Binary Distances Through Similarity-Dissimilarity Measures

Should we want to compare two objects U and V as far as their similarity is concerned, then in the binary case there are only four possible outcomes, which are presented in the table below:

		Object U		
	Outcome	1	0	Total
	1	а	b	a + b
Object V	0	С	d	c+d
	Total	a + c	b+d	a+b+c+d

Table 1: A 2 \times 2 Contingency table between two partitions (objects) U and V.

What we can say about these two objects is how similar or dissimilar they are. If they share the same number of ones and zeroes (or α and d in the table above), then they can be considered similar to each other, if however they do not, then these two objects are characterized as dissimilar. Here are some similarity measures that use the table above (Seung-Seok Choi, Sung-Hyuk Cha, Charles C. Tappert, 2009):

Similarity-Dissimilarity Measures		
Similarity Measure	Formula	
Jaccard or Tanimoto	$\frac{a}{a+b+c}$	
Sørensen–Dice or Czekanowski or Nei & Li	$\frac{2a}{2a+b+c}$	
3W - Jaccard	$\frac{3a}{3a+b+c}$	
Simple Matching or Sokal & Michener	$\frac{a+d}{a+b+c+d}$	
Sokal & Sneath I	$\frac{a}{a+2(b+c)}$	
Sokal & Sneath II	$\frac{2(a+d)}{2a+b+c+2d}$	
Sokal & Sneath III	$\frac{a+d}{b+c}$	

Sokal & Sneath IV	$\frac{1}{4} \left(\frac{a}{a+b} + \frac{a}{a+c} + \frac{d}{b+d} + \frac{d}{c+d} \right)$
Sokal & Sneath V or Ochiai II	$\frac{ad}{\sqrt{(a+b)(a+c)(b+d)(c+d)}}$
Rogers & Tanimoto	$\frac{a+d}{a+2(b+c)+d}$
Faith	$\frac{a+0.5d}{a+b+c+d}$
Gower & Legendre	$\frac{a+d}{a+0.5(b+c)+d}$
Russell & Rao	$\frac{a}{a+b+c+d}$
Ochiai I or Cosine or Otsuka-Ochiai or Fowlkes-Mallows Index	$\frac{a}{\sqrt{(a+b)(a+c)}}$
Forbes I	$\frac{na}{(a+b)(a+c)}$
Forbes II	$\frac{na - (a+b)(a+c)}{n[min(a+b,a+c) - (a+b)(a+c)]}$
Fossum	$\frac{n(a-0.5)^2}{(a+b)(a+c)}$
Sorgenfrei	$\frac{a^2}{(a+b)(a+c)}$
Mountford	$\frac{a}{0.5(ab+ac)+bc}$
McConnaughey	$\frac{a^2 - bc}{(a+b)(a+c)}$

Tarwid	$\frac{na - (a+b)(a+c)}{na + (a+b)(a+c)}$
Kulczynski I	$\frac{a}{b+c}$
Kulczynski II or Driver & Kroeber	$\frac{1}{2} \left(\frac{a}{a+b} + \frac{a}{a+c} \right)$
Johnson	$\frac{a}{a+b} + \frac{a}{a+c}$
Dennis	$\frac{ad - bc}{\sqrt{n(a+b)(a+c)}}$
Simpson or Szymkiewicz-Simpson	$\frac{a}{min(a+b,a+c)}$
Braun-Blanquet	$\frac{a}{max(a+b,a+c)}$
Fager & McGowan	$\frac{a}{\sqrt{(a+b)(a+c)}} - \frac{max(a+b,a+c)}{2}$
Gower	$\frac{a+d}{\sqrt{(a+b)(a+c)(b+d)(c+d)}}$
Pearson I	$\frac{n(ad-bc)^2}{(a+b)(a+c)(b+d)(c+d)}$
Pearson II	$\sqrt{\frac{\frac{n(ad-bc)^{2}}{(a+b)(a+c)(b+d)(c+d)}}{n+\frac{n(ad-bc)^{2}}{(a+b)(a+c)(b+d)(c+d)}}}$

Pearson III	$\sqrt{\frac{ad - bc}{\sqrt{(a+b)(a+c)(b+d)(c+d)}}}$ $\sqrt{n + \frac{ad - bc}{\sqrt{(a+b)(a+c)(b+d)(c+d)}}}$
Pearson & Heron I	$\frac{ad - bc}{\sqrt{(a+b)(a+c)(b+d)(c+d)}}$
Pearson & Heron II	$cos\left(\frac{\pi\sqrt{bc}}{\sqrt{ad}+\sqrt{bc}}\right)$
Cole	$\frac{\sqrt{2}(ad-bc)}{\sqrt{(ad-bc)^2-(a+b)(a+c)(b+d)(c+d)}}$
Stiles	$log_{10}\left(\frac{n(ad-bc -n/2)^{2}}{(a+b)(a+c)(b+d)(c+d)}\right)$
Yule's Q	$\frac{ad - bc}{ad + bc}$
Yule's ω or Yule's Y	$\frac{\sqrt{ad} - \sqrt{bc}}{\sqrt{ad} + \sqrt{bc}}$
Dispersion	$\frac{ad - bc}{(a+b+c+d)^2}$
Hamann	$\frac{(a+d)-(b+c)}{a+b+c+d}$
McEwen & Michael	$\frac{4(ad - bc)}{(a+d)^2 + (b+c)^2}$
Goodman & Kruskal's Lambda	$\frac{\sigma - \sigma'}{2n - \sigma'}$ with: $\sigma = max(a, b) + max(a, c) + max(b, d) + max(c, d)$ $\sigma' = max(a + c, b + d) + max(a + b, c + d)$

Anderberg	$\frac{\sigma - \sigma'}{2n}$ with: $\sigma = max(a,b) + max(a,c) + max(b,d) + max(c,d)$ $\sigma' = max(a+c,b+d) + max(a+b,c+d)$
Baroni-Urbani & Buser I	$\frac{\sqrt{ad} + a}{\sqrt{ad} + a + b + c}$
Baroni-Urbani & Buser II	$\frac{\sqrt{ad} + a - (b+c)}{\sqrt{ad} + a + b + c}$
Peirce	$\frac{ad - bc}{(a+b)(c+d)}$
Eyraud	$\frac{a-(a+b)(a+c)}{(a+b)(a+c)(b+d)(c+d)}$
Tarantula	$\frac{\frac{a}{a+b}}{\frac{a}{a+b} + \frac{c}{c+d}}$

<u>Table 2:</u> A table that contains all the similarity measures that will be calculated, for binary data.

The distance is calculated in the following manner:

$$Distance = 1 - Similarity Measure$$

How one picks a suitable distance depends on how they want to treat the zero-zero similarity matches (or d in the table above), as well as the dissimilarities between the objects (meaning b and c in the table above).

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