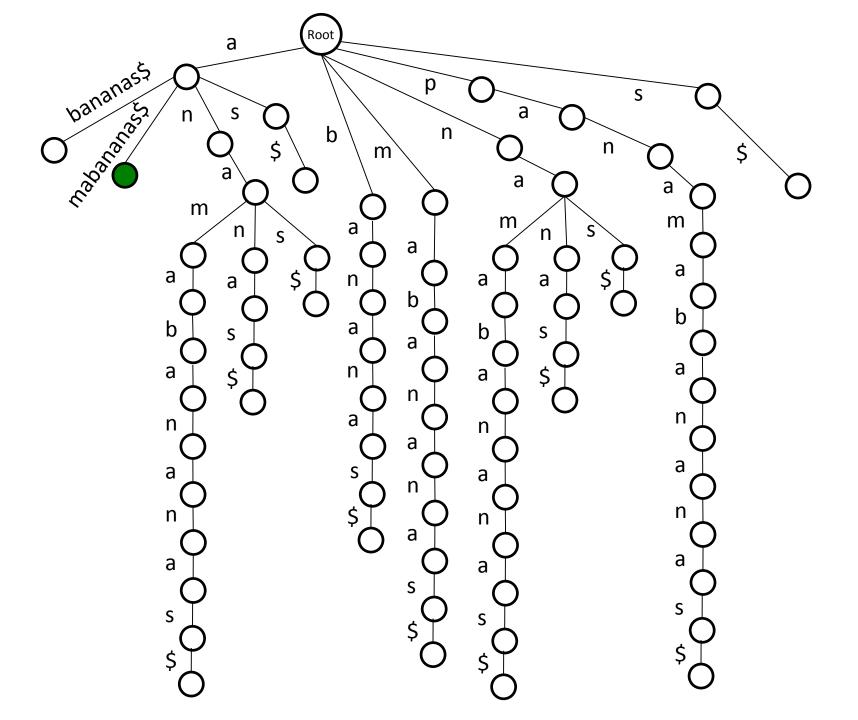
Outline

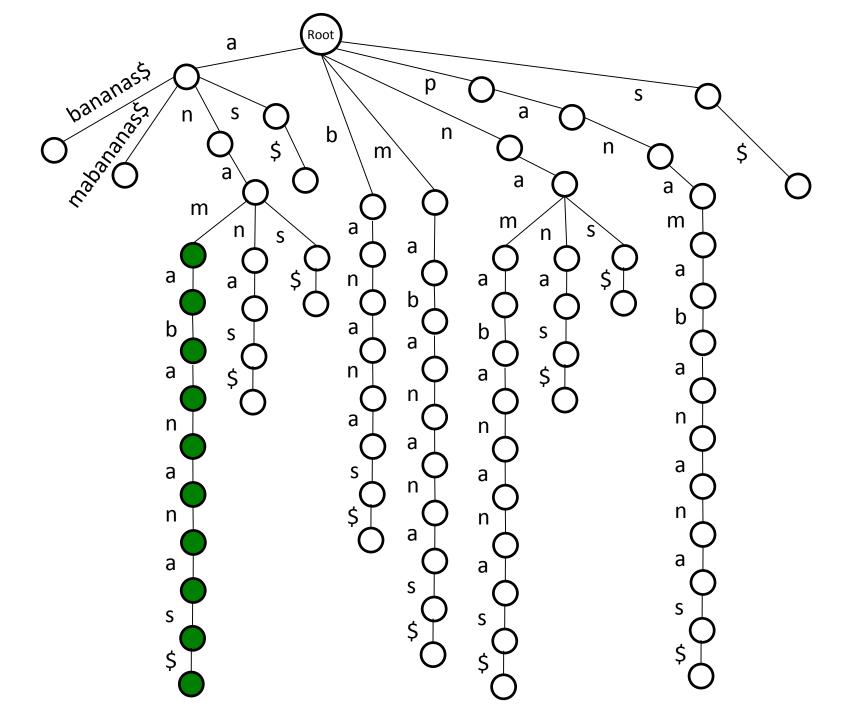
- From Genome Sequencing to Pattern Matching
- Brute Force Approach to Pattern Matching
- Herding Patterns into Trie
- Herding Text into Suffix Trie
- From Suffix Tries to Suffix Trees

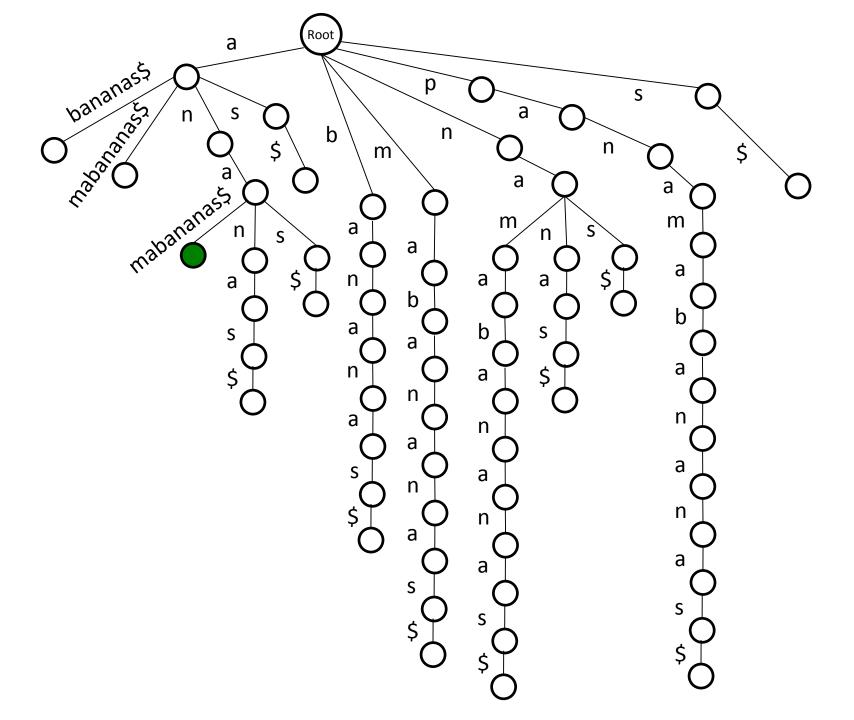


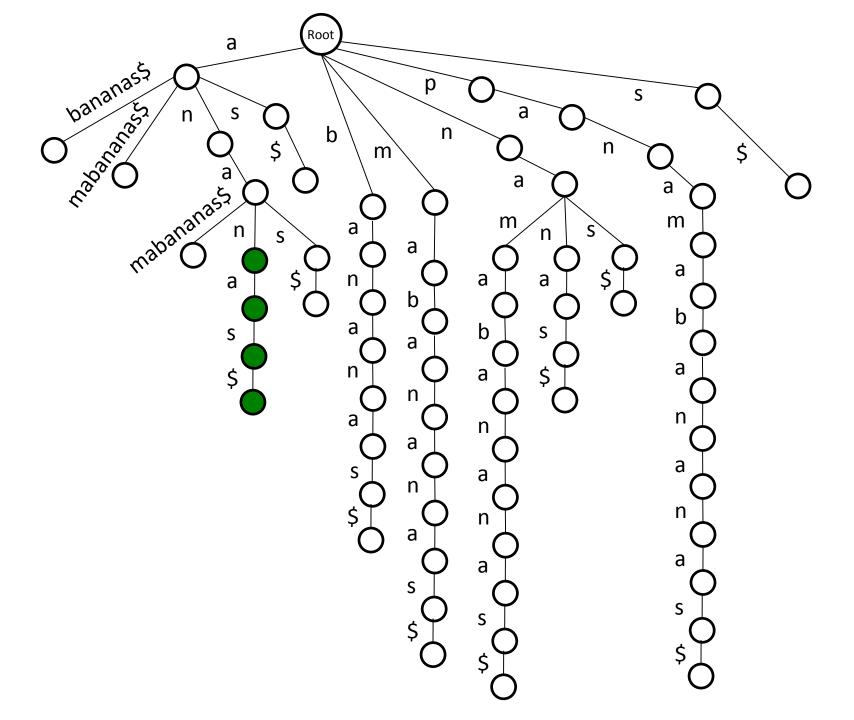


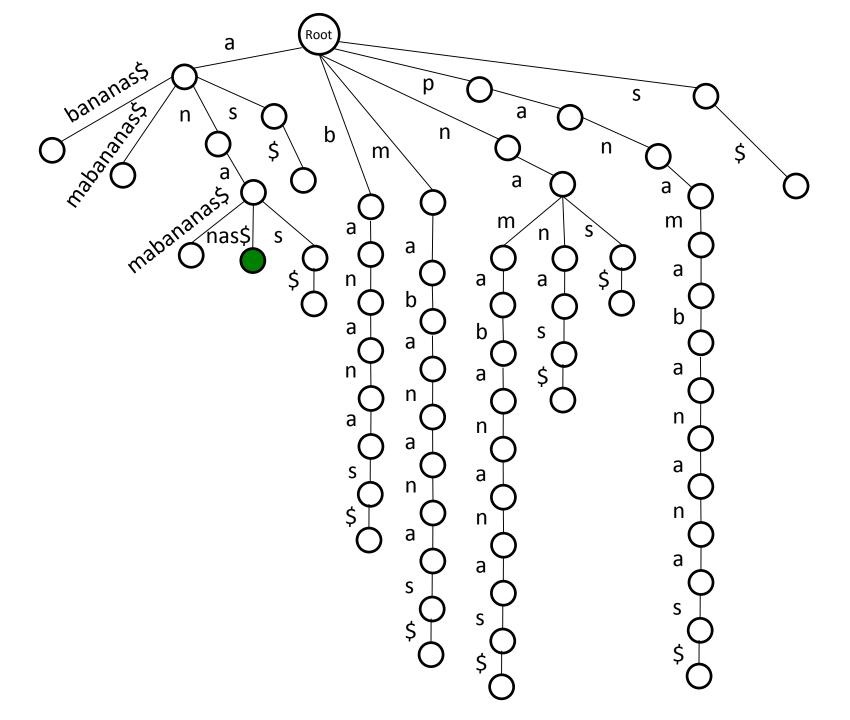


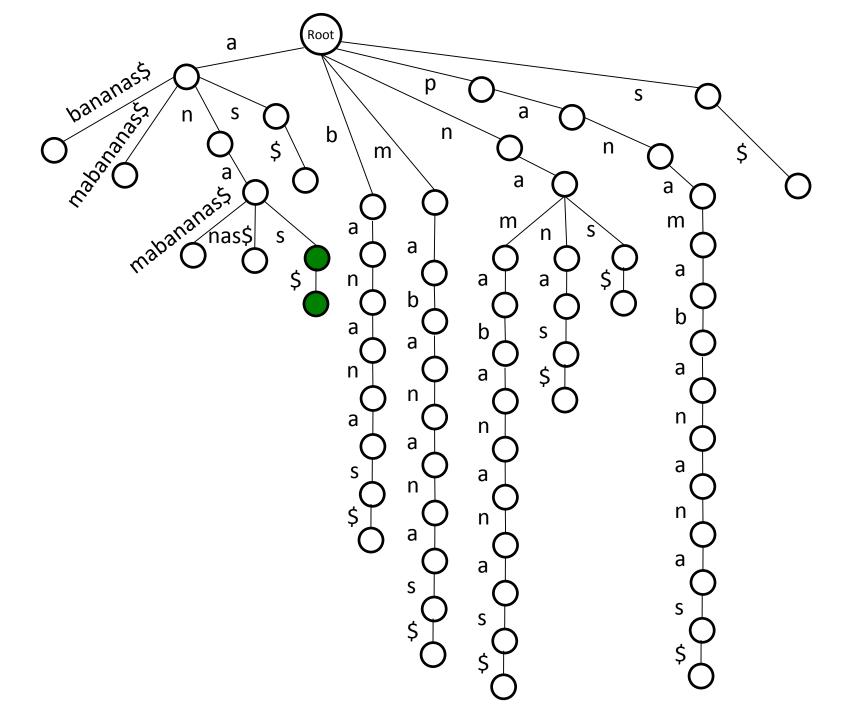


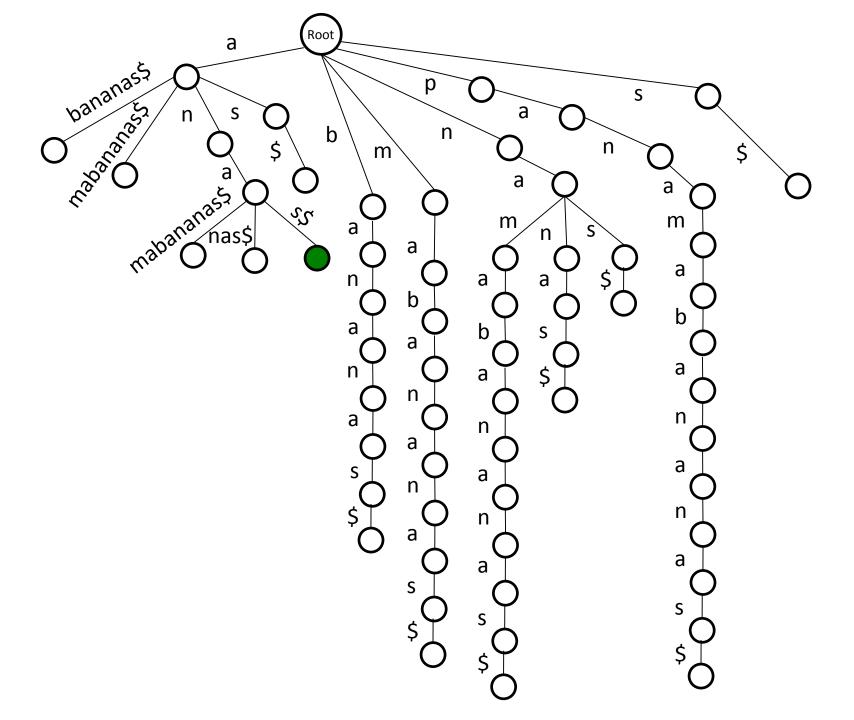


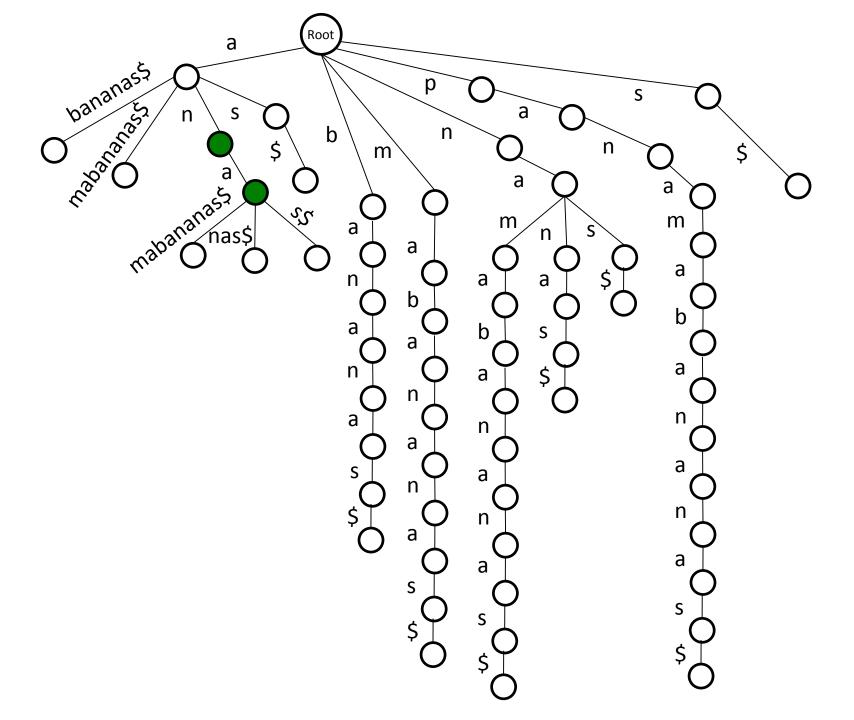


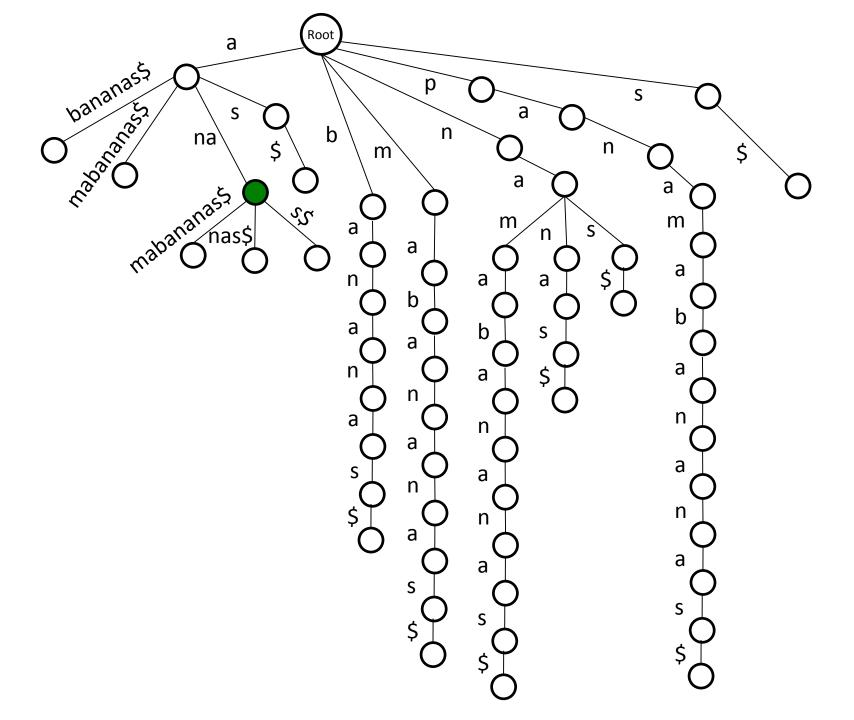


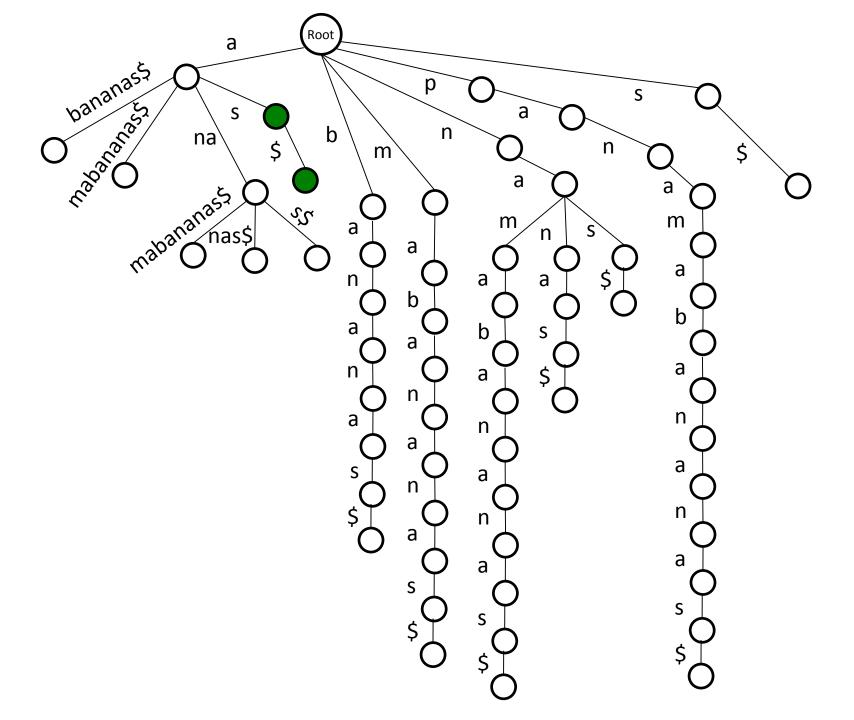


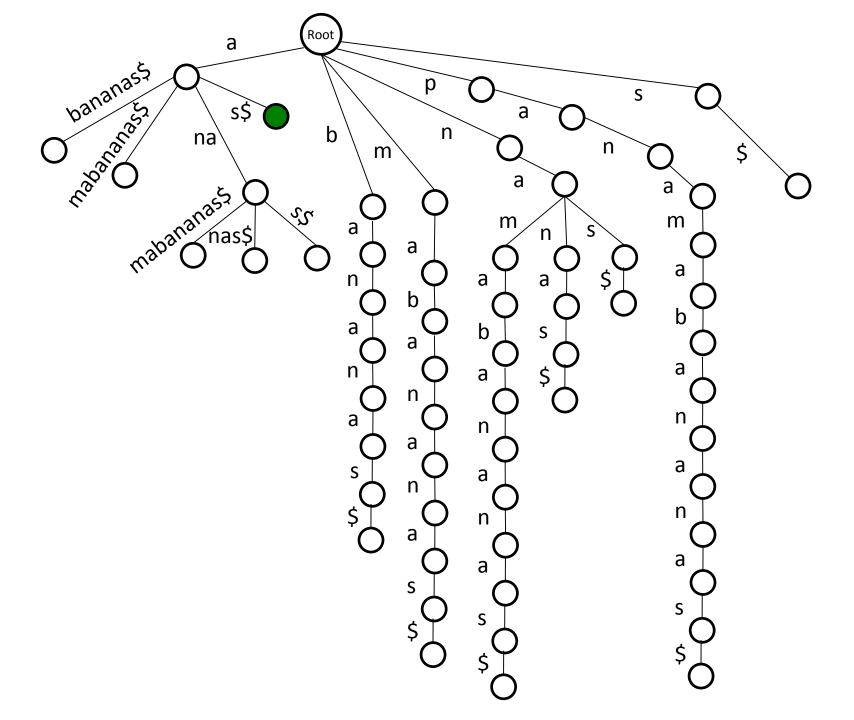


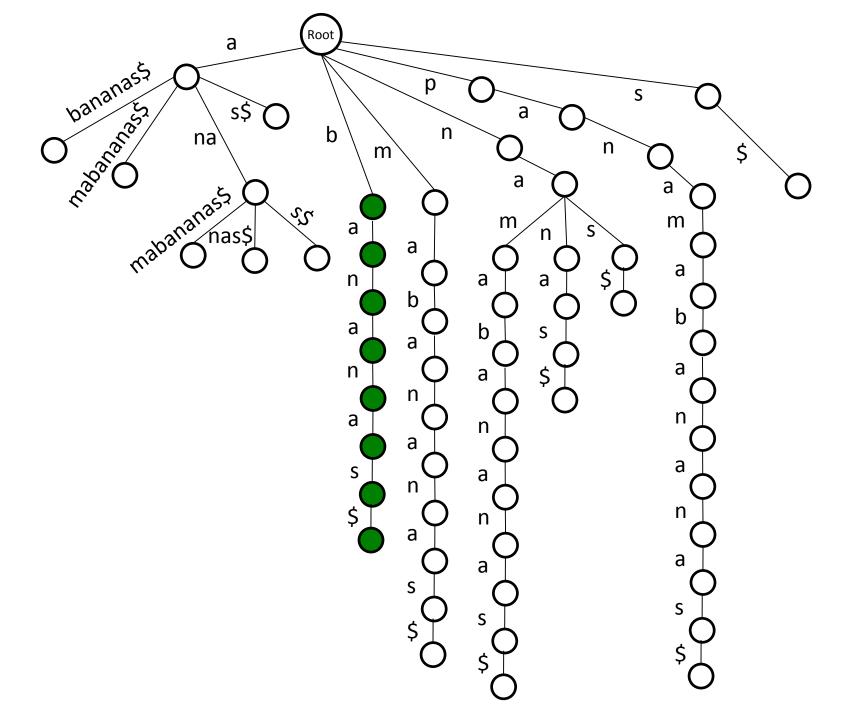


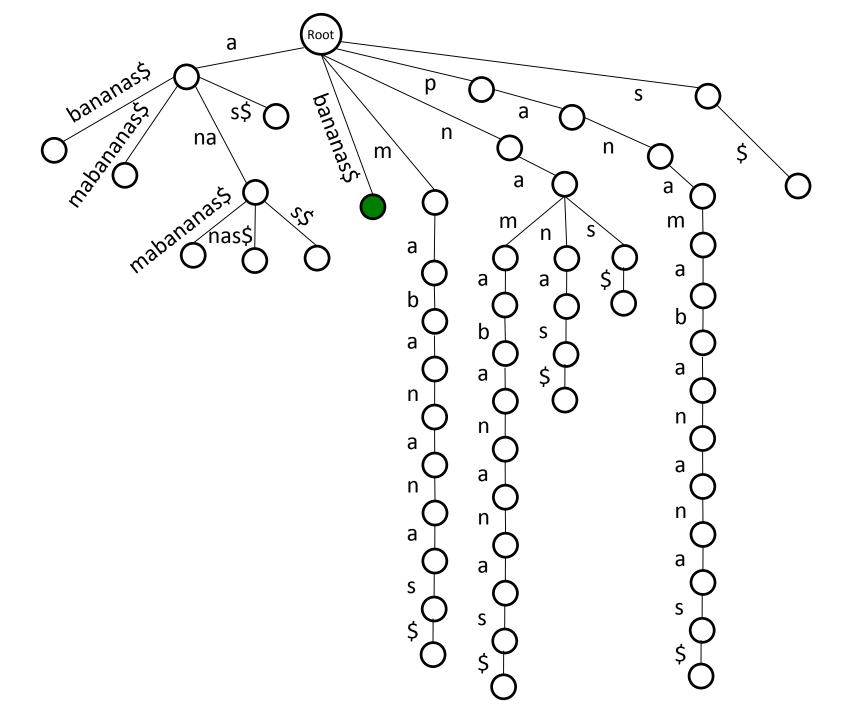


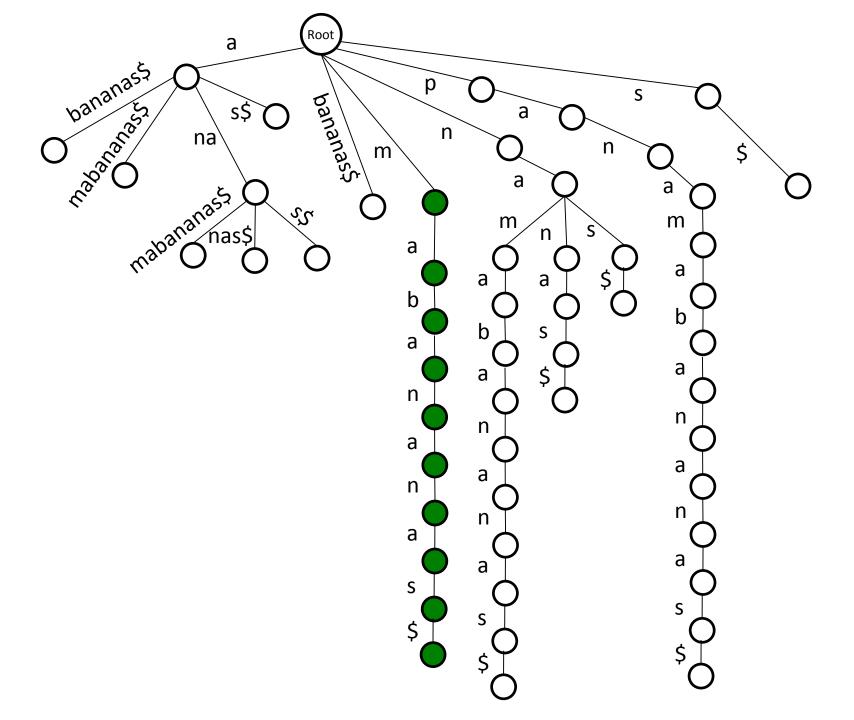


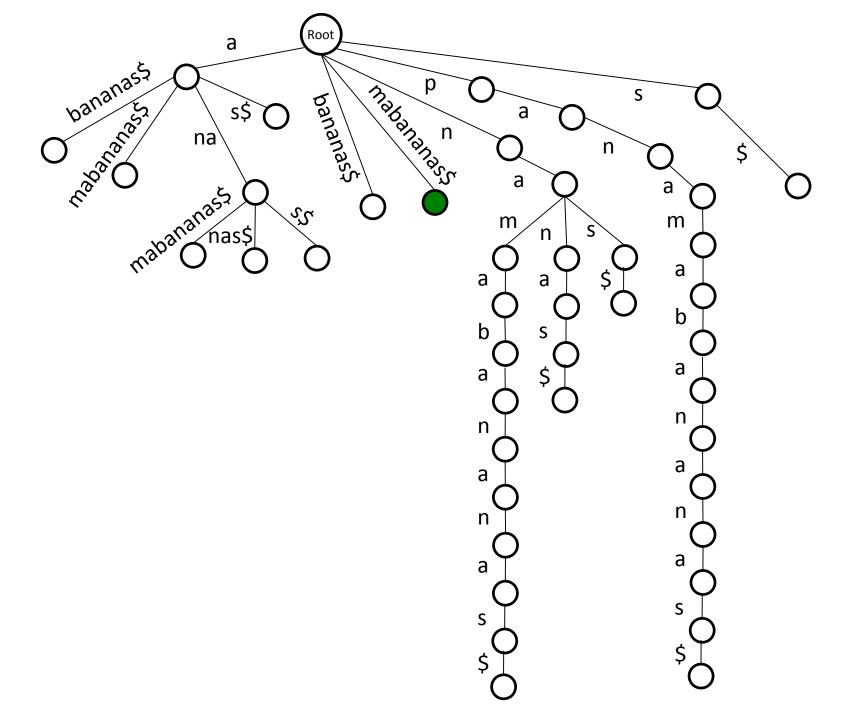


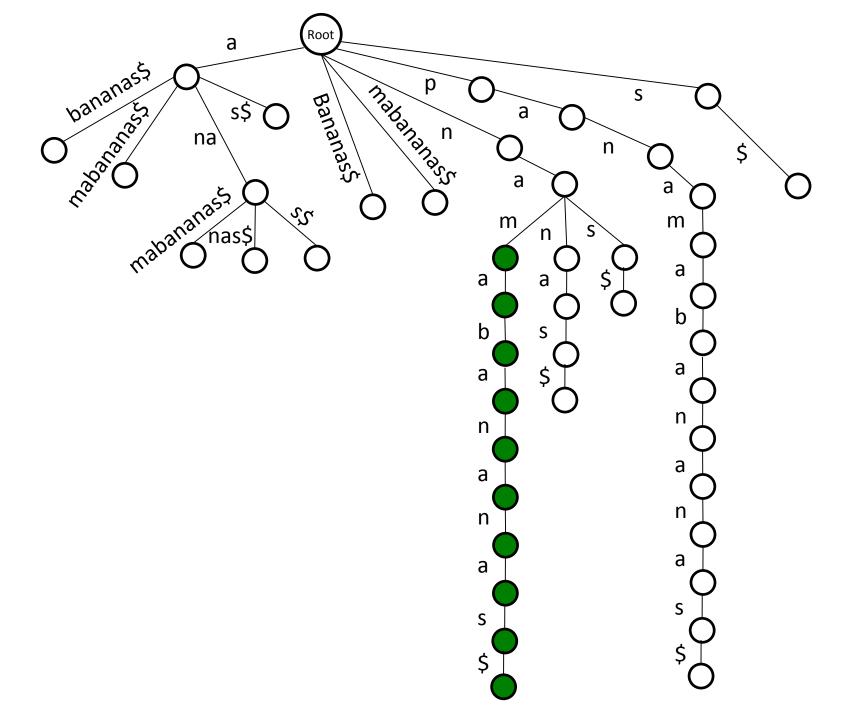


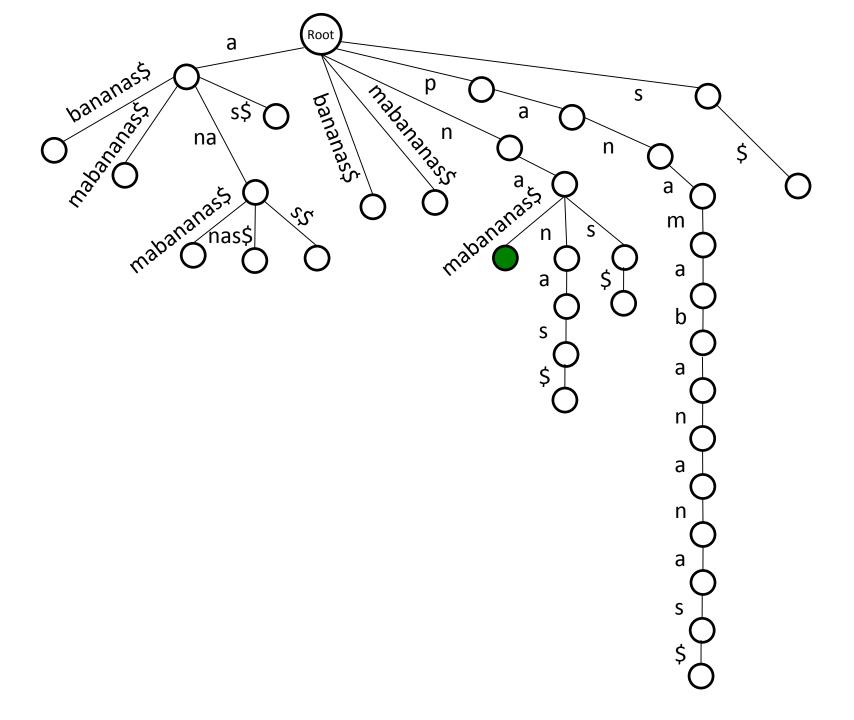


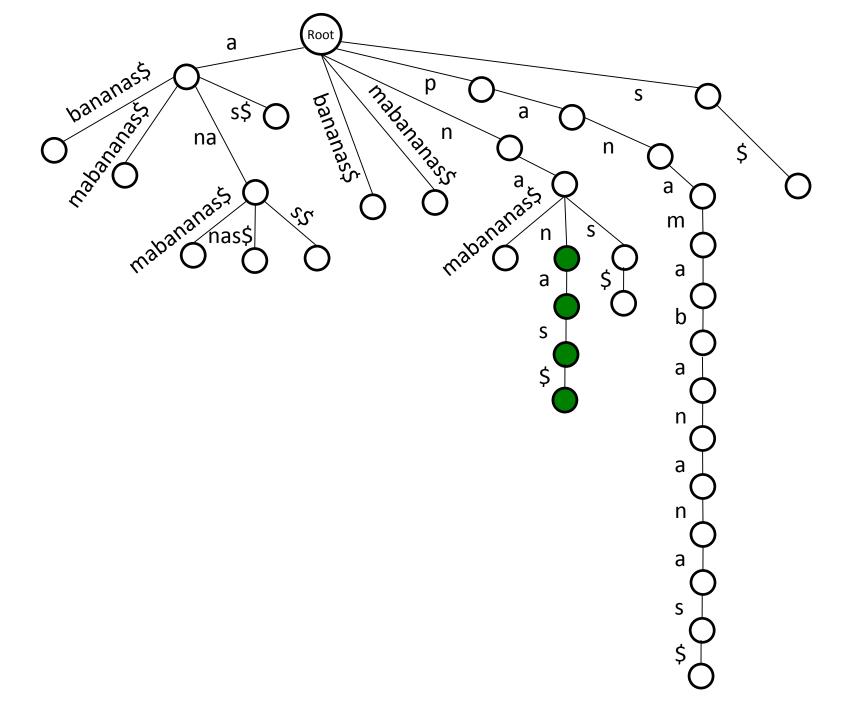


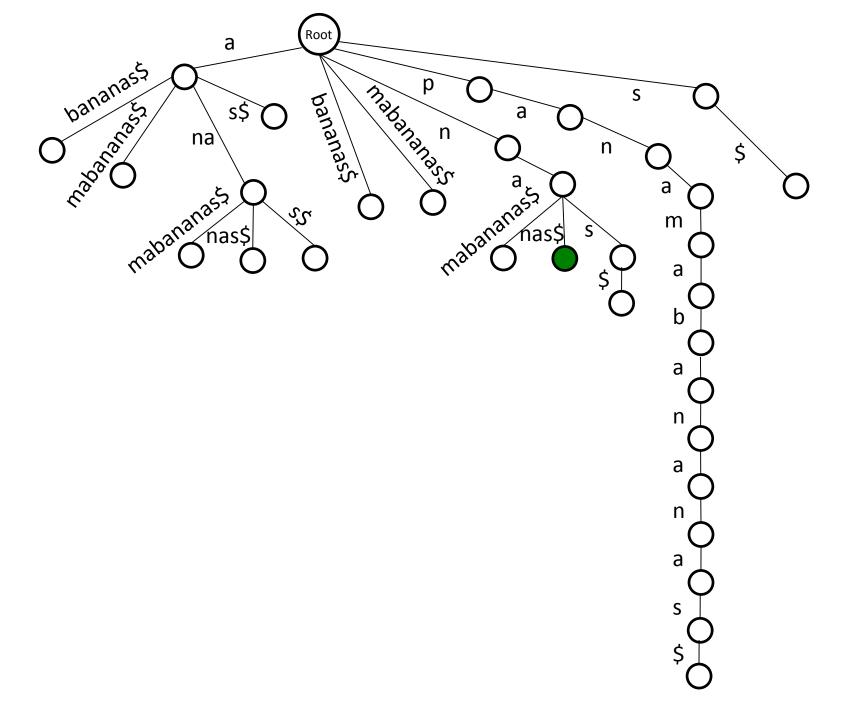


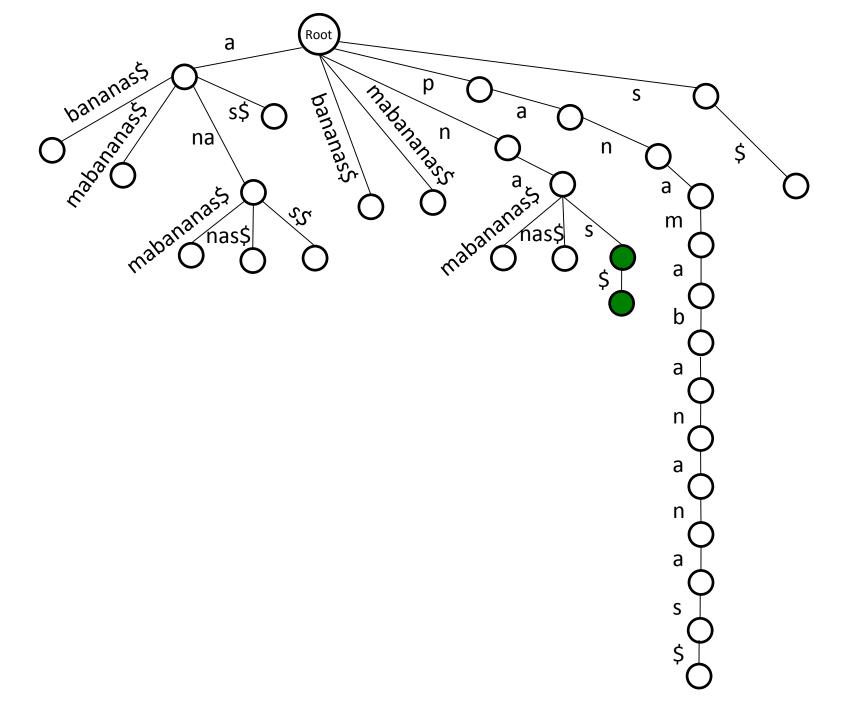


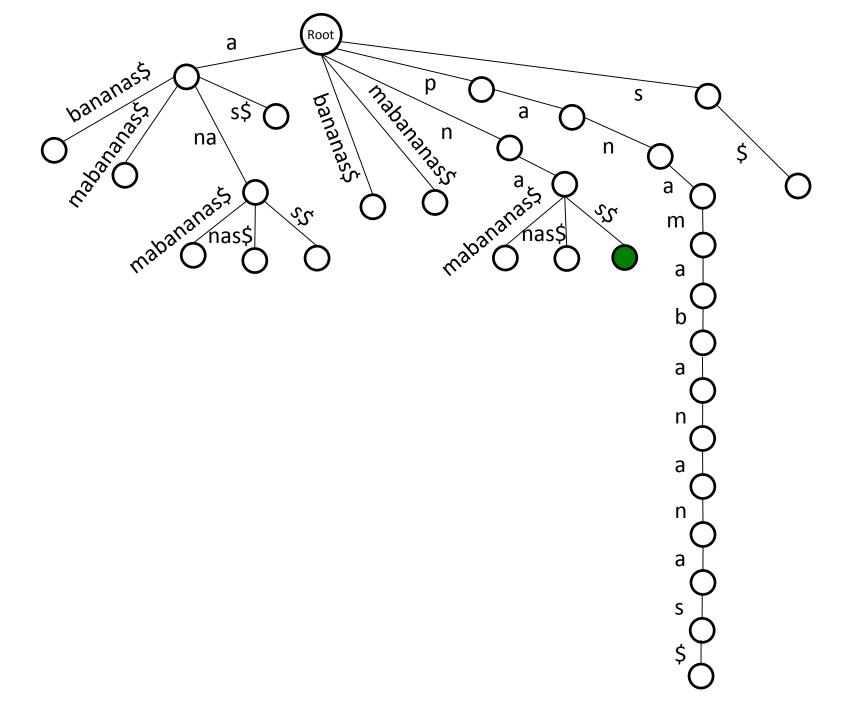


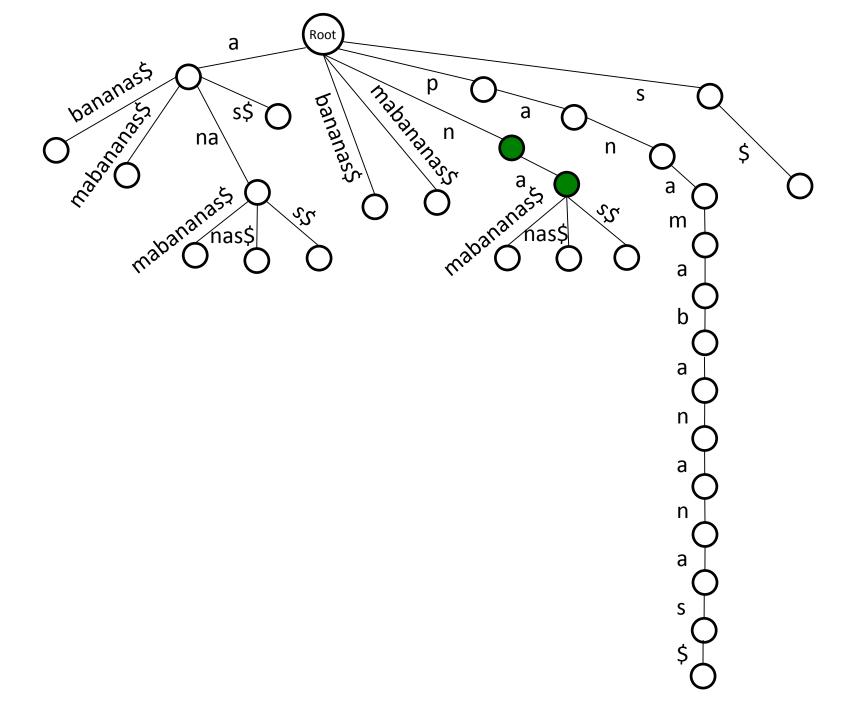


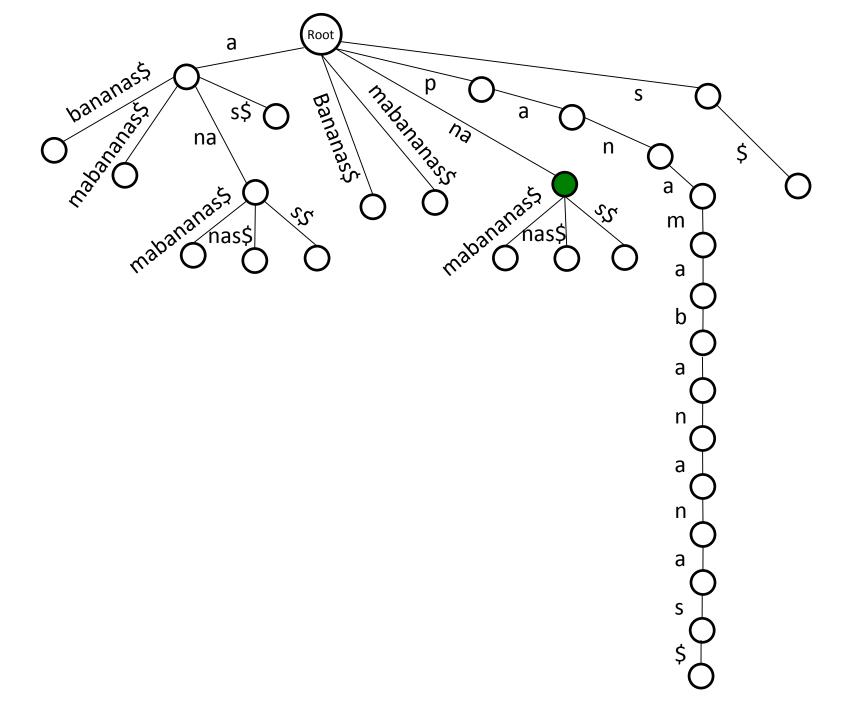


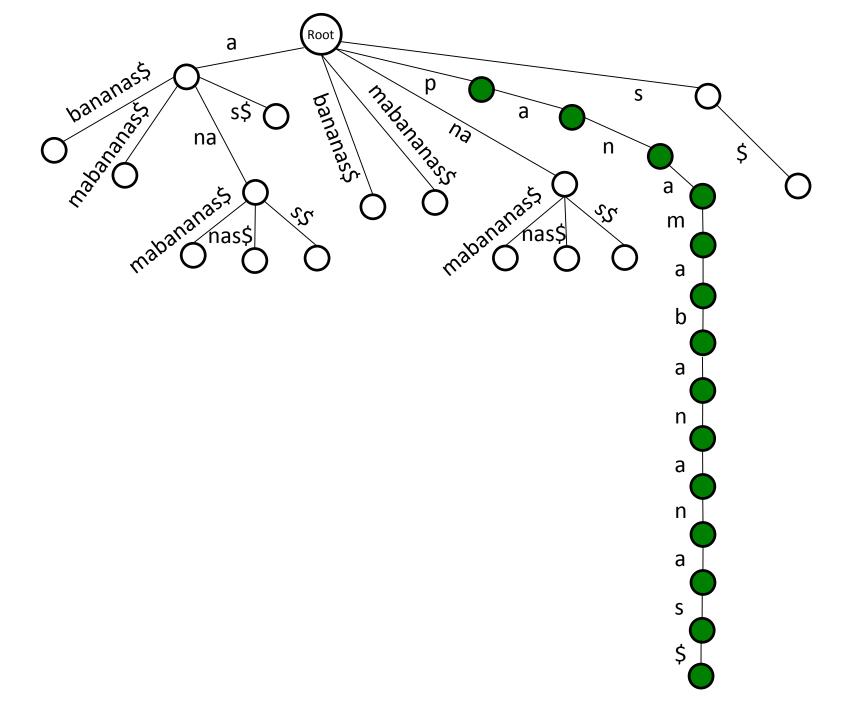


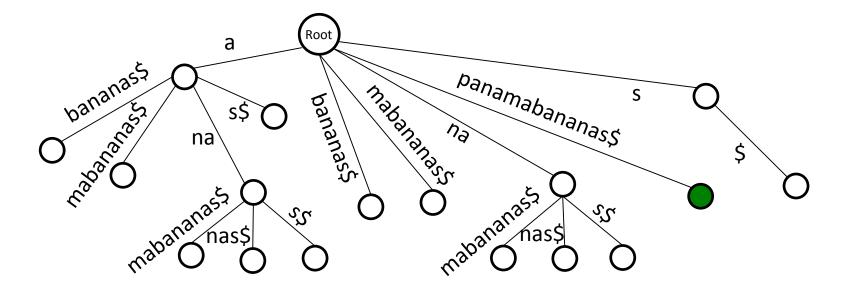


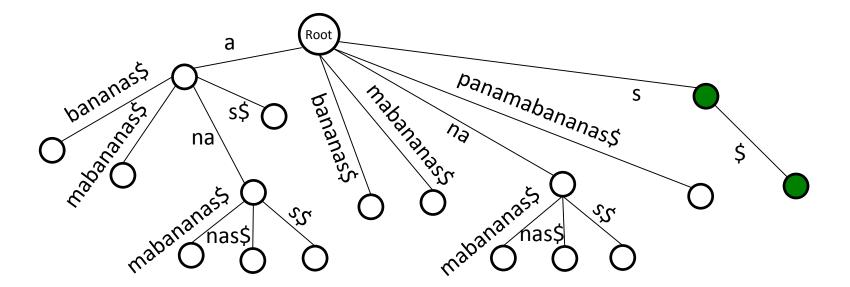


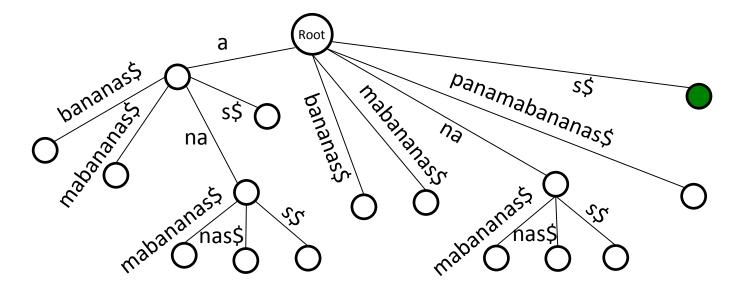




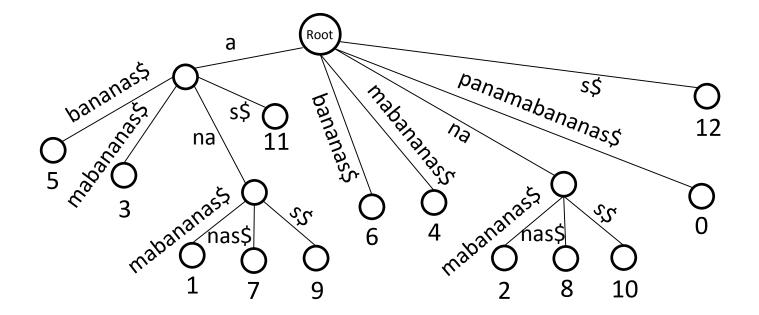




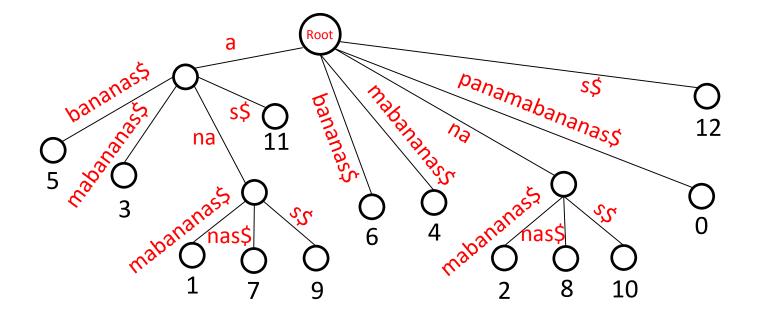




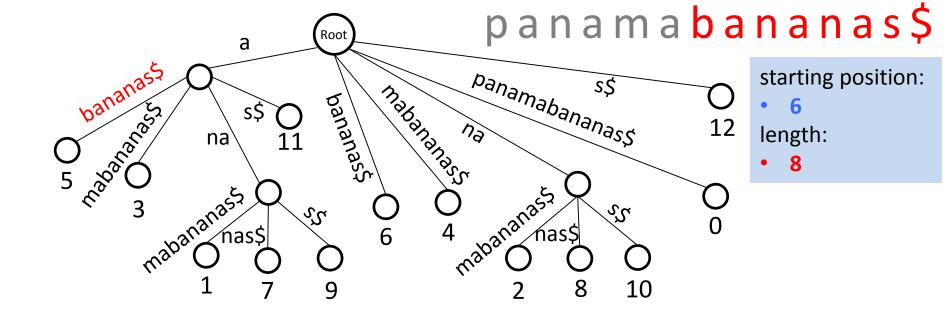
SuffixTree(*Text*)



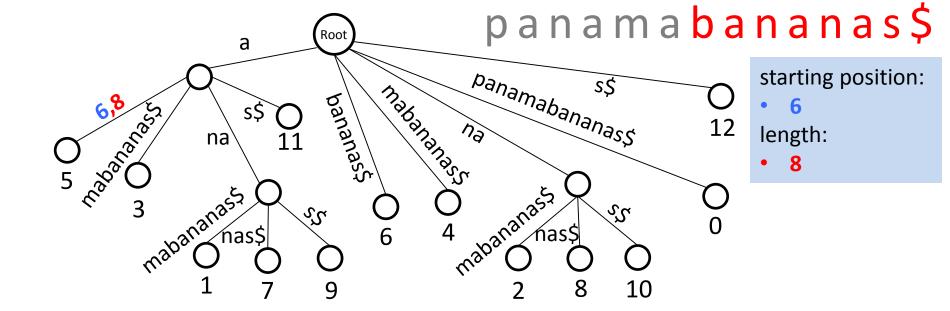
- # vertices < 2 | *Text* |
- memory footprint of the suffix tree: O(|Text|)



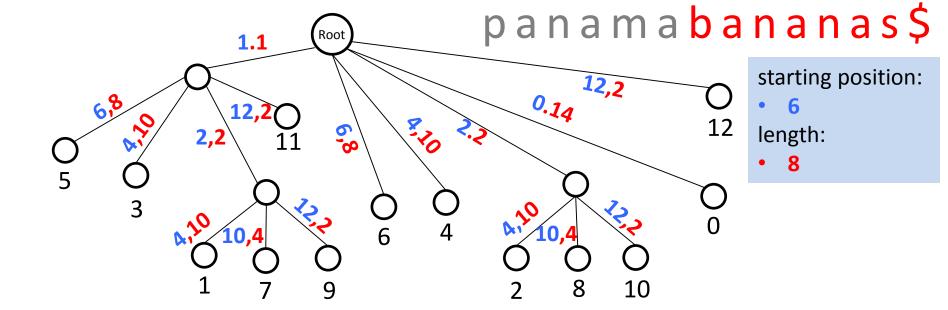
- # vertices < 2 | Text |
- memory footprint of the suffix tree: O(|Text|)
- Cheating!!! how do we store all edge labels?



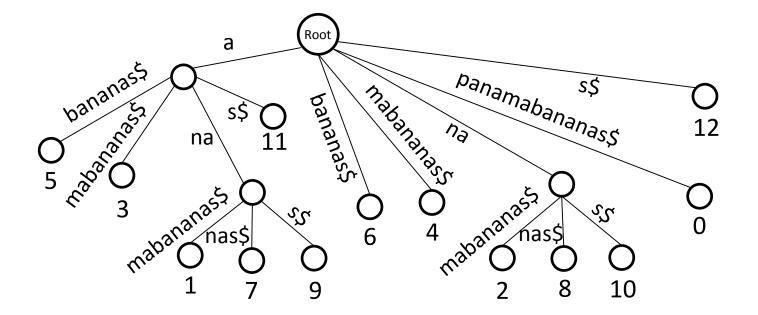
- # vertices < 2 | *Text* |
- memory footprint of the suffix tree: O(|Text|)
- Cheating!!! how do we store all edge labels?



- # vertices < 2 | *Text* |
- memory footprint of the suffix tree: O(|Text|)
- storing edge labels



- # vertices < 2 | *Text* |
- memory footprint of the suffix tree: O(|Text|)
- storing edge labels



Why did we bother to add "\$" to "panamabananas"?

- to make sure that each suffix corresponds to a leaf Why do we want to make sure that each suffix correspond to a leaf?
- construct suffix tree for "papa" (without adding "\$")
 and compare it with the suffix tree for "papa\$"

Constructing Suffix Tree: Naive Approach

• Quadratic runtime:

 $- O(|Text|^2)$

O(|Genome| + |Patterns|) to find pattern matches

Constructing Suffix Tree: Linear-Time Algorithm

• Linear runtime (for a constant-size alphabet):

- O(|*Text*|)



Linear-time algorithm (Weiner, 1973) was simplified by Ukkonen (1995) but it is still too complex to cover in this course

Big Secret of the Big O Notation

- Suffix trees enable fast Exact Multiple Pattern Matching:
 - Runtime: O(|Text| + |Patterns|)
 - Memory: O(|*Text*|)
- However, big-O notation hides constants!
 - suffix tree algorithms has large memory footprint
 20• | Text | for long texts like human genome
- We want to find mutations!
 - it is unclear how to develop fast Approximate
 Multiple Pattern Matching using suffix trees