Outline for Today

Review from Last Time

A quick refresher on tries.

Suffix Tries

• A simple data structure for string searching.

Suffix Trees

• A compact, powerful, and flexible data structure for string algorithms.

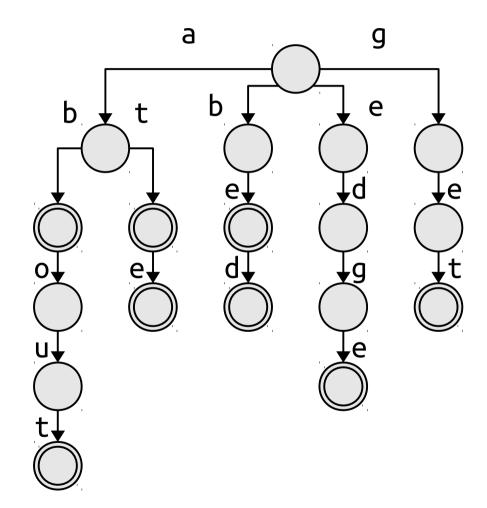
Generalized Suffix Trees

An even more flexible data structure.

Review from Last Time

Tries

- A *trie* is a tree that stores a collection of strings over some alphabet Σ .
- Each node corresponds to a prefix of some string in the set.
- Tries are sometimes called prefix trees, since each node in a trie corresponds to a prefix of one of the words in the trie.



Aho-Corasick String Matching

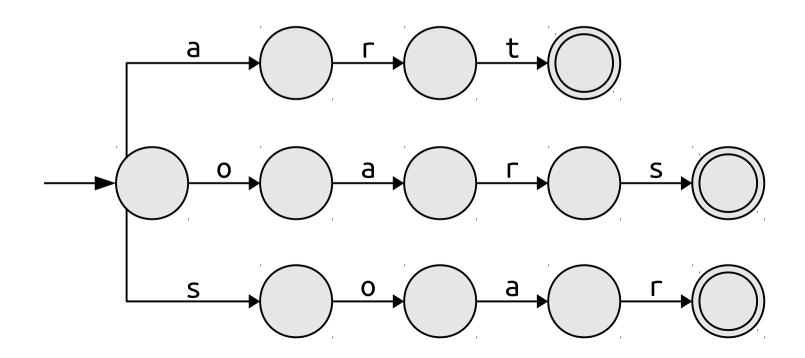
- The *Aho-Corasick string matching algorithm* is an algorithm for finding all occurrences of a set of strings $P_1, ..., P_k$ inside a string T.
- Runtime is $\langle O(n), O(m+z) \rangle$, where
 - m = |T|,
 - $n = |P_1| + ... + |P_k|$, and
 - z is the number of matches.
- Great for the case where the patterns are fixed and the text to search changes.

Genomics Databases

- Many string algorithms these days are developed for or used extensively in computational genomics.
- Typically, we have a *huge* database with many very large strings (genomes) that we'll preprocess to speed up future operations.
- *Common problem:* given a fixed string T to search and changing patterns $P_1, ..., P_k$, find all matches of those patterns in T.
- *Question:* Can we instead preprocess *T* to make it easy to search for variable patterns?

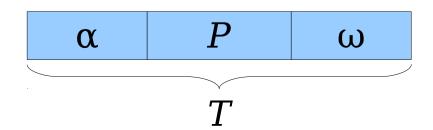
Substrings, Prefixes, and Suffixes

• **Useful Fact 1:** Given a trie storing a set of strings $S_1, S_2, ..., S_k$, it's possible to determine, in time O(|Q|), whether a query string Q is a prefix of any S_i .

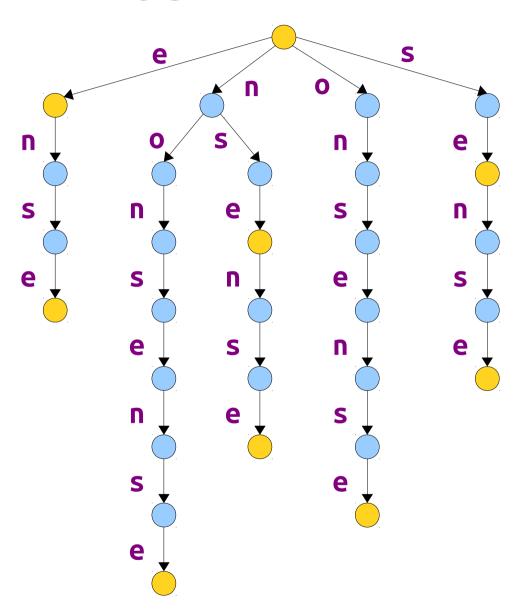


Substrings, Prefixes, and Suffixes

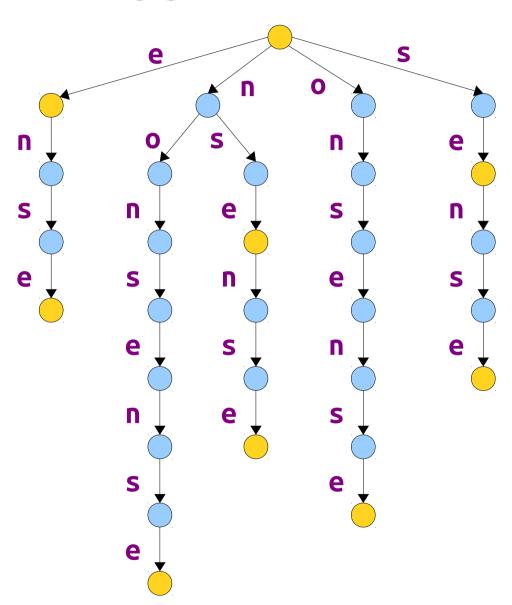
- Useful Fact 1: Given a trie storing a set of strings $S_1, S_2, ..., S_k$, it's possible to determine, in time O(|Q|), whether a query string Q is a prefix of any S_i .
- *Useful Fact 2:* A string *P* is a substring of a string *T* if and only if *P* is a prefix of some suffix of *T*.
 - Specifically, write $T = \alpha P \omega$; then T is a prefix of the suffix $P \omega$ of T.



- A *suffix trie* of *T* is a trie of all the suffixes of *T*.
- Given any pattern string *P*, we can check in time O(|*P*|) whether *P* is a substring of *T* by seeing whether *P* is a prefix in *T*'s suffix trie.
 - (Because that means that *P* is a prefix of a suffix of *T*.)

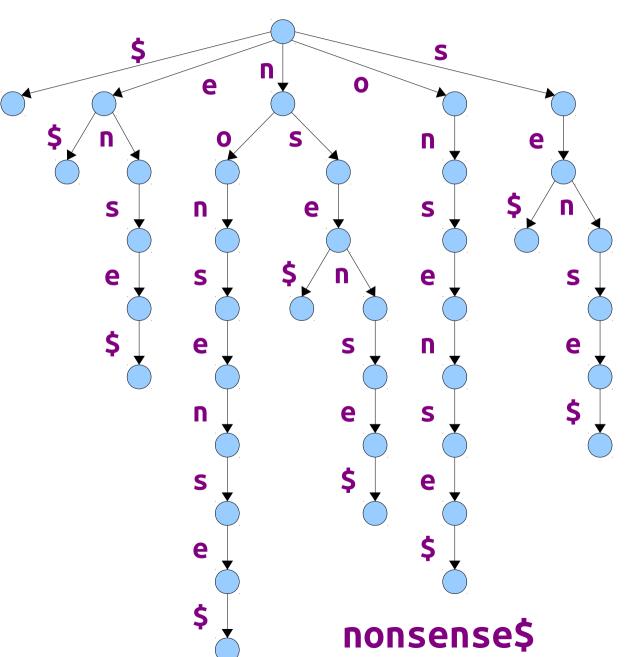


- A *suffix trie* of *T* is a trie of all the suffixes of *T*.
- More generally, given any nonempty patterns $P_1, ..., P_k$ of total length n, we can detect how many of those patterns are substrings of T in time O(n).
- (Finding all matches is a bit trickier; more on that later.)



A Typical Transform

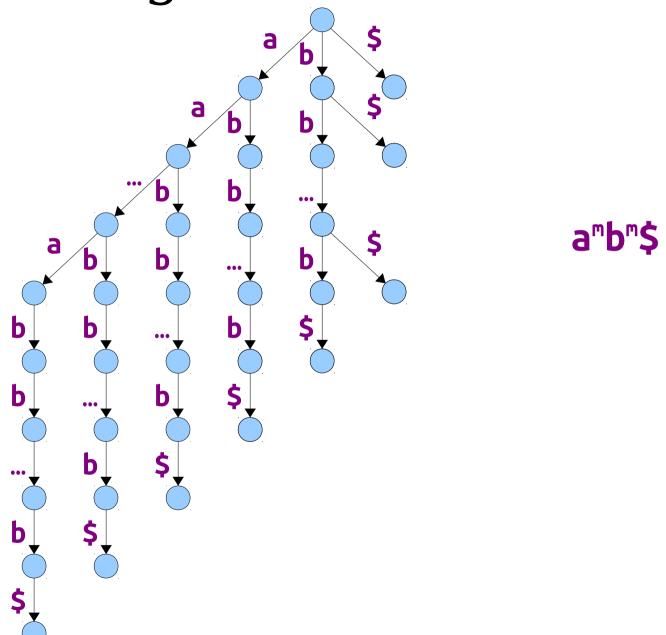
- Typically, we append some new character $\$ \notin \Sigma$ to the end of T, then construct the trie for T\$.
- Leaf nodes correspond to suffixes.
- Internal nodes correspond to prefixes of those suffixes.



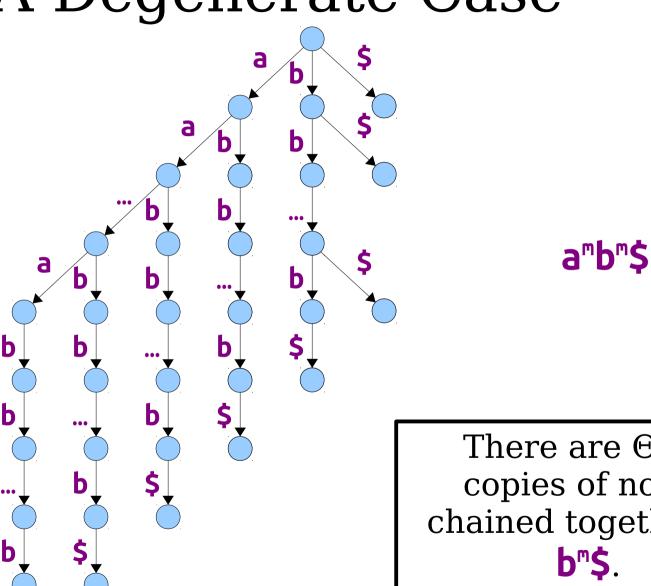
Constructing Suffix Tries

- Once we build a single suffix trie for string T, we can efficiently detect whether patterns match in time O(n).
- *Question:* How long does it take to construct a suffix trie?
- **Problem:** There's an $\Omega(m^2)$ lower bound on the worst-case complexity of *any* algorithm for building suffix tries.

A Degenerate Case

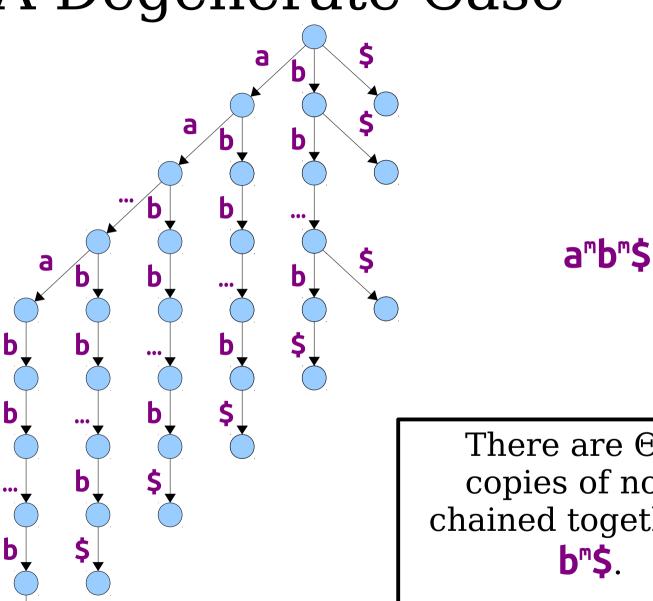


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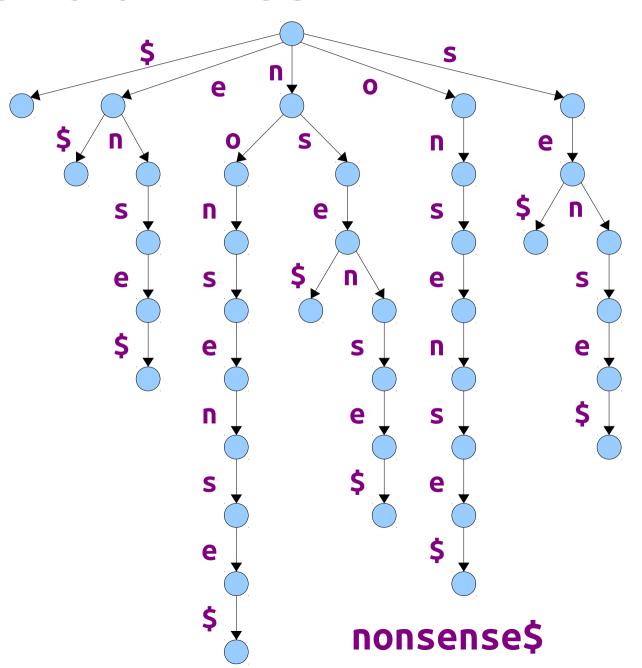
Space usage: $\Omega(m^2)$.

Correcting the Problem

- Because suffix tries may have $\Omega(m^2)$ nodes, all suffix trie algorithms must run in time $\Omega(m^2)$ in the worst-case.
- Can we reduce the number of nodes in the trie?

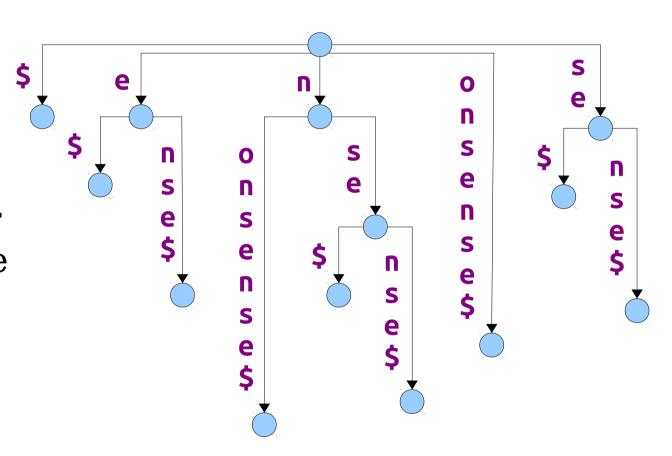
Patricia Tries

- A "silly" node in a trie is a node that has exactly one child.
- A *Patricia trie* (or *radix trie*) is a trie where all "silly" nodes are merged with their parents.

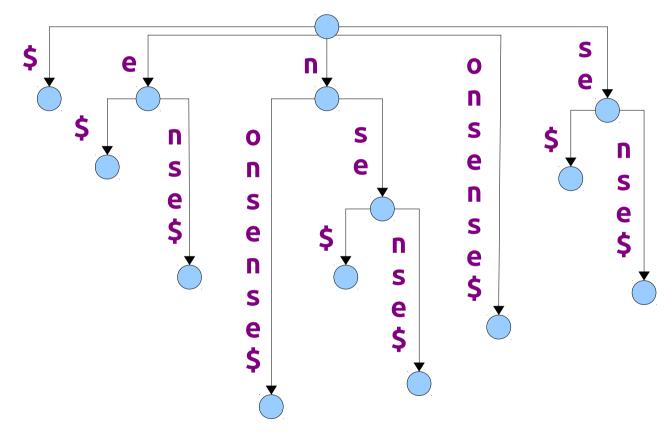


Patricia Tries

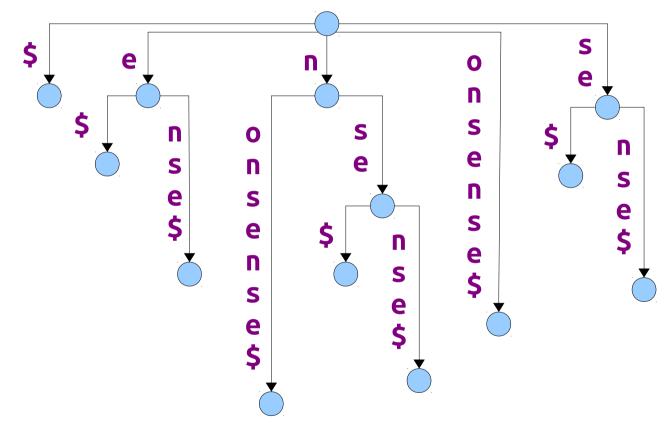
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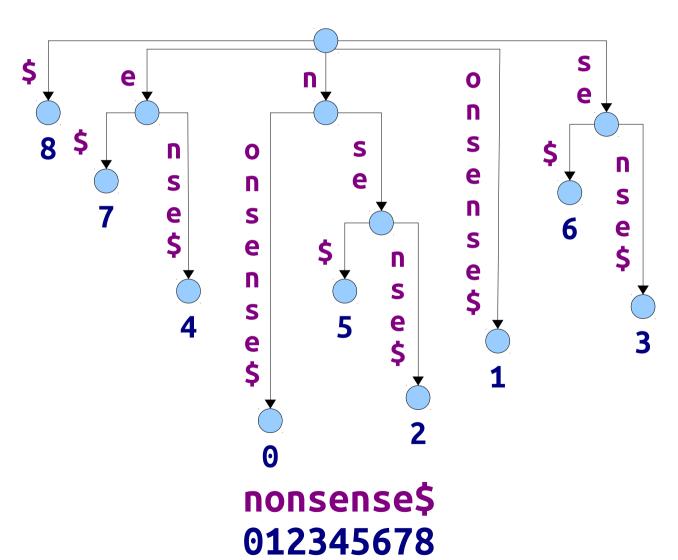
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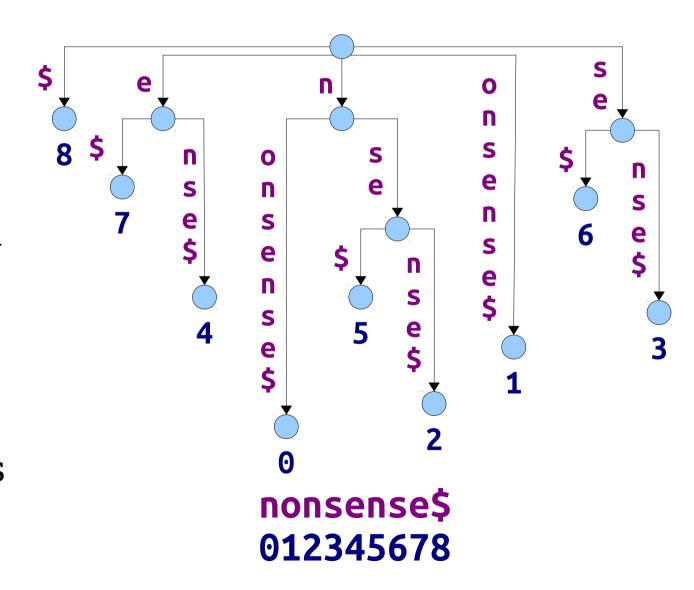


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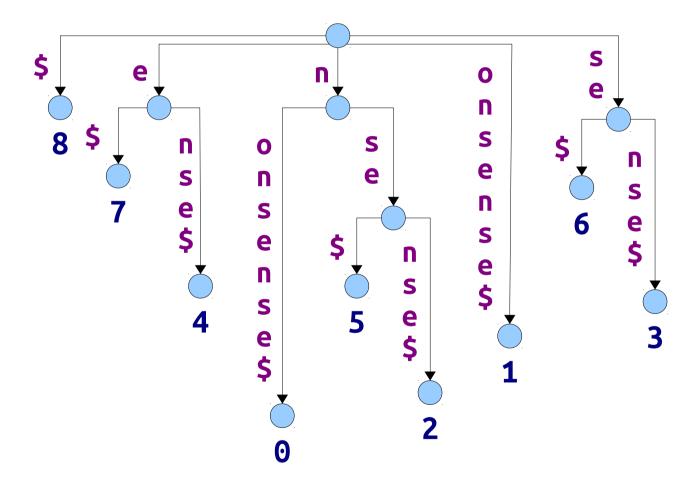


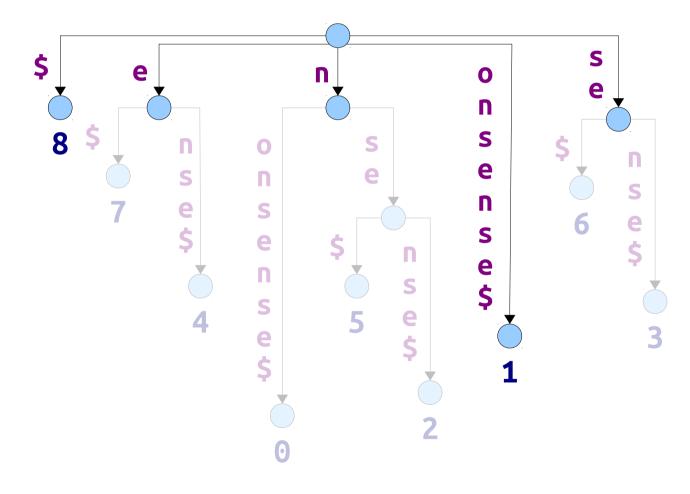
Properties of Suffix Trees

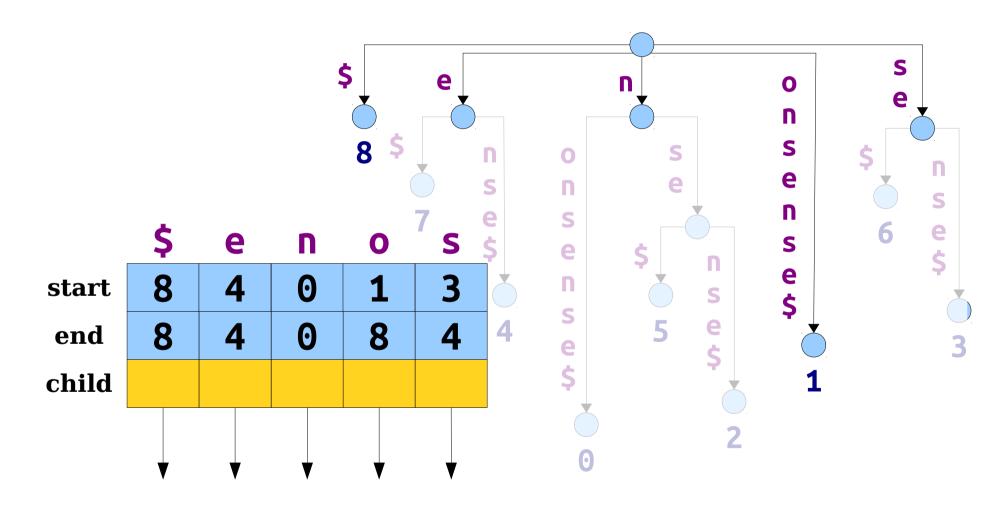
- If |T| = m, the suffix tree has exactly m + 1 leaf nodes.
- For any T ≠ ε, all internal nodes in the suffix tree have at least two children.
- Number of nodes in a suffix tree is $\Theta(m)$.



- Suffix trees may have $\Theta(m)$ nodes, but the labels on the edges can have size $\omega(1)$.
- This means that a naïve representation of a suffix tree may take $\omega(m)$ space.
- *Useful fact:* Each edge in a suffix tree is labeled with a consecutive range of characters from *w*.
- Trick: Represent each edge labeled with a string α as a pair of integers [start, end] representing where in the string α appears.





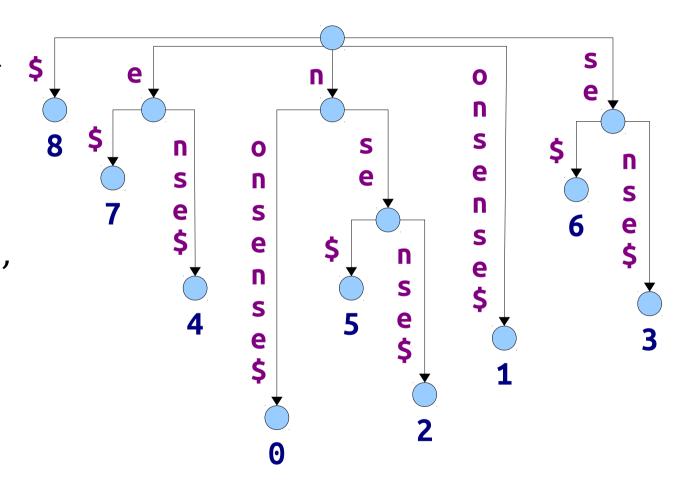


Building Suffix Trees

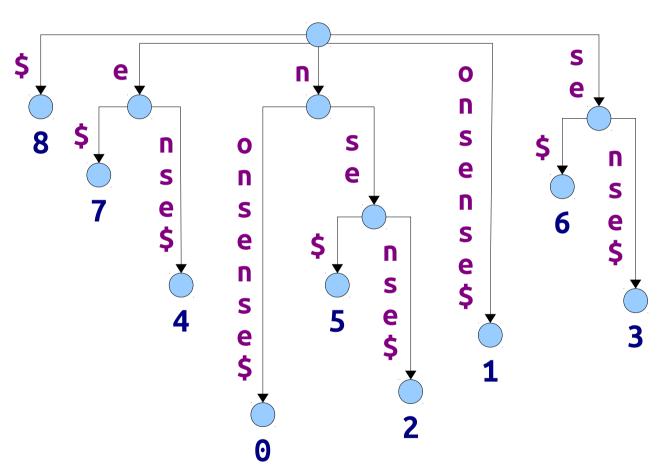
- Using this representation, suffix trees can be constructed using space $\Theta(m)$.
- *Claim:* There are $\Theta(m)$ -time algorithms for building suffix trees.
- These algorithms are not trivial! We'll discuss one of them next time.

Application: Multi-String Matching

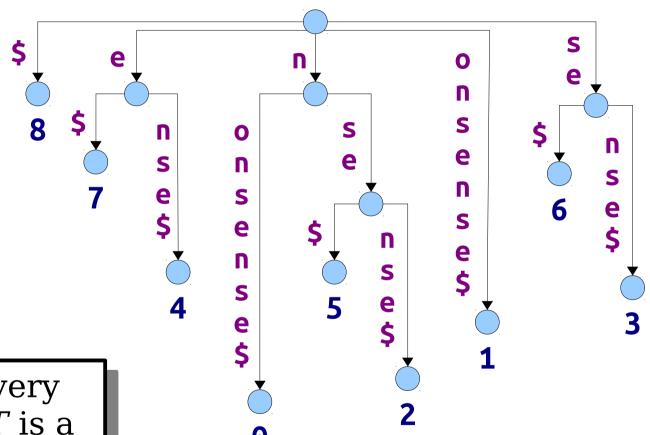
- Suppose we preprocess a string
 T by building a suffix tree for it.
- Given any pattern string P of length n, we can determine, in time O(n), whether n is a substring of P by looking it up in the suffix tree.



• Claim: After spending O(m) time preprocessing T\$, can find all matches of a string P in time O(n + z), where z is the number of matches.



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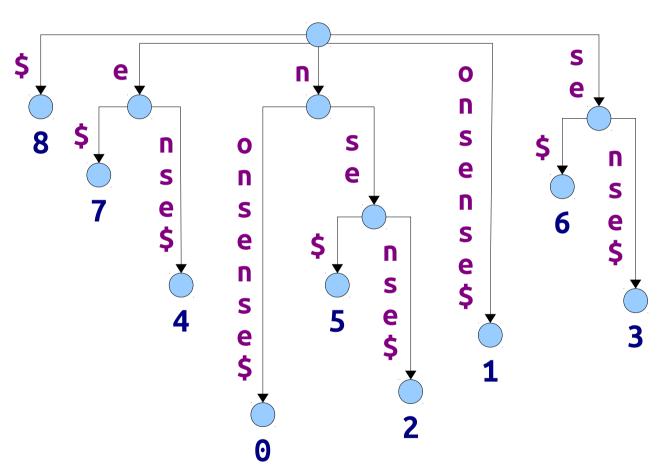


Observation 1: Every occurrence of P in T is a prefix of some suffix of T.

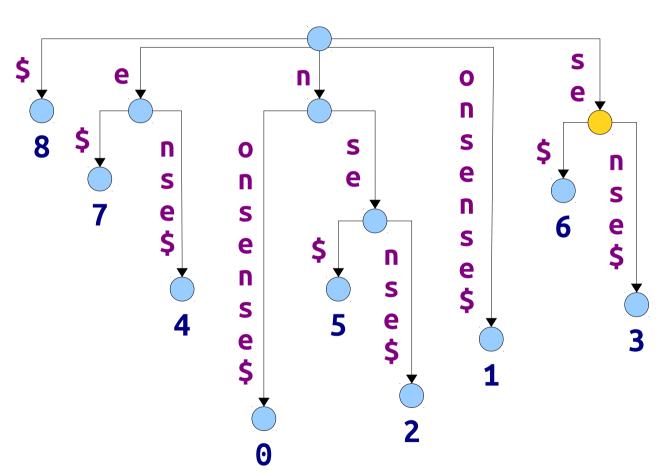
• Claim: After spending O(m) time preprocessing T\$, can find all matches of a string P in time O(n + z), where z is the number of matches.

Observation 2: Every suffix of T\$ beginning with some pattern P appears in the subtree found by searching for P.

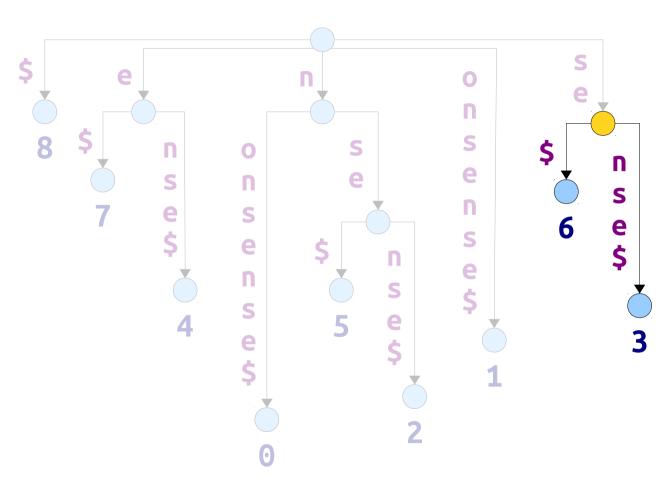
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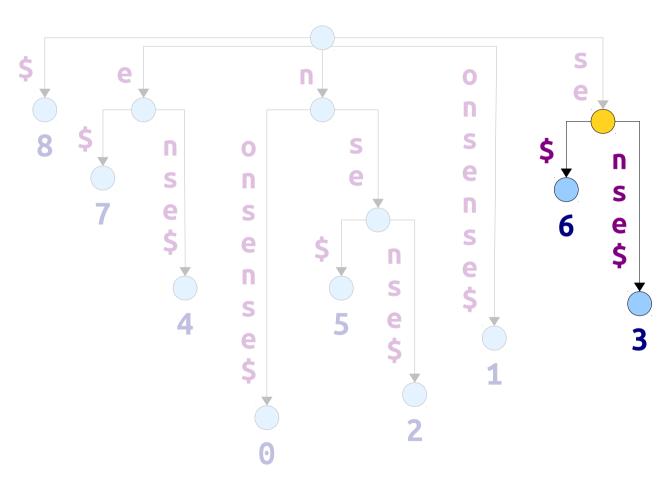
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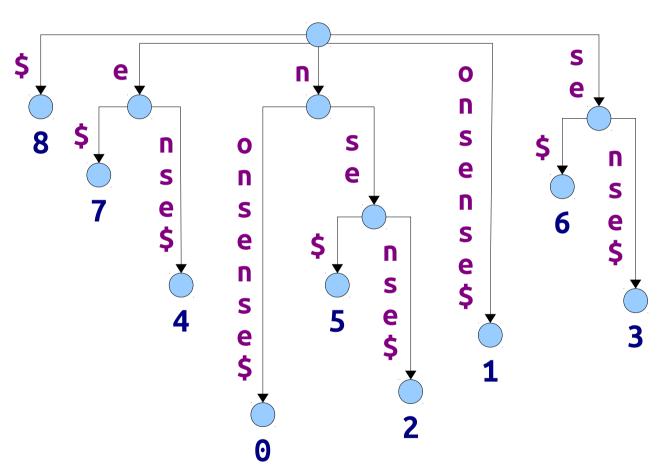


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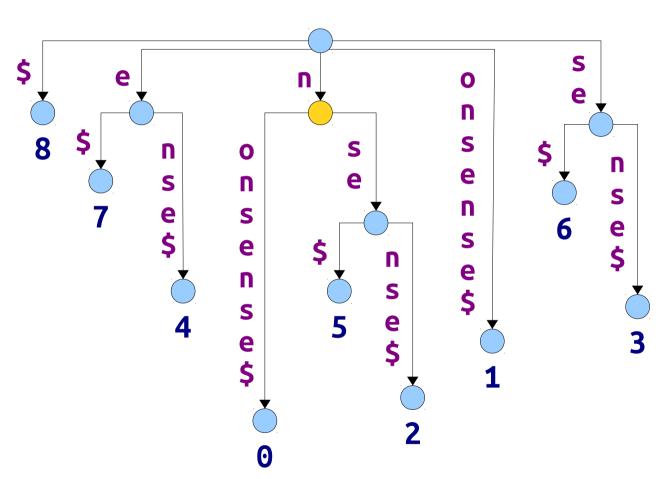


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012345678

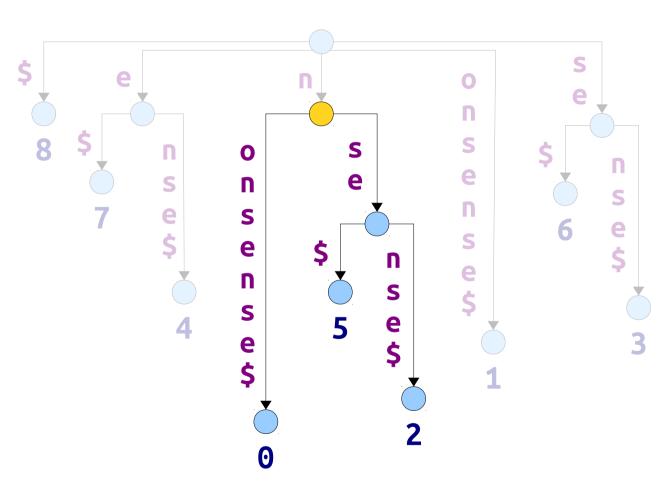
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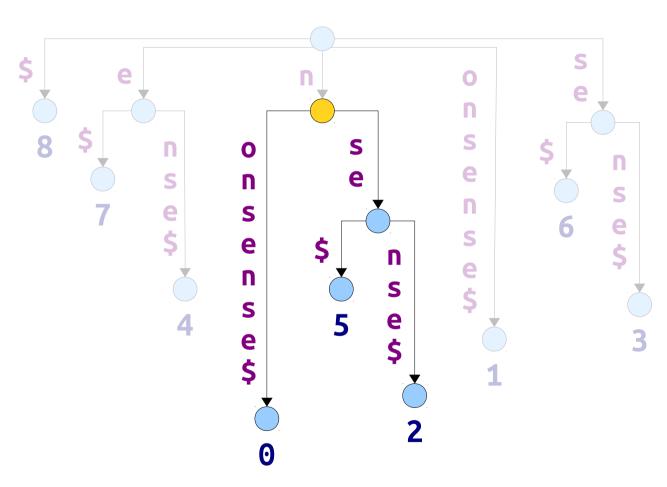
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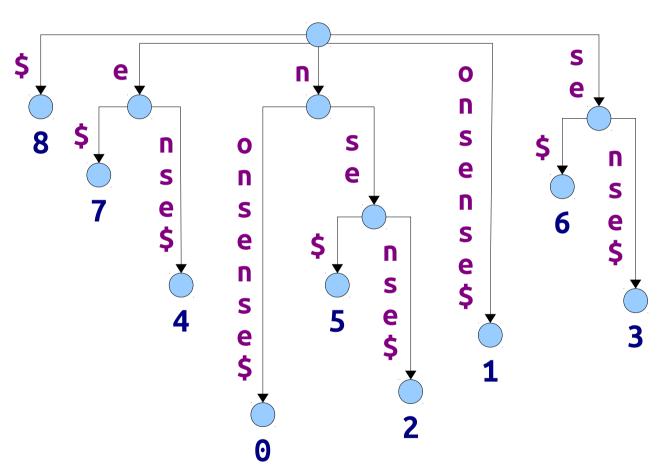


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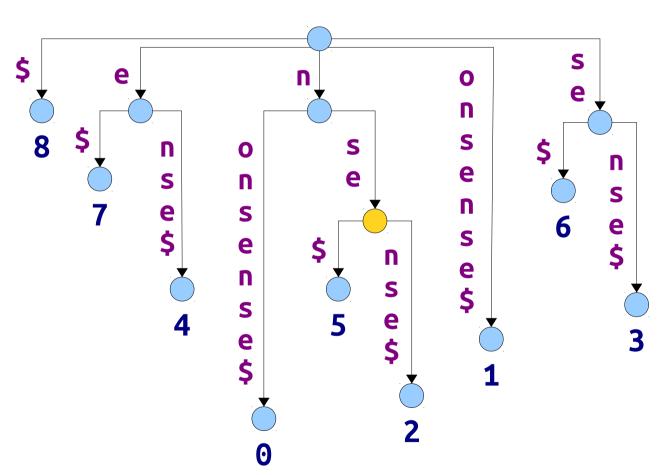


<u>nonsense\$</u> 012345678

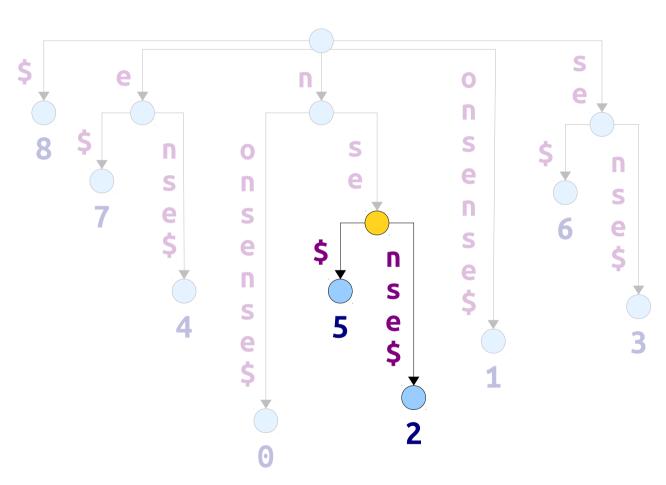
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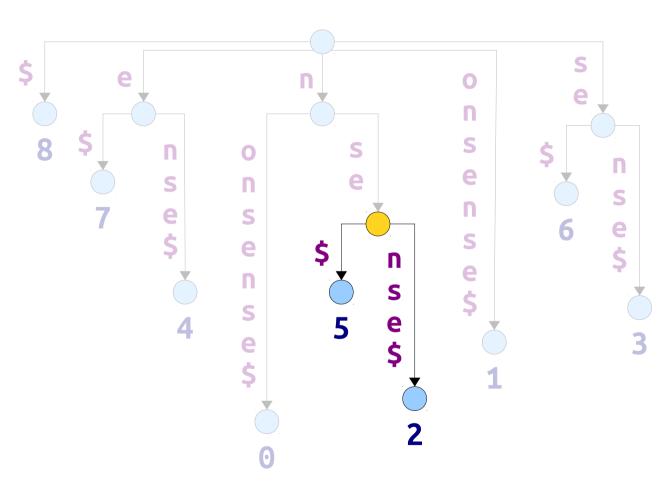
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Finding All Matches

- To find all matches of string *P*, start by searching the tree for *P*.
- If the search falls off the tree, report no matches.
- Otherwise, let *v* be the node at which the search stops, or the endpoint of the edge where it stops if it ends in the middle of an edge.
- Do a DFS and report the numbers of all the leaves found in this subtree. The indices reported this way give back all positions at which *P* occurs.

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• Do a DFS and report the numbers of all the leaves found in this subtree. The indices reported this way give back all positions at which *P* occurs.

How fast is this step?

Claim: The DFS to find all leaves in the subtree corresponding to prefix P takes time O(z), where z is the number of matches.

Proof: If the DFS reports z matches, it must have visited z different leaf nodes.

Since each internal node of a suffix tree has at least two children, the total number of internal nodes visited during the DFS is at most z - 1.

During the DFS, we don't need to actually match the characters on the edges. We just follow the edges, which takes time O(1).

Therefore, the DFS visits at most O(z) nodes and edges and spends O(1) time per node or edge, so the total runtime is O(z).

Reverse Aho-Corasick

- Given patterns P_1 , ... P_k of total length n, suffix trees can find all matches of those patterns in time O(m + n + z).
 - Search for all matches of each P_i ; total time across all searches is O(n + z).
- Acts as a "reverse" Aho-Corasick:
 - Aho-Corasick string matching runs in time $\langle O(n), O(m+z) \rangle$
 - Suffix tree string matching runs in time $\langle O(m), O(n+z) \rangle$

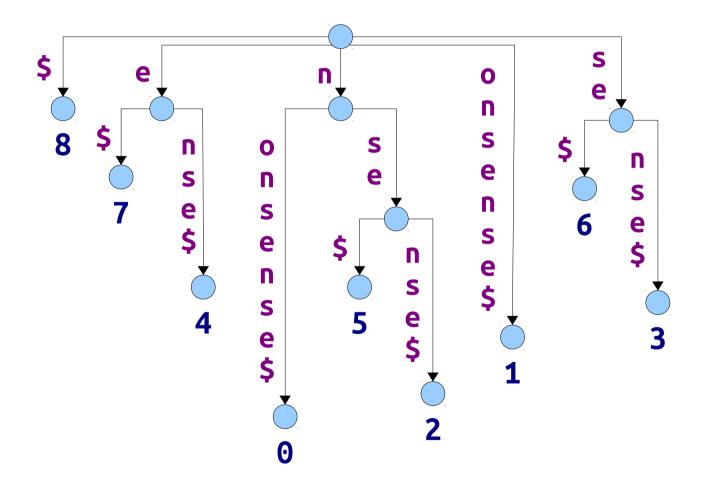
Another Application:

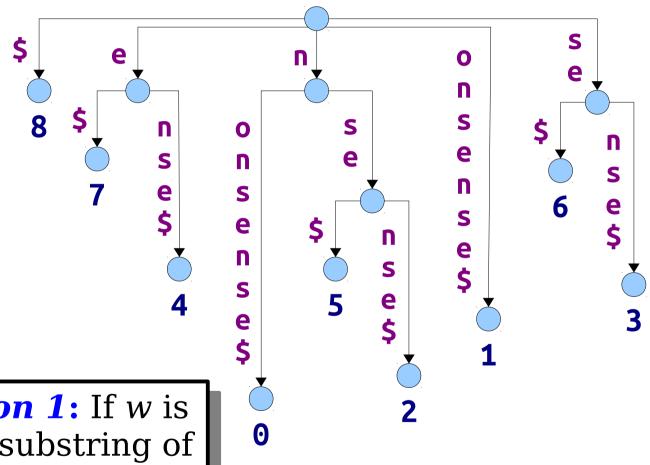
Longest Repeated Substring

Consider the following problem:

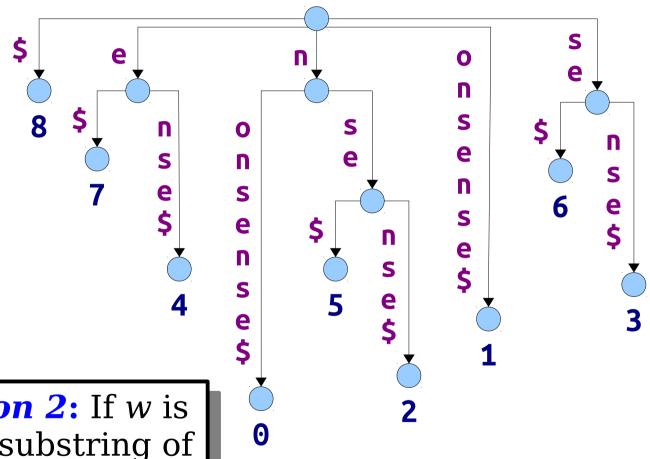
Given a string T, find the longest substring w of T that appears in at least two different positions.

 Applications to computational biology: more than half of the human genome is formed from repeated DNA sequences!

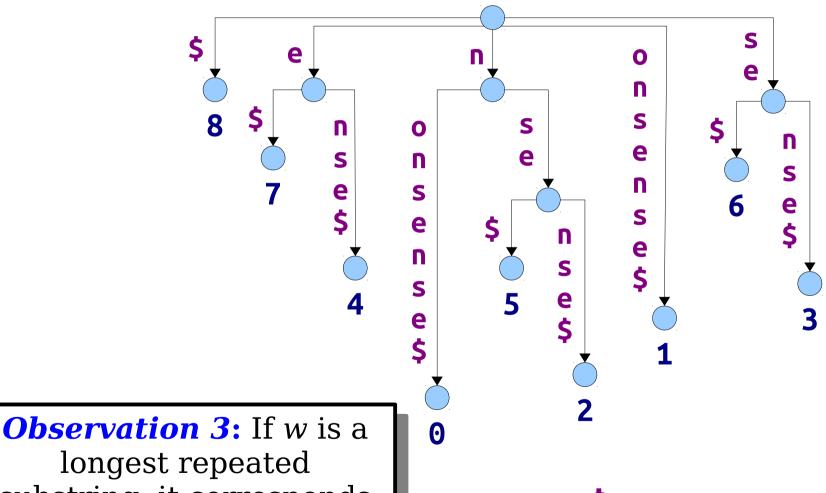




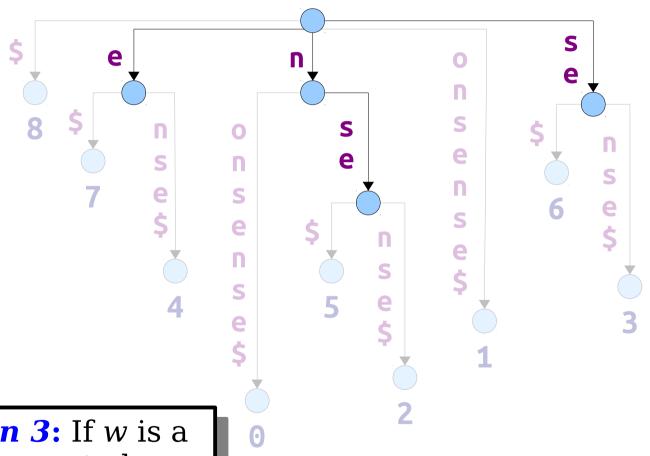
Observation 1: If w is a repeated substring of T, it must be a prefix of at least two different suffixes.



Observation 2: If w is a repeated substring of *T*, it must correspond to a prefix of a path to an internal node.



longest repeated substring, it corresponds to a full path to an internal node.



Observation 3: If w is a longest repeated substring, it corresponds to a full path to an internal node.

- For each node v in a suffix tree, let s(v) be the string that it corresponds to.
- The *string depth* of a node v is defined as |s(v)|, the length of the string v corresponds to.
- The longest repeated substring in *T* can be found by finding the internal node in *T* with the maximum string depth.

- Here's an O(m)-time algorithm for solving the longest repeated substring problem:
 - Build the suffix tree for T in time O(m).
 - Run a DFS over *T*, tracking the string depth as you go, to find the internal node of maximum string depth.
 - Recover the string T corresponds to.
- *Good exercise:* How might you find the longest substring of *T* that repeats at least *k* times?

Challenge Problem:

Solve this problem in linear time without using suffix trees (or suffix arrays).

Time-Out for Announcements!

Problem Set One

- Problem Set One was due today at 3:00PM.
 - Want to use your late days? Submit by Saturday at 3:00PM.
- Solutions will go out on Tuesday.
- Problem Set Two goes out on Tuesday have a good weekend!

Talk Today

- Jon Kleinberg (who authored *Algorithm Design* along with Eva Tardos) is giving a talk today at 4:15PM in the Mackenzie Boardroom.
- Focus is on algorithms for solving problems with agents who don't plan rationally.
- Sounds really fun hopefully we'll finish with a little buffer time.

Back to CS166!

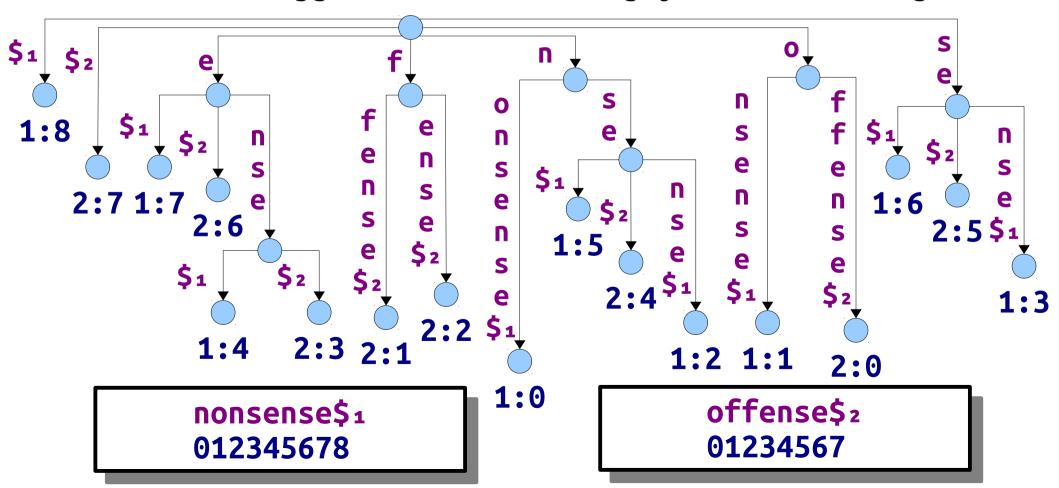
Generalized Suffix Trees

Suffix Trees for Multiple Strings

- Suffix trees store information about a single string and exports a huge amount of structural information about that string.
- However, many applications require information about the structure of multiple different strings.

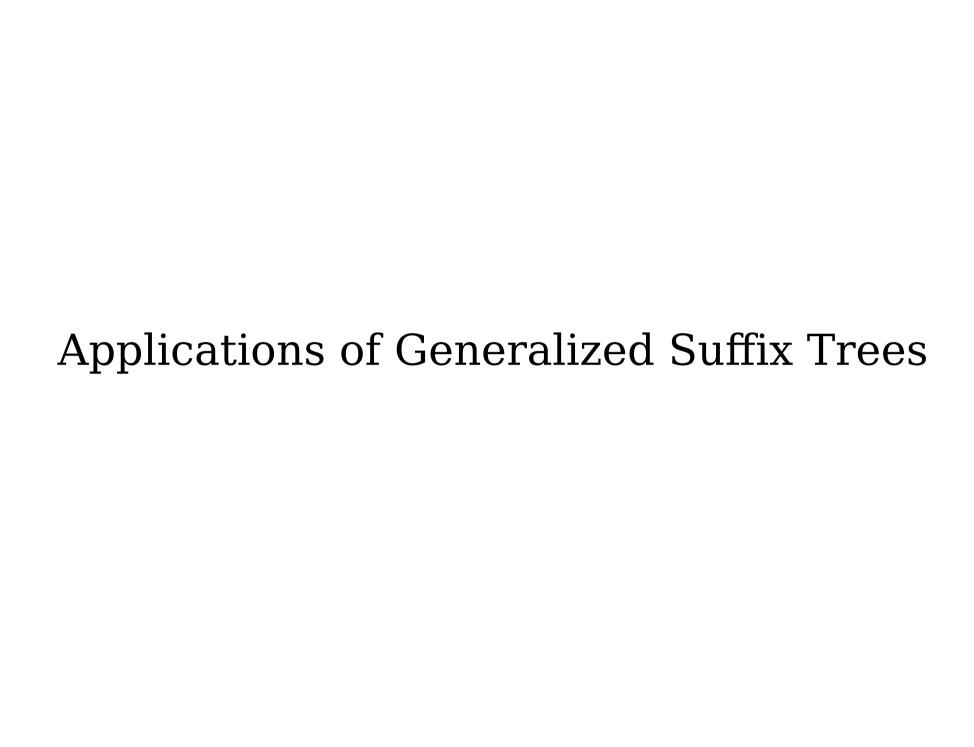
Generalized Suffix Trees

- A *generalized suffix tree* for $T_1, ..., T_k$ is a Patricia trie of all suffixes of T_1 \$1, ..., T_k \$k. Each T_i has a unique end marker.
- Leaves are tagged with i:j, meaning "jth suffix of string T_i "



Generalized Suffix Trees

- *Claim:* A generalized suffix tree for strings $T_1, ..., T_k$ of total length m can be constructed in time $\Theta(m)$.
- Use a two-phase algorithm:
 - Construct a suffix tree for the single string T_1 \$1 T_2 \$2 ... T_k \$k in time $\Theta(m)$.
 - This will end up with some invalid suffixes.
 - Do a DFS over the suffix tree and prune the invalid suffixes.
 - Runs in time O(m) if implemented intelligently.



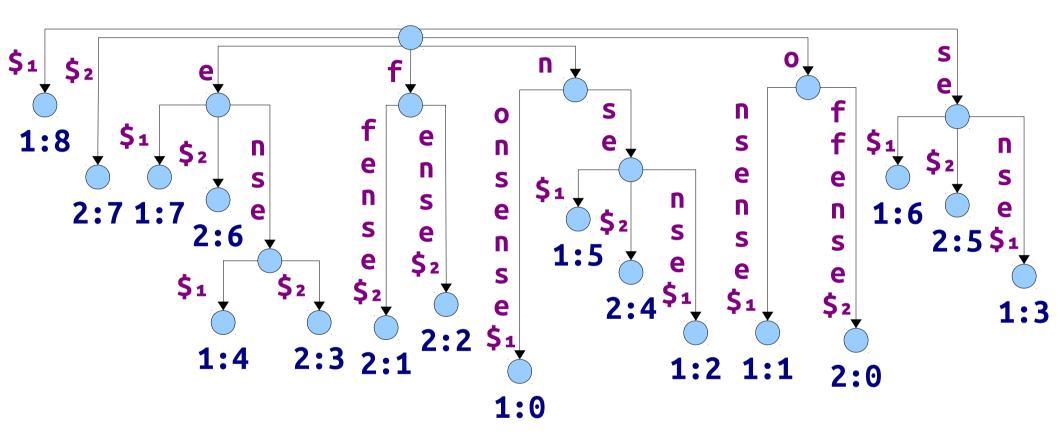
Longest Common Substring

Consider the following problem:

Given two strings T_1 and T_2 , find the longest string w that is a substring of both T_1 and T_2 .

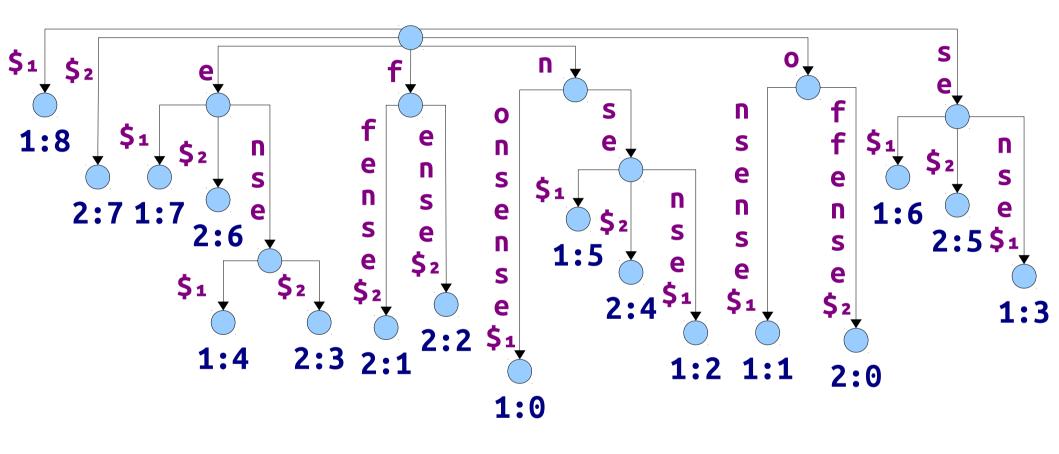
- Can solve in time $O(|T_1| \cdot |T_2|)$ using dynamic programming.
- Can we do better?

Longest Common Substring



nonsense\$1 012345678 offense\$2 01234567

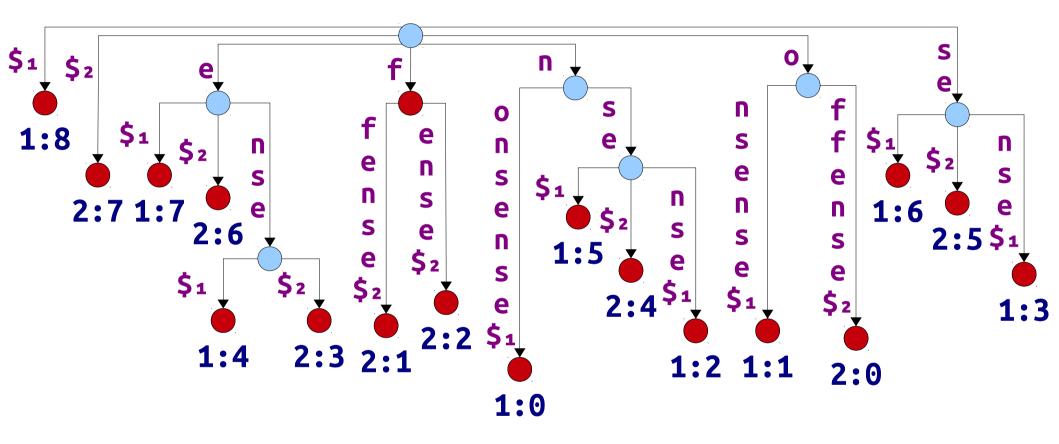
Longest Common Substring



nonsense\$1 012345678 **Observation:** Any common substring of T_1 and T_2 will be a prefix of a suffix of T_1 and a prefix of a suffix of T_2 .

se\$₂ ·567

Longest Common Substring



nonsense\$1 012345678

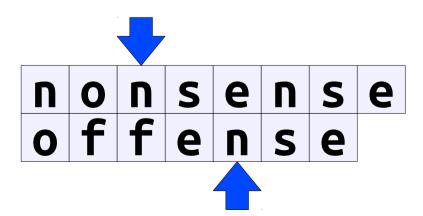
Longest Common Substring

- Build a generalized suffix tree for T_1 and T_2 in time O(m).
- Annotate each internal node in the tree with whether that node has at least one leaf node from each of T_1 and T_2 .
 - Takes time O(m) using DFS.
- Run a DFS over the tree to find the marked node with the highest string depth.
 - Takes time O(m) using DFS
- Overall time: O(m).

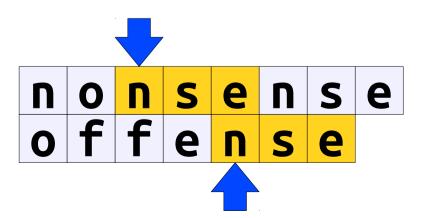
- Given two strings T_1 and T_2 and start positions i and j, the **longest common extension** of T_1 and T_2 , starting at positions i and j, is the length of the longest string w that appears at position i in T_1 and position j in T_2 .
- We'll denote this value by $LCE_{T_1, T_2}(i, j)$.
- Typically, T_1 and T_2 are fixed and multiple (i, j) queries are specified.

n	0	n	S	e	n	S	e
0	f	f	e	n	S	e	

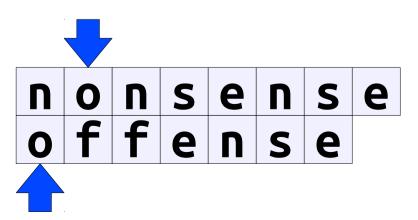
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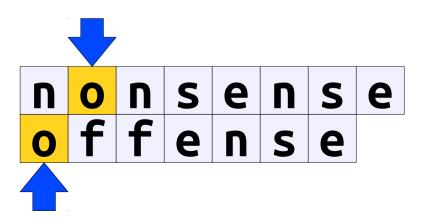
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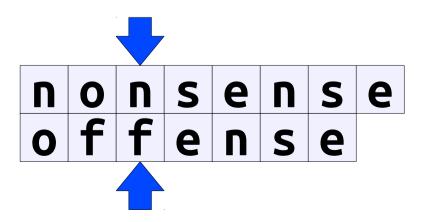
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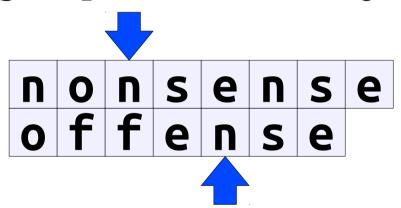
- Given two strings T_1 and T_2 and start positions i and j, the **longest common extension** of T_1 and T_2 , starting at positions i and j, is the length of the longest string w that appears at position i in T_1 and position j in T_2 .
- We'll denote this value by $LCE_{T_1, T_2}(i, j)$.
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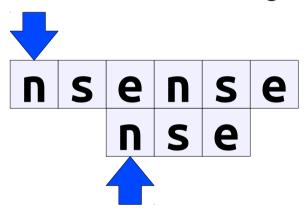
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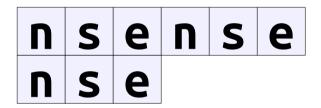
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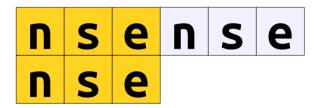
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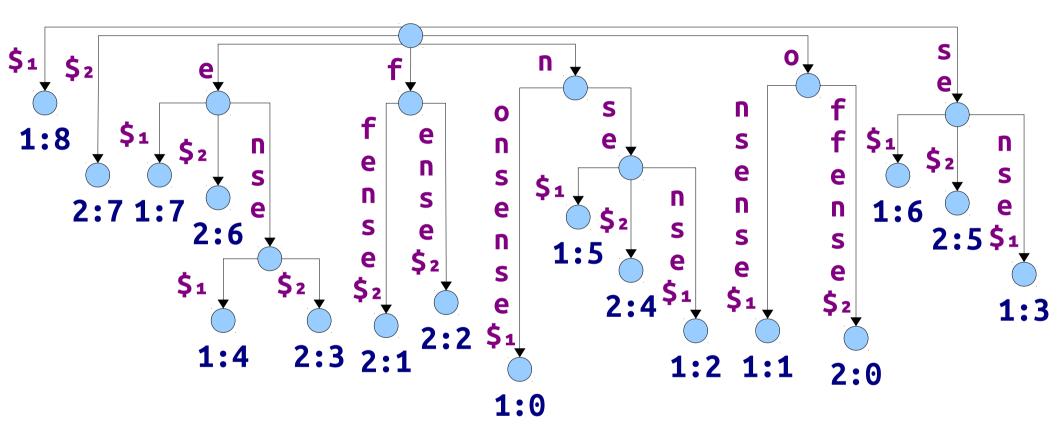


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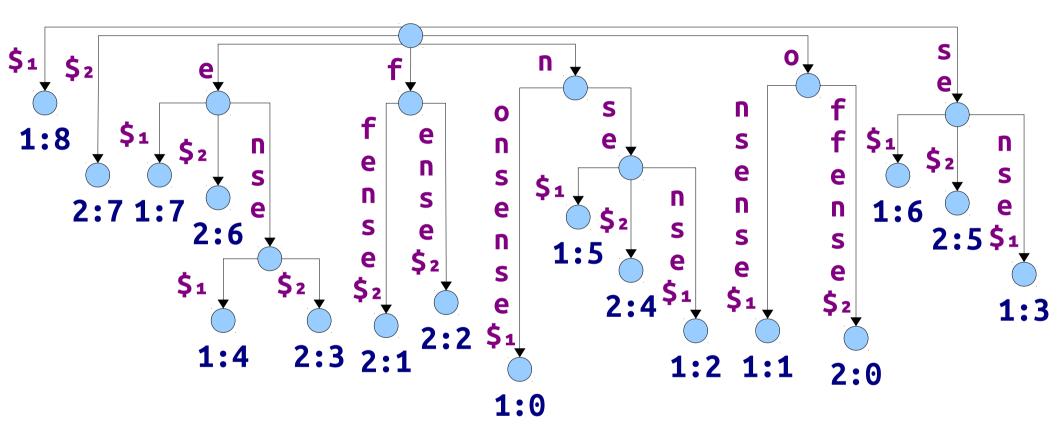


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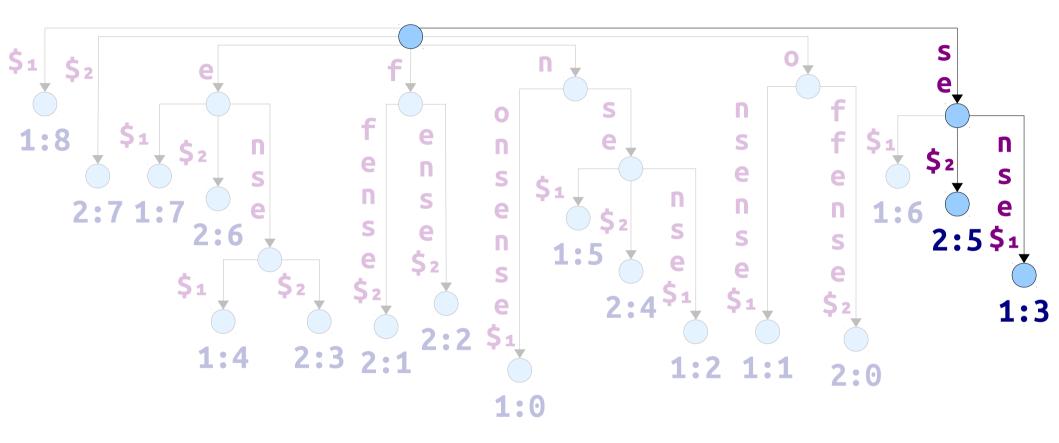




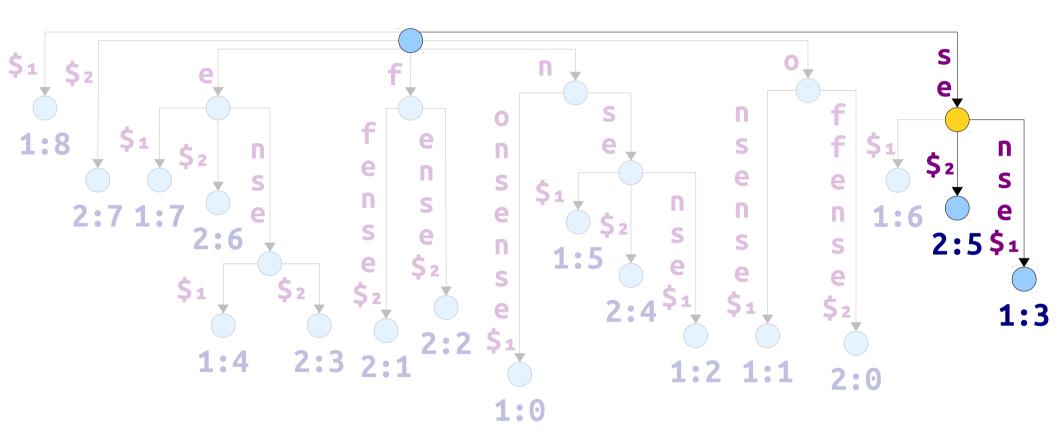
nonsense\$1 012345678



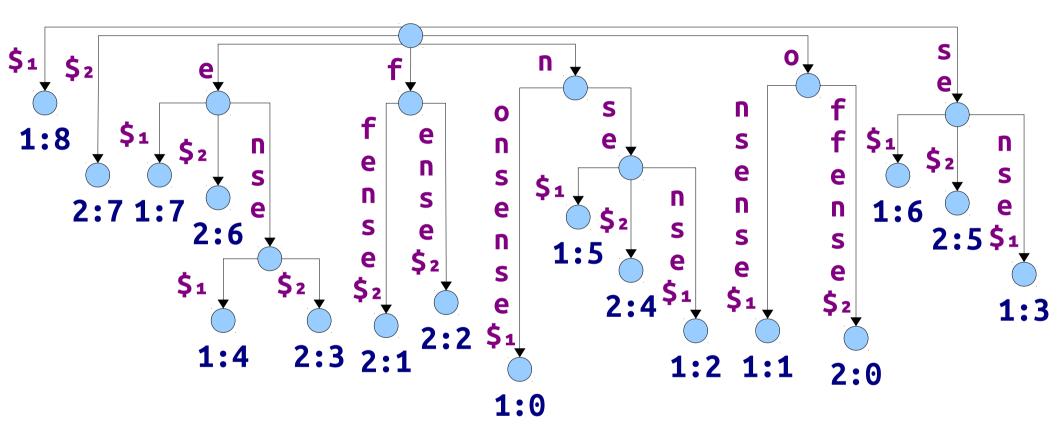
nonsense\$1 012345678



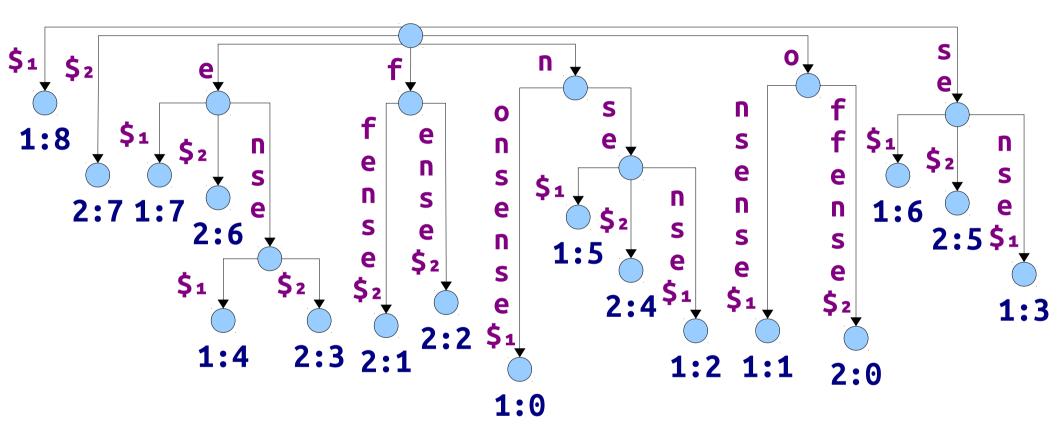
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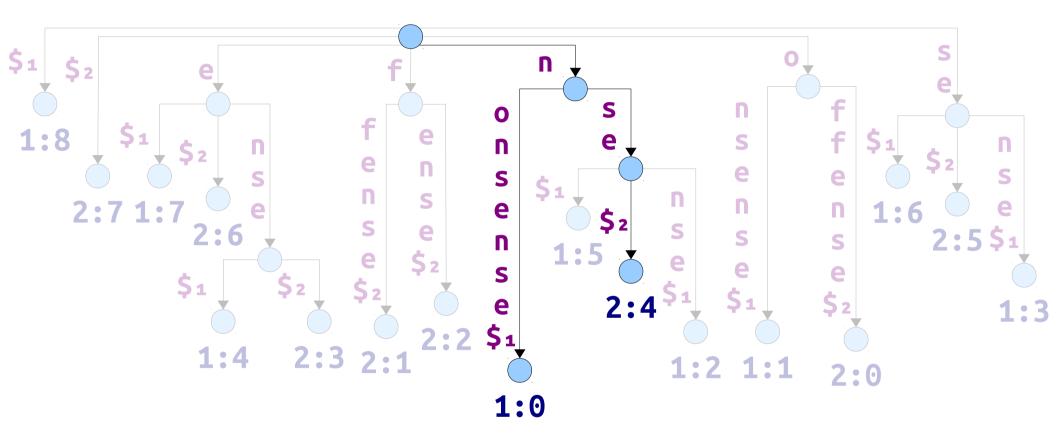
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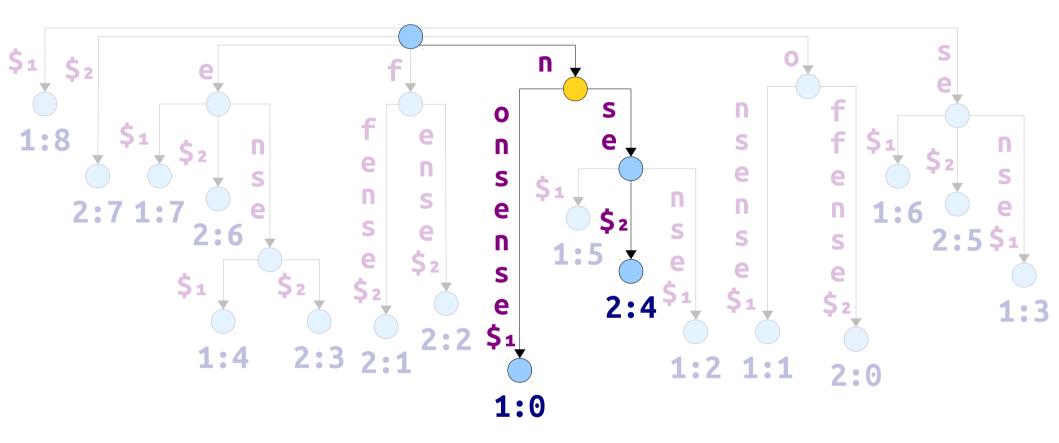
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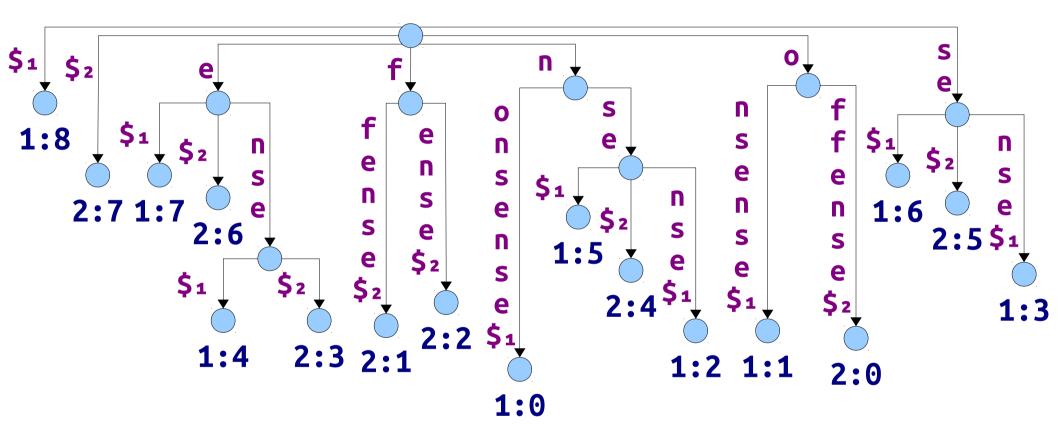


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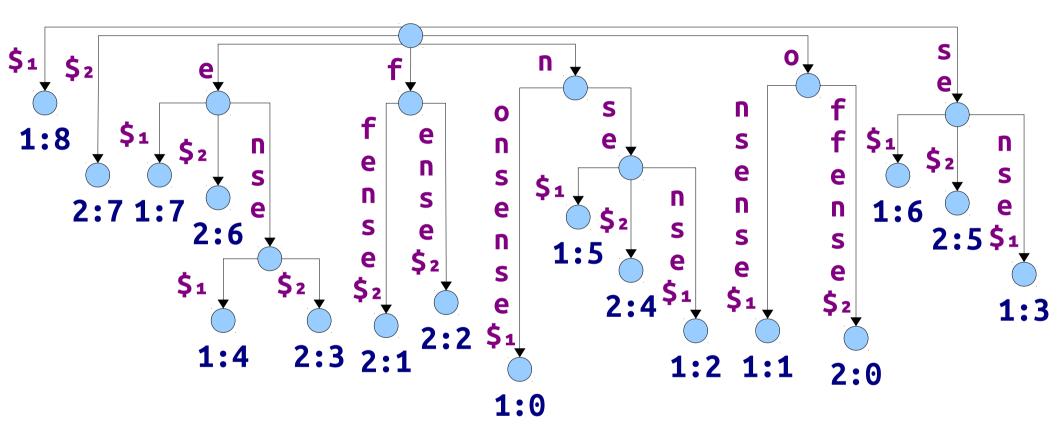
- *Notation:* Let S[i:] denote the suffix of string S starting at position i.
- Claim: LCE_{T_1, T_2}(i, j) is given by the string label of the LCA of $T_1[i:]$ and $T_2[j:]$ in the generalized suffix tree of T_1 and T_2 .
- And hey... don't we have a way of computing these in time O(1)?

Computing LCE's

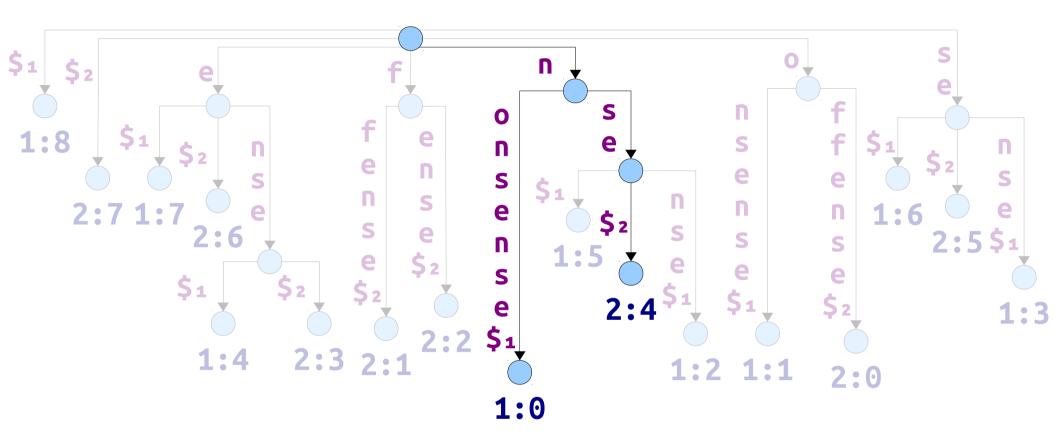
- Given two strings T_1 and T_2 , construct a generalized suffix tree for T_1 and T_2 in time O(m).
- Construct an LCA data structure for the generalized suffix tree in time O(m).
 - Use Fischer-Heun plus an Euler tour of the nodes in the tree.
- Can now query for the node representing the LCE in time O(1).



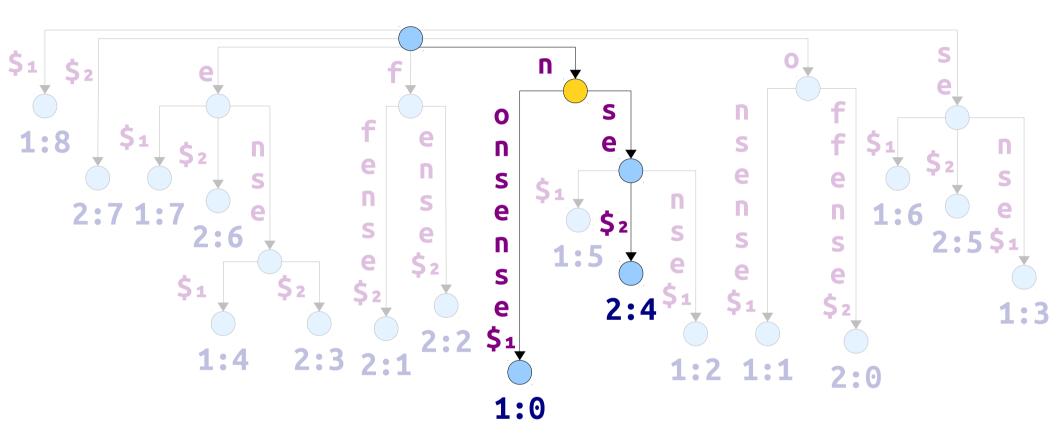
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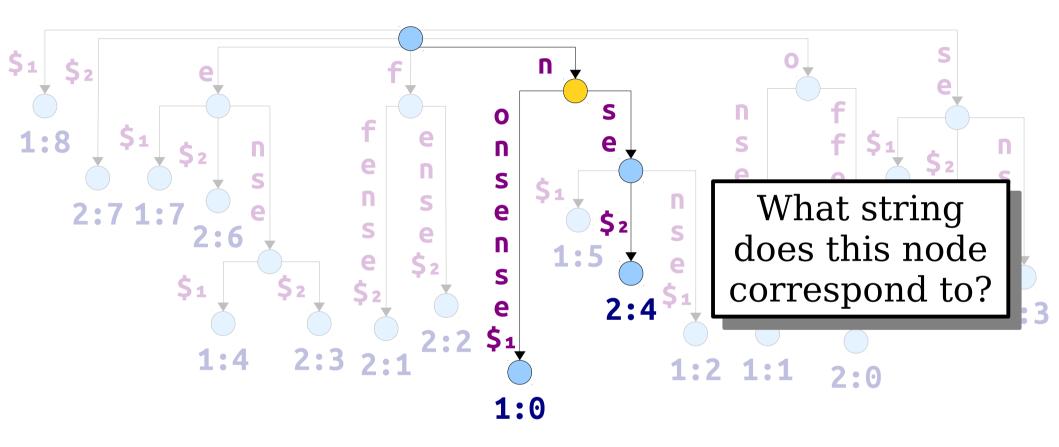
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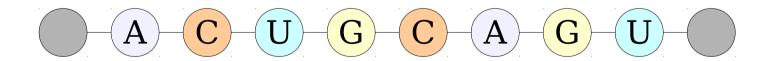
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The Overall Construction

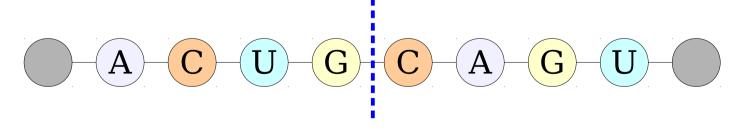
- Using an O(m)-time DFS, annotate each node in the suffix tree with its string depth.
- To compute LCE:
 - Find the leaves corresponding to $T_1[i:]$ and $T_2[j:]$.
 - Find their LCA; let its string depth be *d*.
 - Report $T_1[i:i+d-1]$ or $T_2[j:j+d-1]$.
- Overall, requires O(m) preprocessing time to support O(1) query time.

An Application: Longest Palindromic Substring

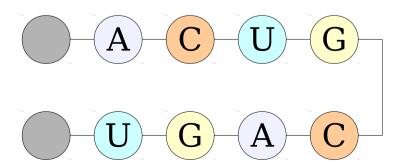
- A palindrome is a string that's the same forwards and backwards.
- A *palindromic substring* of a string *T* is a substring of *T* that's a palindrome.
- Surprisingly, of great importance in computational biology.



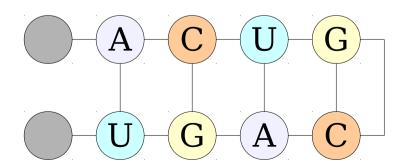
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Longest Palindromic Substring

• The *longest palindromic substring* problem is the following:

Given a string *T*, find the longest substring of *T* that is a palindrome.

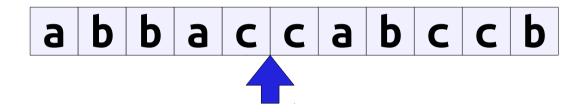
How might we solve this problem?

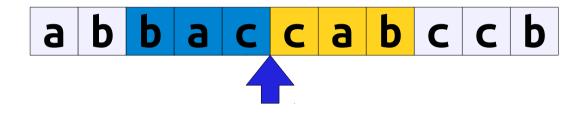
An Initial Idea

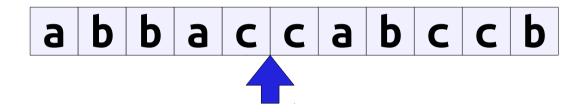
- To deal with the issues of strings going forwards and backwards, start off by forming T and T^R , the reverse of T.
- *Initial Idea*: Find the longest common substring of T and T^R .
- Unfortunately, this doesn't work:
 - T = abcdabaadbcabb
 - $T^{\mathbb{R}} = \mathsf{bbabcdaabadcba}$
 - Longest common substring: abcda
 - Longest palindromic substring: aa

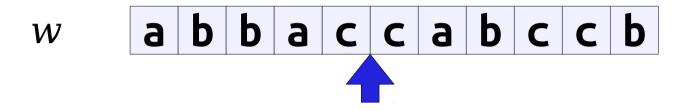
- For now, let's focus on even-length palindromes.
- An even-length palindrome substring ww^R of a string T has a center and radius:
 - *Center:* The spot between the duplicated center character.
 - *Radius:* The length of the string going out in each direction.
- *Idea*: For each center, find the largest corresponding radius.

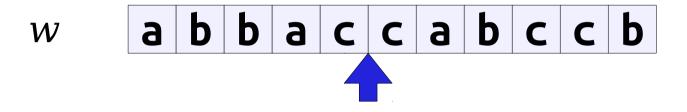
a b b a c c a b c c b



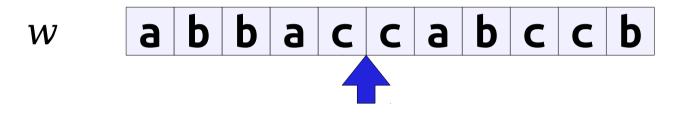


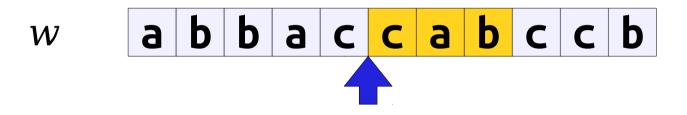






$$W^R$$
 b c c b a c c a b b a



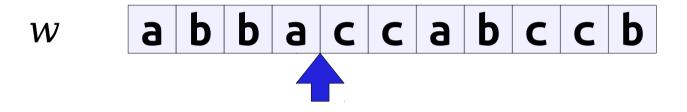




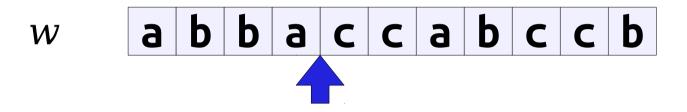
$$W^R$$
 b c c b a c c a b b a







$$W^R$$
 b c c b a c c a b b a



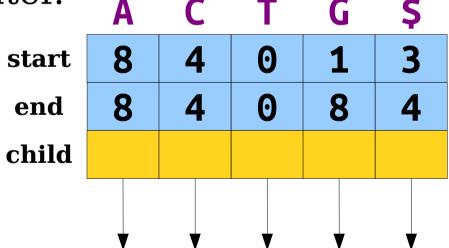
An Algorithm

- In time O(m), construct T^R .
- Preprocess T and T^R in time O(m) to support LCE queries.
- For each spot between two characters in T, find the longest palindrome centered at that location by executing LCE queries on the corresponding locations in T and T^R.
 - Each query takes time O(1) if it just reports the length.
 - Total time: O(m).
- Report the longest string found this way.
- Total time: O(m).

Suffix Trees: The Catch

Space Usage

- Suffix trees are memory hogs.
- Suppose $\Sigma = \{A, C, G, T, \$\}$.
- Each internal node needs 15 machine words: for each character, words for the start/end index and a child pointer.



This is still O(m), but it's a *huge* hidden constant!

Combating Space Usage

- In 1990, Udi Manber and Gene Myers introduced the *suffix array* as a space-efficient alternative to suffix trees.
- Requires one word per character; typically, an extra word is stored as well (details next Tuesday)
- Can't support all operations permitted by suffix trees, but has much better performance.
- Curious? Details are next time!

Summary

- Given a string, it's possible to build a suffix tree for it in time $\Theta(m)$. Suffix trees support
 - efficient detection of all matching substrings, efficient detection of duplicated substrings, efficient detection of common substrings, efficient detection of common extensions,
 - and a lot more!
- Suffix trees use space $\Theta(m)$, but with a huge hidden constant factor.
- Building suffix trees is hard. We'll see how to do it next time.

Next Time

Suffix Arrays

A space-efficient alternative to suffix trees.

LCP Arrays

 A useful auxiliary data structure for speeding up suffix arrays.

Constructing Suffix Trees

• How on earth do you build suffix trees in time O(m)?

Constructing Suffix Arrays

• Start by building suffix arrays in time O(m)...

Constructing LCP Arrays

• ... and adding in LCP arrays in time O(m).