



Algorithmic exploration of axiom spaces for efficient similarity search at large scale

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Outline

- nonmetric similarities
- indexing nonmetric similarities – related work
- motivation
 - ptolemaic indexing
- SIMDEX overview
 - main goals
 - framework stages
- preliminary experiments

Nonmetric similarities

- assuming nonmetric (unconstrained) similarity for complex measures
 - robustness (e.g., noise suppressed)
 - locality (partial matching)
 - comfort of modeling
 - domain expert not stressed by math
 - complex/algorithmic similarities undecidable

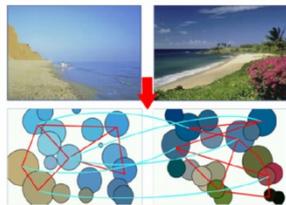
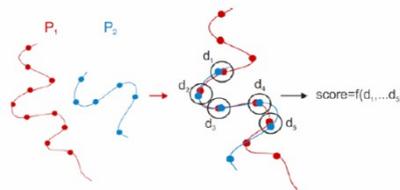
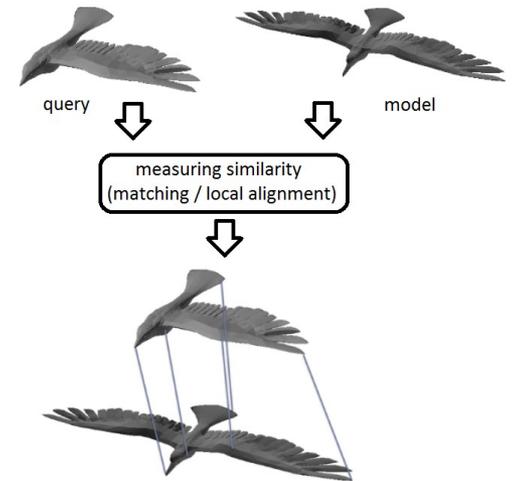
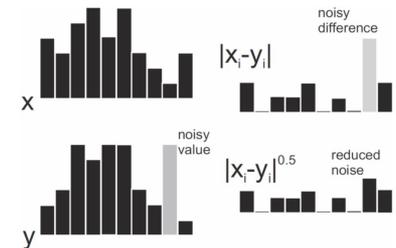
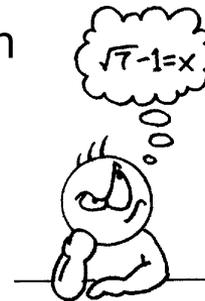


Image similarity

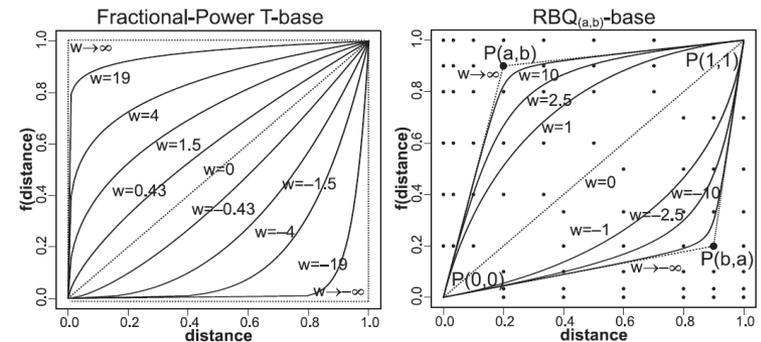


Protein similarity



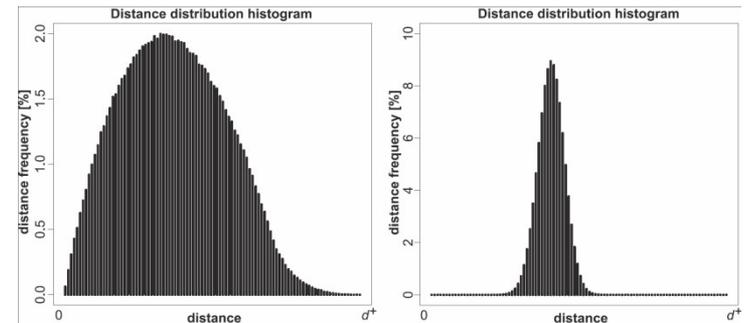
Indexing nonmetric similarities

- specific indexing (e.g., inverted index)
- general indexing
 - usually transformation into “simpler” space + indexing
 - Euclidean space + spatial access methods
 - NMDS, FastMap, MetricMap, SparseMap, BoostMap, ...
 - mapping = altering the universe + distance function
 - metric space + MAMs
 - TriGen algorithm
 - mapping = universe is the same, just the distance function altered



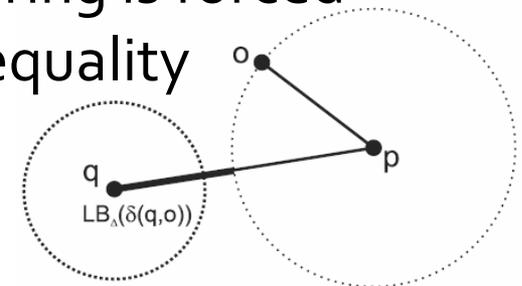
Any problem so far?

- is “metrization” of a nonmetric problem the best solution?
 - it is quite elegant solution, but the “devil lives in detail”
 - the target metric space is usually “overinflated” (high intrinsic dimensionality)
- why?



- complex behavior of a similarity measuring is forced to comply with the “stupid” triangle inequality and simple filtering

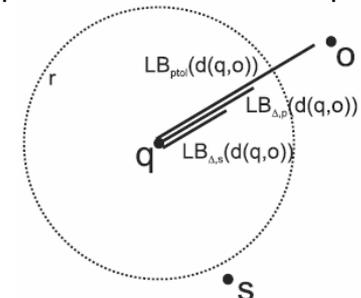
$$LB_{\Delta}(\delta(q, o_i)) = |\delta(q, p) - \delta(p, o_i)|$$



Motivation: Ptolemaic Indexing

- previous approaches
 - “rape data” to comply with an indexing formalism (metric space model)
- opposite approach
 - find an indexing formalism that comply with “data” the best
 - fuzzy similarity indexing [SISAP 2009 & 2011] – didn’t work ☹️
 - **ptolemaic indexing** [SISAP 2011] – worked! 😊
 - ptolemaic inequality instead of (together with) the triangle one
 - works with for (signature) quadratic form distances (other practical distances? open problem)

$$LB_{\text{ptol}}(\delta(q, o)) = \frac{|\delta(q, p) \cdot \delta(o, s) - \delta(q, s) \cdot \delta(o, p)|}{\delta(p, s)}$$



SIMDEX idea

- so, we have metric indexing and ptolemaic indexing
 - we have a different way to construct the lower bounds to the original distance (or upper bound to similarity)

$$LB_{\Delta}(\delta(q, o_i)) = |\delta(q, p) - \delta(p, o_i)|$$

$$LB_{\text{ptol}}(\delta(q, o)) = \frac{|\delta(q, p) \cdot \delta(o, s) - \delta(q, s) \cdot \delta(o, p)|}{\delta(p, s)}$$

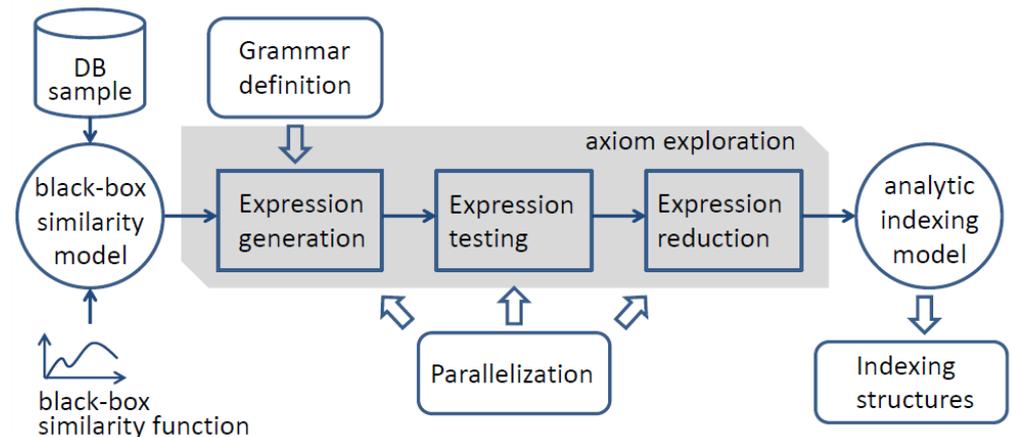
- how about to develop a framework that will discover (for a particular similarity model) an unknown **axiom**

$$LB(\delta(q, o)) = ?$$

such that the generated axiom will be computationally cheap and will perform better than any of the known (and named) axioms

SIMDEX framework

- no parameterized canonical forms but syntactically generated expressions
 - most general solution but very complex to handle
- stages
 - S₁ – grammar definition
 - S₂ – expression generation
 - S₃ – expr. testing
 - S₄ – expr. reduction
 - S₅ – indexing
 - S₆ – parallelization



SIMDEX framework

- S_1 – Grammar definition
 - used to generate right-side lowerbound expressions
 - generally L_3 /Type-3 in Chomsky hierarchy
 - however, restriction specifics turn it into context-dependent language! (next slide)
 - terminals (combined)
 - descriptor variables (q, o, p_1, \dots, p_i) and descriptor constants c_i used in the distance $\delta(\cdot, \cdot)$
 - functions f_i
 - standard arithmetic operators $+, -, *, /$, numeric constants
 - using the grammar a universe of expressions can be generated

SIMDEX framework

- S2 – Expression generation
 - exponential even when the grammar and recursion are limited
 - exploration of the expression universe
 - FIFO, LIFO, random, heuristic traversal
 - interleaved
 - restrictions complicating the language (context-dependent)
 - require $\delta(q, p_i), \delta(p_i, o)$
 - avoid $\delta(q, o)$
 - avoid duplicates (lexical but also semantics, e.g., p_i, p_j the same)
 - avoid useless arithmetic operations (e.g., $\delta(p_i, o) - \delta(p_i, o)$)

SIMDEX framework

- S_3 – Expression testing
 - testing each generated expression as an axiom candidate
 - application on the input distance/similarity matrix
 - either full axiom (all tests pass), or a partial
- S_4 – Expression reduction
 - discarding weaker expressions (producing larger lowerbounds)
 - merging a set of expressions into a compound tighter form

SIMDEX framework

- S5 – Indexing
 - verifying the real usefulness of the passed expressions
 - Pivot table-like index can be always used (direct LB filter)
 - some expressions might be interpreted as “nestable” regions in the similarity space and so applicable to hierarchical indexing
 - such as the ball-regions for triangle inequality are
- S6 – Parallelization
 - the axiom space is huge even after all the optimization stages, so massive parallelization is critical
 - multicore CPU, manycore GPU, Map-Reduce on CPU farm

SIMDEX initial implementation

- covering stages S₁-S₃
- expressions generated by heuristics (fingerprints optimization)

Algorithm 1 SIMDEX (G, C, T, S, δ)

Require: Grammar definition G , validation conditions C , threshold probability value T , database sample S , distance function δ

```
1:  $M_{\delta,S} \leftarrow$  new distance matrix ( $\delta, S$ )
2:  $expressions \leftarrow$  ExpressionGeneration( $G, C$ )
3: for all  $E_i$  in  $expressions$  do
4:   if  $validate_C(E_i)$  equals false then
5:      $expressions.Remove(E_i)$  {validity check fails}
6:     continue {skip further testing of the expression  $E_i$ }
7:   end if
8:   if  $E\_test(E_i, M_{\delta,S}) < T$  then
9:      $expressions.Remove(E_i)$  {probability test fails}
10:  end if
11: end for
12: return  $expressions$  {remaining expressions compose the result set}
```

Preliminary experiments

| Dataset | Expression | Success Ratio | MIN | MAX | AVG |
|-------------|---|---------------|--------|--------|--------|
| Corel | triangle inequality | 99 % | 0.0034 | 0.9983 | 0.3764 |
| | $ \delta(q, p) \cdot \delta(o, p) \cdot (\delta(o, p) - \delta(q, p)) $ | 100 % | 0.1059 | 0.9991 | 0.5020 |
| | $(\delta(q, p) - \delta(o, p))^2$ | 100 % | 0.1352 | 0.9999 | 0.5054 |
| | $ (\delta(q, p_1) - \delta(o, p_1))(\delta(q, p_1) - \delta(o, p_2)) $ | 100 % | 0.0420 | 0.9999 | 0.5161 |
| CoPhIR | triangle inequality | 97.5 % | 0.0021 | 0.9736 | 0.2696 |
| | $(\delta(q, p) - \delta(o, p))^2$ | 100 % | 0.0718 | 0.9979 | 0.3808 |
| | $ (\delta(q, p_1) - \delta(o, p_1))(\delta(q, p_2) - \delta(o, p_2)) $ | 100 % | 0.0845 | 0.9969 | 0.3935 |
| Ratings | triangle inequality | 100 % | 0.6067 | 1.0 | 0.9037 |
| | $\frac{1}{2 \cdot \delta(o, p)}$ | 100 % | 0.0119 | 0.5 | 0.4254 |
| | $(\delta(q, p_1) + \delta(o, p_1)) \cdot \frac{\delta(q, p_1)}{\delta(p_1, p_2)}$ | 100 % | 0.0103 | 0.5845 | 0.4254 |
| Listeria | triangle inequality | 99 % | 0 | 0.9559 | 0.1388 |
| | $\delta(p_1, p_2) \cdot \frac{1}{\delta(p_1, p_2) + \delta(o, p_2)}$ | 100 % | 0.0075 | 0.9994 | 0.2393 |
| | $\delta(q, p_1)^2 \cdot \frac{1}{\delta(o, p_2) \cdot \delta(q, p_2)}$ | 100 % | 0.0008 | 0.9985 | 0.2401 |
| | $(\delta(q, p_1) + \delta(o, p_1)) \frac{\delta(q, p_1)}{\delta(p_1, p_2)}$ | 100 % | 0.0032 | 0.9970 | 0.2555 |
| Spectometry | triangle inequality | 100 % | 0.1823 | 0.93 | 0.7329 |
| | $\delta(o, p) - \delta(o, p)^2$ | 100 % | 0.0009 | 0.8758 | 0.6638 |
| | $ (\delta(q, p_1) \cdot \delta(o, p_2)) - \delta(q, p_2)^2 $ | 100 % | 0.0148 | 0.9399 | 0.7054 |

Conclusions and future work

- SIMDEX sketched
 - universal algorithmical framework for discovering axioms suitable for indexing specific similarity models
 - breaking the metric space paradigm
- a lot of future work ahead!
 - all the stages need to be optimized

Challenges

- two challenges for the SISAP community
 - join us for developing the SIMDEX stages!
(the axiom space is really huge to search by the current unoptimized implementation)
 - answer/prove the holy grail “SIMDEX spoiler” problem:

Is the metric space model the “killer model” for general indexing, so that anything else (found by SIMDEX) is worse?

Thank you...

... for your attention!

questions?