

Partner MU-Brno
WP5 – complex queries

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Complex Queries

- Several different attributes (features) of every object
 - similarity measure for every attribute is different
- Complex queries
 - multiple simple similarity queries on attributes
 - ranks combined by an aggregation function
- Find the best matches of *circular* **shape** objects with *red* **color**
 - the best match for circular shape or red color needs not be the best match combined!!!

Complex Queries – Definition

- Assume that object $o \in D$ has m attributes (o_1, o_2, \dots, o_m)
 - every attribute is comparable by a distance function d_i
 - the value of an aggregation function t

$$t(d_1(q_1, o_1), d_2(q_2, o_2), \dots, d_m(q_m, o_m))$$

represents the “score” of object o with respect to query object q

- function t must be monotonous
- normalized similarity grades can be used instead of distances

$$x_i \in [0;1], i = 1, \dots, m$$

- represents how similar is the object o to query q in respective attribute

The \mathcal{A}_0 Algorithm

- Retrieve k top objects with respect to $q = (q_1, q_2, \dots, q_m)$
- For each attribute i
 - objects delivered in decreasing similarity to q_i
 - incrementally build sets X_i with best matches till

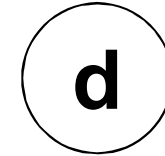
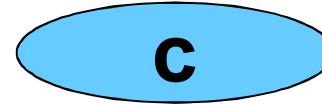
$$\forall i \mid \cap_i X_i \mid = k$$

- For all $o \in \cup_i X_i$
 - compute function $\mathbf{t}(o)$ – needs more dist. comp.
 - sort results according to values of $\mathbf{t}(o)$
 - return k first objects

Threshold Algorithm TA

- Retrieve k top objects with respect to $q = (q_1, q_2, \dots, q_m)$
- Incrementally retrieve objects in every attribute i
 - objects in decreasing similarity stored in lists X_i
- Let μ_i be the maximal grade (distance) seen in list X_i
- The **threshold value** is defined as $t(\mu_1, \mu_2, \dots, \mu_m)$
- For every object o retrieved in any list X_i
 - compute the score $t(o)$
 - if the score belongs to the best k scores seen so far
 - remember o and $t(o)$ - only first k objects are stored
- Stop if at least k objects with scores up to the threshold are found

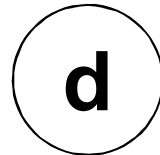
Example



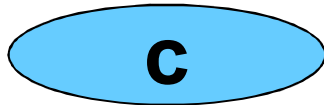
$t(o) = \text{avg}(d_1, d_2)$: $\text{avg}(3, 2) = 2.5$ $\text{avg}(1, 3) = 2$ $\text{avg}(2, 4) = 3$ $\text{avg}(4, 1) = 2.5$

X_1 list (color)

X_2 list (shape)



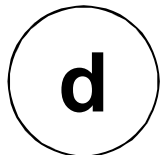
$t(\mu) = 1$



$t(\mu) = 2$



$t(\mu) = 3$



$t(\mu) = 4$

TA Properties

- Sorted access to objects for every attribute
 - index structure for every attribute
 - e.g. an index for color histograms and another one for shapes
 - incremental nearest neighbor search is needed
- Random access to objects
 - ability to compute score of object o in other attributes
 - e.g. for a particular object $o = (color, shape)$, find its color similarity $d(q_1, o_1)$ and shape similarity $d(q_2, o_2)$
- Special variants of threshold algorithm
 - restricting random accesses
 - restricting sorted accesses

Challenge for SAPIR

- Multilayer architecture of MESSI
- Expensive sorted access – incremental nearest neighbor is needed
- Efficient random access
- Minimization of the network communication costs is needed
- Inter- and intra-query parallelism tradeoff

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WP7 – social networks

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