Strings SFWRENG 2CO3: Data Structures and Algorithms

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Strings over alphabets

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Examples

- A typical string over the roman alphabet $\{a', \ldots, z', \cdot'\}$: "hello", "hello␣world", "strings␣over␣alphabets", and "".
- \triangleright A *bit string* is a sequence over $\{0, 1\}$: "0011", "1101100", "", and "0".
- \blacktriangleright A DNA String is a sequence over $\{A, C, G, T\}$: "", "AACATG", "AGT", and "AAACCCAAATTT".
- \triangleright A Unicode string is a sequence over the unicode code points (149 186 symbols and counting).
- \blacktriangleright A byte string is a sequence over bytes.

Operations on strings and alphabets

We assume the following basic operations:

- \triangleright We can sequentially iterate over the symbols in a string.
- \triangleright We can look up the *i*-th symbol in a string in $\Theta(1)$. (This can be hard in some practical settings: UTF-8 and UTF-16 strings do not support this).
- \triangleright We assume that each alphabet $\mathcal A$ is an ordered list L of symbols.
- **►** For each $\sigma \in \mathcal{A}$, we can determine its position in L.

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- 3: buckets[v] := buckets[v] + 1.
- 4: $k := 0$
- 5: for all $i := 0$ upto $M 1$ do
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Algorithm $GBUCKETSORT(L, r)$:

- 1: buckets := $[[] | 0 \le i \le M 1].$
- 2: for all $v \in I$ do
- 3: Append v to buckets $[r(v)]$.
- 4: $k := 0$
- 5: for all $i := 0$ upto $M 1$ do
- 6: for all $j := 0$ upto $|buckets[i]|$ do
- 7: $L[k] = \text{buckets}[i][j].$
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Generalization

Assume we have values that "represent" $0, \ldots, M-1$ via some function r.

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Notice that GBUCKETSORT is *stable*.

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Algorithm RADIXSORT (L) :

- 1: for $d := k 1$ downto 0 do
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3: GBUCKETSORT(*L*, *r_d*) with *r_d*(*S*) = *S*[*d*].

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$$
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Correctness Invariant: In L, the suffix of the last $k - (d + 1)$ symbols is sorted.

Generalization: strings with variable lengths up-to-k Let S be a string of length $|S| < k$. Interpret $S[|S|], \ldots, S[k-1]$ as symbols that come before all other symbols.

The book calls this least-significant-digit string sort.

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- \triangleright Comparing two strings of length k costs at-most $\Theta(k)$.
- **►** For $|L|$ random strings, comparisons are expected to cost $\Theta(\log_2(|L|)).$
- $\Theta(k(|L| + |\mathcal{A}|))$ versus $\Theta(k|L|\log(|L|))$ (or $\Theta(|L|\log^2(|L|))$ expected).

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We will apply this idea *recursively!*

Algorithm $MSD-SORT(L, d)$:

- 1: if $d < k$ and $|L| > 1$ then
- 2: GBUCKETSORT (L, r_d) with $r_d(S) = S[d]$, during which we further sort each individual bucket separately.

$\mathsf{Algorithm} \; \mathsf{MSD-SORT}(L, d)$:

- 1: if $d < k$ and $|L| > 1$ then
- 2: *buckets* := $[[] | 0 \le i \le |\mathcal{A}| 1].$
- 3: for all $v \in L$ do
- 4: Append v to buckets [v[d]].
- 5: $k := 0$
- 6: for all $i := 0$ upto $|\mathcal{A}| 1$ do
- 7: $k_{\text{start}} := k$.
- 8: **for all** $j := 0$ upto $|buckets[i]|$ **do**
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Finetuning

We end up with many arrays buckets that each hold $|\mathcal{A}|$ lists!

Algorithm $MSD-SORT(L, d)$:

- 1: if $|L| \leq |\mathcal{A}|$ then
- 2: Use another algorithm to sort L.
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Complexity At-most $\Theta(k(|L| + |\mathcal{A}|)).$

Sorting: Best practices

So which sort algorithm is the best?

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Often, your standard sort algorithm will be sufficient.

Assumption We have an alphabet \mathcal{A} with $M = |\mathcal{A}|$ symbols.

A Trie is a set representation that can hold strings over $\mathcal A$ such that:

- \triangleright strings of length N can be *added* in $\Theta(N)$;
- \triangleright strings of length N can be removed in $\Theta(N)$;
- \triangleright checking whether a string of length N is in the set costs $\Theta(N)$;
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We have seen tries with $\mathcal{A} = \{0, 1\} \rightarrow \text{BSSET}$ in Example Assignment 3.

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Each node *n* in a *trie* T over $\mathcal{A} = {\sigma_1, \ldots, \sigma_M}$ has

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Adding a string

- ▶ Follow-or-make a path according to the string symbols.
- \triangleright Set *n.end* on the last node *n* on this path.

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Removing a string

- \blacktriangleright Follow a path according to the string symbols to node *n* and unset *n.end.*
- Remove n if n has no children.
- Recurse to the ancestors m: remove m if m has no children and m.end is unset.

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Printing all strings in-order

Perform a pre-order traversal starting at the root. For each node n:

- \triangleright print the path from root to node *n* if *n.end* is set;
- ▶ pre-order traverse all children in-order of alphabet symbols.

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Printing all strings in-order with prefix W

- \blacktriangleright Follow a path according to the string symbols of W to node m.
- \blacktriangleright Perform a pre-order traversal starting at the node *m*.

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Finetuning

 \blacktriangleright To deal with big alphabets:

use a dictionary with A -symbols as keys at each node to store all edges.

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use a dictionary with A -symbols as keys at each node to store all edges.

▶ To compress non-branching paths: nodes can represent strings of symbols.

Input

Lossless compression: The input must be equivalent to the output!

Theorem No algorithm A can compress every input I.

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Conceptually: We need structure in the input to be able to reliably compress that input!

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An usual DNA string S represented by N characters takes up $NB = 8Nbit$.

How many bytes do we need to represent S ?

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From bits to bytes

We can store \widetilde{f} our DNA characters in one byte. Can we store $\mathcal S$ in $\frac{2N}{8}$ B using the above?

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From bits to bytes

We can store \widetilde{f} our DNA characters in one byte. Can we store $\mathcal S$ in $\frac{2N}{8}$ B using the above?

No! Where in the last byte would our string end? E.g., "ACTGA" takes 10 bit (1.25 B).

A common structure: Repetition

Consider the following string of bits:

00000000000000011111110000000011111111111000000000011

A common structure: Repetition

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Consider the following string of bits:

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s 1s 0s 1s 0s 1s

From $15 + 7 + 8 + 11 + 10 + 2 = 53$ bit to $6 \cdot 4 = 24$ bit.

Consider the following string of bits:

Consider the following string of bits:

From $17 + 7 + 8 + 11 + 10 + 2 = 55$ bit to $8 \cdot 4 = 32$ bit.

Consider the following string of bits:

Run-length encoding: simple idea with good results on bitmaps.

Consider simple text written in the English language.

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- \triangleright Digits 0123456789: 10 symbols.
- ▶ Lower-case letters "a"-"z": 26 symbols.
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Stored normally, *each* symbol occupies $1 B = 8 bit$.

Even in these "frequent" symbols, some are much rarer than others: "x" versus "e". Idea. Use fewer bits for frequent characters, more for rare characters.

Consider the string "anna can scan a can!".

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The string has 6 distinct symbols: at-least 3 bits if all the same length.

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anna can scan a can! 0 1 1 0 00 01 0 1 00 10 01 0 1 00 0 00 01 0 1 11

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anna can scan a can! 0 1 1 0 00 01 0 1 00 10 01 0 1 00 0 00 01 0 1 11 01100001010010010100000010111

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01100001010010010100000010111 ← 29 bit *instead of at-least* 60 bit.

Issue. The bit pattern of one symbol (e.g., a) is a prefix of other symbols!

Consider the string "anna can scan a can!".

Attempt 2. The most-frequent symbols get the shortest prefix-free bit patterns.

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Attempt 2. The most-frequent symbols get the shortest *prefix-free* bit patterns.

anna can scan a can! 0 10 10 0 110 1110 0 10 110 11110 1110 0 10 110 0 110 1110 0 10 11111

Consider the string "anna can scan a can!".

Attempt 2. The most-frequent symbols get the shortest *prefix-free* bit patterns.

anna can scan a can! 0 10 10 0 110 1110 0 10 110 11110 1110 0 10 110 0 110 1110 0 10 11111 01010011011100101101111011100101100110111001011111 ← 50 bit.

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 $6 \cdot 1 + 5 \cdot 2 + 4 \cdot 3 + 3 \cdot 4 + 1 \cdot 5 + 1 \cdot 5 = 50$.

Consider the string "anna can scan a can!".

Questions. How to construct the bit patterns and are these patterns optimal?

 $6 \cdot 2 + 5 \cdot 2 + 4 \cdot 2 + 3 \cdot 3 + 1 \cdot 4 + 1 \cdot 4 = 47$

Problem

Given an alphabet A and symbol-frequencies $f : \mathcal{A} \rightarrow [0, 1]$,

Produce prefix-free bit patterns for all symbols in $\mathcal A$ such that these patterns are optimal.

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Algorithm $HUFFMANPFTRIE(f)$:

- 1: $Q := a$ min-priority queue.
- 2: for all $\sigma \in \mathcal{A}$ do
- 3: Make a leaf-node *n* labeled σ .
- 4: Add $(n, f(\sigma))$ to Q with priority $f(\sigma)$.

5: while $|Q| \ge 2$ do

- 6: $(n_0, p_0) := \text{DELMIN}(Q), (n_1, p_1) := \text{DELMIN}(Q).$
- 7: Create a node *n* with children n_0 labeled 0, n_1 labeled 1.
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Let $\sigma_1, \sigma_2 \in \mathcal{A}$ be symbols represented by children n_0, n_1 of node *n* in trie T. Let T' be the prefix-free trie for \mathcal{A}', f' with

- $\blacktriangleright \mathcal{A}' = \mathcal{A} \setminus \{\sigma_2\};$
- ► $f' = {\sigma \mapsto f(\sigma) | \sigma \in \mathcal{A} \setminus \{\sigma_1, \sigma_2\}} \cup {\sigma_1 \mapsto f(\sigma_1) + f(\sigma_2)};$ and

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The trie T is optimal for \mathcal{A}, f if and only if T' is optimal for \mathcal{A}', f' .

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The HUFFMANPFTRIE algorithm builds an optimal prefix-free code.

Proof (sketch) HUFFMANPFTRIE follows Property 1-3.

Beyond Huffman: Frequent strings

Huffman looks at frequent symbols from an alphabet.

- \triangleright We can generalize these ideas to *frequent* sequences of symbols.
- ▶ Tries can be used to efficiently manage *frequency* data for substrings in an input.
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Many variations of this idea used in practice, e.g., .zip, .gif,

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Given strings S (the haystack) and P (the needle or pattern), return the first position in S at which P occurs (if any).

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Searching
$$
P = \text{``example''}
$$

\n" a n e x a m p l e of word s"

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Searching
$$
P = \n\begin{cases} \n\text{er}_1 & \text{if } P = \n\end{cases}
$$
\n

\nFor example, $P = \n\begin{cases} \n\text{er}_2 & \text{if } P = \n\end{cases}$ \n

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Searching
$$
P = \text{``great''}
$$

\n"a n e x am p le of word s"

\ng re a t

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Searching
$$
P = \frac{w}{x}
$$
 and $P = \frac{w}{x}$ and $P = x$ and $P = x$

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Searching
$$
P = \n\begin{array}{c}\n\text{``great''} \\
\text{`` a n} \\
\text{`` a n} \\
\text{`` a n} \\
\text{`` a n} \\
\text{`` a m p l e} \\
\text{``
$$

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Searching
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Searching
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Searching
$$
P = \frac{w}{s}
$$
 and $P = \frac{w}{s}$ is the same as $P = \frac{w}{s}$.

Problem

Given strings S (the haystack) and P (the needle or pattern), return the first position in S at which P occurs (if any).

Algorithm BASICSTRINGSEARCH (S, P) :

- 1: for $i := 0$ upto $|S| |P|$ do
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Complexity

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Given strings S (the haystack) and P (the needle or pattern), return the first position in S at which P occurs (if any).

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- 1: for $i := 0$ upto $|S| |P|$ do
- 2: if MATCHSTRING (S, P, i) then
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Complexity

Algorithm MATCHSTRING (S, P, i) :

- 4: for $j := 0$ upto $|P| 1$ do
- 5: if $S[i + i] \neq P[i]$ then

6: return false.

7: return true.

Problem

Given strings S (the haystack) and P (the needle or pattern), return the first position in S at which P occurs (if any).

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Complexity

Algorithm MATCHSTRING (S, P, i) : 4: for $j := 0$ upto $|P| - 1$ do 5: if $S[i+j] \neq P[j]$ then 6: **return** false. $\Theta(|P|)$ 7: return true.

Problem

Given strings S (the haystack) and P (the needle or pattern), return the first position in S at which P occurs (if any).

Algorithm BASICSTRINGSEARCH (S, P) :

- 1: for $i := 0$ upto $|S| |P|$ do
- 2: if MATCHSTRING (S, P, i) then $\Theta\left(\left(|\mathcal{S}|-|P|\right)|P|\right)$
- $3:$ return i .

Complexity

 $\Theta((|S| - |P|)|P|) = \Theta(|S||P|)$

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$$
3: \qquad \textbf{return } i.
$$

Complexity

Searching
$$
P =
$$
 "string"

\n"a string" is through the following:

\n"a string is through the following:

\n[Insert the image is the image is the image is the provided HTML.

Searching
$$
P =
$$
 "string"

\n" a $s \, t \, r \, o \, n \, g \, s \, t \, r \, i \, n \, g$

\n" a $s \, t \, r \, o \, n \, g \, s \, t \, r \, i \, n \, g$

Searching
$$
P =
$$
 "string"

\n" a $s \, t \, r \, o \, n \, g \, s \, t \, r \, i \, n \, g$

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Searching
$$
P =
$$
 "string"

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\n"a $s \, t \, r \, i \, n \, g$

Searching
$$
P =
$$
 "string"

\n"a $s \, t \, r \, o \, n \, g \, s \, t \, r \, i \, n \, g$

\n"a $s \, t \, r \, i \, n \, g$
Searching
$$
P =
$$
 "string"

\n"a $s \, t \, r \, o \, n \, g \, s \, t \, r \, i \, n \, g$

\n"a $s \, t \, r \, i \, n \, g$

Searching
$$
P =
$$
 "string"

\n"a $s \text{ t} \cap n g$ s $t \cap i n g$ "

\ns $t \cap i n g$

Searching
$$
P =
$$
 "string"

\n"a $s \, t \, r \, o \, n \, g \, s \, t \, r \, i \, n \, g$

\n"a $s \, t \, r \, i \, n \, g$

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$$
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Searching
$$
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\n"a $s \, t \, r \, o \, n \, g \, s \, t \, r \, i \, n \, g$

\n"a $s \, t \, r \, i \, n \, g$

Searching
$$
P =
$$
 "string"

\n"a $s \, t \, r \, o \, n \, g \, s \, t \, r \, i \, n \, g$

\n"b $s \, t \, r \, i \, n \, g$

Searching
$$
P =
$$
 "string"

\n"a string is through the following equation:

\na $f(x) = f(x)$

\nii $f(x) = f(x)$

\niii $f(x) = f(x)$

\niv If $f(x) = f(x)$

\nvi if $f(x) = f(x)$

\n

Searching
$$
P =
$$
 "string"

\n"a $stron g$ string" $strin g$ "

\n"b $stron g$ string

$$
Searching \nP = "ACACGT" \n "A C A C A C A C G T"
$$

Searching
$$
P = \text{``ACACGT''}
$$

\n'' A C A C A C A C G T

\nA C A C G T

Searching
$$
P = \text{``ACACGT''}
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\n'' A C A C A C A C G T

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Finite automata

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Typically, we refer to nodes as states and edges as transitions.

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We can use a string as *input* to the automaton to decide which path to follow. For efficiency: we want a *deterministic* automaton: an automaton without choices!

 $18/20$

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Complexity of running a *deterministic* finite automaton with input S

- ▶ Always in exactly one state.
- \triangleright We perform at-most $\Theta(|\mathcal{S}|)$ state transitions in the automaton.
- Need efficient representation of the *transitions* (per state): e.g., hash table.

19/20

A regular expression describes a set of strings.

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Now let e, e_1 , e_2 be regular expressions describing sets R, R₁, R₂.

- \blacktriangleright (e) desribes R.
- \triangleright e₁e₂ describes {CONCATENATE(S₁, S₂) | S₁ ∈ R₁ ∧ S₂ ∈ R₂}.
- \triangleright e₁ | e₂ describes $R_1 \cup R_2$.
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Examples

moose | mouse

sub∗ section

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Claim: Every regular expression is equivalent to a deterministic finite automaton See SFWRENG 2FA3: Discrete Mathematics with Applications II.

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Idea. Keep track of the *set of states* we can be in while walking to the string.

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- \blacktriangleright Automata-based searching in string can be very fast: core component in lexers, parsers, and compilers.
- \triangleright Many RegExp libraries are regular expression like: they support non-regular features. Consequently, many such libraries use shamefully slow backtracking algorithms: Worst-case exponential complexity, even when searching for simple patterns.

Searching $P = "great"$ " a n e x a m p l e o f w o r d s " g r e a t

How can I skip checking most letters in S ?

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How can I skip checking most letters in \mathcal{S} ? Assume we checked up-till position i.

Observation. If we compare the last symbol from our pattern with $S[i + |P| - 1]$, then

 \triangleright We see a symbol that is not even in *P*: *P* cannot occur in S[i... i + |*P*|).

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When it does not matches: another opportunity to jump!

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How can I skip checking most letters in S ?

With some preprocessing on P one can precompute how to jump around optimally: the Boyer-Moore algorithm.