

List of other possible dissimilarities

To compare histograms $H = (h_1, h_2, \dots, h_n)$ and $K = (k_1, k_2, \dots, k_n)$

Bin-by-bin dissimilarity measures

Only pairs of bins that have the same index are matched – dissimilarity is a combination of all the pairwise differences:

L_1 distance $d_{L_1}(H, K) = \sum_i |h_i - k_i|$

Kullback-Leibler divergence $d_{KL}(H, K) = \sum_i h_i \log \frac{h_i}{k_i}$

Jeffrey divergence

$$d_J(H, K) = \sum_i \left(h_i \log \frac{h_i}{m_i} + k_i \log \frac{k_i}{m_i} \right), \quad m_i = \frac{h_i + k_i}{2}$$

Aitchison distance $d_A^2(H, K) = \sum_i \left(\log \frac{h_i}{g(H)} - \log \frac{k_i}{g(K)} \right)^2$
 $g(H), g(K)$ - the geometric mean.

Cross-bin dissimilarity measures

Information on a ground distance are incorporated.

Qadratic-form distance $d_S^2(H, K) = (H - K)^T S (H - K)$

similarity matrix

$$S = \left[1 - \frac{d_{ij}}{d_{\max}} \right], \quad d_{ij} = |i - j|, \quad d_{\max} = \max_{ij} d_{ij}$$

$$S = \begin{bmatrix} 1 & 1 - \frac{1}{4} & 1 - \frac{2}{4} & 1 - \frac{3}{4} & 0 \\ 1 - \frac{1}{4} & 1 & 1 - \frac{1}{4} & 1 - \frac{2}{4} & 1 - \frac{3}{4} \\ 1 - \frac{2}{4} & 1 - \frac{1}{4} & 1 & 1 - \frac{1}{4} & 1 - \frac{2}{4} \\ 1 - \frac{3}{4} & 1 - \frac{2}{4} & 1 - \frac{1}{4} & 1 & 1 - \frac{1}{4} \\ 0 & 1 - \frac{3}{4} & 1 - \frac{2}{4} & 1 - \frac{1}{4} & 1 \end{bmatrix} = \frac{1}{4} \begin{bmatrix} 4 & 3 & 2 & 1 & 0 \\ 3 & 4 & 3 & 2 & 1 \\ 2 & 3 & 4 & 3 & 2 \\ 1 & 2 & 3 & 4 & 3 \\ 0 & 1 & 2 & 3 & 4 \end{bmatrix}$$

Match distance

$$d_M(H, K) = \sum_i |\hat{h}_i - \hat{k}_i|$$

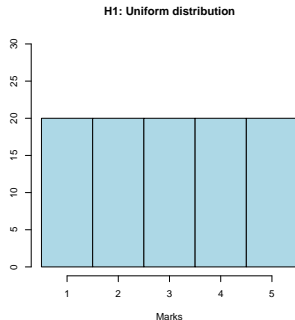
$\hat{h}_i = \sum_{j \leq i} h_j$ the cumulative histogram of H

$\hat{k}_i = \sum_{j \leq i} k_j$ the cumulative histogram of K

This distance represents a special case of the Earth Mover's Distance (EMD) for one-dimensional histograms with equal areas.

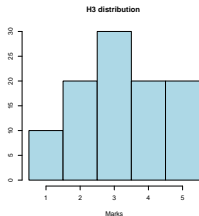
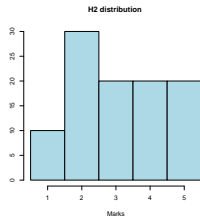
Test data sets and perceptual dissimilarity

$$H1 = (20, 20, 20, 20, 20)$$

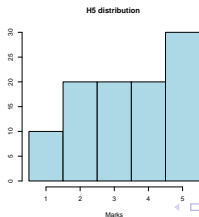
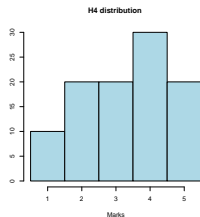


Test data sets and perceptual dissimilarity

$$H2 = (10, 30, 20, 20, 20) \quad H3 = (10, 20, 30, 20, 20)$$



$$H4 = (10, 20, 20, 30, 20) \quad H5 = (10, 20, 20, 20, 30)$$



Perceptual dissimilarities are **increasing**

$$d(H1, H2) < d(H1, H3) < d(H1, H4) < d(H1, H5) \quad (1)$$

$$d(H2, H3) < d(H2, H4) < d(H2, H5) \quad (2)$$

$$d(H3, H4) < d(H3, H5) \quad (3)$$

Dissimilarity matrices

L_1 distance is not increasing:

d_{L_1}	H2	H3	H4	H5
H1	20	20	20	20
H2		20	20	20
H3			20	20
H4				20

Dissimilarity matrices

Kullback-Leibler divergence is not increasing:

d_{KL}	H2	H3	H4	H5
H1	5.75	5.75	5.75	5.75
H2		4.05	4.05	4.05
H3			4.05	4.05
H4				4.05

Dissimilarity matrices

Quadratic form distance is increasing:

d_S^2	H2	H3	H4	H5
H1	50	100	150	200
H2		50	100	150
H3			50	100
H4				50

Match distance is increasing:

d_M	H2	H3	H4	H5
H1	10	20	30	40
H2		10	20	30
H3			20	30
H4				10

Dissimilarity matrices

Directed dissimilarity is increasing:

<i>D</i>	H2	H3	H4	H5
H1	0.04	0.08	0.012	0.16
H2		0.05	0.09	0.13
H3			0.05	0.09
H4				0.05

Discussion and Suggestions

Quadratic form distance, Match distance and Directed dissimilarity are increasing in accordance with perceptual dissimilarities.

Further work:

- To select the most appropriate dissimilarity by applying other test sets of distributions;
- To perform comparison of achievements through grading (school vs. national grades, teachers vs. national grades, teacher vs. teacher in the same subject in a school, subject vs. subject . . .) with selected dissimilarity and adequate classification method and
- to define school's, teacher's, student's or class *profile*.