Similarity Measures for Sets of Strings

R.L. Kashyap\* and B.J. Oommen\*\*

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- \* Department of Electrical Engineering, Purdue University, W. Lafayette, IN, 47907, USA.
- \*\* School of Computer Science, Carleton University, Ottawa, Ontario, K1S 5B6, Canada.

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## SIMILARITY MEASURES FOR SETS OF STRINGS

Kashyap, R.L., and Oommen, B.J.,

#### ABSTRACT

In the companion paper[3], we have presented a common basis for many of similarity and dissimilarity measures involving a pair of strings. In this paper, we extend the results to capture various numerical and nonnumerical measures involving more than two strings. A measure D(X,Y,...,Z) has been defined involving the set of strings  $\{X,Y,...Z\}$  in terms of two abstract operators  $\boldsymbol{\theta}$  and  $\boldsymbol{\theta}$  and a function  $\delta(\boldsymbol{\cdot},\boldsymbol{\cdot})$  which has as many arguments as there are strings in the set  $\{X,Y,...,Z\}$ . The quantity D(X,Y,...,Z)represents various numerical and nonnumerical quantities involving {X,Y,...,Z} such as Length of their Longest Common Subsequence, (LLCS) the Length of their Shortest Common Supersequence, (LSCS) the set of their common subsequences, the set of their common supersequences and the set of their shuffles. The computational properties of D(X,Y,...,Z) have also been discussed.

#### I. Introduction

In the companion paper [3], we presented a common abstract foundation for many of the known similarity and dissimilarity measures that involve a pair of strings. Some of these measures include the LLCS of two strings, their LSCS and the set of Longest Common Subsequences (LCS). Many of the quantities defined for a pair of strings can easily be extended for a set of strings. In fact, all the quantities discussed in the companion paper [3]

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<u>Index Terms</u>: Similarity measures for sets of strings, dissimilarity measures for sets of strings, common basis for properties involving sets of strings, subsequences, supersequences and shuffles of sets of strings.

with the exception of the Generalized Levenshtein Distance GLD(X,Y) and the probability P(Y/X), lend themselves to natural generalizations. For example, a subsequence common to X, Y and Z, of maximum length, would be one of their longest common subsequences. In this paper we consider the problem of defining and computing a family of similarity measures that involve a set of strings.

Since the theory of similarity measures for a pair of strings has been fairly well developed, one possible approach to studying more than two strings is to study them in a pairwise fashion. However, even simple numerical measures involving a set of strings cannot be obtained as a function of their corresponding pairwise measures. For example, even the LLCS of three strings X, Y, and Z, cannot be expressed in terms of three pairwise quantities, LLCS(X, Y), LLCS(Y, Z) and LLCS(X, Z). Specifically, if LLCS (X, Y), LLCS(Y, Z) are known, the LLCS(X, Y, Z) is not determined. Similarly, if X, Y, Z A, it is easy to define the set of LCS of the three strings. Currently, there is no algorithm to compute this set. In fact, very little effort has gone in to efficiently compute or even to define similarity measures over a set of strings.

One of the applications of similarity measures involving sets of strings is in the clustering of records identified by their keywords [2]. Further, comparison of more than two molecular proteins can be achieved [1] by using any reasonable similarity measure involving them.

Considering all this, a unifying theory involving properties common to sets of strings is desirable. The abstract measure D(X,Y), introduced in the companion paper [3], has the advantage that it can be generalized for K strings  $\{X,Y,\ldots,Z\}$ , yielding a measure involving them,  $D(X,Y,\ldots,Z)$ . As in the previous case, all the various measures between the K strings such as

their LLCS, LSCS, the set of their common subsequences etc. can be seen to be particular cases of D(X,Y,...,Z). Further, since D(X,Y,...,Z) can be recursively computed all the above quantities can be computed using a common computational algorithm.

Among the new contributions of this paper are :

- (1) An algorithm to compute numerical indices of similarity, such as the LLCS and LSCS, involving more than two strings.
- (2) Algorithms to compute the set of all the LCS and the set of all the Shortest Common Supersequences (SCS) involving more than two strings. The algorithms presented in this case are also one-pass algorithms, that require no backtracking.

#### I.1. Notation

The notation and the terminology of the companion paper [3] is assumed. A is a finite alphabet, and  $\lambda$  is the null symbol, distinct from  $\mu$  the null string.  $\tilde{A}$ = A U { $\lambda$ } is called the Appended Alphabet. • and • are two abstract operators, whose properties are discussed in [3]. Three structures that we will repeatedly use are the following [3]:

(a) 
$$\tau_3 = (z_0, MAX, +, 0, 0)$$

(b) 
$$\tau_4 = (\tau_A, U, -, \emptyset, \{\mu\})$$

(c) 
$$t_5 = (Z_p, MIN, +, \omega, 0)$$

where  $Z_p^1$  is the set consisting of the nonnegative integers and  $\infty$ ,  $T_A$  is the power set of  $A^*$ , and  $\emptyset$  is the null set.

### II. The Abstract System and the Abstract Measure for K Strings

Let  $\beta$  be a set of K strings,  $\{X,Y,...,Z\}$ . The abstract measure, D(X,Y,...,Z) involving the K strings is said to be induced by the system  $(A,T,\delta)$ , explained below:

- (i) T is the algebraic structure defined in II.1 of [3].
- (ii)  $\delta$  is a map from  $\tilde{A}^K$  to T, where T is the set defined in  $\tau$ .  $\delta$  has as many arguments as there are strings. The quantity  $\delta(\lambda,\lambda,\ldots,\lambda)$  is undefined and is not needed.

The set  $G_{X,Y,...,Z}$  is defined as:

$$G_{X,Y,...,Z} = \{(X',Y',...,Z') \mid (X',Y',...,Z') \in \widetilde{A}^{K}, \text{and obeys (i)-(iii)}\}$$

(i) 
$$C(X') = X$$
,  $C(Y') = Y$ ,...,  $C(Z') = Z$ 

(ii) 
$$|x'| = |y'| = ... = |z'|$$

(iii) In no (
$$X', Y', ..., Z'$$
) is  $x_i' = y_i' = ... = z_i' = \lambda$ ,  $1 \le i \le |X'|$ .

The set  $G_{X,Y}$  defined in [3] for a pair of strings represents all the ways by which Y can be edited into X using only the edit operations of substitution, deletion and insertion.  $G_{X,Y,\ldots,Z}$  defined above is a direct generalization of  $G_{X,Y}$ .

The measure D(X,Y,...,Z), induced by  $(A,T,\delta)$  is defined by (1).

D(X,Y,...,Z) = I, the identity for  $\emptyset$ , if  $X = Y = ... = Z = \mu$ 

$$= \underbrace{(x',y',\ldots,z') \in G_{X,Y,\ldots,z'}}_{(x,y',\ldots,z')} \begin{bmatrix} |x'| \\ \widehat{x} \\ |z| \end{bmatrix} d(x'_i,y'_i,\ldots,z'_i)$$

otherwise

(1)

The fact that D(X,Y,...,Z) is a unique element in T follows again directly from the associativity of  $\Theta$  and the associativity and commutativity of  $\Theta$ . We shall now state some theorems regarding D(X,Y,...,Z) and observe some of the properties of the set  $\{X,Y,...,Z\}$  which it captures. These theorems are direct generalizations of the corresponding theorems for a pair of strings,

and the reasoning behind their proofs follows the same reasoning as their earlier counterparts.

#### Theorem 1.

Let  $\tau_3$  be as defined in Section I.1, and let  $\delta_1$  be a function mapping  $\tilde{A}^K$  to  $Z_p^I$  as:

$$\delta_1(a_1, a_2, \dots, a_K) = 1$$
 if  $a_1 = a_2 = \dots a_K = a$ ,  $a_i \in A$ 

$$= 0$$
 otherwise
 $\delta_1(A_1, A_2, \dots, A_K)$  is undefined.

Then the measure induced by the system  $(A, \tau_3, \delta_1)$  is exactly the LLCS of the set  $\{X, Y, \ldots, Z\}$ .

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### Theorem 2.

Let  $\tau_4$  be as defined as in Section I.1. Let  $\delta_2$  be a function mapping  $\tilde{A}^K$  to the power set of  $A^*$  as:

$$\delta_2(a_1,a_2,...,a_K) = \{a\}$$
 if  $a_1 = a_2 = a_3 = ... a_K = a, a_i \in A$ 

$$= \{\mu\}$$
 otherwise
$$\delta_2(\lambda,\lambda,...,\lambda)$$
 is undefined

The measure induced by the system  $(A,T_4,\delta_2)$  is the <u>set</u> of <u>all</u> the subsequences common to <u>all</u> the strings  $\{X,Y,\ldots,Z\}$ .

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### Theorem 3.

Let  $\tau_5$  be as defined in Section I.1 and let  $\delta_3$  be a function mapping  $\tilde{A}^K$  to  $Z_p^I$  as:

 $\delta_3(a_1,a_2,...,a_K) = \infty$  if any two non-A arguments are distinct.

=1 if all non-A arguments are equal.

 $δ_3$ (λ,...,λ) is undefined.

The measure induced by the system  $(A, T_5, \delta_3)$  is exactly the LSCS of the set  $\{X,Y,\ldots Z\}$ .

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### Theorem 4.

Let  $\tau_4$  be the structure defined in Section I.1. Let  $\delta_4$  be a function mapping  $\tilde{A}^K$  to the power set of  $A^*$  as:

$$\delta_4(a_1,a_2,...,a_K) = \emptyset$$
 if any two non-A arguments are distinct.  
={a} if all non-A  $a_i$ 's are equal to a.

 $\delta_4(\lambda,\lambda,...,\lambda)$  is undefined.

Then the measure induced by the system  $(A, \tau_4, \delta_4)$  is the set of all the supersequences common to all the K strings  $\{X, Y, \dots Z\}$ .

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### Theorem 5.

Let  $\tau_4$  be as defined in Section I.1. Let  $\delta_5$  be a function defining the set of elementary measures, and which maps  $\tilde{A}^K$  to the power set of  $A^*$  as:

$$\delta_5(a_1,a_2,...,a_k) = \{a_i\}$$
 iff exactly one argument  $a_i$  is non-A otherwise

The measure induced by the system  $(A, T_5, \delta_5)$  is the set of all the shuffles of all the K strings  $\{X, Y, ..., Z\}$ .

## III. Computation of the Abstract Measure for More Than Two Strings

The computation of the abstract measure for more than two strings follows arguments similar to the computation of the abstract measure for a pair of strings. The measure has analogous iterative properties, which are K- dimensional generalizations of the properties valid for a pair of strings. For the sake of simplicity, we shall consider the case when K = 3. We shall state a theorem regarding the measure D(X,Y,Z) in terms of the measures associated with their left derivatives, X, Y, and Z respectively, and their last symbols.

#### Theorem 6.

Let  $X = x_1 \dots x_N$ ,  $Y = y_1 \dots y_M$  and  $Z = z_1 \dots z_R$ ,  $M, N, R \ge 1$ . Then, D(X,Y,Z) =

$$\left[D(\underline{X},Y,Z) \oplus \delta(x_{N},A,A)\right] \oplus \left[D(X,\underline{Y},Z) \oplus \delta(A,y_{M},A)\right] \oplus \left[D(X,Y,\underline{Z}) \oplus \delta(A,A,z_{R})\right]$$

$$\bullet \left[ D(\underline{X},\underline{Y},Z) \bullet \delta(x_N,y_M,A) \right] \bullet \left[ D(\underline{X},Y,\underline{Z}) \bullet \delta(x_N,A,z_R) \right] \bullet \left[ D(X,\underline{Y},\underline{Z}) \bullet \delta(A,y_M,z_R) \right]$$

$$\bullet \left[D(\underline{X},\underline{Y},\underline{Z}) \bullet \delta(x_{N},y_{M},z_{R})\right]$$

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The proof of the theorem is identical to that of Theorem 9 of [3]. By partitioning  $G_{X,Y,Z}$  into 7 mutually exclusive and exhaustive sets, D(X,Y,Z) is split up as seven terms operated on by  $\Theta$ . After stripping off the last elements of the triples in  $G_{X,Y,Z}$ , these sets can be identified as  $G_{X,Y,Z}$ ,  $G_{X,Y,Z}$ . The theorem follows directly thereafter by using the associativity of  $\Theta$  and the distributivity of  $\Theta$  over  $\Theta$ . Using Theorem 6 and the 3-dimensional version of Theorem 9 of [3] we present Algorithm I to compute the measure D(X,Y,Z).

#### Algorithm I.

<u>Input</u>: Three strings  $X = x_1 \dots x_N$ ,  $Y = y_1 \dots y_M$ ,  $Z = z_1 \dots z_R$ 

Output: The Abstract measure D(X,Y,Z) induced by the system (A,T,S)

#### Method:

 $D(u_{n}u_{n}u_{n}) = I$ 

for i=1 to N do 
$$D(X_{i-1}) = D(X_{i-1}) \otimes \delta(x_{i})$$

for j=1 to M do 
$$D(\mu_i Y_j) = D(\mu_i Y_{j-1}) \otimes \delta(\lambda_i y_j \lambda)$$

for k=1 to R do D(
$$\mu_{\lambda}\mu_{\lambda}Z_{k}$$
) = D( $\mu_{\lambda}\mu_{\lambda}Z_{k-1}$ )  $\theta$   $\delta(\lambda_{\lambda}\lambda_{\lambda}Z_{k})$ 

for i=1 to N do

for 
$$j = 1$$
 to M do

$$\begin{split} D(X_{i},Y_{j},\mu) &= \begin{bmatrix} D(X_{i-1},Y_{j},\mu) & \theta & \delta(x_{i},\lambda,\lambda) \end{bmatrix} & \theta & \left[ D(X_{i},Y_{j-1},\mu) & \theta & \delta(\lambda,y_{j},\lambda) \right] \\ & & \left[ D(X_{i-1},Y_{j-1},\mu) & \theta & \delta(x_{i},y_{j},\lambda) \right] \end{split}$$

end

end

for i=1 to N 
$$\underline{do}$$
 for k=1 to R  $\underline{do}$  D(X<sub>i</sub>,  $\mu$ ,  $z_k$ ) =  $\begin{bmatrix} D(X_{i-1}, \mu_{i}, z_k) & \delta(x_{i}, \lambda_{i}, z_k) \end{bmatrix} \oplus \begin{bmatrix} D(X_{i}, \mu_{i}, z_{k-1}) & \delta(\lambda_{i}, \lambda_{i}, z_k) \end{bmatrix} \oplus \begin{bmatrix} D(X_{i-1}, \mu_{i}, z_{k-1}) & \delta(\lambda_{i}, \lambda_{i}, z_{k}) \end{bmatrix}$  end end for j=1 1 to M  $\underline{do}$ 

for k=1 to R do

$$\begin{split} \mathsf{D}(\mu,\mathsf{Y}_{j},\mathsf{Z}_{k}) &= \left[ \mathsf{D}(\mu,\mathsf{Y}_{j-1},\mathsf{Z}_{k}) \otimes \delta(\lambda,\mathsf{Y}_{j},\lambda) \right] \otimes \left[ \mathsf{D}(\mu,\mathsf{Y}_{j},\mathsf{Z}_{k-1}) \otimes \delta(\lambda,\lambda,\mathsf{Z}_{k}) \right] \\ & \otimes \left[ \mathsf{D}(\mu,\mathsf{Y}_{j-1},\mathsf{Z}_{k-1}) \otimes \delta(\lambda,\mathsf{Y}_{j},\mathsf{Z}_{k}) \right] \end{split}$$

end

end

for i=1 to N do

for j=1 to M do

for k=1 1 to R do

$$\begin{split} \mathsf{D}(\mathsf{X}_{i},\mathsf{Y}_{j},\mathsf{Z}_{k}) &= \left[\mathsf{D}(\mathsf{X}_{i-1},\mathsf{Y}_{j},\mathsf{Z}_{k}) \,\, \boldsymbol{\theta} \,\, \delta(\mathsf{X}_{i},\mathsf{A},\mathsf{A}_{r})\right] \,\, \boldsymbol{\theta} \, \left[\mathsf{D}(\mathsf{X}_{i},\mathsf{Y}_{j-1},\mathsf{Z}_{k}) \,\, \boldsymbol{\theta} \,\, \delta(\mathsf{A}_{r},\mathsf{Y}_{j},\mathsf{A})\right] \\ &= \left[\mathsf{D}(\mathsf{X}_{i},\mathsf{Y}_{j},\mathsf{Z}_{k-1}) \,\, \boldsymbol{\theta} \,\, \delta(\mathsf{A}_{r},\mathsf{A}_{r},\mathsf{Z}_{k})\right] \,\, \boldsymbol{\theta} \, \left[\mathsf{D}(\mathsf{X}_{i-1},\mathsf{Y}_{j-1},\mathsf{Z}_{k}) \,\, \boldsymbol{\theta} \,\, \delta(\mathsf{X}_{i},\mathsf{Y}_{j},\mathsf{A})\right] \\ &= \left[\mathsf{D}(\mathsf{X}_{i-1},\mathsf{Y}_{j},\mathsf{Z}_{k-1}) \,\, \boldsymbol{\theta} \,\, \delta(\mathsf{X}_{i},\mathsf{A}_{r},\mathsf{Z}_{k})\right] \,\, \boldsymbol{\theta} \, \left[\mathsf{D}(\mathsf{X}_{i},\mathsf{Y}_{j-1},\mathsf{Z}_{k-1}) \,\, \boldsymbol{\theta} \,\, \delta(\mathsf{A}_{r},\mathsf{Y}_{j},\mathsf{Z}_{k})\right] \\ &= \left[\mathsf{D}(\mathsf{X}_{i-1},\mathsf{Y}_{j-1},\mathsf{Z}_{k-1}) \,\, \boldsymbol{\theta} \,\, \delta(\mathsf{X}_{i},\mathsf{Y}_{j},\mathsf{Z}_{k})\right] \\ &= \mathsf{end} \end{split}$$

end

end

return  $D(X_N,Y_M,Z_R)$ 

END Algorithm I.

Just like Algorithm I of [3], this algorithm too can be understood in the context of a three dimensional graph—a three dimensional trellis. The nodes of the trellis are triples, which correspond to the prefixes of the strings. The edges of the trellis are directed, and have elementary measures associated with the transformations they represent. The quantity  $D(X_i,Y_j,Z_k)$  is the measure associated with the node  $(X_i,Y_i,Z_k)$ .

Initially the measures associated with the X, Y and Z axes are obtained, and then using the extended results of Theorem 9 of [3], the measures associated with the nodes on the X-Y, X-Z and Y-Z planes are computed. Finally the 3-dimensional trellis is traversed, the measure associated with a node being computed using the computed measures associated with the seven adjacent nodes from which it can be reached tracing at most one edge.

#### III.1. Remarks

- (i) From the last set of for-loops, it is evident that Algorithm I requires O(MNR) symbol comparisons. It performs the operations and O(MNR) times, and thus has a time complexity of O(MNR) if it is used to compute any of the numerical quantities discussed in this paper.
- (ii) In particular cases (like the LCS problem) many of the individual elementary terms will have a magnitude of zero, and so for these particular cases, the algorithm can be simplified considerably.

(iv) Consider the problem of computing the set of all LCS between X, Y, and Z. From Theorem 2, the measure induced between X, Y and Z by the system  $(A, \tau_4, \delta_2)$  is the set of all their common subsequences. By extracting from this set, the elements of longest length, the <u>set of all LCS</u> of the strings X, Y, Z can be obtained. Since the set of common subsequences can be iteratively computed, by pruning the individual measures between the prefixes of the strings and retaining only the elements of maximum length, the set of LCS can also be iteratively computed.

An alternate method of computing the set of LCS of X, Y and Z is by a pairwise comparison of the three strings. We have already shown [3], that  $D^4(X,Y)$  and  $D^4(Y,Z)$  are the sets of the common subsequences between X and Y, and Y and Z respectively. Hence, the intersection of the two sets,  $D^4(X,Y)$  and  $D^4(Y,Z)$  is the set of the subsequences common to X, Y and Z. From the latter set, the set of LCS of X, Y and Z can be trivially obtained. However, the three dimensional algorithm which processes X, Y and Z simultaneously is more efficient than the latter because, at every stage in the computation, it will retain only the LCS between the prefixes of all the three strings. Strings eliminated on intersecting  $D^4(X,Y)$  and  $D^4(Y,Z)$  will not even appear in the computation. We shall conclude this section with an example of the computation of the set of LCS between three strings.

Example 1. Let X = atoms, Y = tames and Z = atm, where  $A = \{a,e,m,o,s,t\}$ . The computation of the set of LCS of X, Y and Z is given in Fig. 1. The 3-dimensional trellis is represented by a sequence of arrays parallel to the X-Y plane. In the array corresponding to  $Z_k$ , the entry  $X_i, Y_j$  is the set of LCS between  $X_i$ ,  $Y_i$  and  $Z_k$ . The details of the computation are omitted.

<sup>†</sup> As explained in Section V.2 of [3], the set of LCS of  $\{X,Y,\ldots,Z\}$  can be obtained as a special case of  $D(X,Y,\ldots,Z)$  by formalizing this extraction process using the operator (U) described in [3].

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	а	a,t	a,t	a,t	a,t
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Fig. 1: The trellis for computing the LCS of X, Y and Z of Example 1. The required set is {am,tm} in the right bottom corner.

#### IV. Conclusions

In the companion paper [3] we presented a unifying theory for similarity properties involving a pair of strings, X and Y, and which required that the order of the symbols of both the strings are preserved. In this paper, we have extended these results, and given a common basis for many of the similarity measures involving more than two strings. We have also proposed a common computational scheme to compute many numerical and nonnumerical indices of similarity involving subsequences, supersequences and shuffles of a set of strings.

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