String Algorithms

• **Sequence** and **string** are synonymous
  – Finite number of letters\(^1\) from an alphabet, written contiguously from left to right \( S_{i..j} \)
    \[ A C T G G T C A \]

• **Subsequence**
  – an ordered subset obtained by removing letters from a sequence
    \[ A C T G G T C A \]
    \[ \rightarrow A T G G T A \] (Both C’s are removed)

• **Substring**
  – Consecutive and complete string of letters found in the target string
    \[ A C T G G T C A \]
    \[ \boxed{C T G G T} \]

\(^1\)Sometimes called symbols or characters
Prefix and Suffix

For \( S_{1 \leq i \leq j \leq n} \)

Prefix : \( S_{1 \ldots j} \)

S, SL, SLI, SLIT, SLITH, SLITHY

Suffix : \( S_{i \ldots n} \)

Y, HY, THY, ITHY, LITHY, SLITHY

SLITHY \( N=6 \)
String Algorithms

• Algorithms for String Matching
  – No preprocessing
    • Naïve
    • Boyer-Moore
  – Preprocessing of target string
    • Aho-Corasick
  – Preprocessing of query string, target string or fixed DB
    • Rabin-Karp
    • Suffix Tree

• Algorithms for String Alignment
  – No preprocessing
    • Needleman Wunsch
    • Smith Waterman
  – Preprocessing of ‘fixed’ Database
    • Complicated heuristics such as BLAST, FASTA
STRING MATCHING

Looking for an EXACT substring match
Naïve algorithm
Operates in quadratic time

Complexity $\sim mn$
Boyer-Moore

- ‘Smarter’. Operates in linear time
- Pattern string $s$ doesn’t ‘walk’ the target string $t$; it ‘skips along’ the target string
- At each iteration, it chooses the better of:
  - Jump the distance derived from the bad character heuristic
  - Jump the distance derived from the good suffix heuristic
Now o’er the one halfworld
Nature seems dead, and wicked dreams abuse
The curtain’d sleep; witchcraft celebrates
Pale Hecate’s offerings, and wither’d murder,
Alarum’d by his sentinel, the wolf,
Whose howl’s his watch, thus with his stealthy pace
With Tarquin’s ravishing strides, towards his design
Moves like a ghost
Bad Character Heuristic

Use naïve algorithm to align last character in s with corresponding character in t. (this is the letter ‘h’)

Then, extend match leftwards for increasing suffixes of both strings (letters ‘s’, then ‘i’). Stop with a mismatch, identifying the bad character in t (the letter ‘v’)

Slide s to the right to align the rightmost occurrence of a same bad character in s (find a letter ‘v’) with the bad character in t (the ‘v’ identified earlier). This alignment is the rightmost character in the new good suffix.

Jump s 15 characters so that old bad character in t aligns with a mate in s

Note that we don’t use this new good suffix just yet; first we must look at the results from the good suffix heuristic (next slide)
GOOD SUFFIX Heuristic

As before, use naïve algorithm to align last character in s with corresponding character in t. (this is the letter ‘h’)

Then, extend match leftwards for increasing suffixes of both strings (letters ‘s’, then ‘i’). Stop with a mismatch, identifying the bad character in t (the letter ‘v’)

Find the rightmost suffix in s to the left of the original good suffix in s that matches this good suffix

Slide s to align it with a corresponding suffix in t.

Jump 4 characters

New bad character

New good suffix
Choose the larger jump!
AHO-CORASICK

2 Phases

• Preprocess: Build a finite automaton based on the query strings (quadratic time)

• Use the target string as input to the automaton (linear time)
### The Constructed Automaton for the string ‘CATCH’

<table>
<thead>
<tr>
<th>input</th>
<th>A</th>
<th>C</th>
<th>H</th>
<th>T</th>
</tr>
</thead>
<tbody>
<tr>
<td>state</td>
<td>q₀</td>
<td>q₁</td>
<td>q₀</td>
<td>q₀</td>
</tr>
<tr>
<td>q₀</td>
<td>q₀</td>
<td>q₁</td>
<td>q₀</td>
<td>q₀</td>
</tr>
<tr>
<td>q₁</td>
<td>q₂</td>
<td>q₀</td>
<td>q₀</td>
<td>q₀</td>
</tr>
<tr>
<td>q₂</td>
<td>q₀</td>
<td>q₀</td>
<td>q₀</td>
<td>q₃</td>
</tr>
<tr>
<td>q₃</td>
<td>q₀</td>
<td>q₄</td>
<td>q₀</td>
<td>q₀</td>
</tr>
<tr>
<td>q₄</td>
<td>q₀</td>
<td>q₀</td>
<td>q₅</td>
<td>q₀</td>
</tr>
</tbody>
</table>

$q₀ = \text{start}$  
$q₅ = \text{stop}$

Note that this abstract machine is specifically an acceptor for the word ‘catch’. The machine simply halts if ‘catch’ is input, otherwise it remains in $q₀$. 
Deterministic Finite Automaton

Q: Could this scheme work for MISSISSIPPI?
Aho-Corasick

The deterministic automaton can be represented by a rooted prefix tree. Each edge is a character in the target string and each node represents a prefix formed by the edges along the path from the root to that node. When there is a match the query string maps to a leaf of the tree.

Here is a prefix tree of our simple example
Generalized Aho-Corasick

Many patterns can be represented on a single prefix tree (Keyword Tree)
Such trees can be created for entire dictionaries in linear time in a preprocessing step. The tree can then be traversed with a query word as needed

Not shown here are the failure links
Pro’s and Con’s of String Matching Algorithms Thus Far

• Pro
  – No false negatives, no false positives

• Con
  – Significant overhead/slowdown by comparing characters
  – If characters are converted to numbers, the numbers can become quite large
    • Large arithmetic computational overhead (slowdown)
    • Very large numbers can blow RAM
String Searching with ‘Fingerprints’
The Rabin-Karp Algorithm

• Pro
  – Because the algorithm uses modular arithmetic, numbers and intermediate calculations are constrained, so…
    • Very fast
    • No threat to RAM limits
  – No false negatives
  – The probability of a false positive is extremely low

• Con
  – False positives are nonetheless possible
    • If the probability of a false positive is not deemed low enough, or any false positive whatsoever is unacceptable, then some small overhead is required for checking
Essence of the Algorithm

• Every possible substring in the target of the same length as the query string is identified.
• The query string and every target substring so identified is converted to an integer
• Each integer is mapped into its remainder modulo p, where p is a carefully chosen prime number
• The query string (new) integer is compared to every possible target substring (new) integer in the target string.
  – When the query and target numbers are not the same, there is no match
  – When the query and target numbers are the same, there is a very high probability of a match
RABIN-KARP SEARCH

Suppose we have an alphabet consisting only of the letters AFPQRSTVWY, coded by decimal digits. These digits represent the 10 amino acids found in life forms in Star System TGR-75 (our own planet has 20 naturally occurring amino acids)

<table>
<thead>
<tr>
<th>Amino Acid</th>
<th>Letter</th>
<th>Numeric Code</th>
</tr>
</thead>
<tbody>
<tr>
<td>ALANINE</td>
<td>Ala</td>
<td>A</td>
</tr>
<tr>
<td>PHENYLALANINE</td>
<td>Phe</td>
<td>F</td>
</tr>
<tr>
<td>PROLINE</td>
<td>Pro</td>
<td>P</td>
</tr>
<tr>
<td>GLUTAMINE</td>
<td>Gln</td>
<td>Q</td>
</tr>
<tr>
<td>ARGinine</td>
<td>Arg</td>
<td>R</td>
</tr>
<tr>
<td>SERINE</td>
<td>Ser</td>
<td>S</td>
</tr>
<tr>
<td>THREONINE</td>
<td>Thr</td>
<td>T</td>
</tr>
<tr>
<td>VALINE</td>
<td>Val</td>
<td>V</td>
</tr>
<tr>
<td>TRYPTOPHAN</td>
<td>Trp</td>
<td>W</td>
</tr>
<tr>
<td>TYROSINE</td>
<td>Tyr</td>
<td>Y</td>
</tr>
</tbody>
</table>

These amino acids combine in a linear chain to form proteins. There are thousands of proteins, each with a unique sequence of amino acids in the chain, some chains very long (1000’s), some short (5). All variations of these proteins, and all proteins, are all ordered sequentially in a single searchable database, totaling seven million entries. For convenience and storage economy, the numeric code is used instead of the alphabet letter.
The Computational Problem:

We have just synthesized a new protein (or so we think). The sequence of amino acids is 5 long: Val-Tyr-The-Tyr-Ser (code is 31415).

We wish to determine whether this protein (sequence of amino acids) exists already. We agree that it is better to work with the AA codes, rather than the names, so we are searching the DB for at least one instance of 31415 as our target integer.
Computational Solution

- Strategy 1: Combine each successive 5 digits in the target into a 5 digit integer, then walk the set of target integers, seeking the query integer
  - Advantage: No false positives
  - Disadvantage: Could be faster: Must walk the DB and make a complex comparison each time (polynomial)

- Strategy 2: Preprocess the database, building a hash table of hash values, then seek the hash value matching the hash value of the query
  - Advantage: Fast. Preprocess only once in linear time, search with multiple queries in linear time. Smaller database to search
  - Disadvantage: Must deal with hash collisions (false positives), although still fast
Desired string is 31415
Hash function is $x \mod p \quad p=13$
Desired hash value is 7

Computing the hash value while walking the target (strategy #1) or preprocess DB building a hash table (strategy #2)

Efficiently (linearly) generating a new hash value:
- Take away the high order digit
- Shift everything left with zero-fill
- Add value of new right target digit
- Apply hash function

Hash collisions are bound to occur

Snippet of the DB

Raw Code

Hash Values

14152 \equiv (31415 - 3 \cdot 10000) \cdot 10 + 2 \pmod{13}
= (7 - 3 \cdot 3) \cdot 10 + 2 \pmod{13}
= 8 \pmod{13}
About Efficiency

The efficiency is that working in modulo arithmetic saves combining 5 digits into a 5 digit integer, then working with a decimal shift and finding a remainder. Here is a naïve way to slide along:

Let’s slide along the DB from 31415 to the next letter (2), ending in 14152

$$31415 = 30,000 + 1415$$

So, we need to get to 14150, then add 2

$$14152 = (31415 - (3 \times 10000)) \times 10 + 2$$

This is 1415

This is 14150

This is 14152

$$14152 \mod 13 = 8$$
Working efficiently in modulo arithmetic

We already know this number mod13 is 7; we don’t need to re-compute it.

It is easy to compute that 3 mod13 is 3.

By theorem: $[xy]_k \equiv [x]_k [y]_k$,
$[3 \times 10000]_{mod\ 13} \equiv [3]_{mod\ 13} \times [10000]_{mod\ 13}$
$\equiv 3_{mod\ 13} \times 3_{mod\ 13} = 9$

So, walking the DB in modulo arithmetic is highly efficient.

$14152 = (31415 - (3 \times 10000)) \times 10 + 2$

This value is fixed by the problem at 10000 mod 13 = 3

This is fixed at 10 mod 13 = 10

It is easy to compute that 2 mod 13 = 2

Note that if the modulus is greater than 10, it is not necessary to do any modulo calculation on the single base-10 digits- a further efficiency.

$x_{mod\ 13} = (7 - 3 \times 3) \times 10 + 2 \mod 13$
$= (7 - 9) \times 10 + 2 \mod 13$
$= -18 \mod 13 = 8$
What modulus?

Answer: A prime number!

- A prime modulus limits the probability of a false positive
- Prime moduli in a modular hashing function enhance diffusion and prevent clumping of false positives (hash collisions) in a hash table
Probability of false positives*……..

\( n \) is the length of the query

\( m \) is the length of the target

*Let \( \pi(x) \) be the number of prime numbers \( \leq \) any positive integer \( x \), and \( \pi(mn) \) likewise

Ensure that \( n \cdot m \geq 29 \)

Choose a prime \( p \leq x \) at random for any positive integer \( x \)

Theorem: \( \text{prob}(\text{spurious hit}) \leq \frac{\pi(n \cdot m)}{\pi(x)} \)

*Development follows Gusfield
False Positives

We need to pick an \( x \) such that a random prime number \( \leq x \) is neither too big (slows down things) nor too small (too many false positives)

Without getting into the nuances of number theory, let us make an empirical* choice and let \( x = n m^2 \).

In that case, the upper limit on the probability of a false positive is \( 2.53/m \).

So, for our target snippet of length 19 and our query of length 5, the probability of a false positive is \( 2.53/19 \), or \( \sim 13\% \). Not all that great, but we have a tiny database

But consider our entire Star System TGR-75 database. If we were to pick any prime number \( \leq 5 \times (7,000,000)^2 \) at random as our modulus, the probability of a false positive with that choice would be \( 3.614 \times 10^{-7} \).

*If interested, perhaps as a final project, you might explore the prime number counting functions of Gauss and Chebychev
Suffix Trees

• Preprocessing of the target (Tree building: linear)
• Linear Search time
• Can generalize to multiple targets
  – Allows other functions besides exact matching
  – Longest Common Substring
  – Wildcards
Suffix Trees: An example

And, has thou slain the Jabberwock?
Come to my arms, my beamish boy!
O frabjous day! Callooh! Callay!'
He chortled in his joy.

-Lewis Carroll
# Suffixes of *CalloohCallay*

<table>
<thead>
<tr>
<th>CalloohCallay</th>
<th>Callay</th>
</tr>
</thead>
<tbody>
<tr>
<td>alloohCallay</td>
<td>allay</td>
</tr>
<tr>
<td>lloohCallay</td>
<td>llay</td>
</tr>
<tr>
<td>loohCallay</td>
<td>lay</td>
</tr>
<tr>
<td>oohCallay</td>
<td>ay</td>
</tr>
<tr>
<td>ohCallay</td>
<td>y</td>
</tr>
<tr>
<td>hCallay</td>
<td></td>
</tr>
</tbody>
</table>
### Suffixes of *CalloohCally*: ordered

<p>| | |</p>
<table>
<thead>
<tr>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>ay</td>
<td>lay</td>
</tr>
<tr>
<td>allay</td>
<td>llay</td>
</tr>
<tr>
<td>alloohCallay</td>
<td>lloohCallay</td>
</tr>
<tr>
<td>Callay</td>
<td>loohCallay</td>
</tr>
<tr>
<td>CalloohCallay</td>
<td>ohCallay</td>
</tr>
<tr>
<td>hCallay</td>
<td>oohCallay</td>
</tr>
<tr>
<td>y</td>
<td></td>
</tr>
</tbody>
</table>
Constructing the Tree

Common prefix
Suffix Tree for ‘CalloohCallay’