

6 Text Indexing – Searching whole genomes

16 March 2021

Sebastian Wild

6 Text Indexing

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- 6.2 Suffix Trees
- 6.3 Applications
- 6.4 Longest Common Extensions
- 6.5 Suffix Arrays
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6.1 Motivation

Text indexing

- ▶ *Text indexing* (also: *offline text search*):

- ▶ case of string matching: find $P[0..m)$ in $T[0..n)$

- ▶ but with *fixed* text \rightsquigarrow preprocess T (instead of P)

- \rightsquigarrow expect many queries P , answer them without looking at all of T

- \rightsquigarrow essentially a data structuring problem: “building an *index* of T ”

Latin: “one who points out”

- ▶ application areas

- ▶ web search engines

- ▶ online dictionaries

- ▶ online encyclopedia

- ▶ DNA/RNA data bases

- ▶ ... searching in any collection of text documents (that grows only moderately)

Inverted indices

same as "indexes"

- ▶ original indices in books: list of (key) words \mapsto page numbers where they occur
- ▶ assumption: searches are only for **whole** (key) **words**
- ~> often reasonable for natural language text

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Inverted index:

- ▶ collect all words in T
 - ▶ can be as simple as splitting T at whitespace
- (▶ actual implementations typically support *stemming* of words)
goes \rightarrow go, cats \rightarrow cat
- ▶ store mapping from words to a list of occurrences \rightsquigarrow how?

like a dictionary!

keys = words

values = list of occurrence,

BST
but $O(\log n)$
time

Clicker Question



Do you know what a *trie* is?

- ☐ **A** A what? No!
- ☐ **B** I have heard the term, but don't quite remember.
- ☐ **C** I remember hearing about it in a module.
- ☐ **D** Sure.

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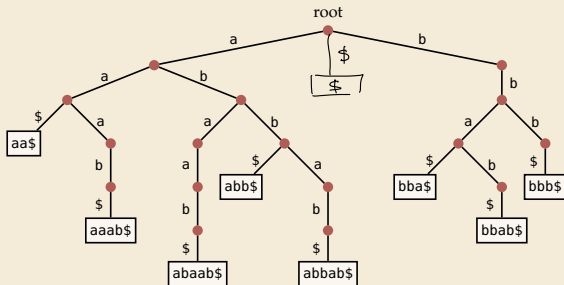
Click on "Polls" tab

Tries

- ▶ efficient dictionary data structure for strings
- ▶ name from retrieval, but pronounced “try” \approx free
- ▶ tree based on symbol comparisons
- ▶ **Assumption:** stored strings are prefix-free (no string is a prefix of another)
 - ▶ strings of same length ✓ some character $\notin \Sigma$
 - ▶ strings have “end-of-string” marker \$ ✓

▶ Example:

$\Sigma = \{a, b\}$
{aa\$, aaab\$, abaab\$, abb\$,
abbab\$, bba\$, bbab\$, bbb\$, \$}



Clicker Question

Suppose we have a trie that stores n strings over $\Sigma = \{A, \dots, Z\}$. Each stored string consists of m characters.

We now search for a query string Q with $|Q| = q$. ($q \leq m$)

How many **nodes** in the trie are **visited** during this **query**?



A $\Theta(\log n)$

F $\Theta(\log m)$

B $\Theta(\log(nm))$

G $\Theta(q)$

C $\Theta(m \cdot \log n)$

H $\Theta(\log q)$

D $\Theta(m + \log n)$

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E $\Theta(m)$

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Suppose we have a trie that stores n strings over $\Sigma = \{A, \dots, Z\}$. Each stored string consists of m characters.

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A $\Theta(n)$

D $\Theta(n \log m)$

B $\Theta(n + m)$

E $\Theta(m)$

C $\Theta(n \cdot m)$

F $\Theta(m \log n)$

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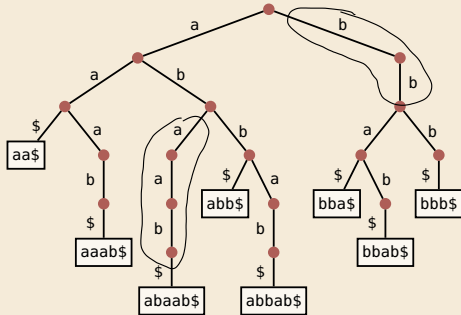
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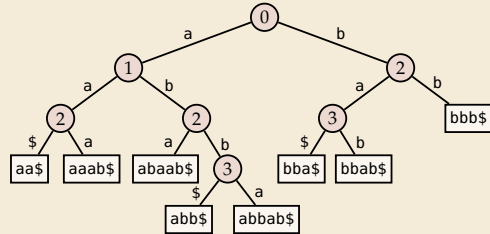
Compact tries

- ▶ compress paths of unary nodes into single edge
- ▶ nodes store index of next character

standard trie



compact trie



↪ searching slightly trickier, but same time complexity as in trie

- ▶ all nodes ≥ 2 children ↪ $\#nodes \leq \#leaves = \#strings$ ↪ linear space $O(n)$

not $O(n \cdot m)$

Tries as inverted index



simple



fast lookup



cannot handle more general queries:

- ▶ search part of a word
- ▶ search phrase (sequence of words)

Tries as inverted index

👍 simple

👍 fast lookup

👎 cannot handle more general queries:

- ▶ search part of a word
- ▶ search phrase (sequence of words)

👎 what if the 'text' does not even have words to begin with?!

- ▶ biological sequences

```
ACAAGATGCCATTGTCCCCGGCCTCTGCTGCTGCTGCTCTCCGGGGCCACGGCCACCGCTGCCCTGCCCTGGAGGGTGGCCCCACCGGC  
CGAGACAGCGAGCATATGCAGGAAGCGGCAGGAATAAGGAAAAGCAGCCTCCTGACTTTCTCGCTTGGTGGTTTGAGTGGACCTCCAGGC  
CAGTGCCGGGCCCCCTCATAGGAGAGGAAGCTCGGGAGGTGGCCAGGCGGCAGGAAGGCGCACCCCCCAGCAATCCGCGCGCCGGGACAGAA  
TGCCCTGCAGGAACCTTCTTCTGGAAGACCTTCTCCTCTGCAAATAAAACCTCACCCATGAATGCTCACGCAAGTTTAATTACAGACCTGAA
```

- ▶ binary streams

```
00000010101001111010111000001111100011111011111001101101000011100010011011110000010001101010  
011011000011010110100000001000000011101011000001000011110101110110010001100101101110111111  
110001010001011001010000001110101010011000000001101100001100111110000101 0101011101111000011  
10101110010010101010100000111110100110000001111001101010000000100100100000101100011000110111
```

~> need new ideas

6.2 Suffix Trees

Suffix trees – A ‘magic’ data structure

Appetizer: Longest common substring problem

- ▶ Given: strings S_1, \dots, S_k **Example:** $S_1 = \text{superiorcalifornialives}$, $S_2 = \text{sealiver}$
- ▶ Goal: find the longest substring that occurs in all k strings

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Enter: *suffix trees*

- ▶ versatile data structure for index with full-text search
- ▶ linear time (for construction) and linear space
- ▶ allows efficient solutions for many advanced string problems

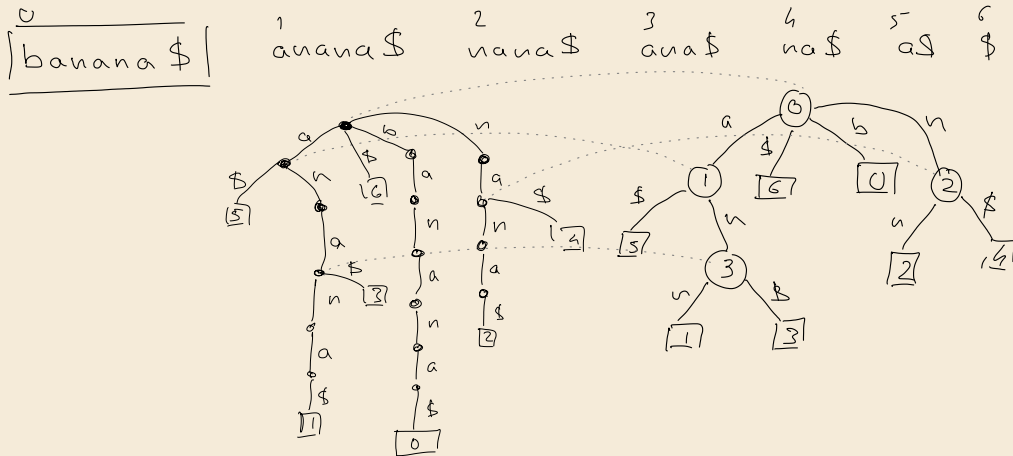


“Although the longest common substring problem looks trivial now, given our knowledge of suffix trees, it is very interesting to note that in 1970 Don Knuth conjectured that a linear-time algorithm for this problem would be impossible.”

[Gusfield: Algorithms on Strings, Trees, and Sequences (1997)]

Suffix trees – Definition

- suffix tree \mathcal{T} for text $T = T[0..n)$ = compact trie of all suffixes of T (set $T[n] := \$$)



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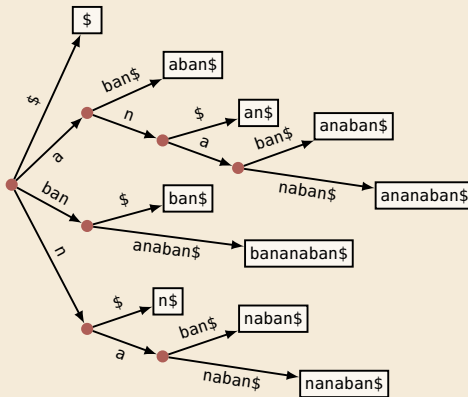
- suffix tree \mathcal{T} for text $T = T[0..n)$ = compact trie of all suffixes of $T\$$ (set $T[n] := \$$)

Example:

$T = \text{bananaban\$}$

suffixes: { $\text{bananaban\$}$, $\text{ananaban\$}$, $\text{nanaban\$}$,
 $\text{anaban\$}$, $\text{naban\$}$, $\text{aban\$}$, $\text{ban\$}$, $\text{an\$}$, $\text{n\$}$, $\text{\$}$ }

	0	1	2	3	4	5	6	7	8	9
$T =$	b	a	n	a	n	a	b	a	n	\$



Suffix trees – Definition

- ▶ suffix tree \mathcal{T} for text $T = T[0..n)$ = compact trie of all suffixes of $T\$$ (set $T[n] := \$$)
- ▶ except: in leaves, store *start index* (instead of actual string)

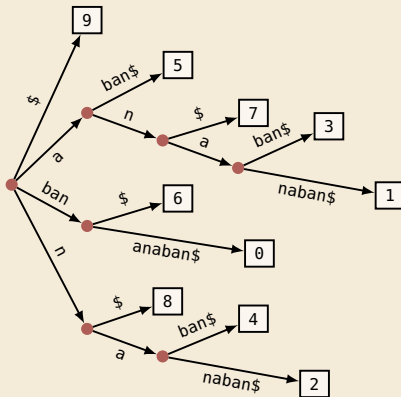
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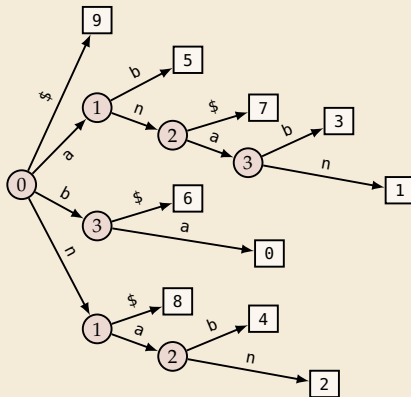
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b	a	n	a	n	a	b	a	n	\$

$T =$

- ▶ also: edge labels like in compact trie
- ▶ (more readable form on slides to explain algorithms)



Suffix trees – Construction

- ▶ $T[0..n)$ has $n + 1$ suffixes (starting at character $i \in [0..n)$)
- ▶ We can build the suffix tree by inserting each suffix of T into a compressed trie.
But that takes time $\Theta(n^2)$. \rightsquigarrow not interesting!

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same order of growth as reading the text!

Amazing result: Can construct the suffix tree of T in $\Theta(n)$ time!

- ▶ algorithms are a bit tricky to understand
- ▶ but were a theoretical breakthrough
- ▶ and they are efficient in practice (and heavily used)!

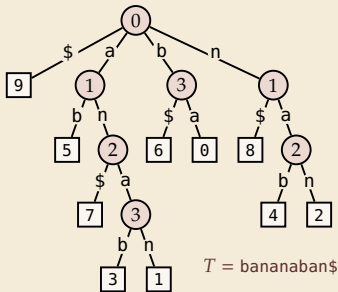
\rightsquigarrow for now, take linear-time construction for granted. What can we do with them?

6.3 Applications

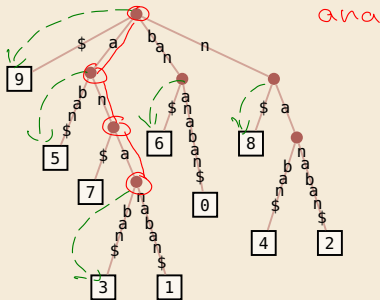
Applications of suffix trees

- In this section, always assume suffix tree \mathcal{T} for T given.

Recall: \mathcal{T} stored like this:

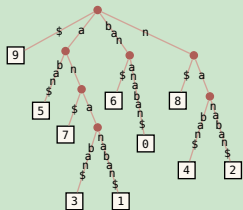


but think about this:



- Moreover: assume internal nodes store pointer to leftmost leaf in subtree.
- Notation: $T_i = T[i..n]$ (including \$)

Clicker Question



What does T 's suffix tree (on the left) tell you about the question whether T contains the pattern $P = \text{ana}$?

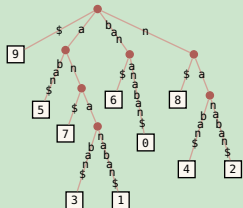
Check all that apply to this example.

- ☐ **A** Nothing.
- ☐ **B** P occurs in T .
- ☐ **C** P does not occur in T .
- ☐ **D** P occurs once in T .
- ☐ **E** P occurs twice in T .
- ☐ **F** P starts at index 0.
- ☐ **G** P starts at index 1.
- ☐ **H** P starts at index 2.
- ☐ **I** P starts at index 3.
- ☐ **J** P starts at index 4.
- ☐ **K** P starts at index 7.

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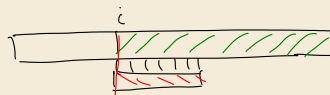
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Application 1: Text Indexing / String Matching

- ▶ P occurs in $T \iff \underline{P}$ is a prefix of a suffix of T
- ▶ we have all suffixes in \mathcal{T} !



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↪ (try to) follow path with label P , until

1. we get stuck

at internal node (no node with next character of P) n^b
or inside edge (mismatch of next characters) $b a_\alpha$

$\rightsquigarrow P$ does not occur in T

2. we run out of pattern

we run out of pattern
reach end of P at internal node v or inside edge towards v

→ P occurs at all leaves in subtree of v

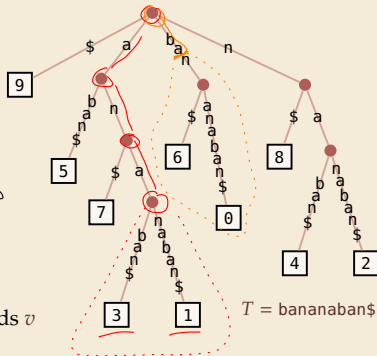
- ### 3. we run out of tree

reach a leaf ℓ with part of P left \rightsquigarrow compare P to ℓ .



This cannot happen when testing edge labels since $\$ \notin \Sigma$, but needs check(s) in compact trie implementation!

- Finding first match (or NO MATCH) takes $O(|P|)$ time!



not possible/relevant
text indexing

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1. **we get stuck**

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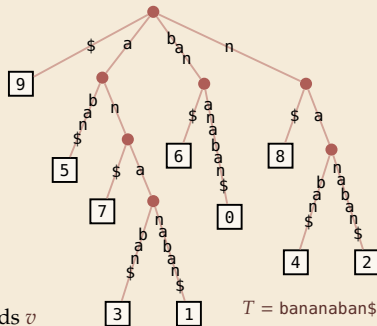
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Examples:

► $P = \text{ann}$

► $P = \text{ana}$

► $P = \text{briar}$

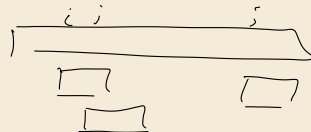
Application 2: Longest repeated substring

► **Goal:** Find longest substring $T[i..i + \ell)$ that occurs also at $j \neq i$: $T[j..j + \ell) = T[i..i + \ell)$.



How can we efficiently check *all possible substrings*?

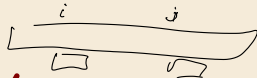
e.g. for compression \rightsquigarrow Unit 7



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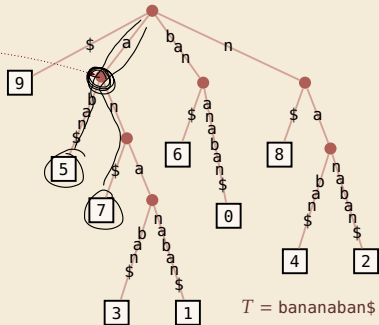
Repeated substrings = shared paths in *suffix tree*



- $T_5 = \underline{a}ban\$$ and $T_7 = \underline{a}n\$$ have longest common prefix 'a'

$\rightsquigarrow \exists$ internal node with path label 'a'

here single edge, can be longer path



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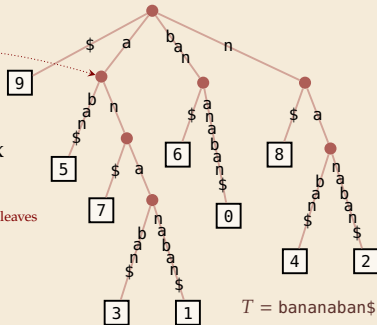
- $T_5 = \text{aban\$}$ and $T_7 = \text{an\$}$ have *longest common prefix 'a'*

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\rightsquigarrow longest repeated substring = longest common prefix (LCP) of two suffixes

actually: adjacent leaves



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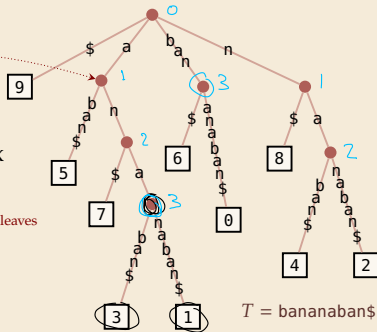
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actually: adjacent leaves

- Algorithm:

1. Compute string depth (=length of path label) of nodes
2. Find internal nodes with maximal string depth

- Both can be done in depth-first traversal $\rightsquigarrow \Theta(n)$ time



Generalized suffix trees

- ▶ longest *repeated* substring (of one string) feels very similar to longest *common* substring of several strings $T^{(1)}, \dots, T^{(k)}$ with $T^{(j)} \in \Sigma^{n_j}$
- ▶ can we solve that in the same way?
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Enter: *generalized suffix tree*

- ▶ Define $T := T^{(1)}\$1 T^{(2)}\$2 \dots T^{(k)}\$k$ for k new end-of-word symbols
- ▶ Construct suffix tree \mathcal{T} for T

\rightsquigarrow $\$j$ -edges always leads to leaves $\rightsquigarrow \exists$ leaf (j, i) for each suffix $T_i^{(j)} = T^{(j)}[i..n_j]$



Clicker Question



What is the longest common substring of the strings
bcabcac, aabca and bcaa?

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Application 3: Longest common substring

► With that new idea, we can find longest common superstrings:

1. Compute generalized suffix tree \mathcal{T} .
2. Store with each node the *subset of strings* that contain its path label:
 - 2.1. Traverse \mathcal{T} bottom-up.
 - 2.2. For a leaf (j, i) , the subset is $\{j\}$.
 - 2.3. For an internal node, the subset is the union of its children.
3. In top-down traversal, compute string depths of nodes. (as above)
4. Report deepest node (by string depth) whose subset is $\{1, \dots, k\}$.

② stores set of j so that there is a leaf (j, i) in the subtree

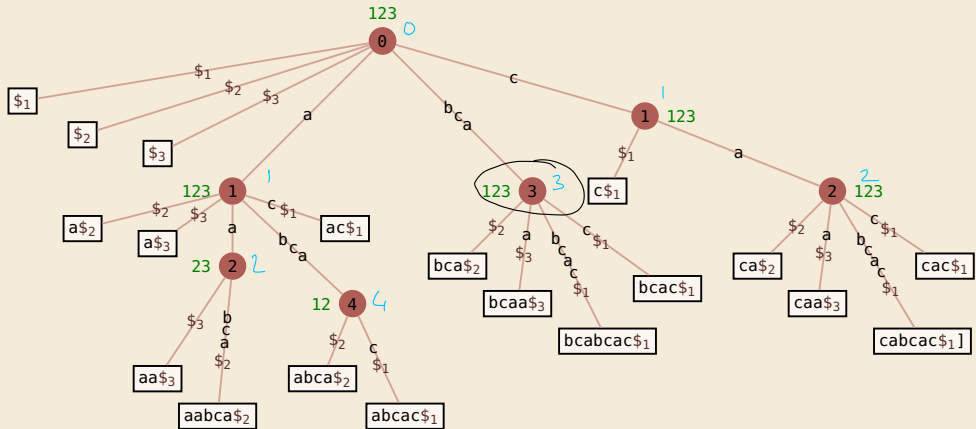
► Each step takes time $\Theta(n)$ for $n = n_1 + \dots + n_k$ the total length of all texts.

“Although the longest common substring problem looks trivial now, given our knowledge of suffix trees, it is very interesting to note that in 1970 Don Knuth conjectured that a linear-time algorithm for this problem would be impossible.”

[Gusfield: Algorithms on Strings, Trees, and Sequences (1997)]

Longest common substring – Example

$T^{(1)} = \text{bcabcac}$, $T^{(2)} = \text{aabca}$, $T^{(3)} = \text{bcaa}$



6.4 Longest Common Extensions

Application 4: Longest Common Extensions

- We implicitly used a special case of a more general, versatile idea:

The *longest common extension (LCE)* data structure:

- **Given:** String $T[0..n)$
- **Goal:** Answer LCE queries, i. e.,
given positions i, j in T ,
how far can we read the same text from there?
formally: $\text{LCE}(i, j) = \max\{\ell : T[i..i + \ell) = T[j..j + \ell)\}$

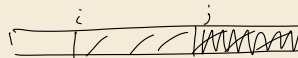


Application 4: Longest Common Extensions

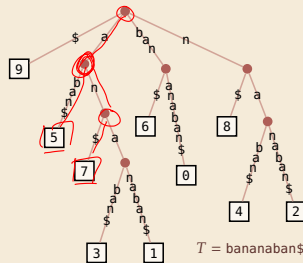
- We implicitly used a special case of a more general, versatile idea:

The *longest common extension (LCE)* data structure:

- **Given:** String $T[0..n)$
- **Goal:** Answer LCE queries, i. e.,
 given positions i, j in T ,
 how far can we read the same text from there?
 formally: $\text{LCE}(i, j) = \max\{\ell : T[i..i + \ell) = T[j..j + \ell)\}$



gban\$
an\$



$T = \text{bananaban\$}$



↪ use suffix tree of T !

- ▶ In \mathcal{T} : $\text{LCE}(i, j) = \text{LCP}(T_i, T_j) \rightsquigarrow$ same thing, different name!
 = string depth of
lowest common ancestor (LCA) of
 leaves \boxed{i} and \boxed{j}
- ▶ in short: $\text{LCE}(i, j) = \text{LCP}(T_i, T_j) = \text{stringDepth}(\text{LCA}(\boxed{i}, \boxed{j}))$



Efficient LCA

How to find lowest common ancestors?

- ▶ Could walk up the tree to find LCA $\rightsquigarrow \Theta(n)$ worst case 
- ▶ Could store all LCAs in big table $\rightsquigarrow \Theta(n^2)$ space and preprocessing 

Efficient LCA

How to find lowest common ancestors?

- ▶ Could walk up the tree to find LCA $\rightsquigarrow \Theta(n)$ worst case 🙄
- ▶ Could store all LCAs in big table $\rightsquigarrow \Theta(n^2)$ space and preprocessing 🙄



Amazing result: Can compute data structure in $\Theta(n)$ time and space that finds any LCA is **constant(!) time**.

- ▶ a bit tricky to understand
- ▶ but a theoretical breakthrough
- ▶ and useful in practice

and suffix tree construction inside ...



\rightsquigarrow for now, use $O(1)$ LCA as black box.

\rightsquigarrow After linear preprocessing (time & space), we can find **LCEs** in $O(1)$ time.

Application 5: Approximate matching

k -mismatch matching:

► **Input:** text $T[0..n)$, pattern $P[0..m)$, $k \in [0..m)$

► **Output:**

- smallest i so that $T[i..i+m)$ and P differ in at most k characters
- or NO_MATCH if there is no such i

"Hamming distance $\leq k$ "

mismatched characters



↪ searching with typos

► Assume longest common extensions in $T_{\$1}P_{\$2}$ can be found in $O(1)$

↪ generalized suffix tree \mathcal{T} has been built

↪ string depths of all internal nodes have been computed

↪ constant-time LCA data structure for \mathcal{T} has been built

Clicker Question



What is the Hamming distance between heart and beard?

2

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Click on "Polls" tab

Clicker Question



Recap: Check all correct statements about suffix tree \mathcal{T} of $T[0..n)$.

- ☐ **A** We require T to end with \$.
- ☐ **B** The size of \mathcal{T} can be $\Omega(n^2)$ in the worst case.
- ☐ **C** \mathcal{T} is a standard trie of all suffixes of $T\$$.
- ☐ **D** \mathcal{T} is a compact trie of all suffixes of $T\$$.
- ☐ **E** The leaves of \mathcal{T} store (a copy of) a suffix of $T\$$.
- ☐ **F** Naive construction of \mathcal{T} takes $\Omega(n^2)$ (worst case).
- ☐ **G** \mathcal{T} can be computed in $O(n)$ time (worst case).
- ☐ **H** \mathcal{T} has n leaves.

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Clicker Question



Recap: Check all correct statements about suffix tree \mathcal{T} of $T[0..n)$.

- ☒ **A** We require T to end with $\$$. ✓
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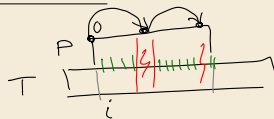
Click on "Polls" tab

Kangaroo Algorithm for approximate matching

easy in $O(n \cdot m)$



```
1 procedure kMismatch( $T[0..n-1], P[0..m-1]$ )  
2   // build LCE data structure  
3   for  $i := 0, \dots, n-m-1$  do  
4     mismatches := 0;  $t := i$ ;  $p := 0$   
5     while mismatches  $\leq k \wedge p < m$  do  
6        $\ell := \text{LCE}(t, p)$  // jump over matching part  
7        $t := t + \ell + 1$ ;  $p := p + \ell + 1$   
8       mismatches := mismatches + 1  
9     if  $p == m$  then  
10      return  $i$ 
```



► **Analysis:** $\Theta(n + m)$ preprocessing + $O(n \cdot k)$ matching

↪ very efficient for small k

► State of the art

- $O\left(n \frac{k^2 \log k}{m}\right)$ possible with complicated algorithms
- extensions for edit distance $\leq k$ possible

Application 6: Matching with wildcards

- ▶ Allow a wildcard character in pattern

stands for arbitrary (single) character

unit*	P
in_unit5_we_will	T

- ▶ similar algorithm as for k -mismatch $\rightsquigarrow O(n \cdot k + m)$ when P has k wildcards

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unit*	P
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* * *

Many more applications, in particular for problems on biological sequences

20+ described in Gusfield, *Algorithms on strings, trees, and sequences* (1999)

Suffix trees – Discussion

- ▶ Suffix trees were a threshold invention

👍 linear time and space

👍 suddenly many questions efficiently solvable in theory



Suffix trees – Discussion

► Suffix trees were a threshold invention

👍 linear time and space

👍 suddenly many questions efficiently solvable in theory

👎 construction of suffix trees:
linear time, but significant overhead

👎 construction methods fairly complicated

👎 many pointers in tree incur large space overhead

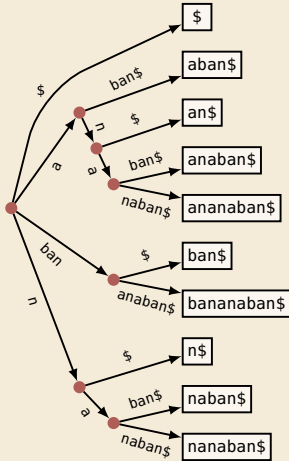


⌊ Σ = ASCII 6 = 128

6.5 Suffix Arrays

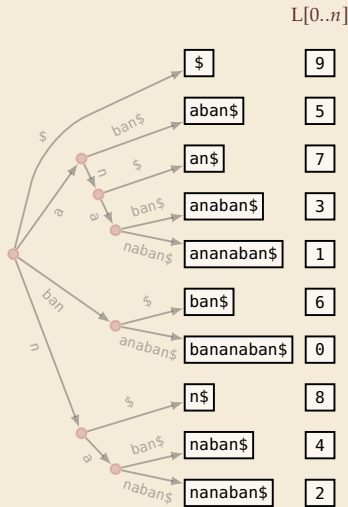
Putting suffix trees on a diet

- **Observation:** order of leaves in suffix tree
= suffixes lexicographically *sorted*



- all occurrences: 2 binary searches
- ana# # > a ∈ Σ

Putting suffix trees on a diet



► **Observation:** order of leaves in suffix tree
= suffixes lexicographically *sorted*

► Idea: only store list of leaves $L[0..n]$

► Enough to do efficient string matching!

1. Use binary search for pattern P
2. check if P is prefix of suffix after position found

► **Example:** $P = \text{ana}$

~> $L[0..n]$ is called *suffix array*:

$L[r] = (\text{start index of } r\text{th suffix in sorted order})$

► using L , can do string matching with
 $\leq (\lg n + 2) \cdot m$ character comparisons

Clicker Question



Check all correct statements about suffix array $L[0..n]$ and suffix tree \mathcal{T} of text $T[0..n]$.

- ☐ **A** $L[0..n]$ lists the start indices of leaves of \mathcal{T} in left-to-right order.
- ☐ **B** $T[L[r]..n]$ is the path label in \mathcal{T} to the leaf storing r .
- ☐ **C** $T[L[r]..n]$ is the path label to the r th leaf in \mathcal{T} .
- ☐ **D** $T_{L[r]}$ is the r th smallest suffix of T (lexicographic order).
- ☐ **E** In terms of Θ -classes, \mathcal{T} needs more space than L .
- ☐ **F** L (and T) suffice to solve the text indexing problem.

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Click on “Polls” tab

Clicker Question



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Click on "Polls" tab

Suffix arrays – Construction

How to compute $L[0..n]$?

- ▶ from suffix tree
 - ▶ possible with traversal . . .
 - 👎 but we are trying to avoid constructing suffix trees!
- ▶ sorting the suffixes of T using general purpose sorting method
 - 👍 trivial to code!
 - ▶ but: comparing two suffixes can take $\Theta(n)$ character comparisons
 - 👎 $\Theta(n^2 \log n)$ time in worst case

Suffix arrays – Construction

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 - 👍 trivial to code!
 - ▶ but: comparing two suffixes can take $\Theta(n)$ character comparisons
 - 👎 $\Theta(n^2 \log n)$ time in worst case
- ▶ We do better!

Fat-pivot radix quicksort – Example *(corrected version)*

she

sells

seashells

by

the

sea

shore

the

shells

she

sells

are

surely

seashells

Fat-pivot radix quicksort – Example

she
sells
seashells
by
the
sea
shore
the
shells
she
sells
are
surely
seashells

Fat-pivot radix quicksort – Example

she	by
sells	are
seashells	she
by	sells
the	seashells
sea	sea
shore	shore
the	shells
shells	she
she	sells
sells	surely
are	seashells
surely	the
seashells	the

Fat-pivot radix quicksort – Example

she	by	are
sells	are	by
seashells	she	
by	sells	
the	seashells	
sea	sea	
shore	shore	
the	shells	
shells	she	
she	sells	
sells	surely	
are	seashells	
surely	the	
seashells	the	

Fat-pivot radix quicksort – Example

she	by	are
sells	are	by
seashells	she	sells
by	sells	seashells
the	seashells	sea
sea	sea	sells
shore	shore	seashells
the	shells	she
shells	she	shore
she	sells	shells
sells	surely	she
are	seashells	surely
surely	the	
seashells	the	

Fat-pivot radix quicksort – Example

she	by	are
sells	are	by
seashells	she	sells
by	sells	seashells
the	seashells	sea
sea	sea	sells
shore	shore	seashells
the	shells	she
shells	she	shore
she	sells	shells
sells	surely	she
are	seashells	surely
surely	the	the
seashells	the	the

Fat-pivot radix quicksort – Example



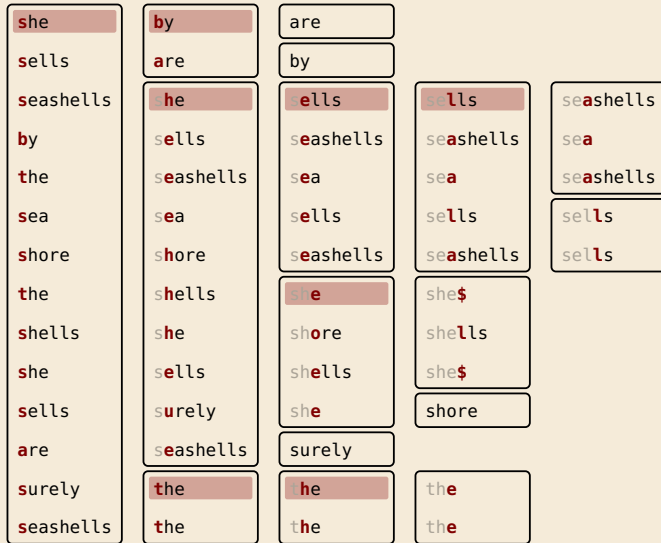
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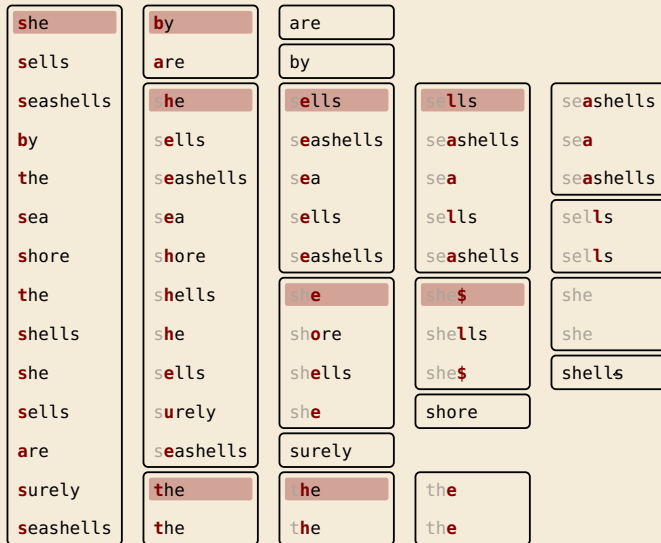
Fat-pivot radix quicksort – Example

she	by	are	
sells	are	by	
seashells	she	sells	sells
by	sells	seashells	seashells
the	seashells	sea	sea
sea	sea	sells	sells
shore	shore	seashells	seashells
the	shells	she	she\$
shells	she	shore	shells
she	sells	shells	she\$
sells	surely	she	shore
are	seashells	surely	
surely	the	the	the
seashells	the	the	the

Fat-pivot radix quicksort – Example



Fat-pivot radix quicksort – Example



Fat-pivot radix quicksort – Example



Fat-pivot radix quicksort – Example



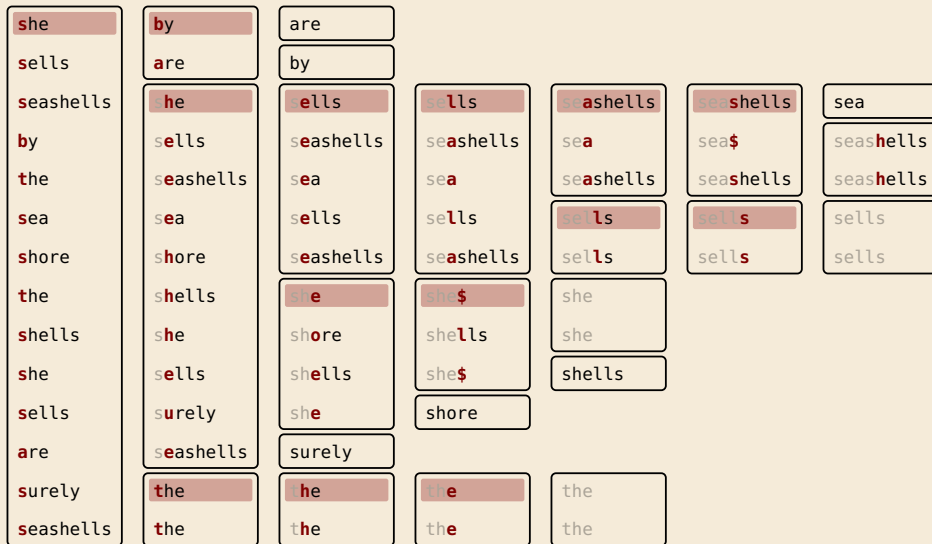
Fat-pivot radix quicksort – Example



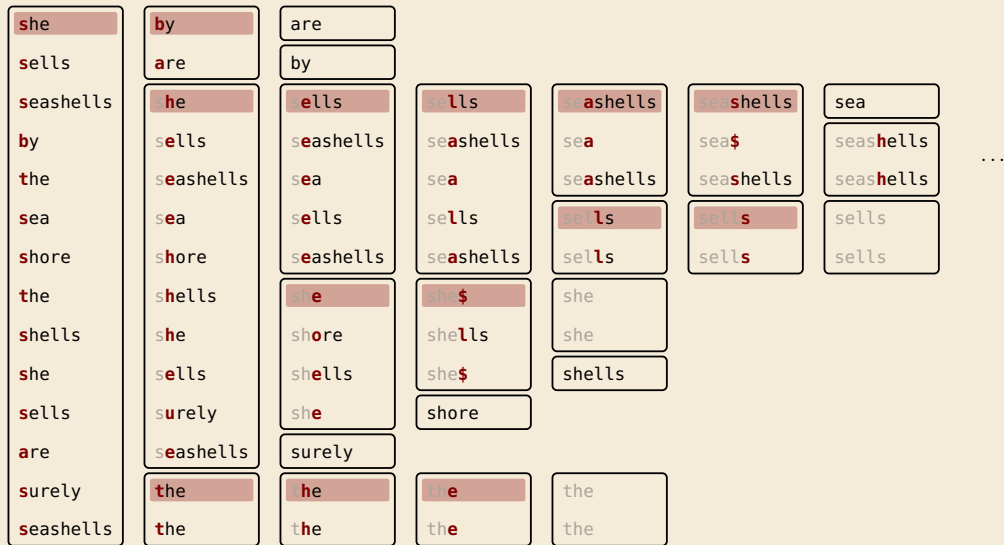
Fat-pivot radix quicksort – Example



Fat-pivot radix quicksort – Example



Fat-pivot radix quicksort – Example



Fat-pivot radix quicksort

details in §5.1 of Sedgewick, Wayne *Algorithms 4th ed.* (2011), Pearson

- ▶ **partition** based on d th character only (initially $d = 0$)
 - ↪ 3 segments: smaller, equal, or larger than d th symbol of pivot
- ▶ recurse on smaller and large with same d , on equal with $d + 1$
 - ↪ never compare equal prefixes twice

Fat-pivot radix quicksort

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↪ can show: $\sim 2 \ln(2) \cdot n \lg n \approx 1.39n \lg n$ character comparisons on average

for random strings

👍 simple to code

👍 efficient for sorting many lists of strings

random string

- ▶ fat-pivot radix quicksort finds suffix array in $O(n \log n)$ expected time

Fat-pivot radix quicksort

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for random strings

👍 simple to code

choice of pivot
and random strings

👍 efficient for sorting many lists of strings

random string

- ▶ fat-pivot radix quicksort finds suffix array in $O(n \log n)$ expected time

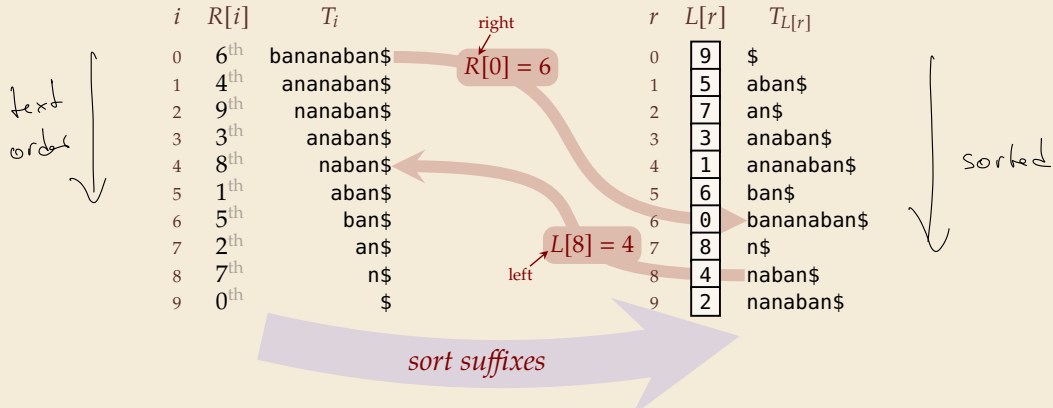
but we can do $O(n)$ time *worst case*!

6.6 Linear-Time Suffix Sorting

Inverse suffix array: going left & right

► to understand the fastest algorithm, it is helpful to define the *inverse suffix array*:

- $R[i] = r \iff L[r] = i$ $L = \text{leaf array}$
 - \iff there are r suffixes that come before T_i in sorted order
 - $\iff T_i$ has (0-based) *rank* $r \rightsquigarrow$ call $R[0..n]$ the rank array



Linear-time suffix sorting

DC3 / Skew algorithm

1. Compute rank array $R_{1,2}$ for suffixes T_i starting at $i \not\equiv 0 \pmod{3}$ *recursively*. *not a multiple of 3*
↓ $\frac{2}{3} \text{ } \simeq$
2. Induce rank array R_3 for suffixes $T_0, T_3, T_6, T_9, \dots$ from $R_{1,2}$.
3. Merge $R_{1,2}$ and R_0 using $R_{1,2}$.
 \rightsquigarrow rank array R for entire input

Linear-time suffix sorting

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3. Merge $R_{1,2}$ and R_3 using $R_{1,2}$.
 \rightsquigarrow rank array R for entire input

► We will show that steps 2. and 3. take $\Theta(n)$ time .

\rightsquigarrow Total complexity is $\underbrace{n}_1 + \underbrace{\frac{2}{3}n}_{\text{second level}} + \left(\frac{2}{3}\right)^2 n + \left(\frac{2}{3}\right)^3 n + \dots \leq n \cdot \sum_{i \geq 0} \left(\frac{2}{3}\right)^i = 3n = \underline{\Theta(n)}$

Linear-time suffix sorting

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\rightsquigarrow Total complexity is $n + \frac{2}{3}n + \left(\frac{2}{3}\right)^2 n + \left(\frac{2}{3}\right)^3 n + \dots \leq n \cdot \sum_{i \geq 0} \left(\frac{2}{3}\right)^i = 3n = \Theta(n)$

► **Note:** L can easily be computed from R in one pass, and vice versa.

\rightsquigarrow Can use whichever is more convenient.

DC3 / Skew algorithm – Step 2: Inducing ranks

- ▶ **Assume:** rank array $R_{1,2}$ known:

$$\text{▶ } R_{1,2}[i] = \begin{cases} \text{rank of } T_i \text{ among } T_1, T_2, T_4, T_5, T_7, T_8, \dots & \text{for } i = 1, 2, 4, 5, 7, 8, \dots \\ \text{undefined} & \text{for } i = 0, 3, 6, 9, \dots \end{cases}$$

- ▶ **Task:** sort the suffixes $T_0, T_3, T_6, T_9, \dots$ in linear time (!)

DC3 / Skew algorithm – Step 2: Inducing ranks

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- **Task:** sort the suffixes $T_0, T_3, T_6, T_9, \dots$ in linear time (!)

- Suppose we want to compare T_0 and T_3 .

$$T_0 = a \overline{T_1}$$

$$T_3 = c \overline{T_4}$$

- Characterwise comparisons too expensive
- but: after removing first character, we obtain T_1 and T_4
- these two can be compared in *constant time* by comparing $\underline{R_{1,2}[1]}$ and $\underline{R_{1,2}[4]}$!

\rightsquigarrow T_0 comes before T_3 in lexicographic order
iff pair $(T[0], R_{1,2}[1])$ comes before pair $(T[3], R_{1,2}[4])$ in lexicographic order

DC3 / Skew algorithm – Inducing ranks example

$T = \text{hannahbansbananasman}\$ \$ \$$

(append 3 \$ markers)

T_0 hannahbansbananasman\$\$\$
 T_3 nahbansbananasman\$\$\$
 T_6 bansbananasman\$\$\$
 T_9 sbanasman\$\$\$
 T_{12} nanasman\$\$\$
 T_{15} asman\$\$\$
 T_{18} an\$\$\$
 T_{21} \$\$

T_1 annahbansbananasman\$\$\$	$R_{1,2}[22] = 0$	T_{22} \$
T_2 nnahbansbananasman\$\$\$	$R_{1,2}[20] = 1$	T_{20} \$\$\$
T_4 ahbansbananasman\$\$\$	$R_{1,2}[4] = 2$	T_4 ahbansbananasman\$\$\$
T_5 hbansbananasman\$\$\$	$R_{1,2}[11] = 3$	T_{11} ananasman\$\$\$
T_7 ansbananasman\$\$\$	$R_{1,2}[13] = 4$	T_{13} anasman\$\$\$
T_8 nsbananasman\$\$\$	$R_{1,2}[1] = 5$	T_1 annahbansbananasman\$\$\$
T_{10} bananasman\$\$\$	$R_{1,2}[7] = 6$	T_7 ansbananasman\$\$\$
T_{11} ananasman\$\$\$	$R_{1,2}[10] = 7$	T_{10} bananasman\$\$\$
T_{13} anasman\$\$\$	$R_{1,2}[5] = 8$	T_5 hbansbananasman\$\$\$
T_{14} nasman\$\$\$	$R_{1,2}[17] = 9$	T_{17} man\$\$\$
T_{16} sman\$\$\$	$R_{1,2}[19] = 10$	T_{19} n\$\$\$
T_{17} man\$\$\$	$R_{1,2}[14] = 11$	T_{14} nasman\$\$\$
T_{19} n\$\$\$	$R_{1,2}[2] = 12$	T_2 nnahbansbananasman\$\$\$
T_{20} \$\$\$	$R_{1,2}[8] = 13$	T_8 nsbananasman\$\$\$
T_{22} \$	$R_{1,2}[16] = 14$	T_{16} sman\$\$\$

$R_{1,2}$ (known)

DC3 / Skew algorithm – Inducing ranks example

$T = \text{hannahbansbananasman}\$ \$ \$$

(append 3 \$ markers)

T_0 hannahbansbananasman\$\$\$
 T_3 nahbansbananasman\$\$\$
 T_6 bansbananasman\$\$\$
 T_9 sbanasman\$\$\$
 T_{12} nanasman\$\$\$
 T_{15} asman\$\$\$
 T_{18} an\$\$\$
 T_{21} \$\$

$\text{sman}\$ \$ \$ = T_{16}$

T_0 h05
 T_3 n02
 T_6 b06
 T_9 s07
 T_{12} n04
 T_{15} a14
 T_{18} a10
 T_{21} \$00

$R_{1,2}[16] = 14$

T_1 annahbansbananasman\$\$\$
 T_2 nnahbansbananasman\$\$\$
 T_4 ahbansbananasman\$\$\$
 T_5 hbansbananasman\$\$\$
 T_7 ansbananasman\$\$\$
 T_8 nsbananasman\$\$\$
 T_{10} bananasman\$\$\$
 T_{11} ananasman\$\$\$
 T_{13} anasman\$\$\$
 T_{14} nasman\$\$\$
 T_{16} sman\$\$\$
 T_{17} man\$\$\$
 T_{19} n\$\$\$
 T_{20} \$\$\$
 T_{22} \$

$R_{1,2}[22] = 0$ T_{22} \$
 $R_{1,2}[20] = 1$ T_{20} \$\$\$
 $R_{1,2}[4] = 2$ T_4 ahbansbananasman\$\$\$
 $R_{1,2}[11] = 3$ T_{11} ananasman\$\$\$
 $R_{1,2}[13] = 4$ T_{13} anasman\$\$\$
 $R_{1,2}[1] = 5$ T_1 annahbansbananasman\$\$\$
 $R_{1,2}[7] = 6$ T_7 ansbananasman\$\$\$
 $R_{1,2}[10] = 7$ T_{10} bananasman\$\$\$
 $R_{1,2}[5] = 8$ T_5 hbansbananasman\$\$\$
 $R_{1,2}[17] = 9$ T_{17} man\$\$\$
 $R_{1,2}[19] = 10$ T_{19} n\$\$\$
 $R_{1,2}[14] = 11$ T_{14} nasman\$\$\$
 $R_{1,2}[2] = 12$ T_2 nnahbansbananasman\$\$\$
 $R_{1,2}[8] = 13$ T_8 nsbananasman\$\$\$
 $R_{1,2}[16] = 14$ T_{16} sman\$\$\$

$R_{1,2}$ (known)

DC3 / Skew algorithm – Inducing ranks example

$T = \text{hannahbansbananasman}\$ \$ \$$

(append 3 \$ markers)

T_0 hannahbansbananasman\$\$\$
 T_3 nahbansbananasman\$\$\$
 T_6 bansbananasman\$\$\$
 T_9 sbanasman\$\$\$
 T_{12} nanasman\$\$\$
 T_{15} asman\$\$\$
 T_{18} an\$\$\$
 T_{21} \$\$

$\text{smans} = T_{16}$

T_0 h05
 T_3 n02
 T_6 b06
 T_9 s07
 T_{12} n04
 T_{15} a14
 T_{18} a10
 T_{21} \$00

$R_{1,2}[16] = 14$

T_1 annahbansbananasman\$\$\$
 T_2 nnahbansbananasman\$\$\$
 T_4 ahbansbananasman\$\$\$
 T_5 hbansbananasman\$\$\$
 T_7 ansbananasman\$\$\$
 T_8 nsbananasman\$\$\$
 T_{10} bananasman\$\$\$
 T_{11} ananasman\$\$\$
 T_{13} anasman\$\$\$
 T_{14} nasman\$\$\$
 T_{16} sman\$\$\$
 T_{17} mans\$\$\$
 T_{19} n\$\$\$
 T_{20} \$\$\$
 T_{22} \$

$R_{1,2}[22] = 0$ T_{22} \$
 $R_{1,2}[20] = 1$ T_{20} \$\$\$
 $R_{1,2}[4] = 2$ T_4 ahbansbananasman\$\$\$
 $R_{1,2}[11] = 3$ T_{11} ananasman\$\$\$
 $R_{1,2}[13] = 4$ T_{13} anasman\$\$\$
 $R_{1,2}[1] = 5$ T_1 annahbansbananasman\$\$\$
 $R_{1,2}[7] = 6$ T_7 ansbananasman\$\$\$
 $R_{1,2}[10] = 7$ T_{10} bananasman\$\$\$
 $R_{1,2}[5] = 8$ T_5 hbansbananasman\$\$\$
 $R_{1,2}[17] = 9$ T_{17} mans\$\$\$
 $R_{1,2}[19] = 10$ T_{19} n\$\$\$
 $R_{1,2}[14] = 11$ T_{14} nasman\$\$\$
 $R_{1,2}[2] = 12$ T_2 nnahbansbananasman\$\$\$
 $R_{1,2}[8] = 13$ T_8 nsbananasman\$\$\$
 $R_{1,2}[16] = 14$ T_{16} sman\$\$\$

$R_{1,2}$ (known)

radix sort

T_{21} \$00 $\rightsquigarrow R_0[21] = 0$
 T_{18} a10 $\rightsquigarrow R_0[18] = 1$
 T_{15} a14 $\rightsquigarrow R_0[15] = 2$
 T_6 b06 $\rightsquigarrow R_0[6] = 3$
 T_0 h05 $\rightsquigarrow R_0[0] = 4$
 T_3 n02 $\rightsquigarrow R_0[3] = 5$
 T_{12} n04 $\rightsquigarrow R_0[12] = 6$
 T_9 s07 $\rightsquigarrow R_0[9] = 7$

DC3 / Skew algorithm – Inducing ranks example

$T = \text{hannahbansbananasman}\$ \$ \$$

(append 3 \$ markers)

T_0 hannahbansbananasman\$\$\$
 T_3 nahbansbananasman\$\$\$
 T_6 bansbananasman\$\$\$
 T_9 sbanasman\$\$\$
 T_{12} nanasman\$\$\$
 T_{15} asman\$\$\$
 T_{18} an\$\$\$
 T_{21} \$\$

$\text{smans} = T_{16}$

T_0 h05
 T_3 n02
 T_6 b06
 T_9 s07
 T_{12} n04
 T_{15} a14
 T_{18} a10
 T_{21} \$00

$R_{1,2}[16] = 14$

T_1 annahbansbananasman\$\$\$
 T_2 nnahbansbananasman\$\$\$
 T_4 ahbansbananasman\$\$\$
 T_5 hbansbananasman\$\$\$
 T_7 ansbananasman\$\$\$
 T_8 nsbananasman\$\$\$
 T_{10} bananasman\$\$\$
 T_{11} ananasman\$\$\$
 T_{13} anasman\$\$\$
 T_{14} nasman\$\$\$
 T_{16} sman\$\$\$
 T_{17} mans\$\$\$
 T_{19} n\$\$\$
 T_{20} \$\$\$
 T_{22} \$

$R_{1,2}[22] = 0$
 $R_{1,2}[20] = 1$
 $R_{1,2}[4] = 2$
 $R_{1,2}[11] = 3$
 $R_{1,2}[13] = 4$
 $R_{1,2}[1] = 5$
 $R_{1,2}[7] = 6$
 $R_{1,2}[10] = 7$
 $R_{1,2}[5] = 8$
 $R_{1,2}[17] = 9$
 $R_{1,2}[19] = 10$
 $R_{1,2}[14] = 11$
 $R_{1,2}[2] = 12$
 $R_{1,2}[8] = 13$
 $R_{1,2}[16] = 14$

T_{22} \$
 T_{20} \$\$\$
 T_4 ahbansbananasman\$\$\$
 T_{11} ananasman\$\$\$
 T_{13} anasman\$\$\$
 T_1 annahbansbananasman\$\$\$
 T_7 ansbananasman\$\$\$
 T_{10} bananasman\$\$\$
 T_5 hbansbananasman\$\$\$
 T_{17} mans\$\$\$
 T_{19} n\$\$\$
 T_{14} nasman\$\$\$
 T_2 nnahbansbananasman\$\$\$
 T_8 nsbananasman\$\$\$
 T_{16} sman\$\$\$

$R_{1,2}$ (known)

radix sort

T_{21} \$00 $\rightsquigarrow R_0[21] = 0$
 T_{18} a10 $\rightsquigarrow R_0[18] = 1$
 T_{15} a14 $\rightsquigarrow R_0[15] = 2$
 T_6 b06 $\rightsquigarrow R_0[6] = 3$
 T_0 h05 $\rightsquigarrow R_0[0] = 4$
 T_3 n02 $\rightsquigarrow R_0[3] = 5$
 T_{12} n04 $\rightsquigarrow R_0[12] = 6$
 T_9 s07 $\rightsquigarrow R_0[9] = 7$
 R_0

DC3 / Skew algorithm – Inducing ranks example

$T = \text{hannahbansbananasman}\$ \$ \$$

(append 3 \$ markers)

T_0 hannahbansbananasman\$\$\$
 T_3 nahbansbananasman\$\$\$
 T_6 bansbananasman\$\$\$
 T_9 sbanasman\$\$\$
 T_{12} nanasman\$\$\$
 T_{15} asman\$\$\$
 T_{18} an\$\$\$
 T_{21} \$\$

$\text{smans} = T_{16}$

T_0 h05
 T_3 n02
 T_6 b06
 T_9 s07
 T_{12} n04
 T_{15} a14
 T_{18} a10
 T_{21} \$00

$R_{1,2}[16] = 14$

T_1 annahbansbananasman\$\$\$
 T_2 nnahbansbananasman\$\$\$
 T_4 ahbansbananasman\$\$\$
 T_5 hbansbananasman\$\$\$
 T_7 ansbananasman\$\$\$
 T_8 nsbananasman\$\$\$
 T_{10} bananasman\$\$\$
 T_{11} ananasman\$\$\$
 T_{13} anasman\$\$\$
 T_{14} nasman\$\$\$
 T_{16} sman\$\$\$
 T_{17} man\$\$\$
 T_{19} n\$\$\$
 T_{20} \$\$\$
 T_{22} \$

$R_{1,2}[22] = 0$ T_{22} \$
 $R_{1,2}[20] = 1$ T_{20} \$\$\$
 $R_{1,2}[4] = 2$ T_4 ahbansbananasman\$\$\$
 $R_{1,2}[11] = 3$ T_{11} ananasman\$\$\$
 $R_{1,2}[13] = 4$ T_{13} anasman\$\$\$
 $R_{1,2}[1] = 5$ T_1 annahbansbananasman\$\$\$
 $R_{1,2}[7] = 6$ T_7 ansbananasman\$\$\$
 $R_{1,2}[10] = 7$ T_{10} bananasman\$\$\$
 $R_{1,2}[5] = 8$ T_5 hbansbananasman\$\$\$
 $R_{1,2}[17] = 9$ T_{17} man\$\$\$
 $R_{1,2}[19] = 10$ T_{19} n\$\$\$
 $R_{1,2}[14] = 11$ T_{14} nasman\$\$\$
 $R_{1,2}[2] = 12$ T_2 nnahbansbananasman\$\$\$
 $R_{1,2}[8] = 13$ T_8 nsbananasman\$\$\$
 $R_{1,2}[16] = 14$ T_{16} sman\$\$\$

$R_{1,2}$ (known)

radix sort

T_{21} \$00 $\rightsquigarrow R_0[21] = 0$
 T_{18} a10 $\rightsquigarrow R_0[18] = 1$
 T_{15} a14 $\rightsquigarrow R_0[15] = 2$
 T_6 b06 $\rightsquigarrow R_0[6] = 3$
 T_0 h05 $\rightsquigarrow R_0[0] = 4$
 T_3 n02 $\rightsquigarrow R_0[3] = 5$
 T_{12} n04 $\rightsquigarrow R_0[12] = 6$
 T_9 s07 $\rightsquigarrow R_0[9] = 7$
 R_0

► sorting of pairs doable in $O(n)$ time
by 2 iterations of counting sort

\rightsquigarrow Obtain R_0 in $O(n)$ time