Suffix-based indexing

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- So far we've discussed substring indices
- Today we'll cover data structures that index the suffixes of a string: suffix tries, suffix tree, suffix array

Tries



A trie (pronounced "try") is a tree representing a collection of strings with one node per common prefix

Smallest tree such that:

Each edge is labeled with a character $c \in \Sigma$

A node has at most one outgoing edge labeled c, for $c \in \Sigma$

Each key is "spelled out" along some path starting at the root

Natural way to represent a *set* or a *map* where keys are strings





Make this map into a trie:

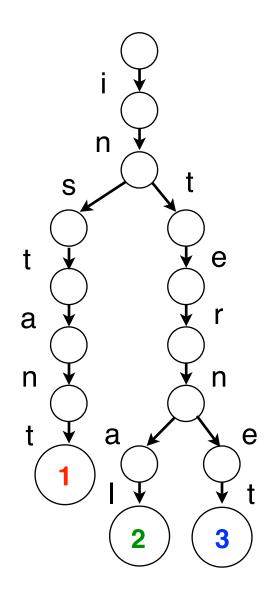
Key	Value
instant	1
internal	2
internet	3

The smallest tree such that:

Each edge is labeled with a character $c \in \Sigma$

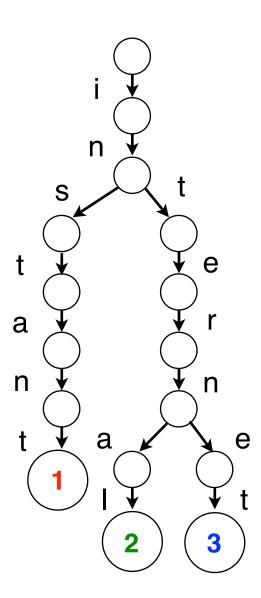
A node has at most one outgoing edge labeled c, for $c \in \Sigma$

Each key is "spelled out" along some path starting at the root



Tries: example





Checking for presence of a key P, where n = |P|, is O(n) time

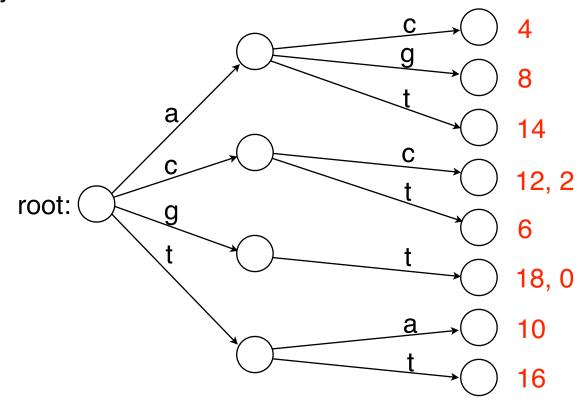
If total length of all keys is N, trie has O(N) nodes





We can implement a substring index of *T* with a trie. The trie maps substrings to offsets where they occur

ac	4
ag	8
at	14
СС	12
СС	2
ct	6
gt	18
gt	0
ta	10
tt	16

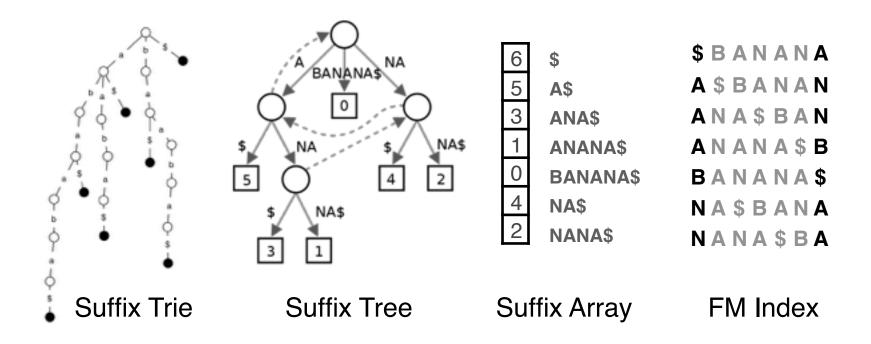






Until now, our indexes have been based on extracting substrings from T

A very different approach is to extract *suffixes* from *T.* This will lead us to some interesting and practical index data structures:







Build a **trie** containing all **suffixes** of a text *T*

```
T:
 GTTATAGCTGATCGCGGCGTAGCGG$
 GTTATAGCTGATCGCGGCGTAGCGG$
  TTATAGCTGATCGCGGCGTAGCGG$
   TATAGCTGATCGCGGCGTAGCGG
     ATAGCTGATCGCGGCGTAGCGG$
      TAGCTGATCGCGGCGTAGCGG
       AGCTGATCGCGGCGTAGCGG
        GCTGATCGCGGCGTAGCGG
         CTGATCGCGGCGTAGCGG$
          TGATCGCGGCGTAGCGG
           GATCGCGGCGTAGCGG
                                m(m+1)/2
            ATCGCGGCGTAGCGG$
             TCGCGGCGTAGCGG$
                                chars
              CGCGGCGTAGCGG
               GCGGCGTAGCGG$
                 CGGCGTAGCGG$
                  GGCGTAGCGG$
                   GCGTAGCGG
                    CGTAGCGG$
                     GTAGCGG$
                      TAGCGG$
                       AGCGG$
                        GCGG$
                         CGG$
                          GG$
                           G $
```



First add special *terminal character* \$ to the end of T

\$ is a character that does not appear elsewhere in T, and we define it to be less than other characters (for DNA: \$ < A < C < G < T)

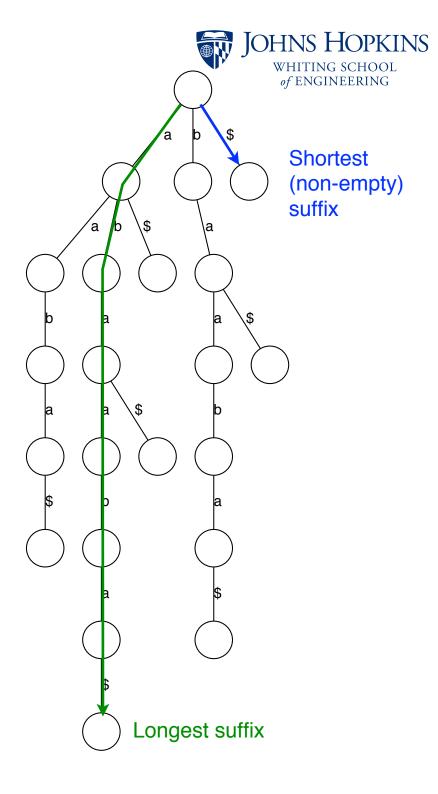
\$ enforces a rule we're all used to using: e.g. "as" comes before "ash" in the dictionary. \$ also guarantees no suffix is a prefix of any other suffix.

```
T:
 GTTATAGCTGATCGCGGCGTAGCGG$
 GTTATAGCTGATCGCGGCGTAGCGG$
  TTATAGCTGATCGCGGCGTAGCGG
   TATAGCTGATCGCGGCGTAGCGG
      TAGCTGATCGCGGCGTAGC
       AGCTGATCGCGGCGTAGCG
        GCTGATCGCGGCGTAGCGG
         CTGATCGCGGCGTAGCG
          TGATCGCGGCGTAGCGG
           GATCGCGGCGTAGCGG
            ATCGCGGCGTAGCGG
              CGCGGCGTAGCG
               GCGGCGTAGCGG
                CGGCGTAGCGGS
                  GCGTAGCGGS
```

T: abaaba T\$: abaaba\$

Each path from root to leaf represents a suffix; each suffix is represented by some path from root to leaf

Would this still be the case if we hadn't added \$?

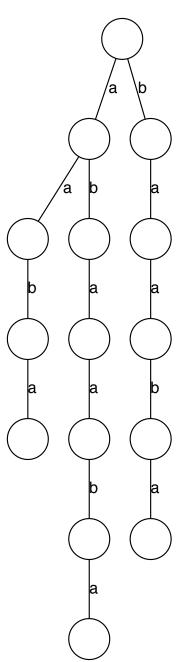




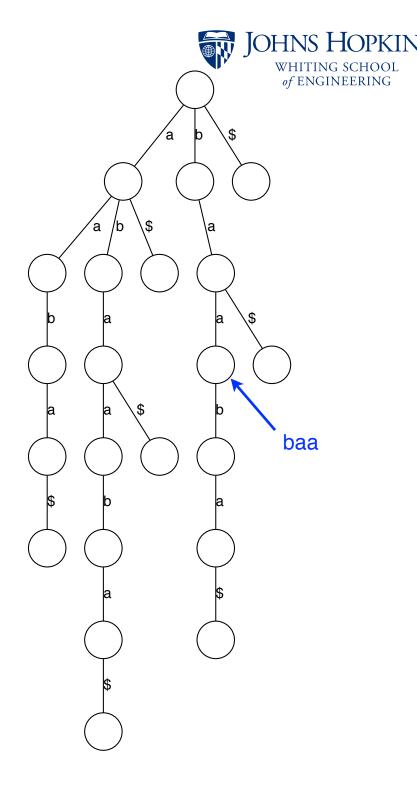
T: abaaba

Each path from root to leaf represents a suffix; each suffix is represented by some path from root to leaf

Would this still be the case if we hadn't added \$? No



We can think of nodes as having **labels**, where the label spells out characters on the path from the root to the node



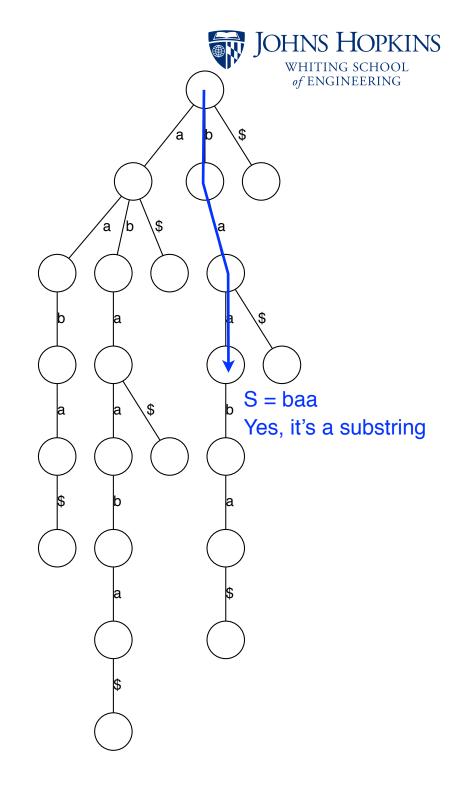
How do we check whether a string S is a substring of T?

Note: Each of *T*'s substrings is spelled out along a path from the root. I.e., every *substring* is a *prefix* of some *suffix* of T.

Start at the root and follow the edges labeled with the characters of *S*

If we "fall off" the trie -- i.e. there is no outgoing edge for next character of S, then S is not a substring of T

If we exhaust S without falling off, S is a substring of T



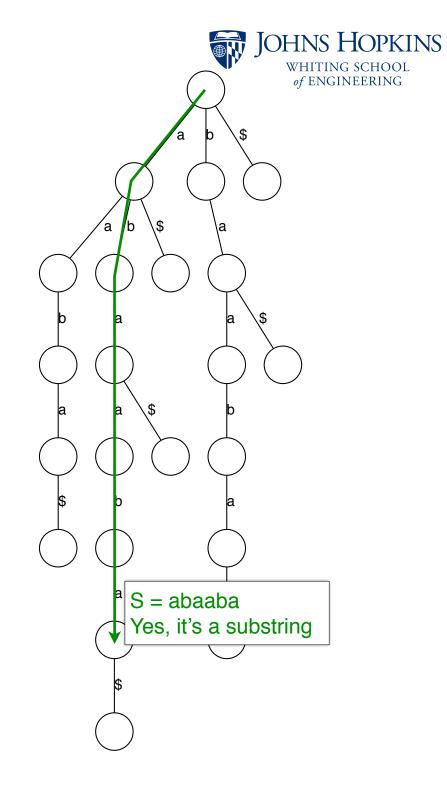
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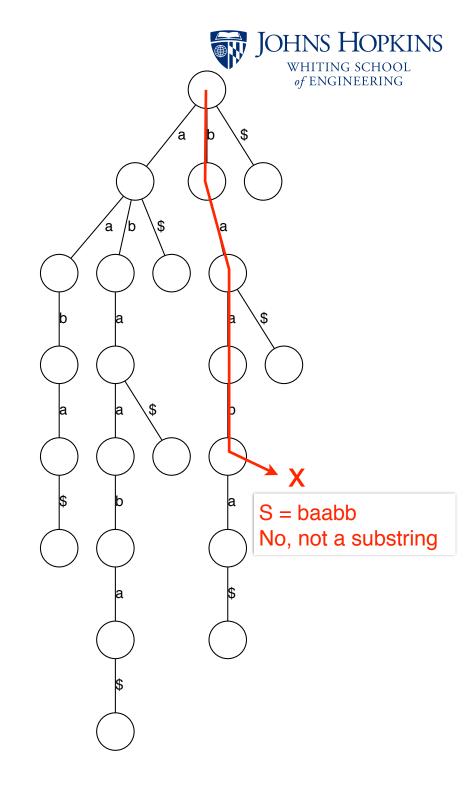
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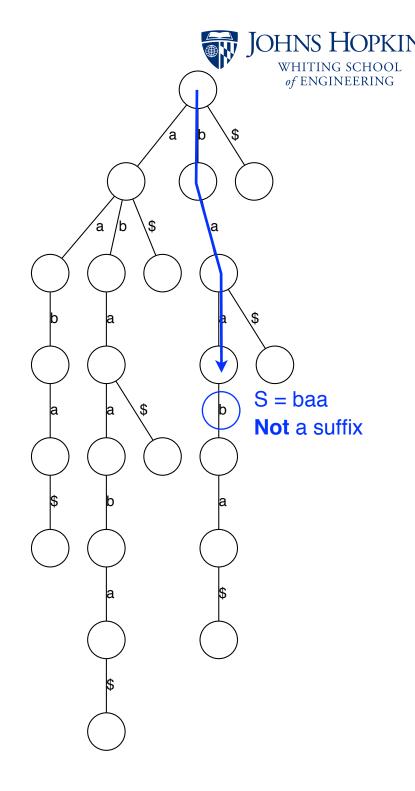
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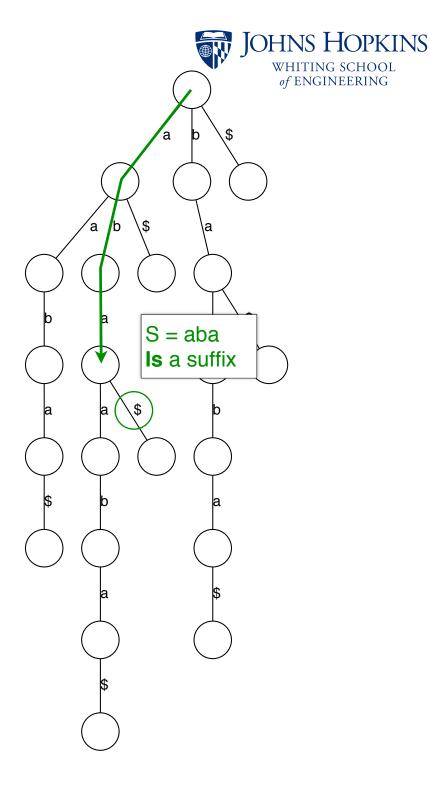
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Same procedure as for substring, but additionally check whether the final node in the walk has an outgoing edge labeled \$



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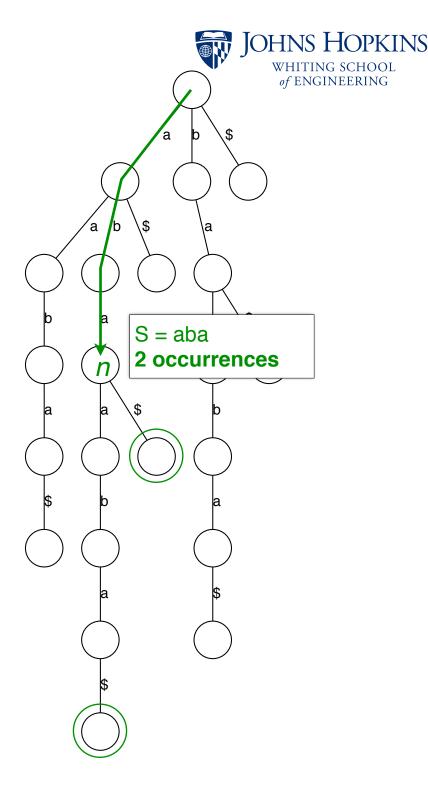
Same procedure as for substring, but additionally check whether the final node in the walk has an outgoing edge labeled \$



How do we count the **number of times** a string *S* occurs as a substring of *T*?

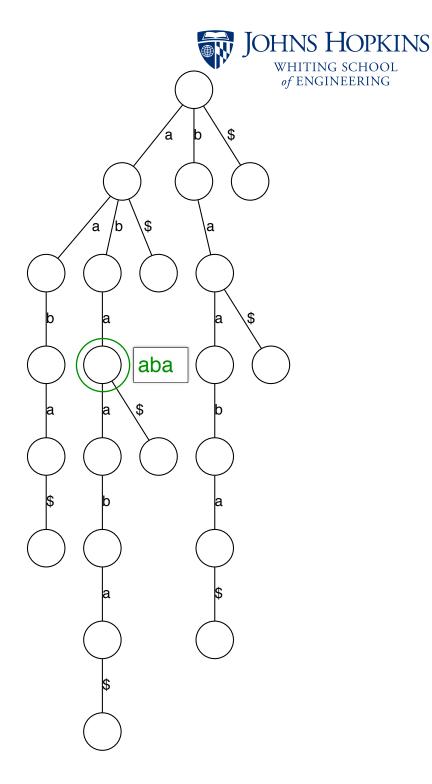
Follow path corresponding to S. Either we fall off, in which case answer is 0, or we end up at node nand the answer = # of leaf nodes in the subtree rooted at n.

Leaves can be counted with depth-first traversal.



How do we find the **longest repeated substring** of *T*?

Find the deepest node with more than one child

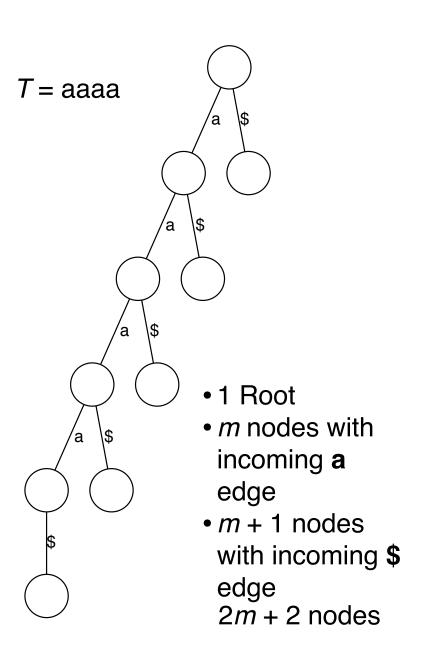




How many nodes does the suffix trie have?

Is there a class of string where the number of suffix trie nodes grows linearly with *m*?

Yes: e.g. a string of m a's in a row (a^m)





Is there a class of string where the number of suffix trie nodes grows with m^2 ?

Yes: aⁿbⁿ

- 1 root
- *n* nodes along "b chain," right
- n nodes along "a chain," middle
- n chains of n "b" nodes hanging off each "a chain" node
- •2n + 1 \$ leaves (not shown)

 $n^2 + 4n + 2$ nodes, where m = 2n

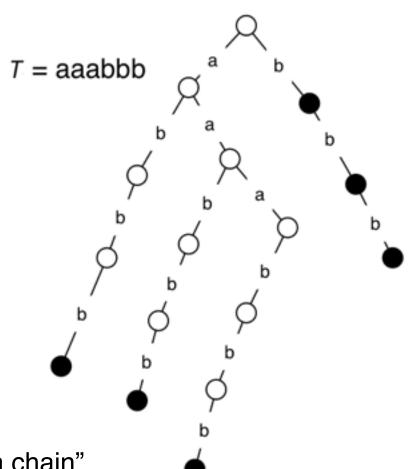
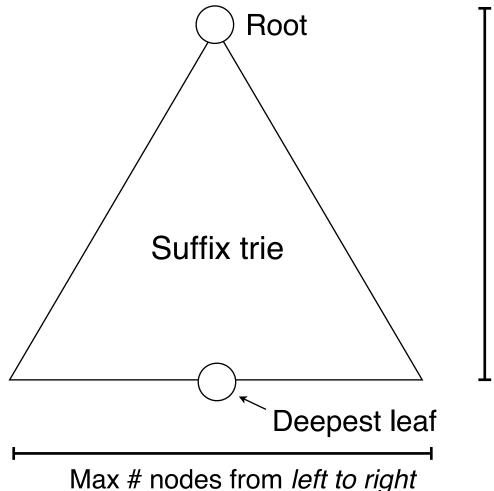


Figure & example by Carl Kingsford



Suffix trie: upper bound on size

Could worst-case # nodes be worse than $O(m^2)$?



Max # nodes from top to bottom = length of longest suffix + 1 = m + 1

Max # nodes from *left to right* = max # distinct substrings of any length ≤ m

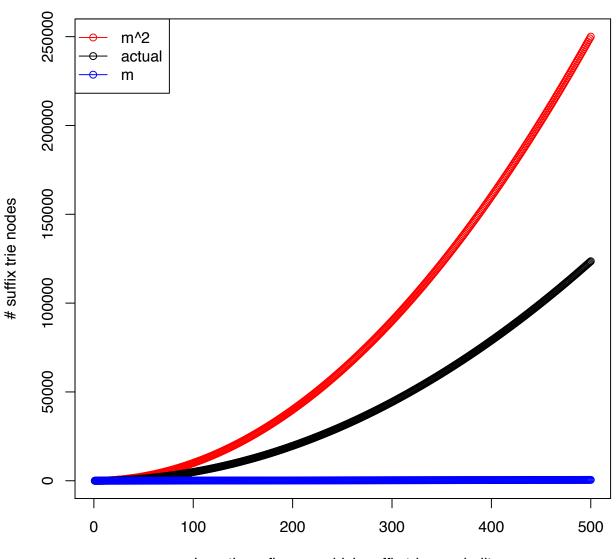
 $O(m^2)$ is worst case





Built suffix tries for the first 500 prefixes of the lambda phage virus genome

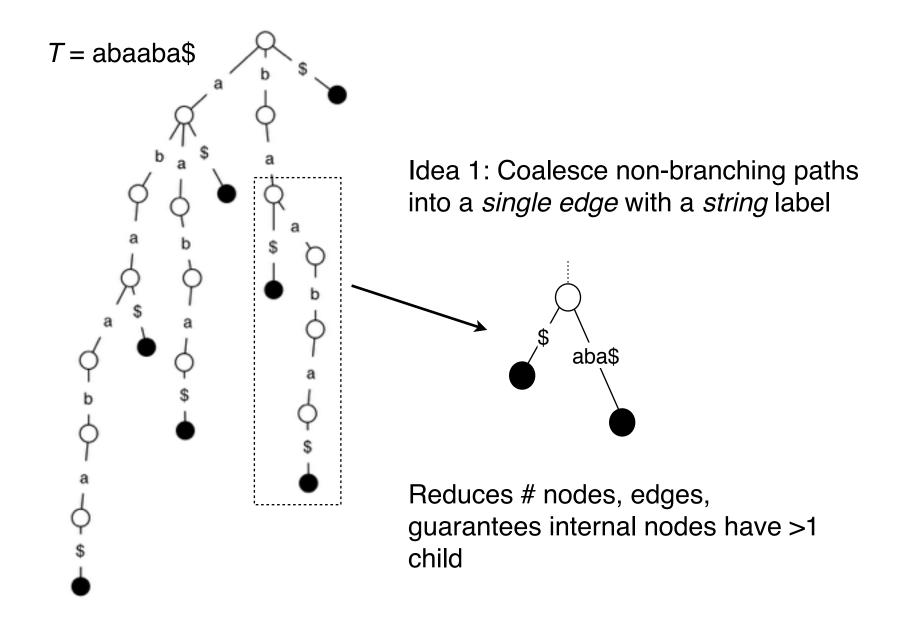
Black curve shows how # nodes increases with prefix length



Length prefix over which suffix trie was built

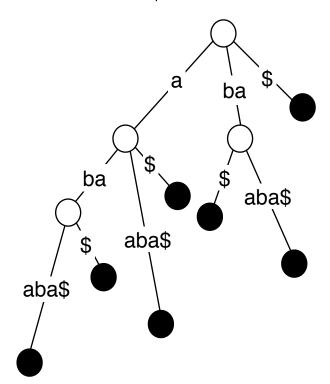


Suffix trie: making it smaller





T = abaaba\$



With respect to *m*:

How many leaves? *m*

How many non-leaf nodes? ≤ m - 1

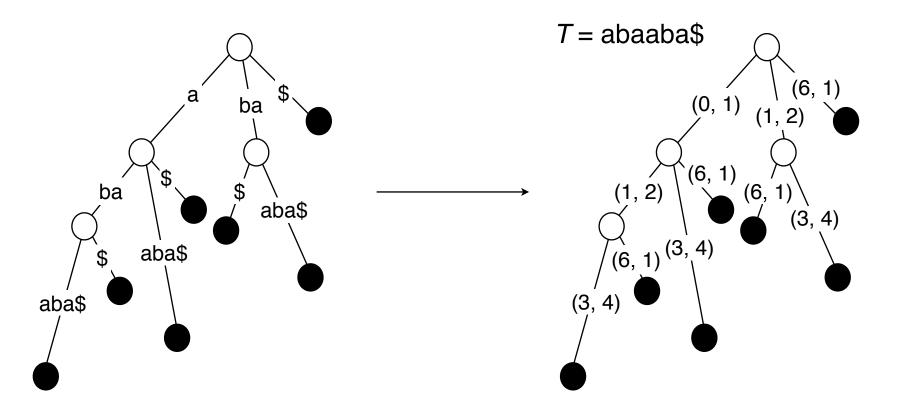
 $\leq 2m$ -1 nodes total, or O(m) nodes

Is the total size O(m) now? No: total length of edge labels is quadratic in m



T = abaaba\$

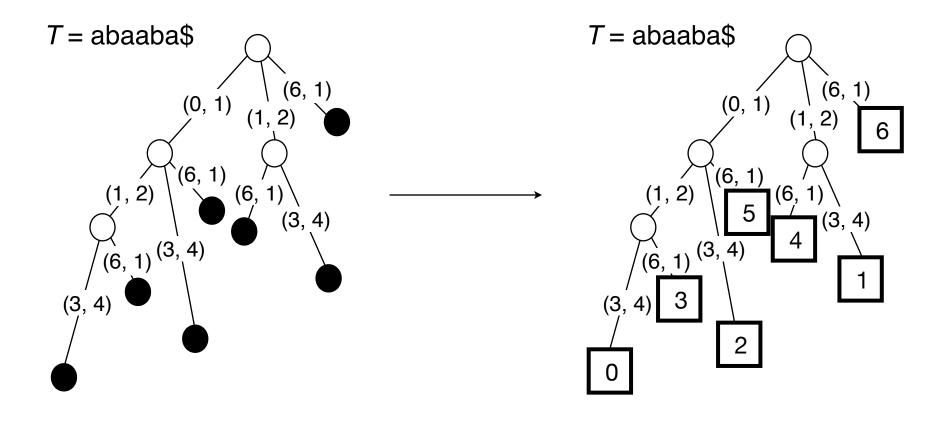
Idea 2: Store *T* itself in addition to the tree. Convert tree's edge labels to (offset, length) pairs with respect to *T*.



Space required for suffix tree is now O(m)

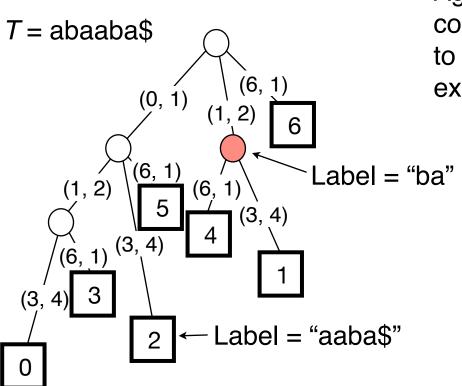


Suffix tree: leaves hold offsets



Suffix tree: labels





Again, each node's *label* equals the concatenated edge labels from the root to the node. These aren't stored explicitly.

Suffix tree: labels



T = abaaba\$ (0, 1) (1, 2) (6, 1) (1, 2) (6, 1) (6, 1) (3, 4) (3, 4) (3, 4) (3, 4) (3, 4) (3, 4) (3, 4) (3, 4)

Because edges can have string labels, we must distinguish two notions of "depth"

- Node depth: how many edges we must follow from the root to reach the node
- Label depth: total length of edge labels for edges on path from root to node

Suffix tree: building

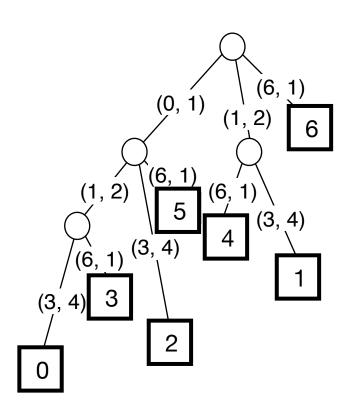


Naive method 1: build a suffix trie, then coalesce non-branching paths and relabel edges

Naive method 2: build a single-edge tree representing only the longest suffix, then augment to include the 2nd-longest, then augment to include 3rd-longest, etc

Both are $O(m^2)$ time, but first uses $O(m^2)$ space while second uses O(m)

Naive method 2 is described in Gusfield 5.4



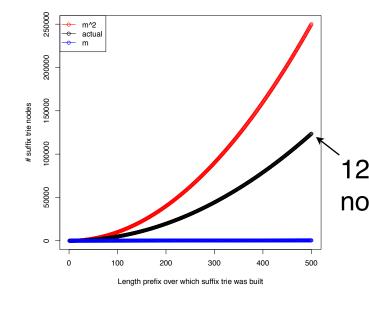


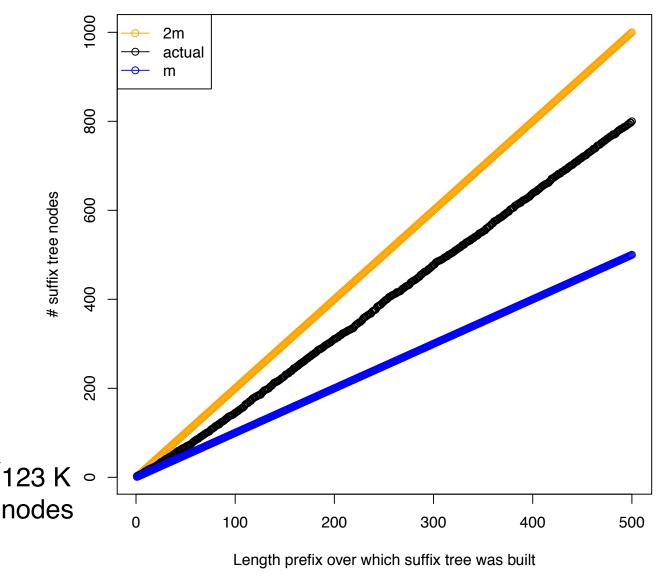


Built suffix trees for the first 500 prefixes of the lambda phage virus genome

Black curve shows # nodes increasing with prefix length

Compare with suffix trie:





Suffix tree: building



Method of choice: Ukkonen's algorithm

Ukkonen, Esko. "On-line construction of suffix trees." *Algorithmica* 14.3 (1995): 249-260.

O(m) time and space

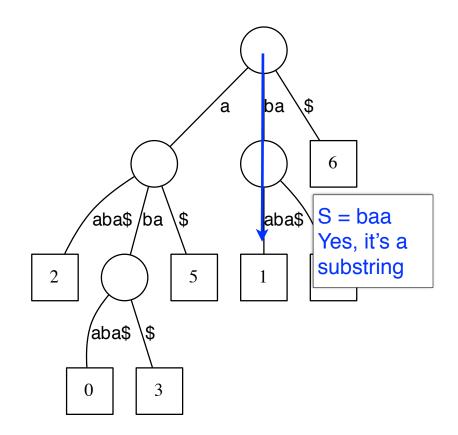
Has *online* property: if *T* arrives one character at a time, algorithm efficiently updates suffix tree upon each arrival

We won't cover it here; see Gusfield Ch. 6 for details



How do we check whether a string S is a substring of T?

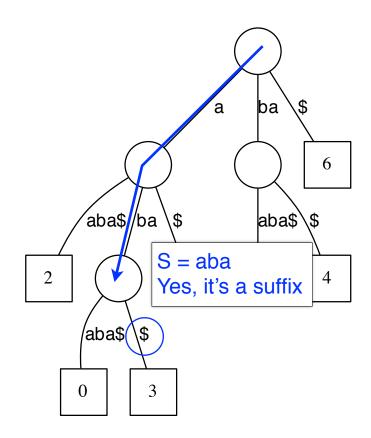
Essentially same procedure as for suffix trie, except we have to deal with coalesced edges





How do we check whether a string S is a suffix of T?

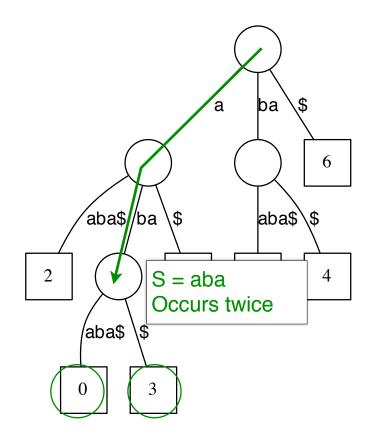
Essentially same procedure as for suffix trie, except we have to deal with coalesced edges

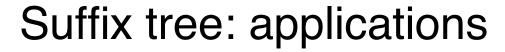




How do we count the **number of times** a string *S* occurs as a substring of *T*?

Same procedure as for suffix trie







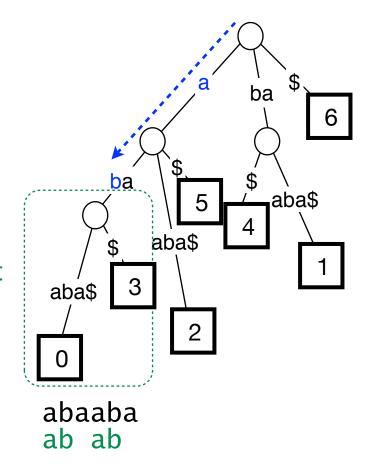
With suffix tree of T, we can find all matches of P to T. Let k = # matches.

E.g.,
$$P = ab$$
, $T = abaaba$ \$

Step 1: walk down ab path O(n) If we "fall off" there are no matches

O(k) Step 2: visit all leaf nodes below Report each leaf offset as match offset

O(n + k) time



Suffix trees in the real world: the constant factor

JOHNS HOPKINS

WHITING SCHOOL

of ENGINEERING

While O(m) is desirable, the constant in front of the m limits wider use of suffix trees in practice

Constant factor varies depending on implementation:

Estimate of MUMmer's constant factor = 3.94 GB / 250 million nt $\approx 15.75 \text{ bytes per node}$

Literature reports implementations achieving as little as 8.5 bytes per node, but no implementation used in practice that I know of is better than \approx 12.5 bytes per node

Kurtz, Stefan. "Reducing the space requirement of suffix trees." *Software Practice and Experience* 29.13 (1999): 1149-1171.





Organizes all suffixes into an incredibly useful, flexible data structure, in O(m) time and space

A naive method (e.g. suffix trie) could easily be quadratic

Used in practice for whole genome alignment, repeat identification, etc

GTTATAGCTGATCGCGGCGTAGCGG\$

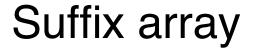
ATAGCTGATCGCGGCGTAGCGG\$

Actual memory footprint (bytes per node) is quite high, limiting usefulness

m chars

m(m+1)/2 chars

A G C G G \$
G C G G \$
C G G \$
G G \$





$$T$$$
 = abaaba\$ \leftarrow As with suffix tree, T is part of index

$$SA(T) = \begin{bmatrix} 6 \\ 5 \\ a \\ 2 \\ a a b a \\ 3 \\ a b a \\ 4 \\ b a \\ 5 \end{bmatrix}$$

$$m + 1$$
integers
$$0 \text{ ab a a b a } \\ 4 \text{ b a } \\ 1 \text{ b a a b a }$$

Suffix array of T is an array of integers in [0, m] specifying the lexicographic order of T\$'s suffixes

Suffix array



O(m) space, same as suffix tree. Is constant factor smaller?

32-bit integer can distinguish characters in the human genome, so suffix array is ~12 GB, smaller than MUMmer's 47 GB suffix tree.





Is *P* a substring of *T*?

- 1. For P to be a substring, it must be a prefix of ≥ 1 of T's suffixes
- 2. Suffixes sharing a prefix are consecutive in the suffix array

Use binary search

6	\$
5	a \$
2	aaba\$
3	aba\$
0	abaaba\$
4	ba\$
1	baaba\$





Python has bisect module for binary search

bisect.bisect_left(a, x): Leftmost offset where we can
insert x into a to maintain sorted order. a is already sorted!

bisect.bisect_right(a, x): Like bisect_left, but returning
rightmost instead of leftmost offset

```
from bisect import bisect_left, bisect_right

a = [1, 2, 3, 3, 3, 4, 5]
print(bisect_left(a, 3), bisect_right(a, 3)) # output: (2, 5)

a = [2, 4, 6, 8, 10]
print(bisect_left(a, 5), bisect_right(a, 5)) # output: (2, 2)
```

Python example: http://nbviewer.ipython.org/6753277



Suffix array: binary search

We can straightforwardly use binary search to find a range of elements in a sorted list that *equal* some query:

```
from bisect import bisect_left, bisect_right

strls = ['a', 'awkward', 'awl', 'awls', 'axe', 'axes', 'bee']

# Get range of elements that equal query string 'awl'
st, en = bisect_left(strls, 'awl'), bisect_right(strls, 'awl')
print(st, en) # output: (2, 3)
```

Python example: http://nbviewer.ipython.org/6753277



Suffix array: binary search

Can also use binary search to find a range of elements in a sorted list with some query as a *prefix*:

```
from bisect import bisect_left, bisect_right

strls = ['a', 'awkward', 'awl', 'awls', 'axe', 'axes', 'bee']

# Get range of elements with 'aw' as a prefix
st, en = bisect_left(strls, 'aw'), bisect_left(strls, 'ax')

print(st, en) # output: (1, 4)
```

Python example: http://nbviewer.ipython.org/6753277



Suffix array: binary search

We can do the same thing for a sorted list of suffixes:

```
6 $
5 a $
2 a a b a $
3 a b a $
0 a b a a b a $
4 b a $
1 b a a b a $
```





Is *P* a substring of *T*?

Do binary search, check whether *P* is a prefix of the suffix there

How many times does *P* occur in *T*?

Two binary searches yield the range of suffixes with P as prefix; size of range equals # times P occurs in T

Worst-case time bound?

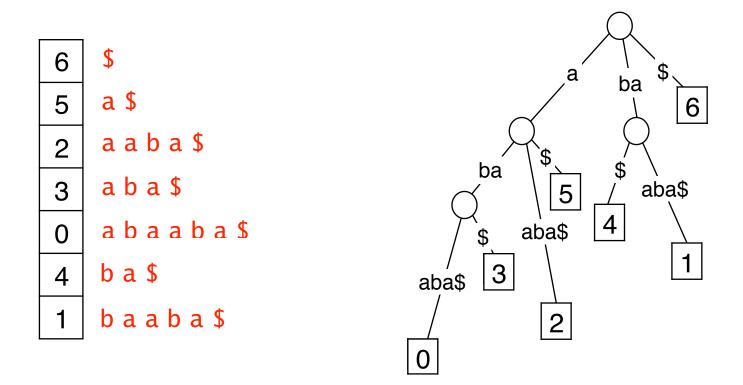
 $O(\log_2 m)$ bisections, O(n) comparisons per bisection, so $O(n \log m)$

aaba\$ aba\$ abaaba\$ b a \$ baaba\$





Contrast suffix array: $O(n \log m)$ with suffix tree: O(n)



But we can improve bound for suffix array...

Suffix array: querying



Consider further: binary search for suffixes with *P* as a prefix

Assume there's no \$ in P. So P can't be equal to a suffix.

Initialize
$$l = 0$$
, $c = floor(m/2)$ and $r = m$ (just past last elt of SA)

 \uparrow

"left" "center" "right"

Notation: We'll use use SA[l] to refer to the suffix corresponding to suffix-array element l. We could write T[SA[l]:], but that's too verbose.

Throughout the search, invariant is maintained:





Throughout search, invariant is maintained:

What do we do at each iteration?

```
Let c = \text{floor}((r+l)/2)
If P < \text{SA}[c], either stop or let r = c and iterate
If P > \text{SA}[c], either stop or let l = c and iterate
```

When to stop?

```
P < SA[c] and c = l + 1 - answer is c

P > SA[c] and c = r - 1 - answer is r
```



Suffix array: querying

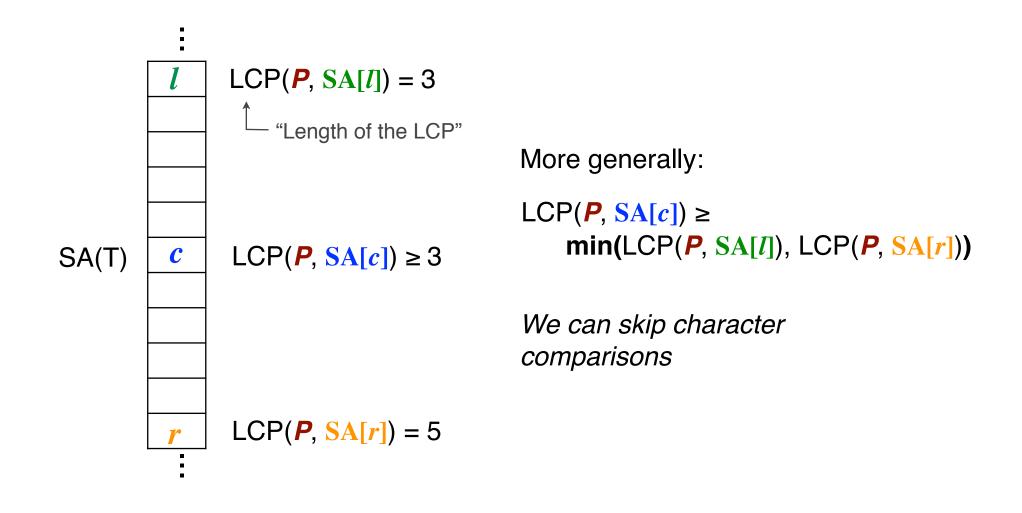
```
def binarySearchSA(t, sa, p):
    assert t[-1] == '$' # t already has terminator
    assert len(t) == len(sa) # sa is the suffix array for t
   if len(t) == 1: return 1
    1, r = 0, len(sa) # invariant: sa[1] < p < sa[r]
   while True:
       c = (1 + r) // 2
       # determine whether p < T[sa[c]:] by doing comparisons</pre>
        # starting from left-hand sides of p and T[sa[c]:]
        plt = True # assume p < T[sa[c]:] until proven otherwise</pre>
        i = 0
       while i < len(p) and sa[c]+i < len(t):
            if p[i] < t[sa[c]+i]:
                                                  # loop iterations ≈ length
                break # p < T[sa[c]:]</pre>
                                                  of Longest Common
            elif p[i] > t[sa[c]+i]:
                                                  Prefix (LCP) of P and
                plt = False
                break # p > T[sa[c]:]
                                                  SA[c]
            i += 1 # tied so far
        if plt:
           if c == 1 + 1: return c
                                          If we already know something about
            r = c
                                          LCP of P and SA[c], we can save
        else:
           if c == r - 1: return r
                                          work
            1 = c
```

Python example: http://nbviewer.ipython.org/6765182





Say we're comparing P to SA[c] and we've already compared P to SA[l] and SA[r] in previous iterations.







```
def binarySearchSA lcp1(t, sa, p):
    if len(t) == 1: return 1
   l, r = 0, len(sa) \# invariant: sa[1] 
   lcp lp, lcp rp = 0, 0
   while True:
        c = (1 + r) // 2
        plt = True
        i = min(lcp_lp, lcp rp)
        while i < len(p) and sa[c]+i < len(t):
            if p[i] < t[sa[c]+i]:
                break # p < T[sa[c]:]</pre>
            elif p[i] > t[sa[c]+i]:
                plt = False
                break # p > T[sa[c]:]
            i += 1 # tied so far
        if plt:
            if c == 1 + 1: return c
            r = c
            lcp rp = i
        else:
            if c == r - 1: return r
            1 = c
            lcp lp = i
```

Worst-case time bound is still $O(n \log m)$, but we're closer

Python example: http://nbviewer.ipython.org/6765182





We saw 2 ways to query (binary search) the suffix array:

- 1. Typical binary search. Ignores LCPs. $O(n \log m)$.
- 2. Binary search with some skipping using LCPs between P and T's suffixes. Still $O(n \log m)$ worst case, but near $O(n + \log m)$ in practice.

Gusfield: "Simple Accelerant"

If we precompute external LCP arrays we can accelerate the search to worst case $O(n + \log m)$ at the additional space cost of 2m integers. "Super Accelerant" algorithm

Suffix array: performance comparison



	Super accelerant	Simple accelerant	No accelerant
python -O	68.78 s	69.80 s	102.71 s
руру -О	5.37 s	5.21 s	8.74 s
# character comparisons	99.5 M	117 M	235 M

Matching 500K 100-nt substrings to the \sim 5 million nt-long *E. coli* genome. Substrings drawn randomly from the genome.

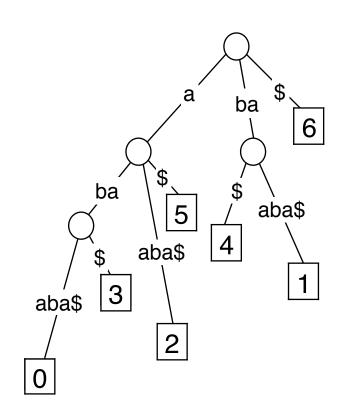
Index building time not included





Given T, how to we efficiently build T's suffix array?

6	\$
5	a \$
2	aaba\$
3	aba\$
0	abaaba\$
4	ba\$
1	baaba\$

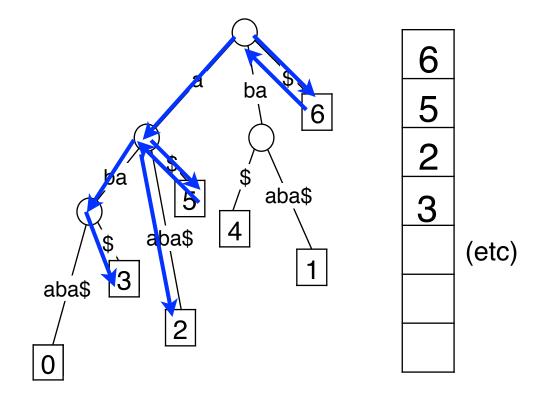


Suffix array: building



Idea: Build suffix tree, do a lexicographic depth-first traversal reporting leaf offsets as we go

Traverse O(m) nodes and emit m integers, so O(m) time assuming edges are already ordered







Suffix trees are big. Given *T*, how do we efficiently build *T*'s suffix array *without* first building a suffix tree?

```
6
5
a $
a a b a $
a b a a b a $
b a $
b a a b a $
```



Suffix array: sorting suffixes

One idea: Use your favorite sort, e.g., quicksort

```
abaaba$
                     def quicksort(q):
                         lt, gt = [], []
baaba$
                         if len(q) <= 1:
aaba$
                             return q
                         for \times in q[1:]:
aba$
                             if x < q[0]:
b a $
                                 lt.append(x)
                             else:
a $
                                 gt.append(x)
                         return quicksort(lt) + q[0:1] + quicksort(gt)
```

Expected time: O($m^2 \log m$)

Not $O(m \log m)$ because a suffix comparison is O(m) time





One idea: Use a sort algorithm that's aware that the items being sorted are strings, e.g. "multikey quicksort"

0	abaaba\$
1	baaba\$
2	aaba\$
3	aba\$
4	ba\$
5	a \$
6	\$

Essentially $O(m^2)$ time

Bentley, Jon L., and Robert Sedgewick. "Fast algorithms for sorting and searching strings." *Proceedings of the eighth annual ACM-SIAM symposium on Discrete algorithms.* Society for Industrial and Applied Mathematics, 1997





Another idea: Use a sort algorithm that's aware that the items being sorted are all suffixes of the same string

Original suffix array paper suggested an $O(m \log m)$ algorithm

Manber U, Myers G. "Suffix arrays: a new method for on-line string searches." SIAM Journal on Computing 22.5 (1993): 935-948.

Other popular $O(m \log m)$ algorithms have been suggested

Larsson NJ, Sadakane K. Faster suffix sorting. Technical Report LU-CS-TR: 99-214, LUNDFD6/(NFCS-3140)/1-43/(1999), Department of Computer Science, Lund University, Sweden, 1999.

More recently O(m) algorithms have been demonstrated!

Kärkkäinen J, Sanders P. "Simple linear work suffix array construction." Automata, Languages and Programming (2003): 187-187.

Ko P, Aluru S. "Space efficient linear time construction of suffix arrays." *Combinatorial Pattern Matching*. Springer Berlin Heidelberg, 2003.

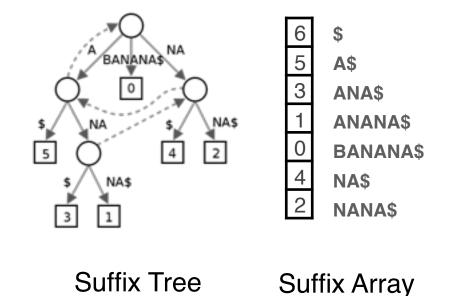
Suffix array: summary



Suffix array gives us index that is:

(a) Just m integers, with $O(n \log m)$ worst-case query time, but close to $O(n + \log m)$ in practice

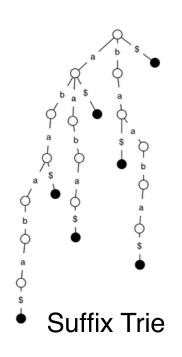
or **(b)** 3m integers, with $O(n + \log m)$ worst case



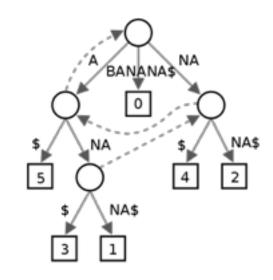
(a) will often be preferable: index for entire human genome fits in ~12 GB instead of > 45 GB

Summary





O(n) queries O(m²) space



O(n) queries O(m) space

Suffix Tree



O(n log m) queries O(m) space

Suffix Array





 Short break then I'll present the main ideas from the MUMmer paper