Week 04b: String Algorithms, Approximation

Strings

Strings 2/86

A string is a sequence of characters.

An alphabet Σ is the set of possible characters in strings.

Examples of strings:

- C program
- HTML document
- DNA sequence
- · Digitised image

Examples of alphabets:

- ASCII
- Unicode
- {0,1}
- {A,C,G,T}

... Strings 3/86

Notation:

- length(P) ... #characters in P
- λ ... *empty* string $(length(\lambda) = 0)$
- Σ^m ... set of all strings of length m over alphabet Σ
- Σ^* ... set of all strings over alphabet Σ

 $v\omega$ denotes the *concatenation* of strings v and ω

Note: $length(v\omega) = length(v) + length(\omega)$ $\lambda \omega = \omega = \omega \lambda$

... Strings 4/86

Notation:

- substring of P ... any string Q such that $P = \nu Q \omega$, for some $\nu, \omega \in \Sigma^*$
- prefix of P ... any string Q such that $P = Q\omega$, for some $\omega \in \Sigma^*$
- suffix of P ... any string Q such that $P = \omega Q$, for some $\omega \in \Sigma^*$

The string a/a of length 3 over the ASCII alphabet has

- how many prefixes?
- how many suffixes?
- how many substrings?
- 4 prefixes: "" "a" "a/" "a/a"
- 4 suffixes: "a/a" "/a" "a" ""
- 6 substrings: "" "a" "/" "a/" "/a" "a/a"

Note:

"" means the same as λ (= empty string)

... Strings 7/86

ASCII (American Standard Code for Information Interchange)

- Specifies mapping of 128 characters to integers 0..127
- The characters encoded include:
 - o upper and lower case English letters: A-Z and a-z
 - o digits: 0-9
 - o common punctuation symbols
 - o special non-printing characters: e.g. newline and space

Ascii	Char	Ascii	Char	Ascii	Char	Ascii	Char
0	Null	32	Space	64	9	96	
1	Start of heading	33	1	65	λ	97	a
2	Start of text	34		66	В	98	b
3	End of text	35	*	67	C	99	c
4	End of transmit	36	s	68	D	100	d
5	Enquiry	37	٠.	69	E	101	e
6	Acknowledge	38	6	70	P	102	£
7	Audible bell	39		71	G	103	g
8	Backspace	40	(72	H	104	h
9	Horizontal tab	41)	73	I	105	<u>i</u>
10	Line feed	42		7.4	J	106	i
11	Vertical tab	43	+	7.5	K	107	k
12	Form feed	44	,	76	L	108	1
13	Carriage return	45	-	77	M	109	n
14	Shift in	46		78	N	110	n
15	Shift out	47	/	79	0	111	0
16	Data link escape	48	0	80	P	112	p
17	Device control 1	49	1	81	Q.	113	q
18	Device control 2	50	2	82	R	114	r
19	Device control 3	51	3	83	s	115	8
20	Device control 4	52	4	84	T	116	t
21	Neg. acknowledge	53	5	85	U	117	u
22	Synchronous idle	54	6	86	V	118	v
23	End trans. block	55	7	87	W	119	w
24	Cancel	56	8	88	x	120	×
25	End of medium	57	9	89	Y	121	у
26	Substitution	58	1	90	Z	122	z
27	Escape	59	1	91	[123	(
28	File separator	60	<	92	1	124	1
29	Group separator	61		93	1	125	}
30	Record separator	62	>	94	^	126	-
31	Unit separator	63	?	95		127	Forward del.

... Strings 8/86

Reminder:

In C a string is an array of chars containing ASCII codes

- these arrays have an extra element containing a 0
- the extra 0 can also be written '\0' (null character or null-terminator)
- convenient because don't have to track the length of the string

Because strings are so common, C provides convenient syntax:

```
char str[] = "hello"; // same as char str[] = {'h', 'e', 'l', 'l', 'o', '\0'}; Note: str[] will have 6 elements
```

... Strings 9/86

C provides a number of string manipulation functions via #include <string.h>, e.g.

```
strlen() // length of string
strncpy() // copy one string to another
strncat() // concatenate two strings
strstr() // find substring inside string
```

Example:

char *strncat(char *dest, char *src, int n)

- appends string src to the end of dest overwriting the '\0' at the end of dest and adds terminating '\0'
- returns start of string dest
- will never add more than n characters
 (If src is less than n characters long, the remainder of dest is filled with '\0' characters. Otherwise, dest is not null-terminated.)

Pattern Matching

Pattern Matching

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Example (pattern checked backwards):



- Text ... abacaab
- Pattern ... abacab

... Pattern Matching

Given two strings T (text) and P (pattern), the pattern matching problem consists of finding a substring of T equal to P

Applications:

- Text editors
- · Search engines
- Biological research

... Pattern Matching

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Naive pattern matching algorithm

- checks for each possible shift of *P* relative to *T*
 - o until a match is found, or
 - all placements of the pattern have been tried

```
NaiveMatching(T,P):
   Input text T of length n, pattern P of length m
   Output starting index of a substring of T equal to P
          -1 if no such substring exists
   for all i=0..n-m do
                                      // check from left to right
      while j \le m and T[i+j]=P[j] do // test i^{th} shift of pattern
         j=j+1
         if j=m then
            return i
                                      // entire pattern checked
         end if
      end while
   end for
                                      // no match found
   return -1
```

Analysis of Naive Pattern Matching

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Naive pattern matching runs in O(n·m)

Examples of worst case (forward checking):

- T = aaa...ah
- P = aaah
- may occur in DNA sequences
- unlikely in English text

Exercise #2: Naive Matching

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Suppose all characters in P are different.

Can you accelerate NaiveMatching to run in O(n) on an n-character text T?

When a mismatch occurs between P[j] and T[i+j], shift the pattern all the way to align P[0] with T[i+j]

⇒ each character in T checked at most twice

Example:

abcdabcdeabcc abcdabcdeabcc abcdexxxxxxxx xxxxabcde

Boyer-Moore Algorithm

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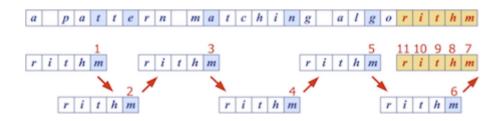
The *Boyer–Moore* pattern matching algorithm is based on two heuristics:

- Looking-glass heuristic: Compare P with subsequence of T moving backwards
- Character-jump heuristic: When a mismatch occurs at T[i]=c
 - o if P contains $c \Rightarrow \text{shift } P \text{ so as to align the last occurrence of } c \text{ in } P \text{ with } T[i]$
 - o otherwise \Rightarrow shift P so as to align P[0] with T[i+1] (a.k.a. "big jump")

... Boyer-Moore Algorithm

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Example:



... Boyer-Moore Algorithm

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Boyer–Moore algorithm preprocesses pattern P and alphabet Σ to build

- last-occurrence function L
 - \circ L maps Σ to integers such that L(c) is defined as
 - the largest index i such that P[i]=c, or
 - -1 if no such index exists

Example: $\Sigma = \{a,b,c,d\}, P = acab$

С	a	b	С	d	
L(c)	2	3	1	-1	

L can be represented by an array indexed by the numeric codes of the characters

• L can be computed in O(m+s) time $(m \dots \text{ length of pattern}, s \dots \text{ size of } \Sigma)$

... Boyer-Moore Algorithm

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BoyerMooreMatch(T,P,Σ):

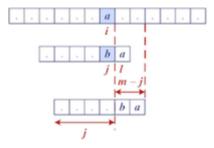
```
text T of length n, pattern P of length m, alphabet \Sigma
Input
Output starting index of a substring of T equal to P
       -1 if no such substring exists
L=lastOccurenceFunction(P,\Sigma)
i=m-1, j=m-1
                               // start at end of pattern
repeat
   if T[i]=P[j] then
      if j=0 then
                               // match found at i
         return i
      else
         i=i-1, j=j-1
      end if
                               // character-jump
   else
      i=i+m-min(j,1+L[T[i]])
      j=m-1
   end if
until i≥n
return -1
                               // no match
```

• Biggest jump (m characters ahead) occurs when L[T[i]] = -1

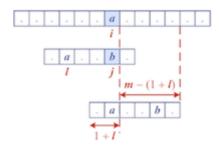
... Boyer-Moore Algorithm

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Case 1: $j \le 1 + L[c]$



Case 2: 1+L[c] < i



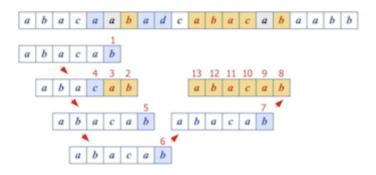
Exercise #3: Boyer-Moore algorithm

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For the alphabet $\Sigma = \{a,b,c,d\}$

- 1. compute last-occurrence function L for pattern $P = \mathbf{abacab}$
- 2. trace Boyer-More on P and text T = abacaabadcabacabaabb
 - o how many comparisons are needed?

С	a	b	С	d	
L(c)	4	5	3	-1	



13 comparisons in total

... Boyer-Moore Algorithm

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Analysis of Boyer-Moore algorithm:

- Runs in O(nm+s) time
 - \circ $m \dots$ length of pattern $n \dots$ length of text $s \dots$ size of alphabet
- Example of worst case:
 - \circ T = aaa ... a
 - \circ P = baaa
- Worst case may occur in images and DNA sequences but unlikely in English texts
 - ⇒ Boyer–Moore significantly faster than naive matching on English text

Knuth-Morris-Pratt Algorithm

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The Knuth-Morris-Pratt algorithm ...

- compares the pattern to the text *left-to-right*
- but shifts the pattern more intelligently than the naive algorithm

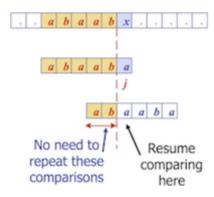
Reminder:

- Q is a prefix of P ... $P = Q\omega$, for some $\omega \in \Sigma^*$
- Q is a suffix of P ... $P = \omega Q$, for some $\omega \in \Sigma^*$

... Knuth-Morris-Pratt Algorithm

When a mismatch occurs ...

- what is the most we can shift the pattern to avoid redundant comparisons?
- Answer: the largest *prefix* of *P*[0..j] that is a *suffix* of *P*[1..j]



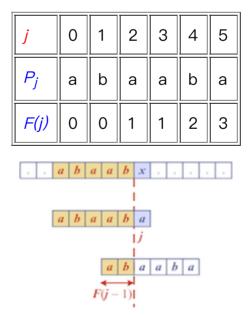
... Knuth-Morris-Pratt Algorithm

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KMP preprocesses the pattern P[0..m-1] to find matches of its prefixes with itself

- Failure function F(j) defined as
 the size of the largest prefix of P[0..j] that is also a suffix of P[1..j] for each position j=0..m-1
- if mismatch occurs at $P_j \Rightarrow$ advance j to F(j-1)

Example: P = abaaba



... Knuth-Morris-Pratt Algorithm

```
KMPMatch(T,P):
```

```
F=failureFunction(P)
j=0
                             // number of characters matched
i=0
                             // scan the text from left to right
while i<n do
   if T[i]=P[j] then
      i=i+1, j=j+1
      if j=m then
                              // all of P matched?
                                // match found at i-j
         return i-j
      end if
   end if
   else
                             // next character does not match
      j=F[j]
   end while
   i=i+1
end while
return -1
                             // no match
```

Exercise #4: KMP-Algorithm

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- 1. compute failure function F for pattern P = abacab
- 2. trace Knuth-Morris-Pratt on P and text T = abacaabaccabacabaabb
 - how many comparisons are needed?

j	0	1	2	3	4	5
P_j	а	b	а	С	а	b
F(j)	0	0	1	0	1	2

```
      a
      b
      a
      c
      a
      b
      a
      c
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```

19 comparisons in total

... Knuth-Morris-Pratt Algorithm

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Construction of the failure function matches pattern against itself:

```
failureFunction(P):
   Input pattern P of length m
   Output failure function for P
   F[0]=0
   i=1, j=0
   while i<m do
      if P[i]=P[j] then
                           // we have matched j+1 characters
         F[i]=j+1
         i=i+1, j=j+1
                           // use failure function to shift P
      else if j>0 then
         j=F[j-1]
      else
                           // no match
         F[i]=0
         i=i+1
      end if
   end while
   return F
```

... Knuth-Morris-Pratt Algorithm

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Analysis of failure function computation:

- At each iteration of the while-loop, either
 - o *i* increases by one, or
 - \circ the "shift amount" i-j increases by at least one (observe that F(j-1)< j)
- Hence, there are no more than $2 \cdot m$ iterations of the while-loop
- \Rightarrow failure function can be computed in O(m) time

... Knuth-Morris-Pratt Algorithm

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Analysis of Knuth-Morris-Pratt algorithm:

- Failure function can be computed in O(m) time
- At each iteration of the while-loop, either
 - o *i* increases by one, or
 - the "shift amount" i-j increases by at least one (observe that F(j-1)< j)
- Hence, there are no more than $2 \cdot n$ iterations of the while–loop
- \Rightarrow KMP's algorithm runs in optimal time O(m+n)

Boyer-Moore vs KMP

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Boyer-Moore algorithm

- decides how far to jump ahead based on the mismatched character in the text
- works best on large alphabets and natural language texts (e.g. English)

- uses information embodied in the pattern to determine where the next match could begin
- works best on small alphabets (e.g. A, C, G, T)

For the keen: The article "Average running time of the Boyer-Moore-Horspool algorithm" shows that the time is inversely proportional to size of alphabet

Word Matching With Tries

Preprocessing Strings

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Preprocessing the *pattern* speeds up pattern matching queries

• After preprocessing *P*, KMP algorithm performs pattern matching in time proportional to the text length

If the text is large, immutable and searched for often (e.g., works by Shakespeare)

we can preprocess the text instead of the pattern

... Preprocessing Strings

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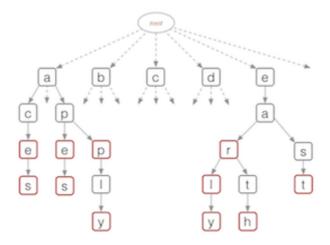
A trie ...

- is a compact data structure for representing a set of strings
 e.g. all the words in a text, a dictionary etc.
- supports pattern matching queries in time proportional to the pattern size

Note: Trie comes from retrieval, but is pronounced like "try" to distinguish it from "tree"

Tries 38/86

Tries are trees organised using parts of keys (rather than whole keys)



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... Tries

Each node in a trie ...

- contains one part of a key (typically one character)
- may have up to 26 children
- may be tagged as a "finishing" node
- but even "finishing" nodes may have children

Depth d of trie = length of longest key value

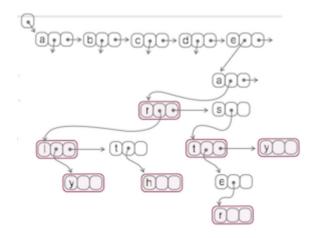
Cost of searching O(d) (independent of n)

... Tries 40/86

Possible trie representation:

... Tries 41/86

Note: Can also use BST-like nodes for more space-efficient implementation of tries



Trie Operations

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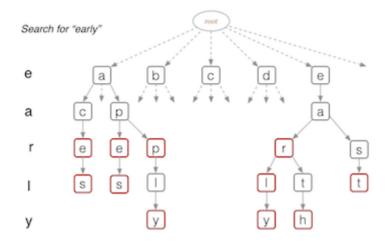
Basic operations on tries:

1. search for a key

2. insert a key

Trie Operations

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... Trie Operations

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Traversing a path, using char-by-char from Key:

```
find(trie,key):
   Input trie, key
   Output pointer to element in trie if key found
          NULL otherwise
   node=trie
   for each char in key do
      if node.child[char] exists then
         node=node.child[char] // move down one level
      else
         return NULL
      end if
   end for
   if node.finish then
                                // "finishing" node reached?
      return node
   else
      return NULL
   end if
```

... Trie Operations

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Insertion into Trie:

```
insert(trie,item,key):
    Input trie, item with key of length m
    Output trie with item inserted
    if trie is empty then
        t=new trie node
    end if
    if m=0 then
```

```
t.finish=true, t.data=item
else
t.child[key[0]]=insert(t.child[key[0]],item,key[1..m-1])
end if
return t
```

... Trie Operations

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Analysis of standard tries:

- O(n) space
- insertion and search in time O(d·m)
 - o n ... total size of text (e.g. sum of lengths of all strings in a given dictionary)
 - o m ... size of the string parameter of the operation (the "key")
 - o d ... size of the underlying alphabet (e.g. 26)

Word Matching With Tries

Word Matching with Tries

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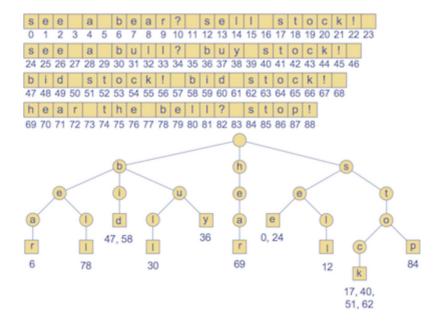
Preprocessing the text:

- 1. Insert all searchable words of a text into a trie
- 2. Each leaf stores the occurrence(s) of the associated word in the text

... Word Matching with Tries

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Example text and corresponding trie of searchable words:

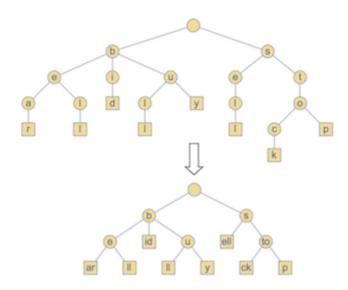


Compressed Tries

Compressed tries ...

- have internal nodes of degree ≥ 2
- are obtained from standard tries by compressing "redundant" chains of nodes

Example:



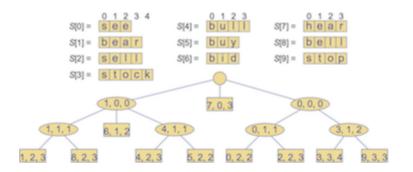
... Compressed Tries

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Possible compact representation of a compressed trie to encode an array S of strings:

- nodes store ranges of indices instead of substrings
 use triple (i,j,k) to represente substring S[i][j..k]
- requires O(s) space (s = # strings in array <math>S)

Example:

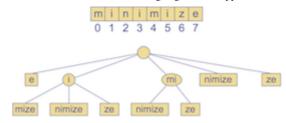


Pattern Matching With Suffix Tries

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The *suffix trie* of a text *T* is the compressed trie of all the suffixes of *T*

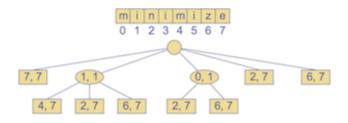
Example:



... Pattern Matching With Suffix Tries

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Compact representation:



... Pattern Matching With Suffix Tries

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Input:

- compact suffix trie for text T
- pattern P

Goal:

• find starting index of a substring of T equal to P

... Pattern Matching With Suffix Tries

```
suffixTrieMatch(trie,P):
   Input compact suffix trie for text T, pattern P of length m
   Output starting index of a substring of T equal to P
          -1 if no such substring exists
   j=0, v=root of trie
   repeat
      // we have matched j+1 characters
      if ∃w∈children(v) such that P[j]=T[start(w)] then
         i=start(w)
                               // start(w) is the start index of w
         x=end(w)-i+1
                               // end(w) is the end index of w
         if m≤x then
                       // length of suffix ≤ length of the node label?
            if P[j..j+m-1]=T[i..i+m-1] then
               return i-j
                               // match at i-j
            else
               return -1
                               // no match
         else if P[j..j+x-1]=T[i..i+x-1] then
            j=j+x, m=m-x
                               // update suffix start index and length
                               // move down one level
                               // no match
         else return -1
         end if
      else
         return -1
      end if
```

until v is leaf node
return -1

// no match

... Pattern Matching With Suffix Tries

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Analysis of pattern matching using suffix tries:

Suffix trie for a text of size $n \dots$

- can be constructed in O(n) time
- uses O(n) space
- supports pattern matching queries in O(s·m) time
 - o m... length of the pattern
 - o s ... size of the alphabet

Text Compression

Text Compression

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Problem: Efficiently encode a given string X by a smaller string Y

Applications:

Save memory and/or bandwidth

Huffman's algorithm

- computes frequency f(c) for each character c
- encodes high-frequency characters with short code
- no code word is a prefix of another code word
- uses optimal encoding tree to determine the code words

... Text Compression

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Code ... mapping of each character to a binary code word

Prefix code ... binary code such that no code word is prefix of another code word

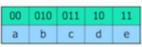
Encoding tree ...

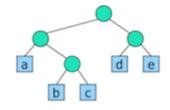
- represents a prefix code
- each leaf stores a character
- code word given by the path from the root to the leaf (0 for left child, 1 for right child)

... Text Compression

Example:

每个字母的编码不能有相同的prefix 当按照prefix线索时遇到相同的编码值即可输出





... Text Compression

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Text compression problem

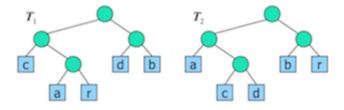
Given a text T, find a prefix code that yields the shortest encoding of T

- short codewords for frequent characters
- long code words for rare characters

... Text Compression

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Example: T = abracadabra



 T_1 requires 29 bits to encode text T,

 T_2 requires 24 bits

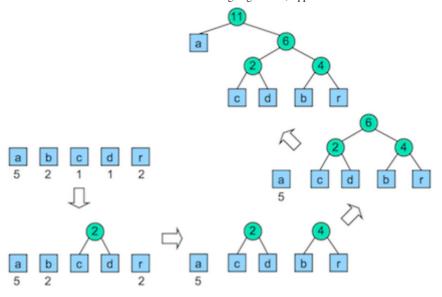
... Text Compression

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Huffman's algorithm

- computes frequency f(c) for each character
- successively combines pairs of lowest–frequency characters to build encoding tree "bottom-up"

Example: abracadabra



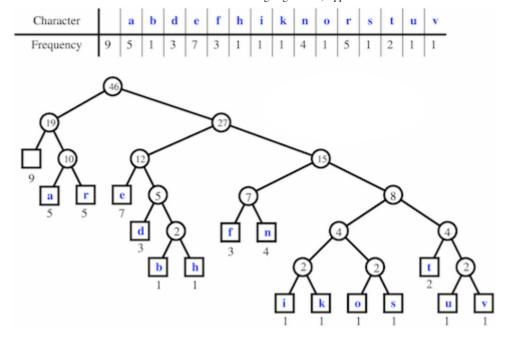
Huffman Code 64/86

Huffman's algorithm using priority queue:

```
HuffmanCode(T):
   Input string T of size n
   Output optimal encoding tree for T
   compute frequency array
   Q=new priority queue
   for all characters c do
      T=new single-node tree storing c
      join(Q,T) with frequency(c) as key
   end for
  while |Q| \ge 2 do
      f_1=Q.minKey(), T_1=leave(Q)
      f_2=Q.minKey(), T_2=leave(Q)
      T=new tree node with subtrees T_1 and T_2
      join(Q,T) with f_1+f_2 as key
   end while
   return leave(Q)
```

... Huffman Code 65/86

Larger example: a fast runner need never be afraid of the dark



... Huffman Code 66/86

Analysis of Huffman's algorithm:

- O(n+d·log d) time
 - ∘ n ... length of the input text T
 - o d... number of distinct characters in T

Approximation

Approximation for Numerical Problems

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Approximation is often used to solve numerical problems by

- solving a simpler, but much more easily solved, problem
- where this new problem gives an approximate solution
- and refine the method until it is "accurate enough"

Examples:

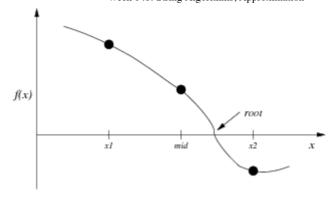
- roots of a function f
- length of a curve determined by a function f
- ... and many more

... Approximation for Numerical Problems

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Example: Finding Roots

Find where a function crosses the x-axis:



Generate and test: move x_1 and x_2 together until "close enough"

... Approximation for Numerical Problems

70/86

A simple approximation algorithm for finding a root in a given interval:

```
bisection(f, x_1, x_2):

| Input function f, interval [x_1, x_2]
| Output x \in [x_1, x_2] with f(x) \cong 0

| repeat
| | mid=(x_1+x_2)/2
| | if f(x_1)*f(mid)<0 then
| | x_2=mid // root to the left of mid
| else
| | x_1=mid // root to the right of mid
| end if
| until f(mid)=0 or x_2-x_1<\epsilon // \epsilon: accuracy end while
| return mid
```

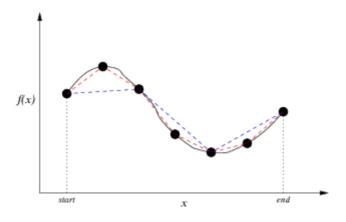
bisection guaranteed to converge to a root if f continuous on $[x_1,x_2]$ and $f(x_1)$ and $f(x_2)$ have opposite signs

... Approximation for Numerical Problems

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Example: Length of a Curve

Estimate length: approximate curve as sequence of straight lines.



... Approximation for Numerical Problems

```
curveLength(f,start,end):
    Input function f, start and end point
    Output curve length between f(start) and f(end)
    length=0, δ=(end-start)/StepSize
    for each x∈[start+δ,start+2δ,..,end] do
        length = length + sqrt(δ² + (f(x)-f(x-δ))²)
    end for
    return length
```

Sidetrack: Function Pointers

73/86

Function pointers ...

- are references to memory address of a function
- are pointer values and can be assigned/passed

Function pointer variables/parameters are declared as:

```
typeOfReturnValue (*fname)(typeOfArguments)
```

... Sidetrack: Function Pointers

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Example:

```
// define a function of type double → double
double myfun(double x) {
   return sqrt(1-x*x);
}

double curveLength(double start, double end, double (*f)(double)) {
   ...
   deltaY = f(x) - f(x-delta);
   length += sqrt(delta*delta + deltaY*deltaY);
   ...
}

printf("%.10f\n", curveLength(-1, 1, myfun));
```

Approximation for Numerical Problems

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Trade-offs in curve length approximation algorithm:

- large step size ...
 - less steps, less computation (faster), lower accuracy
- small step size ...
 - more steps, more computation (slower), higher accuracy

However, too many steps may lead to higher rounding error.

Each f has an optimal step size ...

• but this is difficult to determine in advance

... Approximation for Numerical Problems

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```
Example: length = curveLength(0,\pi,sin);
```

Convergence when using more and more steps

```
steps = 0, length = 0.000000
steps = 10, length = 3.815283
steps = 100, length = 3.820149
steps = 10000, length = 3.820197
steps = 100000, length = 3.820198
steps = 1000000, length = 3.820198
```

Actual answer is 3.820197789...

Approximation for NP-hard Problems

77/86

Approximation is often used for NP-hard problems ...

- computing a near-optimal solution
- in polynomial time

Examples:

- · vertex cover of a graph
- subset-sum problem

Vertex Cover 78/86

Reminder: Graph G = (V, E)

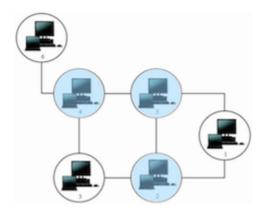
- set of vertices V
- set of edges E

Vertex cover C of G ...

- C ⊆ V
- for all edges $(u,v) \in E$ either $v \in C$ or $u \in C$ (or both)
- \Rightarrow All edges of the graph are "covered" by vertices in C

... Vertex Cover 79/86

Example (6 nodes, 7 edges, 3-vertex cover):



Applications:

- Computer Network Security
 - o compute minimal set of routers to cover all connections
- Biochemistry

... Vertex Cover

size of vertex cover $C \dots \mid C \mid$ (number of elements in C)

optimal vertex cover ... a vertex cover of minimum size

Theorem.

Determining whether a graph has a vertex cover of a given size k is an NP-complete problem.

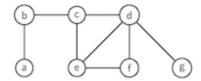
... Vertex Cover

An approximation algorithm for vertex cover:

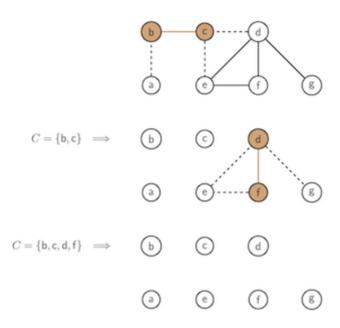
Exercise #5: Vertex Cover

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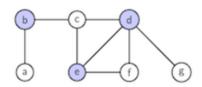
Show how the approximation algorithm produces a vertex cover on:



Possible result:



What would be an optimal vertex cover?



... Vertex Cover

Theorem.

The approximation algorithm returns a vertex cover at most twice the size of an optimal cover.

Proof. Any (optimal) cover must include at least one endpoint of each chosen edge.

Cost analysis ...

- repeatedly select an edge from E
 - add endpoints to C
 - o delete all edges in E covered by endpoints

Time complexity: O(V+E) (adjacency list representation)

Summary

- Alphabets and words
- Pattern matching
 - o Boyer-Moore, Knuth-Morris-Pratt
- Tries
- Text compression
 - o Huffman code
- Approximation
 - o factor-2 approximation for vertex cover
- Suggested reading:
 - o tries ... Sedgewick, Ch. 15.2
 - o approximation ... Moffat, Ch. 9.4

Produced: 6 Jan 2020