Lecture 16

COL 351: Analysis and Design of Algorithms

OneNote

Lecture 16

Given: A string $P = [p_1, ..., p_k]$ of size k.

Find: A Table of size k satisfying

Table[i] := Length of longest (non-trivial) common prefix and suffix of P[1, i]

Prefix Suffix problem— Lemma

Table[i] := Length of longest (non-trivial) common prefix and suffix of P[1, i]

Lemma 1: Suppose $L \ge 1$ satisfy that the L-length prefix and suffix of P[1, i] are identical.

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Then, the length of longest common prefix-suffix of P[1, i] of size just smaller than L is "Table [L]".

Proof Sketch:

6-length prefix 6-length suffix
$$p_1 \quad p_2 \quad p_3 \quad p_4 \quad p_5 \quad p_6 \quad p_7 \quad \dots \quad p_{i-5} \quad p_{i-4} \quad p_{i-3} \quad p_{i-2} \quad p_{i-1} \quad p_i$$

Prefix Suffix problem— Lemma

Table[i] := Length of longest (non-trivial) common prefix and suffix of P[1, i]

Lemma 1: Suppose $L \ge 1$ satisfy that the L-length prefix and suffix of P[1, i] are identical.

Then, the length of longest common prefix-suffix of P[1, i] of size just smaller than L is "Table[L]".

Proof Sketch:

So, Table[i], Table[Table[i]], Table[Table[Table[i]]], is sequence of all common prefixes-suffixes of P[1,i].

Prefix Suffix Algorithm — using Dynamic Programming

Table[i] := Length of longest (non-trivial) common prefix and suffix of P[1, i]

```
Table \leftarrow Array of size k;

Table[1], L = 0;

For (i = 1 \text{ to } k - 1):

/* value of L is Table[i] */

While (L > 0 \text{ and } P[i+1] \neq P[L+1]): L = \text{Table}[L];

If (P[i+1] = P[L+1]): L = L + 1;

Table[i+1] = L;
```

Update L to length of largest common non-trivial Prefix-Suffix of P[1, i] that satisfy P[i+1] = P[L+1].

If no such L exists, then L is just 0.

Prefix Suffix Algorithm — using Dynamic Programming

Table[i] := Length of longest (non-trivial) common prefix and suffix of P[1, i]

```
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Table[1], L = 0;

For (i = 1 \text{ to } k - 1):

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While (L > 0 \text{ and } P[i+1] \neq P[L+1]): L = Table[L];

If (P[i+1] = P[L+1]): L = L + 1;

Table[i+1] = L;
```

Analysis of Time-Complexity

Fact 1: L is incremented at most k times.

Fact 2: Throughout the algorithm L can decrease at most k times, so total number of iterations of While loop is at most k.

Therefore, time complexity is O(k).

Main Problem — String Matching

Given: String $S = [s_1, ..., s_n]$ and a pattern $P = [p_1, ..., p_k]$, represented as arrays of size n, k. (Here k < n).

Find: Does there exists a sub-string of S that is identical to P.

Examples:

P = "hash"
S = "cuckoo hashing is efficient"
Yes

P = "hash-table"

S = "cuckoo hashing is efficient"

No

String Matching problem— Lemma

Table[i] := Length of longest (non-trivial) common prefix and suffix of P[1, i]

Lemma 2: Suppose $i, L \ge 1$ satisfy that the L-length prefix of P is identical to L-length suffix of S[1, i].

Then, the length of longest prefix of P that is also a suffix of S[1, i] of size just smaller than L is "Table [L]".

Proof Sketch:

Knuth-Morris-Pratt (KMP) algorithm

Table[i] := Length of longest (non-trivial) common prefix and suffix of P[1, i]

A[i] := Length of longest prefix of P that is also a suffix of S[1, i]

 $A \leftarrow \text{Array of size } n + 1 \text{ with } A[0] = 0;$

 $L \leftarrow 0$;

For (i = 0 to n - 1):

Update L to largest integer satisfying

- *L*-length prefix of P = L-length suffix of S[1,i]
- S[i+1] = P[L+1].

If no such L exists then L is just Ω

```
/* value of L is A[i] */
While (L > 0 and S[i+1] \neq P[L+1]): L = Table[L];

If (S[i+1] = P[L+1]): L = L+1;

A[i+1] = L;

If (L = k) Return True;

Return False;
```

Knuth-Morris-Pratt (KMP) algorithm

Table[i] := Length of longest (non-trivial) common prefix and suffix of P[1, i]

A[i] := Length of longest prefix of P that is also a suffix of S[1, i]

```
A \leftarrow Array of size n+1 with A[0]=0;

L \leftarrow 0;

For (i=0 \text{ to } n-1):

/* value of L is A[i]*/

While (L>0 \text{ and } S[i+1] \neq P[L+1]): L=\text{Table}[L];

If (S[i+1]=P[L+1]): L=L+1;

A[i+1]=L;
```

Analysis of Time-Complexity

Fact 1: L is incremented at most n times.

Fact 2: Throughout the algorithm L can decrease at most n times, so total number of iterations of While loop is at most n.

Therefore, time complexity is O(n).

If (L = k) Return True;

Return False;

Single Source Distances in graph with negative edge-weights

Single Source Distance Problem

Given: A directed weighted graph G = (V, E) with possible negative edge weights, and a source s.

Output: Either a Shortest-path-tree rooted at s, or report that G contains a negative cycle reachable from s.

Examples:

