CSE 5095: Research Topics in Big Data Analytics Lecture 16: 3/25/14

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In this lecture we'll present the linear time algorithm of (Kärkäinen and Sanders 2003) for the construction of the suffix array for any given input string.

Let $T = t_0 t_1 \dots t_{m-1}$ be the given input string. For k = 0, 1, and 2 define $B_k = \{i \in [0, m] : i \mod 3 = k\}$ Let $B = B_1 \cup B_2$. Let S_i stand for the suffix of T starting at position i, for $0 \le i \le m-1$. Let S_c denote the collection of suffixes S_i for each $j \in C$, where $C \subseteq [0, m-1]$.

Algorithm:

- 1. Sort the suffixes S_B ; Let this sorted sequence be Q;
- 2. Using the order obtained in step 1, sort the suffixes S_{B_0} to get Q';
- 3. Merge Q with Q';

Note: It suffices to assume that $\sum = [1, m]$

For k = 1 or 2 define:

 $R_{k} = [t_{k} t_{k+1} t_{k+2}] [t_{k+3} t_{k+4} t_{k+5}] \dots [t_{maxBk} t_{maxBk+1} t_{maxBk+2}]$

In this string, each substring of length 3 enclosed within square brackets is thought of as a single super character. Any such super character is an integer in the range $[1, m^3]$.

Example:

Position	t ₀	<i>t</i> ₁	<i>t</i> ₂	t3	t4	<i>t</i> 5	t ₆	<i>t</i> ₇	t ₈	t9	<i>t</i> ₁₀	t ₁₁
<i>T</i> =	5	2	1	4	3	3	1	5	3	4	4	1

1. $R_1 = [214][331][534][410]$ $R_2 = [143][315][344][100]$

Construct the string $R = R_1 R_2$. In the above example,

R = [214][331][534][410][143][315][344][100]

<u>Observation</u>: The relative ordering of the suffixes in *R* is the same as the relative ordering of the suffixes in S_B .

1a. Sort the super characters in R using radix sort in linear time and replace each super character with its sorted rank. As a result, each super character is replaced with an integer in the range [1, *IRI*]. If the characters in R are now distinct, we are done with sorting S_B .

R	lank	3	5	8	7	2	4	6	1
	<i>R</i> =	214	331	534	410	143	315	344	100

1b. If the characters in R are not distinct, then recursively sort the suffixes in the resultant string (where each character is an integer in the range [1, |R|]).

2. To sort S_{B_0} :

Let rank(S_i) be the rank (among the suffixes in S_B) of the suffix S_i where $i \in B$. Note: $S_i \leq S_k$ where $j, k \in B_0$ if and only if $(t_i, \operatorname{rank}(S_{i+1})) \leq (t_k, \operatorname{rank}(S_{k+1}))$

Example1: $S_3 \le S_0$ since (4, 5) \le (5, 3)

Example 2: $S_3 \le S_9$ since (4, 5) \le (4, 7)

To sort S_{B_0} , sort pairs of the form $(t_j, \operatorname{rank}(S_{j+1}))$ for $j \in B_0$ using integer sort. This takes O(m) time.

3. <u>Merging *Q* and *Q'*</u>: Let S_i and S_j be two suffixes such that $S_i \in B_0$ and $S_j \in B_1$ or B_2 Case 1: $S_j \in B_1$ $S_i \le S_j$ if and only if $(t_i, \operatorname{rank}(S_{i+1})) \le (t_j, \operatorname{rank}(S_{j+1}))$ Case 2: $S_j \in B_2$ $S_i \le S_i$ if and only if $(t_i, t_{i+1}, \operatorname{rank}(S_{i+2})) \le (t_i, t_{i+1}, \operatorname{rank}(S_{i+2}))$

Let T(m) be the RUN TIME of this algorithm on any string of length m.

Then, $T(m) = T(\frac{2}{3}m) + O(m) = O(m)$.

Note: This algorithm is known as the skew algorithm (since the split is not 1/2, 1/2).

DATA MINING

<u>Association Rules Mining</u>: Input: A set of transactions $= \{t_1, t_2, ..., t_n\}$ Let *I* be a set of possible items. Let $I = \{i_1, i_2, ..., i_d\}$. Each $t_i \subseteq I$, $1 \leq i \leq n$.

An <u>association rule</u> is an implication $X \rightarrow Y$ where $X \neq \phi$; $Y \neq \phi$; $X \cap Y = \phi$; $X \subseteq I$; and $Y \subseteq I$.

<u>Definition</u>: An itemset is a set of items. A *k*-itemset is an itemset with *k* items. Let *X* be and itemset. Then the <u>support</u> for *X*, denoted as $\sigma(X)$, is defined as the number of transactions that contain *X*. <u>Support</u> for a rule $X \rightarrow Y$ is $\sigma(X \cup Y)/n$.

<u>Confidence</u> for the rule $X \rightarrow Y$ is $\sigma(X \cup Y) / \sigma(X)$.

Problem:

Given a database *DB* of transactions, and two real numbers *minSupport* and *minConfidence*, find all the rules $X \rightarrow Y$ whose support is \geq *minSupport* and whose confidence is \geq *minConfidence*.

<u>Definition</u>: An itemset is frequent if its support is $\geq n$. *minSupport*.

The problem of finding association rules is normally done in two steps.

- 1. Find all the frequent itemsets; and
- 2. Using the frequent itemsets generate all the relevant association rules.

Example: If X is frequent and $X = X_1 \cup X_2$, then check if $X_1 \rightarrow X_2$ has enough confidence, for every nonempty and proper subset X_1 of X.