

4

String Matching – What's behind Ctrl+F?

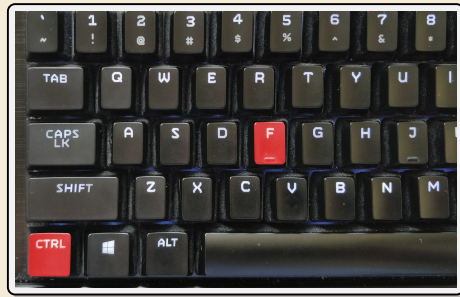
20 October 2023

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Learning Outcomes

1. Know and use typical notions for *strings* (substring, prefix, suffix, etc.).
2. Understand principles and implementation of the *KMP*, *BM*, and *RK* algorithms.
3. Know the *performance characteristics* of the KMP, BM, and RK algorithms.
4. Be able to solve simple *stringology problems* using the *KMP failure function*.

Unit 4: *String Matching*



4 String Matching

- 4.1 String Notation
- 4.2 Brute Force
- 4.3 String Matching with Finite Automata
- 4.4 Constructing String Matching Automata
- 4.5 The Knuth-Morris-Pratt algorithm
- 4.6 Beyond Optimal? The Boyer-Moore Algorithm
- 4.7 The Rabin-Karp Algorithm

4.1 String Notation

Ubiquitous strings

string = sequence of characters

- ▶ universal data type for ... everything!
 - ▶ natural language texts
 - ▶ programs (source code)
 - ▶ websites
 - ▶ XML documents
 - ▶ DNA sequences
 - ▶ bitstrings
 - ▶ ... a computer's memory \rightsquigarrow ultimately any data is a string

\rightsquigarrow many different tasks and algorithms

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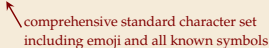
\rightsquigarrow many different tasks and algorithms

- ▶ This unit: finding (exact) **occurrences of a pattern** text.
 - ▶ Ctrl+F
 - ▶ grep
 - ▶ computer forensics (e. g. find signature of file on disk)
 - ▶ virus scanner
- ▶ basis for many advanced applications

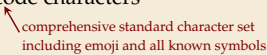
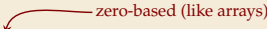
Notations

- ▶ *alphabet* Σ : finite set of allowed **characters**; $\sigma = |\Sigma|$ “a string over alphabet Σ ”
 - ▶ letters (Latin, Greek, Arabic, Cyrillic, Asian scripts, ...)
 - ▶ “what you can type on a keyboard”, Unicode characters $\simeq 130\text{k}$
 - ▶ $\{0, 1\}$; nucleotides $\{A, C, G, T\}$; ...
-  comprehensive standard character set
including emoji and all known symbols

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- ▶ $\Sigma^n = \Sigma \times \cdots \times \Sigma$: strings of **length** $n \in \mathbb{N}_0$ (n -tuples)
- ▶ $\Sigma^* = \bigcup_{n \geq 0} \Sigma^n$: set of **all** (finite) strings over Σ
- ▶ $\Sigma^+ = \bigcup_{n \geq 1} \Sigma^n$: set of **all** (finite) **nonempty** strings over Σ
- ▶ ε $\in \Sigma^0$: the *empty* string (same for all alphabets)

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- ▶ $\varepsilon \in \Sigma^0$: the *empty* string (same for all alphabets)
- ▶ for $S \in \Sigma^n$, write $S[i]$ (other sources: S_i) for **i th** character $(0 \leq i < n)$
zero-based (like arrays)!
- ▶ for $S, T \in \Sigma^*$, write ST = $S \cdot T$ for **concatenation** of S and T
- ▶ for $S \in \Sigma^n$, write $S[i..j]$ or $S_{i,j}$ for the **substring** $S[i] \cdot S[i+1] \cdots S[j]$ $(0 \leq i \leq j < n)$
 - ▶ $S[0..j]$ is a **prefix** of S ; $S[i..n-1]$ is a **suffix** of S
 - ▶ $S[i..j]$ = $S[i..j-1]$ (endpoint exclusive) $\rightsquigarrow S = S[0..n)$

Clicker Question



True or false: $\Sigma^* = \Sigma^+ \cup \{\varepsilon\}$

A True

B False



→ *sli.do/comp526*

Clicker Question



True or false: $\Sigma^* = \Sigma^+ \cup \{\epsilon\}$

strings of length ≥ 0
||
strings of length ≥ 1

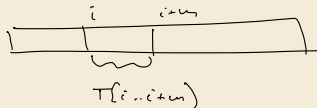
A True ✓ **B** ~~False~~



→ sli.do/comp526

String matching – Definition

Search for a string (pattern) in a large body of text



► Input:

- $T \in \Sigma^n$: The text (haystack) being searched within
- $P \in \Sigma^m$: The pattern (needle) being searched for; typically $n \gg m$

► Output:

- the first occurrence (match) of P in T : $\min\{i \in [0..n-m) : T[i..i+m) = P\}$
 - or NO_MATCH if there is no such i (" P does not occur in T ")
- Variant: Find **all** occurrences of P in T .
 \rightsquigarrow Can do that iteratively (update T to $T[i+1..n)$ after match at i)

► Example:

- $T = \text{"Where is he?"}$
 - $P_1 = \text{"he"} \rightsquigarrow i = 1$
 - $P_2 = \text{"who"} \rightsquigarrow \text{NO_MATCH}$
- string matching is implemented in Java in `String.indexOf`, in Python as `str.find`

Clicker Question



Let $T = \overset{0}{\text{C}}\overset{1}{\text{O}}\overset{2}{\text{M}}\overset{3}{\text{P}}\overset{4}{\text{5}}\overset{5}{\text{2}}\overset{6}{\text{6}}\overset{7}{\text{.}}\overset{8}{\text{i}}\overset{9}{\text{s}}\overset{10}{\text{_}}\overset{11}{\text{f}}\overset{12}{\text{u}}\overset{13}{\text{n}}\overset{14}{\text{.}}$
What is $T[3..7]$?



→ sli.do/comp526

Clicker Question



Let $T = \text{COMP526_is_fun.}$

What is $T[3..7]$?

012345678901234

COMP526_is_fun.



→ *sli.do/comp526*

4.2 Brute Force

Abstract idea of algorithms

String matching algorithms typically use *guesses* and *checks*:

- ▶ A **guess** is a position i such that P might start at $T[i]$.
Possible guesses (initially) are $0 \leq i \leq n - m$.
- ▶ A **check** of a guess is a comparison of $T[i + j]$ to $P[j]$.

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- ▶ A **check** of a guess is a comparison of $T[i + j]$ to $P[j]$.
- ▶ Note: need all m checks to verify a single *correct* guess i ,
but it may take (many) fewer checks to recognize an *incorrect* guess.
- ▶ Cost measure: #character comparisons



\leadsto #checks $\leq n \cdot m$ (number of possible checks)

Brute-force method

```
1 procedure bruteForceSM( $T[0..n]$ ,  $P[0..m]$ )  
2   for  $i := 0, \dots, n - m - 1$  do  
3     for  $j := 0, \dots, m - 1$  do  
4       if  $T[i + j] \neq P[j]$  then break inner loop  
5       if  $j == m$  then return  $i$   
6   return NO_MATCH
```

- try all guesses i
- check each guess (left to right); stop early on mismatch
- essentially the implementation in Java!

► **Example:**

$T = \text{abbbababbab}$

$P = \text{abba}$

a	b	b	b	a	b	a	b	b	a	b
a	b	b	a							
	a									
		a								
			a							
				a	b	b				

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► **Example:**

$T = \text{abbbababbab}$

$P = \text{abba}$

↪ 15 char cmps
(vs $n \cdot m = 44$)
not too bad!

	a	b	b	b	a	b	a	b	b	a	b
a	a	b	b	a							
		a									
			a								
				a							
					a	b	b				
						a					
							a	b	b	a	

Brute-force method – Discussion



Brute-force method can be good enough

- ▶ typically works well for natural language text
- ▶ also for random strings



but: can be as bad as it gets!

a	a	a	a	a	a	a	a	a	a	a
a	a	a	b							
	a	a	a	b						
		a	a	a	b					
			a	a	a	b				
				a	a	a	b			
					a	a	a	b		
						a	a	a	b	
							a	a	a	b

▶ Worst possible input: $P = a^{m-1}b$,
 $T = a^n$

▶ Worst-case performance: $(n - m + 1) \cdot m$

⇒ for $m \leq n/2$ that is $\Theta(mn)$

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a	a	a	b								
	a	a	a	b							
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			a	a	a	b					
				a	a	a	b				
					a	a	a	b			
						a	a	a	b		
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▶ Bad input: lots of self-similarity in T ! ↪ can we exploit that?

▶ brute force does 'obviously' stupid repetitive comparisons ↪ can we avoid that?

Roadmap

- ▶ **Approach 1** (this week): Use *preprocessing* on the **pattern** P to eliminate guesses (avoid 'obvious' redundant work)

- ▶ Deterministic finite automata (DFA)
- ▶ Knuth-Morris-Pratt algorithm
- ▶ Boyer-Moore algorithm
- ▶ Rabin-Karp algorithm

- ▶ **Approach 2** (\rightsquigarrow Unit 8): Do *preprocessing* on the **text** T
Can find matches in time *independent of text size*(!)

- ▶ inverted indices
- ▶ Suffix trees
- ▶ Suffix arrays

4.3 String Matching with Finite Automata

Clicker Question



Do you know what regular expressions, NFAs and DFAs are, and how to convert between them?

- ☐ **A** Never heard of this; are these new emoji?
- ☐ **B** Heard the terms, but don't remember conversion methods.
- ☐ **C** Had that in my undergrad course (memories fading a bit).
- ☐ **D** Sure, I could do that blindfolded!



→ *sli.do/comp526*

Theoretical Computer Science to the rescue!

► string matching = deciding whether $T \in \Sigma^* \cdot P \cdot \Sigma^*$

► $\Sigma^* \cdot P \cdot \Sigma^*$ is *regular* formal language

↪ first occurrence ?

↪ \exists deterministic finite automaton (DFA) to recognize $\Sigma^* \cdot P \cdot \Sigma^*$

↪ can check for occurrence of P in $|T| = \underline{n}$ steps!

automata /
formal
languages (8)

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WTF!?

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\rightsquigarrow can check for occurrence of P in $|T| = n$ steps!



Job done!



WTF!?

We are not quite done yet.

- (Problem 0: programmer might not know automata and formal languages ...)
- Problem 1: existence alone does not give an algorithm!
- Problem 2: automaton could be very big!

String matching with DFA

Σ = alphabet
 Q = set of states

- Assume first, we already have a deterministic automaton
- How does string matching work?

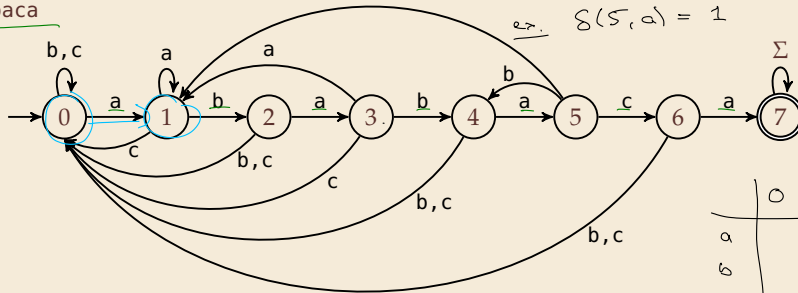
exactly 1 arc transition function
 for each pair (q, c) $\delta : Q \times \Sigma \rightarrow Q$

Example:

$T = \underline{a}abacaababacaa$

$P = \underline{a}babaca$

ex. $\delta(5, a) = 1$



	0	1	2	3	4	5	6	7
a						1		
b								
c								

text:		a	a	b	a	c	a	a	b	a	b	a	c	a	a
state:	0	1	1	2	3	0	1	1	2	3	4	5	6	7	

String matching with DFA

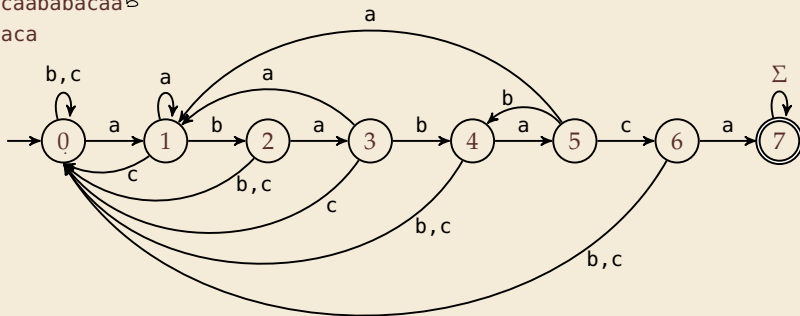
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time to find first occurrence
 $O(n)$

Example:

$T = \text{aabacaababacaab}$

$P = \text{ababaca}$



text:		a	a	b	a	c	a	a	b	a	b	a	c	a	a
state:	0	1	1	2	3	0	1	1	2	3	4	5	6	7	7

b
 7 7 7

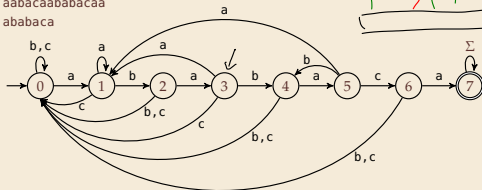
String matching DFA – Intuition

Why does this work?

► Main insight:

State q means:
*“we have seen $P[0..q]$ until here
 (but not any longer prefix of P)”*

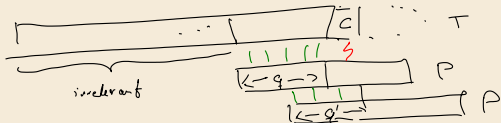
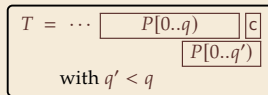
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► If the next text character c does not match, we know:

- (i) text seen so far ends with $P[0...q] \cdot c$
- (ii) $P[0...q] \cdot c$ is not a prefix of P
- (iii) without reading c , $P[0..q)$ was the *longest* prefix of P that ends here.

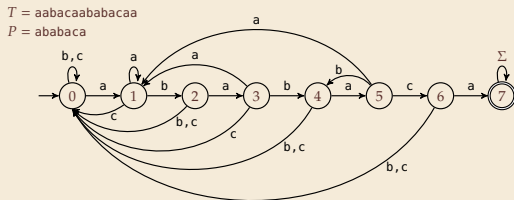


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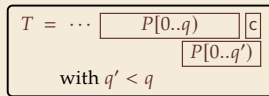
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► If the next text character c does not match, we know:

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- (iii) without reading c , $P[0..q)$ was the *longest* prefix of P that ends here.



↪ New longest matched prefix will be (weakly) shorter than q

↪ All information about the text needed to determine it is contained in $P[0...q) \cdot c!$

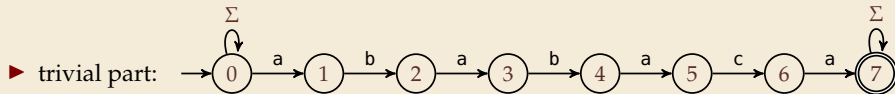
Note: our automata stay in state  forever
once they found the first occurrence
one can also give edges to  to
keep finding occurrences

\Rightarrow DFA can find all occurrences in time $O(n)$

4.4 Constructing String Matching Automata

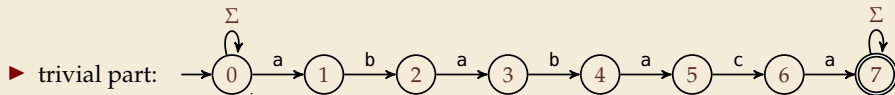
NFA instead of DFA?

It remains to *construct* the DFA.



NFA instead of DFA?

It remains to *construct* the DFA.



► that actually is a *non*deterministic finite automaton (NFA) for $\Sigma^*P\Sigma^*$

↪ We *could* use the NFA directly for string matching:

- at any point in time, we are in a **set** of states
- accept when one of them is final state

Example:

text:		a	a	b	a	c	a	a	b	a	b	a	c	a	a
state:	0	0,1	0,1	0,2	0,1,3	0	0,1	0,1	0,2	0,1,3	0,2,4	0,1,3,5	0,6	0,1,7	0,1,7

But maintaining a whole set makes this slow ... $\Theta(n \cdot m)$ w.r.

Computing DFA directly



You have an NFA and want a DFA?
Simply apply the power-set construction
(and maybe DFA minimization)!

The powerset method has exponential state blow up!
I guess I might as well use brute force ...



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Ingenious algorithm by Knuth, Morris, and Pratt: construct DFA *inductively*:

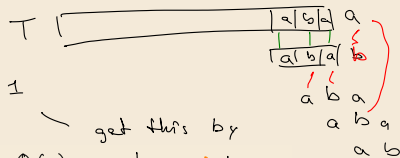
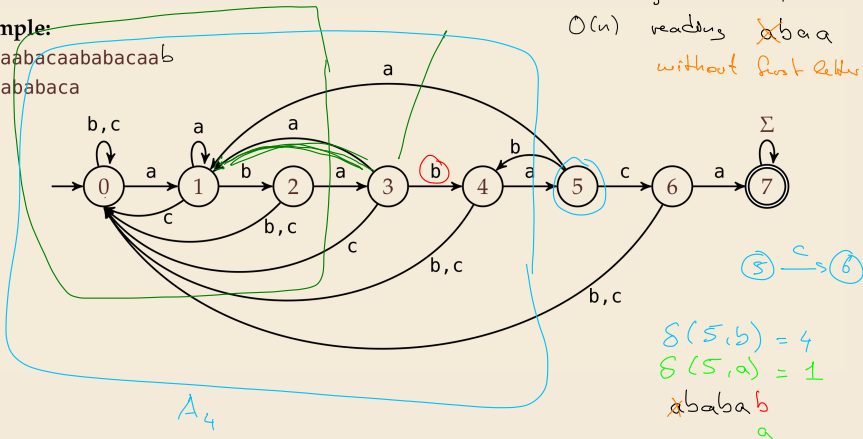
Suppose we add character $P[j]$ to automaton A_{j-1} for $P[0..j)$

- ▶ add new state and matching transition \rightsquigarrow easy
- ▶ for each $c \neq P[j]$, we need $\delta(j, c)$ (transition from \textcircled{j} when reading c)

Example:

$T = aabacaababacaab$

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Suppose we add character $P[j]$ to automaton A_{j-1} for $P[0..j]$

- ▶ add new state and matching transition \rightsquigarrow easy
- ▶ for each $c \neq P[j]$, we need $\delta(j, c)$ (transition from \textcircled{j} when reading c)
- ▶ $\delta(j, c) =$ length of the longest prefix of $P[0..j)c$ that is a suffix of $P[1..j)c$
= state of automaton after reading $P[1..j)c$
 $\leq j \rightsquigarrow$ can use known automaton A_{j-1} for that!

\rightsquigarrow can directly compute A_j from A_{j-1} !



seems to require simulating automata $m \cdot \sigma$ times

State q means:
“we have seen $P[0..q)$ until here
(but not any longer prefix of P)”

Computing DFA efficiently

- ▶ **KMP's second insight:** simulations in one step differ only in last symbol

↪ simply maintain state x , the state after reading $P[1..j]$.

- ▶ copy its transitions
- ▶ update x by following transitions for $P[j]$

Computing DFA efficiently

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1 procedure constructDFA( $P[0..m]$ )
2   //  $\delta[q][c]$  = target state when reading  $c$  in state  $q$ 
3   for  $c \in \Sigma$  do
4      $\delta[0][c] := 0$ 
5    $\delta[0][P[0]] := 1$ 
6    $x := 0$ 
7   for  $j = 1, \dots, m - 1$  do
8     for  $c \in \Sigma$  do // copy transitions
9        $\delta[j][c] := \delta[x][c]$ 
10     $\delta[j][P[j]] := j + 1$  // match edge
11     $x := \delta[x][P[j]]$  // update  $x$ 
```

Example: $P[0..6) = \text{ababac}$

$\delta(c, q)$	0	1	2	3	4	5
a	1	1	3	1	5	1
b	0	2	0	4	0	4
c	0	0	0	0	0	6

$$x = 3$$

Computing DFA efficiently

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String matching with DFA – Discussion

► Time:

- Matching: n table lookups for DFA transitions
- building DFA: $\Theta(m\sigma)$ time (constant time per transition edge).

$\leadsto \Theta(m\sigma + n)$ time for string matching.

► Space:

- $\Theta(m\sigma)$ space for transition matrix.

Oct 2022
Unicode $\sigma = 149,186$

👍 **fast matching** time actually: hard to beat!

👍 total time asymptotically optimal for small alphabet (for $\sigma = O(n/m)$)

👎 substantial **space overhead**, in particular for large alphabets

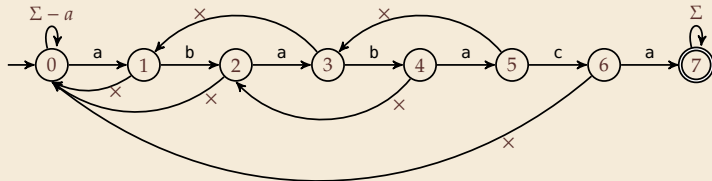
4.5 The Knuth-Morris-Pratt algorithm

Failure Links

- ▶ Recall: String matching with DFA is fast, but needs table of $m \times \sigma$ transitions.
- ▶ in fast DFA construction, we used that all simulations differ only by *last* symbol
- ~> **KMP's third insight:** do this last step of simulation from state x during *matching!*
... but how?

Failure Links

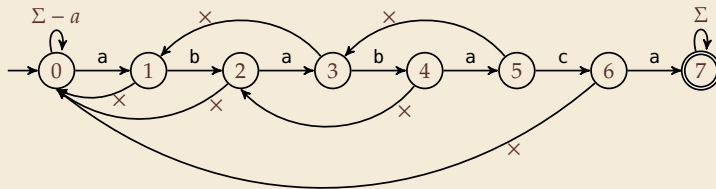
- ▶ Recall: String matching with DFA is fast, but needs table of $m \times \sigma$ transitions.
- ▶ in fast DFA construction, we used that all simulations differ only by *last* symbol
- ~> **KMP's third insight:** do this last step of simulation from state x during *matching*!
... but how?
- ▶ **Answer:** Use a new type of transition, the failure links
 - ▶ Use this transition (only) if no other one fits.
 - ▶ \times does not consume a character. ~> might follow several failure links



~> Computations are deterministic (but automaton is not a real DFA.)

Failure link automaton – Example

Example: $T = \text{abababaaaca}$, $P = \text{ababaca}$



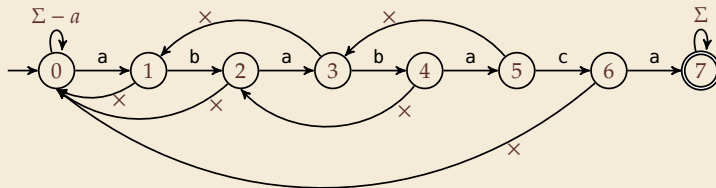
$T:$ a b a b a b a a b a b

 1 2 3 4 5

 3 4

Failure link automaton – Example

Example: $T = \text{abababaaaca}$, $P = \text{ababaca}$



T :	a	b	a	b	a	b	a	a	b	a	b
P :	a	b	a	b	a	×					
			(a)	(b)	(a)	b	a	×			
							a	b	a	b	

to state 3
to state 1

q :	1	2	3	4	5	3,4	5	3,1,0,1	2	3	4
-------	---	---	---	---	---	-----	---	---------	---	---	---

(after reading this character)

Clicker Question



What is the worst-case time to process one character in a failure-link automaton for $P[0..m]$?

A $\Theta(1)$

C $\Theta(m)$

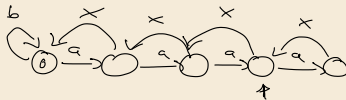
B $\Theta(\log m)$

D $\Theta(m^2)$



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Clicker Question



What is the worst-case time to process one character in a failure-link automaton for $P[0..m]$?

A ~~$\Theta(1)$~~

C $\Theta(m)$ ✓

B ~~$\Theta(\log m)$~~

D ~~$\Theta(m^2)$~~



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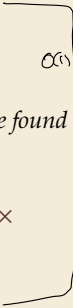
The Knuth-Morris-Pratt Algorithm

```
1 procedure KMP( $T[0..n]$ ,  $P[0..m]$ )
2    $fail[0..m] := failureLinks(P)$ 
3    $i := 0$  // current position in  $T$ 
4    $q := 0$  // current state of KMP automaton
5   while  $i < n$  do
6     if  $T[i] == P[q]$  then
7        $i := i + 1$ ;  $q := q + 1$ 
8     if  $q == m$  then
9       return  $i - q$  // occurrence found
10    else // i.e.  $T[i] \neq P[q]$ 
11      if  $q \geq 1$  then
12         $q := fail[q]$  // follow one  $\times$ 
13      else
14         $i := i + 1$ 
15  end while
16  return NO_MATCH
```

- ▶ only need single array *fail* for failure links
- ▶ (procedure failureLinks later)

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- ▶ only need single array *fail* for failure links
- ▶ (procedure failureLinks later)

Analysis: (matching part)

- ▶ always have $fail[j] < j$ for $j \geq 1$

↪ in each iteration

- ▶ either advance position in text
 $(i := i + 1)$ $\leq n$ steps
- ▶ or shift pattern forward
(guess $i - q$) $\leq n$ steps

- ▶ each can happen at most n times

↪ $\leq 2n$ symbol comparisons!

Computing failure links

► failure links point to error state x (from DFA construction)

↪ run same algorithm, but store $fail[j] := x$ instead of copying all transitions

```
1 procedure failureLinks( $P[0..m]$ )
2    $fail[0] := 0$ 
3    $x := 0$ 
4   for  $j := 1, \dots, m - 1$  do
5      $fail[j] := x$ 
6     // update failure state using failure links:
7     while  $P[x] \neq P[j]$ 
8       if  $x == 0$  then
9          $x := -1$ ; break
10      else
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Computing failure links

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8       if  $x == 0$  then
9          $x := -1$ ; break
10      else
11         $x := fail[x] < \times$ 
12      end while
13       $x := x + 1$ 
14  end for
```

Analysis:

- ▶ m iterations of for loop
- ▶ while loop always decrements x
- ▶ x is incremented only once per iteration of for loop
- $\rightsquigarrow \leq m$ iterations of while loop *in total*
- $\rightsquigarrow \leq 2m$ symbol comparisons

Knuth-Morris-Pratt – Discussion

► Time:

► $\leq 2n + 2m = \underline{O(n + m)}$ character comparisons

► clearly must at least *read* both T and P

~> KMP has optimal worst-case complexity!

► Space:

► $\Theta(m)$ space for failure links



total time asymptotically optimal (for any alphabet size)



reasonable extra space

Clicker Question

What are the main advantages of the KMP string matching (using the failure-link automaton) over string matching with DFAs? Check all that apply.



- ☐ A faster preprocessing on pattern
- ☐ B faster matching in text
- ☐ C fewer character comparisons
- ☐ D uses less space
- ☐ E makes running time independent of σ
- ☐ F I don't have to do automata theory



→ *sli.do/comp526*

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What are the main advantages of the KMP string matching (using the failure-link automaton) over string matching with DFAs? Check all that apply.



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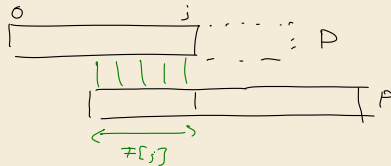
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The KMP prefix function

- It turns out that the failure links are useful beyond KMP
- a slight variation is more widely used: (for historic reasons)
the (KMP) *prefix function* $F : [1..m - 1] \rightarrow [0..m - 1]$:

*$F[j]$ is the length of the longest prefix of $P[0..j]$
that is a suffix of $P[1..j]$.*

- Can show: $fail[j] = F[j - 1]$ for $j \geq 1$, and hence



*$fail[j] = \text{length of the}$
 $\text{longest prefix of } P[0..j]$
 $\text{that is a suffix of } P[1..j]$.*

← memorize this!